

Doctoral thesis

Doctoral theses at NTNU, 2021:107

Tanja Kofod Petersen

Biodiversity dynamics in urban areas under changing land-uses

NTNU
Norwegian University of Science and Technology
Thesis for the Degree of
Philosophiae Doctor
Faculty of Natural Sciences
Department of Biology



Norwegian University of
Science and Technology

Tanja Kofod Petersen

Biodiversity dynamics in urban areas under changing land-uses

Thesis for the Degree of Philosophiae Doctor

Trondheim, April 2021

Norwegian University of Science and Technology
Faculty of Natural Sciences
Department of Biology



Norwegian University of
Science and Technology

NTNU

Norwegian University of Science and Technology

Thesis for the Degree of Philosophiae Doctor

Faculty of Natural Sciences

Department of Biology

© Tanja Kofod Petersen

ISBN 978-82-326-5218-1 (printed ver.)

ISBN 978-82-326-5431-4 (electronic ver.)

ISSN 1503-8181 (printed ver.)

ISSN 2703-8084 (online ver.)

Doctoral theses at NTNU, 2021:107

Printed by NTNU Grafisk senter

“Ultimately we need to recognize that while humans continue to build urban landscapes, we share these spaces with other species.”

- David Suzuki -

Acknowledgements

“Thank you” is an insufficient phrase for expressing how grateful I am to all the people, who have helped me reach this point. Accepting the unexpected offer to complete a doctoral degree, migrating north and settling in an unfamiliar country – all of this would have been a herculean task without the immense support I have been so lucky to receive. Regrettably, I will most definitely have forgotten someone in this inadequate acknowledgements section – rest assure that if you are reading this, you are likely on the long list of people without whom I could not have done this!

Specifically, I want to thank my supervisors for all of their encouragement, guidance and productive discussions during my PhD studies. I am grateful for the great collaborations, their assistance and enthusiasm during the work and writing of the papers included in this thesis. I am indebted to Professor **Gunnar Austrheim**, for being the calm, familiar support I needed, for his kindness and encouragement when the project seemed too overwhelming, and for sparking my interest in the societal relevance of ecology and biodiversity management. Associate Professor **James D. M. Speed**, for his indispensable help with statistical issues, meticulous reading and correction of my English grammar – for making sure that I strived for a higher standard in my work than what my own ambitions allowed, and for involving me in other scientific projects. Associate Professor **Vidar Grøtan**, for his help with the statistical modelling, and for being my link to the Centre for Biodiversity Dynamics.

I am grateful to the employees of the Trondheim Municipality for providing data on present and past land-cover, the fruitful discussions and feedback I have received on the project, and their general assistance and interest in the project. Thank you to my colleagues at the NTNU University Museum for provision of and assistance with additional data.

I want to thank my family for their invaluable support throughout my education and subsequent emigration – you may not have understood all of my “strange” interests, or have agreed with all of my choices and opinions, but you have held my hands through every step nevertheless – I hope that I have made you proud. To Anika: thank you for keeping me

entertained with an endless stream of funny pictures, and promises of dessert-fuelled sister-dates; I am looking forward to more of those! To my late grandfather: thank you for (not so) gently nudging me down the path of biology.

Lastly, but certainly not least, I want to thank my friends and colleagues from the Department of Natural History – I cannot express how invaluable and inspirational you have been and still are to me. Thank you for making the years of my PhD joyful, and for keeping my head up and my back straight when the world was too overwhelming. Moreover, for adding to my collections of funny shirts! The last few years have brought countless fantastic experiences, new friends, interests and hobbies, and a newfound love and appreciation of nature and outdoor experiences – I will always think fondly of all the amazing people who have been part of these adventures.

To my friends, instructors and teammates at “*Trondheim Martial Arts Academy – Kickboxing*”: you have taught me to fight back, in both a figurative and a literal sense.

To each and every one of you: KAMELÅSÅ!

Contents

List of papers.....	vii
Author contributions	vii
Introduction	1
Urban ecology.....	1
Native vs. alien species in the city.....	2
Spatial scale.....	3
Species occurrence data	4
Biodiversity and the Sustainable Development Goals	5
Aims	7
General methods	8
Summary of the main results	14
Discussion	17
Not all datasets are created equal	17
Where the wild things are.....	18
For whom the bell tolls.....	19
Fantastic data and where to find them	20
The crux.....	22
Conclusions	24
Synthesis and future perspectives.....	26
References	31
Appended papers.....	I
Paper I: <i>Species data for understanding biodiversity dynamics: The What, Where and When of species occurrence data collection</i>	I
Paper II: <i>Urban aliens and threatened near-naturals: Land-cover affects the species richness of alien- and threatened species in an urban-rural setting</i>	LXXXI
Paper III: <i>Competitors and ruderals go to town: Plant community composition and function along an urbanisation gradient</i>	CV
Paper IV: <i>Urbanisation and land-cover change affect the functional turnover of bird communities but not the extent of species composition change</i>	CXLIX

List of papers

The doctoral thesis is based on the following papers. These will hereafter be referred to by their Roman numerals:

- I. **Petersen, T.K.**, Speed, J. D. M., Grøtan, V., Austrheim, G. (2021). „Species data for understanding biodiversity dynamics: The What, Where and When of species occurrence data collection“. *Ecological Solutions and Evidence*, 2 (1).
- II. **Petersen, T.K.**, Speed, J. D. M., Grøtan, V., Austrheim, G. (2020). „Urban aliens and threatened near-naturals: Land-cover affects the species richness of alien- and threatened species in an urban-rural setting“. *Scientific Reports*, 10 (8513).
- III. **Petersen, T.K.**, Speed, J. D. M., Grøtan, V., Austrheim, G. (*in press*). „Competitors and ruderals go to town: Plant community composition and function along an urbanisation gradient“. *Nordic Journal of Botany*.
- IV. **Petersen, T.K.**, Speed, J. D. M., Grøtan, V., Frøyen, Y.K., Austrheim, G. (submitted). „Urbanisation and land-cover change affect the functional turnover of bird communities but not the extent of species composition change“. Submitted manuscript to *Journal of Urban Ecology*.

Author contributions

Tanja Kofod Petersen (**TKP**), Gunnar Austrheim (**GA**), James D. M. Speed (**JDMS**), Vidar Grøtan (**VG**), Yngve K. Frøyen (**YKF**). All authors have given their consent to use their work in this thesis.

Paper I, II & III: **TKP**, **GA**, **JDMS** and **VG** conceived the ideas and designed the methodologies; **TKP** retrieved and analysed the data; **TKP** wrote the first draft of the manuscript. All authors contributed critically to the drafts and gave final approval for publication.

Paper IV: **TKP**, GA, JDMS and VG conceived the idea and designed the methodology; **TKP** and YKF retrieved the data, and **TKP** analysed conducted the analyses; **TKP** wrote the first draft of the manuscript. All authors contributed critically to the drafts and gave final approval for publication.

Funding

NTNU Sustainability and the NTNU University Museum, Department of Natural History, funded this PhD-thesis.

Introduction

Since the dawn of civilisation, humans have altered the physical environment to accommodate their own needs; pristine areas have been converted to agricultural land, cities, and roads. In the past decades, especially urbanisation has increased due to a growing human population (IPBES 2019). In addition, the majority of the world's population now lives in cities, and both further population growth and urbanisation are predicted in the future (United Nations 2018). The growth of cities inevitably happens as the cost of other habitats – the homes to numerous other organisms. Over the last decades, the appreciation of the importance of biodiversity as the basis of ecosystem function and services has increased, as has the understanding of the significance of nature's contributions to people (also known as Ecosystem Services) (IPBES 2018). How to both accommodate the increasing human population and to curb deterioration of biodiversity is an issue of prime importance now and in the immediate future.

Urban ecology

Historically, “urban ecology” has been treated as an oxymoron, and in attempts to

study “natural” processes and patterns, researchers actively avoided urban areas (Forman 2014a, Salomon Cavin and Kull 2017). This approach largely adhered to a worldview in which large-scale disturbances (whether they were natural or man-made) were thought to have little long-term impact on a natural system “in balance”; humans were treated as external to biological systems (McDonnell 2011). With growing documentation of anthropogenic impacts on the planet (such as the rise in atmospheric CO₂ levels first published in the early 1960s) and increasing urbanisation, it was acknowledged that such a “humans-outside-of-ecosystems” worldview might hamper our understanding of the ecosphere. With the ever-increasing presence of people, human settlements were recognised as valid foci of ecological studies. From the early 1970s, recognition of the vital part ecology could play in a human context emerged, and a multidisciplinary field reminiscent of current urban ecology budded (McDonnell 2011, Salomon Cavin and Kull 2017, Sukopp 2008).

With the increasing appreciation of urban ecology as an autonomous research

field, various biological responses to urbanisation have been recognised, depending on taxonomic-, spatial- and temporal scale (Aronson et al. 2014). Human actions govern biodiversity patterns in urban areas both directly and indirectly; directly through habitat loss and fragmentation, and by introduction of new species; indirectly by changing climate and physiochemical conditions (Kowarik 2011). Through alterations of the physiochemical environment, conditions are made potentially inhospitable for native, locally adapted species, but suitable for generalist and/or alien species. Urban areas thus have a distinct biogeochemistry compared to non-urban areas, driven by human actions (Kaye et al. 2006). Specifically, the urban environment is drier, warmer, more nutrient-rich, polluted and more frequently disturbed compared to more natural counterparts (Forman 2014b, 2014c, Oke 1988, Pellissier et al. 2008). In general, urbanisation can be thought of as a homogenising force on biodiversity (McKinney 2006).

Native vs. alien species in the city

Alien species are frequently introduced to the urban environment (Padayachee et al. 2017). “Alien species” here refers to species (regardless of taxon) which have been introduced to areas outside of their natural range through human action, either deliberately (for example through import of ornamental species and pets), or through “hitch-hiking” (IUCN 2000). Alien species may initially increase local species richness (α diversity¹), but on a larger, (near-)global scale, cities across biomes become more alike – a decrease in β diversity² (McKinney 2006). Alien species are increasingly recognised as drivers of native species extinctions. As the very definition of alien species includes human activity, the link between introductions and cities is obvious. Especially trade, traffic and horticulture are important sources of alien species (Padayachee et al. 2017); urban areas are junctions of all of these. Urban areas thus serve as hotspots for introductions of alien species, and they are particularly susceptible to invasions (Gaertner et al. 2017,

¹ α diversity: species richness/diversity of a local community, as introduced by Whittaker (1972).

² β diversity: heterogeneity in composition between communities.

van Ham et al. 2013, Padayachee et al. 2017, Salomon Cavin and Kull 2017).

One might be led to believe that cities are depauperate concerning native species, but this is far from the truth. On broad spatial scales, a positive relationship between human population size and biodiversity is seen (Gaston 2005). Cities (and other densely populated areas) can be surprisingly species rich (Cincotta et al. 2000, Gaertner et al. 2017); in some studies, urban areas have even harboured more species than non-urban equivalents, also regarding native and threatened species (Ives et al. 2016, Kühn et al. 2004). It is hypothesised that this is the case as cities generally have been established in productive, species-rich areas – high levels of native biodiversity are thus not necessarily *because of* the urban conditions, but rather *despite of* them (Gaston 2005, Kowarik 2011, Kühn et al. 2004, McKinney 2008). This results in potentially high numbers of native species (including threatened ones) in areas characterised by rapid change, disturbance, and high pressure from alien species – a situation calling for concern. In the worst case, this means that native species inhabiting cities

are in reality “living dead” with an extinction debt to pay.

Spatial scale

Studies of biodiversity in an urban context have been performed at multiple scales, including large-scale global studies (Aronson et al. 2014) and single-city ones (Turrini and Knop 2015). Despite the general importance of large-scale studies (e.g. on country- to global level), such investigations are not necessarily appropriate for conservation action, as local management decisions are more often made on a city/county/municipality-level (Luck 2007). Likewise, whereas a positive correlation between species richness and human population density can be seen at large spatial scales, this might not be the case at finer ones (Pautasso 2007); the same mechanisms are not determining large-scale and fine-scale ecological patterns (Gaston 2000, Münkemüller et al. 2014, Uchida et al. 2020). Thus, studies covering the same spatial extent as the one on which actual management decisions are made are needed to inform authorities on the best ways forward regarding biodiversity conservation.

Species occurrence data

To make any statement on the distribution of species in space and time, data on their whereabouts are obviously needed. Scarcity of data is a constant obstacle to overcome, both in space and time (Magurran et al. 2019). Only an estimated 10% of the World's biocollections are available in digital form (Ball-Damerow et al. 2019).

In the past decades, the amounts and availability of species occurrence data have increased immensely (Amano et al. 2016, Gaiji et al. 2013), in no small part due to technological advancements (Powney and Isaac 2015). With the advent of the World-wide Web, handheld GPS's, smartphones and so on, the quantity, quality and accuracy of species occurrence records have increased vastly compared to the early days of biogeographical surveys. These technological advancements have furthermore made species registrations much more attainable for non-professionals, and have thus allowed for an extensive inclusion of "citizen science" in research (August et al. 2015). Likewise, data on species occurrences are increasingly compiled in online portals and databases (Amano et al. 2016, Ball-Damerow et al. 2019). Such

portals (the main example being the Global Biodiversity Information Facility, GBIF (GBIF.org 2019, Newbold 2010)) serve as a hodgepodge of different datasets and data types with vastly different origins, scopes and utilities (Speed et al. 2018). A recognised problem with such occurrence records is variation in sampling effort; not all areas nor taxa are investigated to the same degree, neither are the available data evenly spread out in time (Newbold 2010, Powney and Isaac 2015, Tiago et al. 2017). Especially with the increased participation of citizen scientists, taxonomic and spatiotemporal skews can be extensive, based on individual collectors' behaviours and motivations (Boakes et al. 2016). Regardless, these open portals are invaluable sources of spatiotemporal data which would have been otherwise impossible to obtain (Miller-Rushing et al. 2012). Knowledge of the nature of these biases is crucial for developing strategies and methods to account for the issues – this is a field under- and in need of rapid development (Robinson et al. 2018).

Biodiversity and the Sustainable Development Goals

All of the points raised here prove even more crucial in the light of the United Nations' Sustainable Development Goals (SDG) (United Nations 2020b) (Figure 1), and the "Cities and Biodiversity Outlook" by the Convention on Biological Diversity (CBD) (CBD 2012). The SDG include specific targets concerning "Sustainable Cities and Communities" (target 11), and "Life on

land" (target 15). In relation to this thesis, three sub-targets of target 15 stand out:

"Take urgent and significant action to reduce the degradation of natural habitats, halt the loss of biodiversity and, by 2020, protect and prevent the extinction of threatened species" (target 15.5),

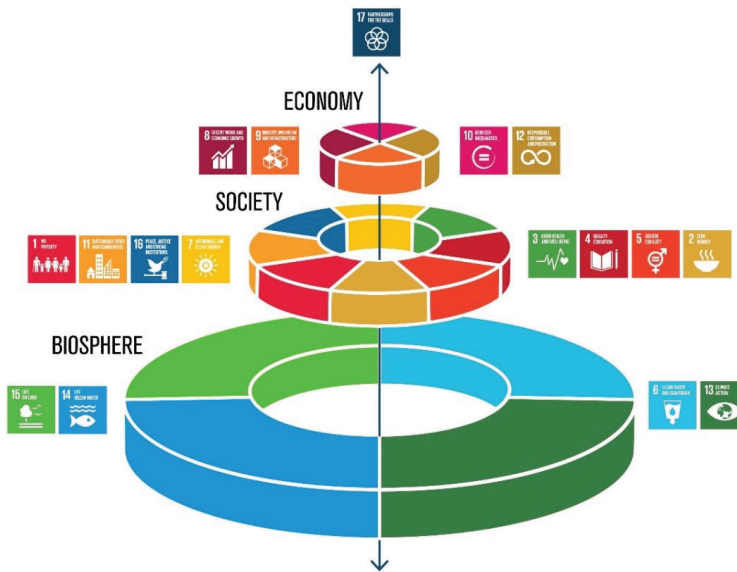


Figure 1. Conceptual figure from Folke et al. (2016) illustrating the interdependence of the United Nation's Sustainable Development Goals. The goals concerning economy and societal aspects are nested within- and are dependent upon sustainable use and management of the biosphere. Illustration: Azote for Stockholm Resilience Centre, Stockholm University.

“By 2020, introduce measures to prevent the introduction and significantly reduce the impact of invasive alien species on land and water ecosystems and control or eradicate the priority species?” (target 15.8),

“By 2020, integrate ecosystem and biodiversity values into national and local planning, development processes, poverty reduction strategies and accounts?” (target 15.9).

Neither of these stated targets have been achieved so far, nor are they on the right track to be so (United Nations 2020a). Along the same lines, the 2020 Aichi Biodiversity Targets set by the CBD were not

reached either (Brondízio et al. 2019). It is clear that none of these targets can be met without involving city development and biodiversity management within urban boundaries. Cities do not only present issues in this regard, but can also provide opportunities for innovative ways to deal with the relationship between humans and nature (Elmqvist et al. 2013, 2018). To further aid in the completion of the Sustainable Development Goals and the Aichi Targets, and to ensure sustainable use and management of the planets’ biological diversity on both global and local levels, understanding in detail how urbanisation and other anthropogenic conversions of land affects other ecosystems and other species is paramount.

Aims

The overall aim of this thesis is to expand our knowledge on the spatiotemporal patterns in biodiversity in urban areas. As the world has experienced increased urbanisation during the last decades, and continue to do so, detailed understanding of how this process affects biodiversity is crucial. As investigations spanning both space and time requires large amounts of (high quality) data, the scope of this thesis is twofold:

- 1) To evaluate the availability and quality of species occurrence data within Norway and the city of Trondheim, and assess the utility of this data in biodiversity modelling;
- 2) Investigate how land-use and land-use change affects biodiversity. Particularly the effects of urbanisation were of interest, as were the responses of threatened and/or alien species.

With the studies included in this thesis, I thus aim to assess the availability of data on species occurrences and distributions, examine effects of urbanisation on biodiversity,

and provide suggestions on how to develop cities and manage biodiversity within them in a more sustainable fashion. All studies were performed within Norway, most of them limited to Trondheim Municipality. To reach the stated goals, I here attempt to answer the following specific questions:

- I. What is the status of openly available species occurrence data in the study area? (**Paper I**)
- II. What characterises biodiversity hotspots on an urban-suburban/municipality scale? (**Paper II & III**)
- III. What are the most important drivers determining the distribution of threatened and alien species? (**Paper II**)
- IV. What are the effects of land-use and land-use change on biodiversity, on threatened species and on alien species? (**Paper II, III & IV**)

General methods

The following is a general description and discussion of the methods used in the four papers included in this thesis. Please refer to the “Methods” sections of the individual, appended papers (page I-CCVII) for detailed descriptions.

Study area

For **paper II, III and IV**, Trondheim municipality served as the study system; for **paper I**, the study area was expanded to encompass all of Norway (Figure 2). Mainland Norway comprises the north-western part of

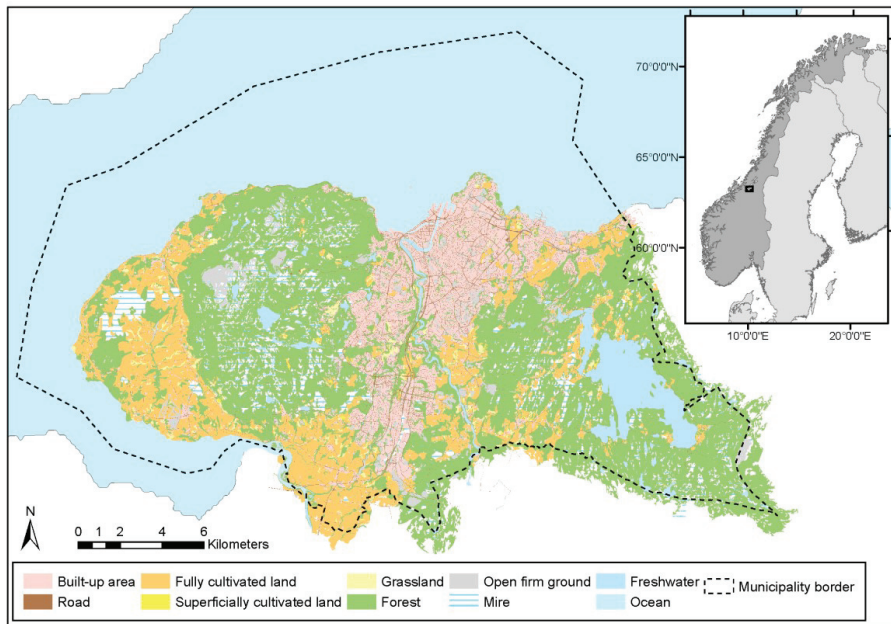


Figure 2. Map of the study area(s) included in all papers. Norway is highlighted in dark grey (upper right inset), and the position of Trondheim Municipality is indicated by a black square. The administrative border of the municipality is illustrated with dashed line, and colours delineate different land-cover types (Norwegian Institute of Bioeconomy Research 2018). Land-cover of the areas surrounding Trondheim has not been included. See **paper I-IV** for more details.

the Scandinavian Peninsula. The country covers $\approx 323\,800\text{ km}^2$, of which approximately 1.7% is built-up area, and it inhabits ca. 5.4 million people (Statistics Norway 2020). Trondheim³ is a southern-boreal, coastal municipality, located around 63.42°N , 10.38°E , and it covers approximately 342 km^2 . It is the third-most populated municipality in Norway, with approximately 195 000 inhabitants (Statistics Norway 2020). Note that these numbers are from prior to the merger with Klæbu municipality pr. 01.01.2020. In the past decades the municipality has experienced a population increase with subsequent conversions of land-use. The municipality contains large variation in habitats, ranging from industrial- and highly urbanised areas to near-natural forests and mires, and thus presents both extremes of an urbanisation gradient (Trondheim Municipality 2020). In addition, due to the long-term presence of academic institutions such as The Royal Norwegian Society of Sciences and Letters (DKNVS 1760), the university (now named NTNU), the NTNU University Museum, and other research- and

management institutions working with biodiversity-related issues (such as NINA (2021) and the Norwegian Biodiversity Information Centre (2021)), Trondheim is fairly well-sampled regarding species occurrence records. Taken together, these circumstances make Trondheim an ideal place to study the effects of urbanisation and land-cover change on biodiversity.

Data

Data on land-cover were based on current and previous land-cover/land-use maps. As we operated on multiple temporal- and spatial scales, several datasets were included, but in all cases the land-cover categories were coded based on land- and tree cover type, timber productivity and soil condition. The outdated DMK (“Digitalt Markslagskart”) map was used in **paper III**, previous and updated versions of the AR5 map were used in **paper II, III** and **IV**, and the AR50 map were used in **paper I** (Norwegian Institute of Bioeconomy Research 2018, 2019b, 2019a).

³ For these studies, Trondheim is defined by the municipal administrative border prior to the merger with Klæbu municipality pr.

01.01.2020. All statistics regarding Trondheim municipality are thus from before this date.

Paper I-IV all included species occurrence data downloaded from the Global Biodiversity Information Facility (GBIF.org 2019). For each paper, different data cleaning steps were applied, generally relating to ensuring data quality, taxonomic- and spatial resolution. DOI's for all downloaded datasets are referenced in the respective papers. For **paper III**, an additional dataset from the NTNU University Museum was included, presenting a structured vegetation survey of 100×100 m² vegetation plots along an urbanisation gradient within Trondheim. The full dataset is available through a GBIF public repository (see **paper III** for reference).

A better mechanistic understanding of species' distributions in urban areas, and thus their responses to anthropogenic pressures and land-use changes, can be obtained by examining the functional characteristics and diversity of biological systems rather than focussing on species identities *per se*. In **paper III** and **IV**, we included relevant functional traits of respectively plants and birds to assess the functional responses of the investigated communities (Figure 3). The data were retrieved from various databases,

depending on taxon (Billerman et al. 2020, Kattge et al. 2011, Myhrvold et al. 2015, Wilman et al. 2014).

Assessment of spatial and taxonomic biases in species occurrence data

As has been pointed out by several other authors (e.g. Newbold (2010) and Speed et al. (2018)), species occurrence records from databases compiling multiple datasets (including both preserved specimens, structured observational surveys and opportunistic recordings) brings inherent biases and challenges. To address these issues, we examined the status and potential biases of GBIF records in Norway (**paper I**). We studied a subset of ten individual datasets from GBIF differing in origin and thus characteristics. These characteristics included (but were not limited to) taxonomic focus, methodology, number- and skill-level of the reporters. We examined multiple sources of potential skew in the data: taxonomic differences, the distribution of red-listed- and alien species within and between the datasets, temporal patterns, and geographic biases. The potential geographic biases were assessed by repeatedly simulating random distributions of occur-

rence records across Norway, and comparing these to the distribution of the GBIF data. Dataset name and conservation status of the simulated points were randomly assigned in the same proportions as in the observed dataset. Generalised Linear Models (GLMs) were fitted to the simulated points, providing the expected ranges of number of records falling within each land-cover type.

Predicting species richness by land-cover

In **paper II**, we examined the influence of land-cover variables on species richness patterns among different taxonomic groups, and groups of conservation concern (red-listed and alien species). We constructed spatially dependent Generalised Linear Mixed Effects Models (GLMMs) predicting the total-, threatened- and alien number of species within 500×500 m² grid cells across the municipality by habitat, habitat heterogeneity and slope aspect. All grid cells were assigned to habitat categories based on hierarchical cluster analysis of the area of each AR5 land-cover category within the grid cells; thus, the dominating land-cover classes within each grid cell determined “habitat”. We identified which variables and variable interactions

were able to predict species richness by backwards stepwise model selection based on ΔAIC_c .

Effects of land-cover variables on functional diversity

To examine a more mechanistic relationship between biological communities and (anthropogenic) land-cover, we assessed the differences in functional traits of plants along an urbanisation gradient in **paper III** (Figure 3). We analysed a vegetation survey performed in fifty 100×100 m² plots along a suggested urbanisation gradient in Trondheim in 2001-2002. The plots were grouped based on plant community composition using hierarchical cluster analysis, and the differences in slope aspect, age of built structures, proportion of developed area and area of multi-layered forest within the plots were assessed. Rather than assessing the distribution of functional traits of all species recorded in the vegetation plots, we performed Indicator Species Analysis to identify species characteristic of different urbanisation levels, and contrasted functional trait values of the Indicator Species.

To evaluate the identified Indicator Species and their functional responses, we compared

the observed patterns with an independent GBIF dataset. In order to evaluate the status of the Indicator Species, we constructed logistic models estimating the probability of presence of an urban- or non-urban indicator species based on the percentage of developed area in 500×500 m² grid cells (i.e. a proxy for urbanisation level). We further aimed to evaluate the differences in plant functional traits along an urbanisation gradient, and therefore contrasted the functional trait values of all plant species identified in urban vs. non-urban 500×500 m² grid cells.

The differences in functional traits identified in the two datasets were compared to identify overall trait syndromes.

Land-cover change effects on functional diversity

Within most ecosystems a degree of “ecological memory” exists. This means that community composition depends not only on the current conditions and interactions, but also by what conditions and disturbances happened in the past. In **paper IV**, we examined the effects the land-cover changes

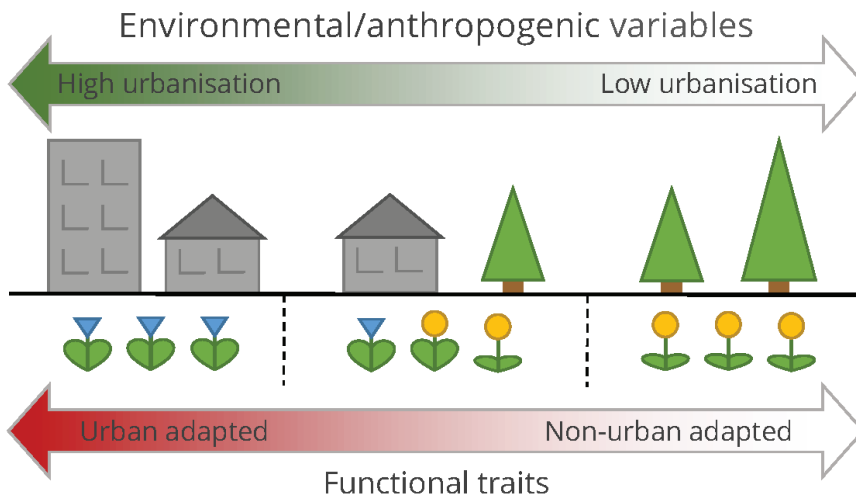


Figure 3. Conceptual figure illustrating the species turnover and interspecific trait variation along a gradient of environmental and anthropogenic variables. Figure modified from **paper III**.

on bird communities in Trondheim. Using land-cover maps from 2011 and 2018, we calculated an index of land-cover change within the 500×500 m² grid cells; changes in bird community composition were denoted as the turnover component of the Jaccard dissimilarity index (β_{turnover}) between the two time steps. We tested whether land-cover change could be used as a reliable predictor of species turnover. However, stability in community level metrics can mask deviations in species composition (MacLean et al. 2018) – measures of β diversity make no claims regarding the characteristics of the communities. To assess the responses of dif-

ferent functional groups of birds, we constructed spatially explicit models of the probability of a species either disappearing from or appearing in a grid cell, based on the functional traits of the species and land-dissimilarity.

To put the results in a conservation context, we evaluated potential future land-cover changes within the municipality, based on the land-use element of the municipal master plan. These potential future land-cover conversions were then held against the identified responses by different avian functional groups.

Summary of the main results

Paper I

Species data for understanding biodiversity dynamics: The What, Where and When of species occurrence data collection

We examined the potential differences in taxonomic-, time-, and land-cover biases in observational GBIF records from ten different datasets, focussing on red-listed and alien species. Specifically, we tested whether the spatial distribution of observational records differed from what could be seen in randomly simulated datasets.

The studied datasets differed taxonomically and temporally, with datasets predominantly consisting of citizen science records being dominated by birds. The number of GBIF records have increased over time, with citizen science records providing the bulk of the increase in the later years. The datasets differed in their geographical coverage, with anthropogenic land-covers having more records than what would be expected by chance in most cases. The prevalence of a roadside bias was underlined by a scarcity

of records from remote areas, compared to what was to be expected by chance.

Paper II

Urban aliens and threatened near-naturals: Land-cover affects the species richness of alien- and threatened species in an urban-rural setting

We examined the effects of fine-scale land-cover variables (land-cover, habitat heterogeneity and mean slope aspect) on species richness of different taxonomic groups (birds, non-avian animals, plants and fungi) and groups of conservation concern (threatened- and alien species) within 500×500 m² grid cells across Trondheim municipality.

The total number of species predicted depended on both mean aspect, habitat and –heterogeneity, all varying by taxon, illustrating highly complex dynamics in the relationships between species richness and the predictor variables. The number of plant species increased with increasing habitat heterogeneity, whereas other taxa showed no clear responses. Across taxa, the highest number of species were seen in relatively undisturbed habitats, and the lowest numbers

were predicted in urban areas. The highest numbers of threatened species were predicted in habitats relatively unaffected by human activity and by mean aspect. On the contrary, the number of alien species depended only on taxon and spatial autocorrelation. Conclusively, the study showed how land-cover variables within an administrative border could be used to predict species richness within taxonomic groups.

Paper III

Competitors and ruderals go to town:

Plant community composition and function along an urbanisation gradient

We explored plant species turnover along an urbanisation gradient by examining how species indicative of diverging urbanisation levels varied in their functional traits. To evaluate the identified patterns in functional traits, we compared the differences in functional traits of urban- and non-urban Indicator Species to the differences in functional trait distributions of all plants registered in urban- and non-urban areas in a GBIF dataset.

We identified differences in community composition between two different ur-

banisation levels, mirroring a shift in environmental- and anthropogenic variables, thus indicating how environmental conditions function as an ecological filter. Species indicative of urban areas had higher Specific Leaf Areas, higher affinities for nitrogen and light, had lower seed dry mass, were taller, were more likely to be annual and were more disturbance tolerant than species indicative of non-urban areas. The probability of presence of a non-urban Indicator Species in a grid cell in the independent dataset showed a linear, negative relationship within increasing percentage of developed area, validating their status as indicators. The probability of presence of an urban Indicator Species showed a hump-shaped relationship with increasing proportion of developed area, peaking at intermediate levels. Large confidence intervals around predictions however weakened the indicative value of these indicators.

The exact same functional traits could not be identified among all plant species registered in urban- vs. non-urban areas in the independent dataset, but traits related to the same functional trait syndrome were observed: species in urban areas had lower

moisture affinity, higher nitrogen and alkaline affinity, were taller and had a larger specific leaf area. In conclusion, disturbance-tolerance and rapid resource acquisition is favoured in highly urbanised areas, compared to species adapted for stress-tolerance and resource conservation.

Paper IV

Urbanisation and land-cover change affect turnover of bird functional groups but not the extent of species composition change

We studied the characteristics of land-cover change seen in Trondheim between 2011 and 2018, and how the species- and functional diversity of birds in 500×500 m² grid cells across Trondheim municipality have been affected by these changes. We examined the probability of a bird species from the local species pool either disappearing from or appearing in a grid cell, based on the land-cover dissimilarity index and functional traits of the bird species.

The changes in land-cover and land-use in the past decade were dominated by urbanisation; in particular, conversion of culti-

vated land and open firm ground to developed area. The degree of species turnover could not be explained by the land-cover change index. However, the probability of (dis-)appearance of a bird species was affected by both the traits of the species (longevity), and interaction between traits and land-cover dissimilarity: forage stratum, habitat association and main dietary component; increasing longevity decreased the probability of disappearance and increased the probability of appearance. Generally, granivorous species and species associated with urban- and open areas showed decreasing probabilities of disappearing with increasing land-cover dissimilarity, whereas urban- and open-associated, non-ground-feeding species showed an increasing probability of appearing with increasing land-cover change. Overall, predicted probabilities of disappearing were much higher than the probabilities of appearing, indicating that increasing land-cover change has the potential of impoverishing local avifauna. Considering the land-cover conversions planned within the municipality, local avian communities will likely be impoverished given continued urbanisation and land-cover change.

Discussion

Not all datasets are created equal

In **paper I**, we evaluated the status of openly available species occurrence data. Specifically, we tested a subset of data available through GBIF to assess whether biases in taxonomic-, temporal- and geographic scope differed between datasets, and related those differences to the origins of the datasets. This study showed, unsurprisingly, that the datasets in GBIF differ in taxonomic and temporal coverage; particularly datasets dominated by citizen science records focused on birds, and had a sharper increase in abundance in recent years. The most interesting highlight of this study was the difference in geographical coverage by the different datasets: anthropogenic land-covers (such as urban/built-up areas) had many more records than expected by chance, whereas remote areas generally had fewer records than expected. Likewise, the different datasets also showed varying biases regarding red-listed- and alien species records, with citizen science-dominated datasets skewed towards red-listed species. This underlines that caution must be taken when

dealing with compiled datasets. The disproportionate taxonomic distribution of records have been documented in previous studies as well (Amano et al. 2016), but relating this to both conservation status of the records and the characteristics of the (likely) collectors emphasises the importance of careful evaluation of data quality and suitability (Ball-Damerow et al. 2019). Developed areas (i.e. cities) generally had more records than would be expected through random sampling for most datasets – particularly datasets dominated by citizen science records, or datasets focussing on alien species. Thus, such compiled datasets have great potential as data sources within urban areas. The main caveats then become the taxonomic skew(s) and disentangling whether the disproportionate number of alien- and red-listed records are due to true abundances, or simply reflects the interests of the collectors. If species occurrence data from open databases are to be used for comparisons across habitats, measures to account for the differences in sampling effort (taxonomically, geographically and by conservation status group) needs to be developed and implemented.

Where the wild things are

In **paper II**, we examined the effects of land-cover and land-cover variables on species richness of four different taxonomic groups, both in total and for threatened- or alien species. This study revealed that the different taxonomic- and conservation groups did not respond similarly to different habitat types. This illustrates the complex relationships between biodiversity and land-cover, and how there is no “silver bullet” approach for predicting species richness across taxonomic- and conservation groups. This is in concordance with the meta-analysis performed by Beninde et al. (2015), who found multiple significant predictors of species richness within urban areas, both at local- and landscape level. Similarly, a study by Matthies et al. (2017) concluded multivariate analyses of biodiversity to be essential, as different predictor variables were retained as significant for different taxonomic groups. Nevertheless, our study showed that on a municipality-wide scale, the highest numbers of species across taxa and conservation groups were found in habitats relatively unaffected by humans, whereas the lowest numbers were seen in urban areas (with few exceptions).

We found a similar result in **paper III**, using a finer spatial scale and focussing on plants: more species were registered in non-urban areas compared to urban areas with a higher anthropogenic pressure. This is in agreement with what has been shown by other studies (e.g. Melliger et al. (2018)). However, dissimilar results have been reported as well, ranging from increasing species richness with urbanisation (e.g. Kühn et al. (2004)) or peaking species richness at moderate levels of urban development (e.g. McKinney (2008)). These responses depend heavily on spatial resolution, taxonomic scope, conservation status, and choice of urbanisation metrics (see for example Pautasso (2007), McKinney (2008), Aronson et al. (2014) and Concepción et al. (2016)).

Nevertheless, some general points regarding the effects of urbanisation and land-cover on biodiversity can be made: overall, intense urbanisation (using sharp increases in built-up area as a proxy) affects species richness negatively. If intermediate levels of urbanisation are found to be most beneficial for biodiversity, this effect can in part be the result of an increase in generalist and/or al-

ien species. Our studies agree with the general consensus that if (native) species richness is to be favoured by management, urban sprawl should be decreased as far as possible, and more natural habitat types should be relieved of anthropogenic pressure whenever possible (Concepción et al. 2016, Geschke et al. 2018, Sushinsky et al. 2013).

For whom the bell tolls

To study the effects of urbanisation on biodiversity in more detail, we examined the effects of land-cover variables and land-cover change on the functional traits of biological communities in **paper III** and **IV** (for plants and birds, respectively). This allowed for a more mechanistic understanding of the factors and processes determining community composition along an urbanisation gradient in Trondheim. Natural systems undergo continuous compositional change (Darwin 1859); our study indicates that increasing levels of land-cover change do not necessarily affect the extent of this compositional change, but rather the functional direction. The degree of land-cover change (measured by a land-cover change index) could not be linked directly to the extent of community composition change (species replacement).

Thus, community level metrics can mask deviations in species compositions (MacLean et al. 2018). We showed that environmental and anthropogenic conditions (indicating different urbanisation levels) filter plant species according to their functional traits. Likewise, the extent of changes in land-cover over time, dominated by increases in built-up area, affects the probability of bird species either appearing in or disappearing from grid-cells depending on their functional traits. The plant species benefitting from high urbanisation levels are species adapted for rapid resource acquisition and disturbed environments. The bird species responding positively to increasing levels of land-cover change (i.e. urbanisation) are likewise pre-adapted for the urban conditions: they are generally associated with urban- or open areas, are non-ground-feeding and granivorous. In a broad sense, urbanisation thus function as an ecological filter across taxa (Figure 4). It is reasonable to assume that the “urban filter” (dis-)favours the same traits across large spatial scales. Our studies thus support the notion that urbanisation can serve as biologically homogenising force over extensive spatial scales (McKinney 2006).

Fantastic data and where to find them

Data availability has always been, and continues to be, a major obstacle to biodiversity research (Dornelas et al. 2013). Despite the large amount of species occurrence records available at the moment, and the increasing number of records continuously being published, these are still merely a drop in the ocean compared to all sampled species data, which are again just samples. There are no

definitive estimates of how much of the world's species data are currently available in digital form; this is in part driven by the fact that no inventory of the world's natural history collections exists (Ariño 2010). Likewise, the amount of data stored in private collections is unknown. Some precautionary estimates have been inferred, such as: 6-10% of specimen-based data held in biological collections are available through GBIF

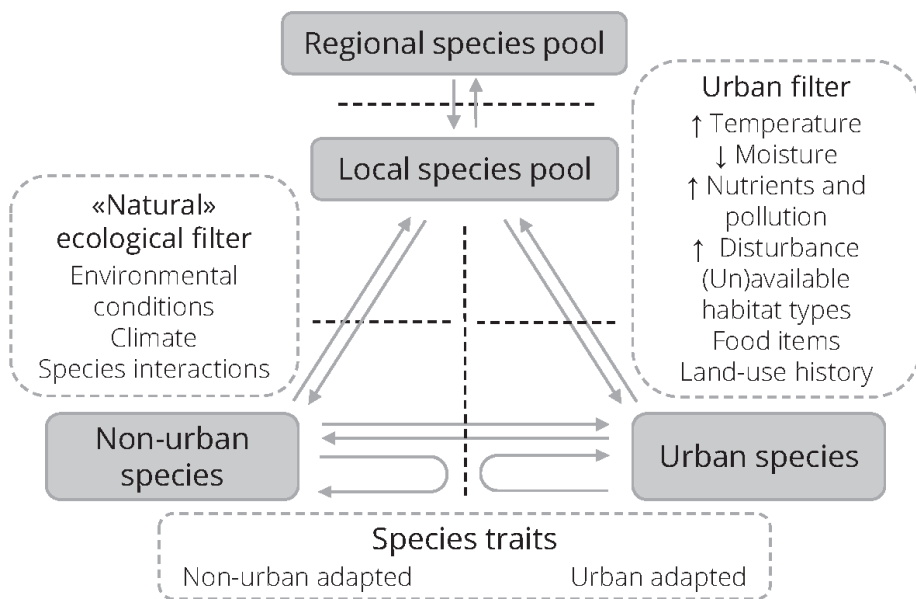


Figure 4. Conceptual figure of the filtering effect of urbanisation. The local species pools are subsets of the regional species pool (solid, grey boxes). The local species pools are filtered (arrows and dashed lines) based on local biogeochemical conditions and existing communities, according to the (functional) traits of the species, which determine whether the species are (pre-)adapted for the local conditions (dashed boxes).

(Townsend Peterson et al. 2018) and 10% of biocollections are digitally available (Ball-Damerow et al. 2019). Much of the data are however incomplete: they miss spatial-, temporal- or taxonomic information (Townsend Peterson et al. 2018). The uneven taxonomic distribution of the digitally available data is of prime importance: in particular, invertebrates and non-vascular plants are underrepresented in databases, despite their abundances in nature (Beck et al. 2012). The excess or shortfall of records for different taxa generally depend on whether the included species are considered “charismatic” (Troudet et al. 2017).

The papers included in this thesis further illustrate these issues by the immense decrease in number of species occurrence records between initial data download and the final analyses. Data shortage hampered both the taxonomic- and the temporal scope of the studies, as well as the habitat coverage (in particular for **paper II** and **paper IV**, and further illustrated in **paper I**).

Along the same lines, the availability of fine-grain land-cover data (and other relevant biogeographical factors) constitutes a

bottleneck for research advancements. Several initiatives for making such data available exist (such as Landsat from NASA (National Aeronautics and Space Administration 2021), the CORINE Land Cover from the European Environment Agency (Copernicus 2021), and the Copernicus Sentinel programs by the European Space Agency (European Space Agency 2021)). See also Randin et al. (2020) for a review on the use of remote sensing for biodiversity monitoring. However, the spatial resolution of such data is often either too coarse to be useful for detailed mechanistic studies, or the land-cover categories are not recognised in sufficient detail. Without a common standard for how to categorise land-cover, comparisons across space and time becomes increasingly difficult. As demonstrated in **paper III**, if the classification scheme within a limited area changes over time, temporal studies are complicated as well. Thus, digital biogeographical studies cannot replace local, field studies with classification of the habitats done in real time.

Unfortunately, detailed knowledge on land-cover and related variables cannot be created retroactively, but future efforts

should be made to ensure available, high-quality land-cover data, preferably adhering to a common standard to aid future research. Currently, a good candidate for future tracking of land-cover and land-cover changes is the Sentinel-2 satellites; the images have already been used in various ecological studies (European Space Agency 2021, Murayama and Ranagalage 2020).

The crux

The four papers included in this thesis tie together to advance the field of urban ecology by multiple strands (Figure 5): the assessment of GBIF data quality and bias in **paper I** lays the foundation for the use of GBIF data in the remaining papers by illustrating the general abundance of available data in urban- and anthropogenic habitats. It further connects with **paper II** by illustrating the importance of examining groups of conservation concern separately. The main contribution to the field of urban ecology by this particular study is the demonstration of the usefulness of compiled GBIF data for urban studies, if appropriate measures are taken to account for taxonomic- and temporal variation in sam-

pling effort. **Paper II, III and IV** jointly illustrate the varying responses to land-cover and land-cover change displayed by different taxa, groups of conservation concern, and functional groups. In particular, the responses to urban areas and urbanisation are highlighted. These papers thus underline how urbanisation functions as an ecological filter. The main contribution to the field is the indication that the filtering effect of urbanisation is evident across taxa and consistent across spatial resolutions, confirming the generality of the concept. **Paper IV** show how increasing land-cover change (i.e. urbanisation) will likely cause impoverishment local avifauna, which in turn might signal biotic homogenisation over large scales. Thus, the papers included in this thesis jointly emphasise the importance of minimising pressure on- and urbanisation of non-anthropogenic habitats, if a management goal is to favour native biodiversity.

The collective findings of the included papers advance the field of urban ecology by showing how land-cover data, as categorised and classified by management

institutions, relate to biodiversity, and how urbanisation functions as an ecological filter in a broad sense. Simultaneously, new questions and directions for future research

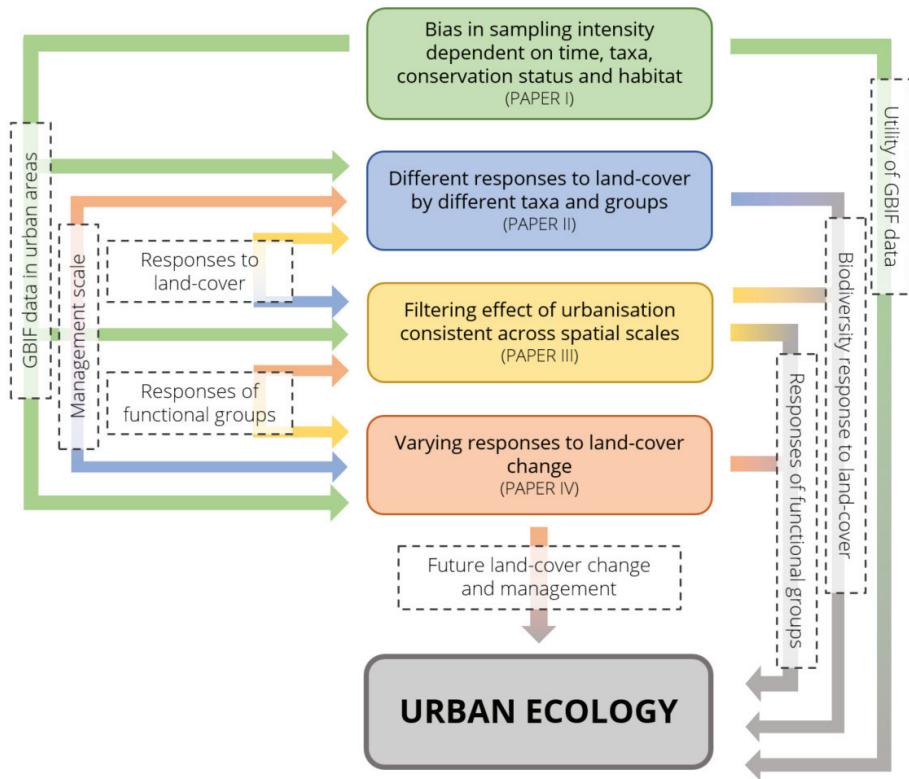


Figure 5. Synthetic diagram illustrating how the four papers included in the thesis tie together, and how they relate to and advance the field of urban ecology in general. The individual papers and their respective main themes and results relevant to the broader scope are indicated by the coloured boxes. The coloured arrows on the left-hand side of the diagram illustrate how the papers are connected, the coloured/grey arrows on the right-hand side of the diagram illustrate how they relate to urban ecology (grey box). Detailed descriptions of the connections and contributions can be found in the main text

within urban ecology arise; whether the results are repeatable over multiple spatial scales and -resolutions warrants further investigation. This includes both fine-scale, mechanistic studies to investigate local drivers, and national- to global scale studies to elucidate the general patterns. Further refinement of the models, including (but not limited to) climatic factors and soil properties, could disentangle biodiversity dynamics even further (McDonnell and Hahs 2013). As pointed out by McDonnell and Hahs (2013), urban ecologists need to translate the questions asked by practitioners (such as “How much urban green space is needed to support biodiversity?”), into ecologically relevant questions (such as:

“What are the ecological processes determining population viability and what are the attributes of green spaces that drive these processes?”).

To include the important temporal aspects evident in urban ecology, the field would benefit from long-term experiments, rather than the use of opportunistic data, as pointed out by Ramalho and Hobbs (2012). Additionally, a greater integration with evolutionary ecology to answer how urbanisation affects selection processes, as pointed out by Rivkin et al. (2019), warrants attention. To handle the shifting biases in species occurrence records, additional work on how to dynamically account for these issues is necessary.

Conclusions

In recent years, Trondheim has experienced increasing urbanisation, illustrated by an increase in built-up area, similar to what is seen on both a national- and global level. The available open-source species occurrence datasets available for Norway and Trondheim

are greatly skewed in their taxonomic representation, with an overabundance of birds in particular. The sharp increase in yearly number of records is mainly driven by an increase in the number of volunteer citizen science records. Likewise, the numbers of available

records are skewed between groups of conservation concern, with a seemingly greater motivation to report red-listed species compared to alien ones. Geographically, anthropogenic land-covers have many more records than would be expected by chance, whereas the opposite is true for remote areas. To bridge the knowledge-gaps, future sampling should be directed towards less sampled areas and for infrequently investigated taxa. Along the same lines, further work on how to account for inherent bias in such opportunistic data is needed.

Land-cover and related environmental variables determine the distribution- and number of species in an urban-rural context, but different taxa and groups of conservation concern do not respond similarly to the same variables. A greater number of threatened species are found in habitats relatively unaffected by humans, whereas alien species are mainly determined by the location of their initial introduction. On a more mechanistic level, the observed responses of plants along an urbanisation gradient indicate that

increasing urbanisation favours species adhering to a disturbance-tolerant, resource-acquisitive life strategy, compared to a more stress-tolerant, conservative strategy seen among plant species characteristic of non-urban areas. In a similar fashion, the probability of bird species either disappearing from or appearing within an area over time depends on the degree of land-cover change, with birds from different functional groups responding contrastingly: granivores, generalists and species associated with urban- and open areas respond positively to increasing land-cover change, whereas the opposite is true for forest- and wetland specialists. However, the overall probabilities for species disappearance are far higher than the probability of colonisation by new species.

Recommendations for future biodiversity management in Trondheim, in the light of continued urban expansion, are to minimise the conversion of vulnerable habitat as much as possible, and to monitor urban areas closely to mitigate the introduction and spread of alien species.

Synthesis and future perspectives

This thesis have documented how the diversity of species, including the diversity within threatened and alien ones, depends on current land-cover and -use, and how land-use history (i.e. changes in land-cover over time) affects the functional composition of biological communities on a municipality scale. Despite the growing availability of species occurrence data, great care must be taken to ensure data quality, and to counteract inherent skews and biases within the compiled data available through open data portals. Despite the contributions from this thesis to the collective knowledge of biodiversity dynamics in urban areas, further studies are still warranted.

Paper I illustrated aspects of the potential issues with using compiled species occurrence records from open databases, due to inherent differences in scope and sampling effort. With the increasing use of such data (Powney and Isaac 2015), new and robust methods to account for these differences would advance multiple research fields. As high-quality data is evidently a constant obstacle to overcome, further efforts

to make biodiversity data available are encouraged. Fortunately, a general paradigm of openness regarding data sharing and reuse is forming within the ecological communities (August et al. 2015), an example of these efforts being the Living Norway Ecological Data Network (2020):

“(...) a joint effort to promote management of ecological data from Norwegian research institutions in agreement with FAIR principles.”

The FAIR principles refer to the efforts of making data “Findable, Accessible, Interoperable and Reusable” (Wilkinson et al. 2016). This thesis (in particular **paper I**) underlined great variation in data availability over time, and thus emphasises the concerns raised by Dornelas et al. (2018, 2013) and Magurran et al. (2019): time-series data are lacking, despite it being of crucial importance to elucidate how the planet’s ecosystem are responding to anthropogenic pressures. Due to the unfortunate uni-directionality of time, we cannot start time-series surveys retroactively, but will have to make

the best of what imperfect data we have available (Dornelas et al. 2013). A relatively small fraction of the world's biocollections are currently digitised (and thus available for reuse) (Ball-Damerow et al. 2019) – it can therefore be advocated that effort be put into digitising specimens from e.g. herbaria and museums (both physical specimens and observational records), and making them more easily available to the ecological research communities.

As already pointed out, further development of (statistical) tools for accounting for differences in sampling effort, whether that be taxonomically, temporally or spatially, is a front worth venturing.

As for local management practices, **paper II** and **paper IV** illustrated how, in broad terms, biodiversity will be affected by further urbanisation. The results presented in this thesis suggest that to favour the diversity of native and/or threatened species, and to prevent a potential impoverishment of local biological communities, the grand themes should be to limit further anthropogenic pressure, particularly when this happens at the cost of vulnerable habitat types. Specifically, **paper II** identified coastal areas

(i.e. ecotones) and mires to be of prime importance for the local biodiversity of Trondheim; thus on a local management level, such areas should be prioritised. To mitigate the spread of alien species, the main tool is to prevent introductions in the first place (Skarpaas and Økland 2009). Hampering urban sprawl while at the same time accommodating human welfare will be a challenging, but essential and interesting field in the future.

The Fifteenth meeting of the Conference of the Parties to the Convention on Biological Diversity is currently in preparation (CBD 2021); these preparations are done in the light of the unfulfilled Aichi Biodiversity Targets and SDGs. This includes the development of a “Post-2020 Global Biodiversity Framework” – this framework builds on the Strategic Plan for Biodiversity 2011-2020, and will be the successor of this and the Aichi Targets (CBD 2020). According to the zero draft of the Post-2020 Global Biodiversity Framework, the action targets in this framework includes specifically:

“Target 11. By 2030, increase benefits from biodiversity and green/blue spaces for

human health and well-being, including the proportion of people with access to such spaces by at least [100%], especially for urban dwellers.”

In this target, the importance of biodiversity for human health and well-being is articulated; if the access to green/blue spaces are to be increased, especially for urban dwellers, the immediate interpretation is that these spaces should be located within city boundaries⁴. Thus, as biodiversity in cities is a UN target in itself, understanding the dynamics of urban biological communities becomes essential, and especially how these dynamics relate to human welfare (CBD 2020).

Urban ecology is still a relatively young field, and its scopes and paradigms need to be refined. A shift from studies of “ecology *in* cities” to “ecology *of* cities”⁵ is slowly

emerging (McDonnell and Hahs 2008), but the paradigm of humans as external to ecosystems is still widely prevalent. Either time must pass for this shift to happen, or an in-depth discussion of under which circumstances either view is appropriate is warranted. Repeating the statement made by McDonnell & Hahs (2013), the questions asked by managers and decision-makers might not match the ones asked by researchers; the former likely asking questions regarding day-to-day decisions. These aspects should be incorporated into urban ecological research, and we should aim at making it possible to translate practical questions into ecologically relevant research questions, and *vice versa*, to aid in evidence-based conservation and biodiversity management. Along those lines, it is worth venturing into more detailed predictors and response variables; rather than focussing on aggregate measures

⁴ In addition, the terms “green/blue spaces” and “green/blue infrastructure” are most frequently used in connection with urban structures and areas (IPBES 2019). The urban aspect of this target is thus implicit.

⁵ As described by Grimm et al. (2000) and Pickett et al. (2008): ecology *in* the city focusses on the physical environment, frequently taking a case study approach ad-

ressing questions akin to “how do ecological patterns and processes differ in vs. outside city boundaries?”. Ecology *of* cities represent a more holistic framework, treating entire cities as ecosystems in themselves, thus often incorporating multiple disciplines; human activities are in this framework not considered as external disturbances to an ecosystem, but as an important driver within the system.

(such as species richness), both site-specific and large-scale studies of mechanistic relationships and processes (e.g. fecundity and recruitment success) could provide direct, useful answers to management questions and concerns (McDonnell and Hahs 2008, 2013).

It is clear that the future of urban ecology must take even more advantage of its already multifaceted nature. Close collaborations between multiple disciplines and stakeholders are necessary to accommodate the needs of both humans and other organisms (Kaye et al. 2006). Specifically, in addition to ecologists, also actors from social sciences are needed to further understand how nature

and society interacts, and thus shed light on the potential problems and synergies (Folke et al. 2016). City planners are needed to put the conclusions of academics into a real-world context. Such involvement of multiple fields has been described as both “multidisciplinary”, “interdisciplinary” and “transdisciplinary”, but no clear definition of these expressions exist, and they are frequently used interchangeably (Figure 6) (Choi and Pak 2006, Jahn et al. 2012, von Wehrden et al. 2019), and are defined with various levels of complexity (Max-Neef 2005). However, attempts at disentangling the concepts are provided by e.g. Choi and Pak (2006) and Jahn

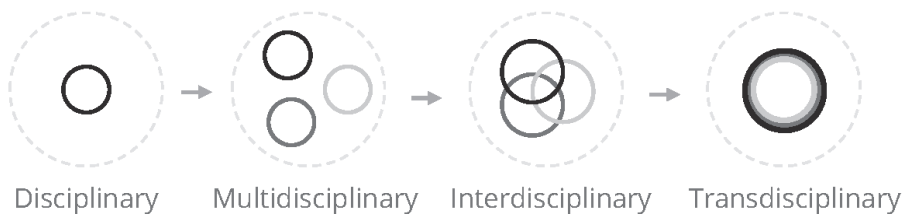


Figure 6. Visualisation of the (simplified) relations between different disciplines in different disciplinary frameworks. In a multidisciplinary approach, the individual fields operate individually, working within the boundaries of the field. In an interdisciplinary framework, the disciplines share knowledge and methodologies, creating synergies and blurring the boundaries; the individual fields are however still recognisable. In a transdisciplinary framework, the collaborations transcend the disciplinary boundaries, forming a new, holistic field. Examples of potential fields relevant to the envisioned field of “*Ecology of cities*” can be ecology, sociology and economy.

et al. (2012), and a version of these interpretations will be the definitions used here: “multidisciplinary” is an interplay and collaboration between disciplines, but each one stays and operates within the boundaries of the respective field. “Interdisciplinary” refers to a synthesis of multiple disciplines in which a transfer of knowledge and methodology happens between fields, blurring the boundaries. “Transdisciplinary” indicates a holistic framework in which both scientists

and stakeholders transcend the boundaries between disciplines and operates in a new, unified field (Choi and Pak 2006) (Figure 6). Despite the fact that transdisciplinarity has been a known concept for a long time (Jahn et al. 2012), practical appliance is still rare (Angelstam et al. 2013). To fully achieve sustainable city development in the future, urban ecology needs to move from multi- and interdisciplinary research and management, to a fully transdisciplinary field.

References

- Ahlström, Anja P. et al. 2014. *AR5 Klassifikasjonssystem. Klassifikasjon Av Arealressurser*.
- Amano, Tatsuya et al. 2016. "Spatial Gaps in Global Biodiversity Information and the Role of Citizen Science." - *BioScience* 66(5): 393–400.
- Angelstam, Per et al. 2013. "Solving Problems in Social-Ecological Systems: Definition, Practice and Barriers of Transdisciplinary Research." - *Ambio* 42: 254–65.
- Ariño, Arturo H. 2010. "Approaches to Estimating the Universe of Natural History Collections Data." - *Biodiversity Informatics* 7(2): 81–92.
- Aronson, Myla F J et al. 2014. "A Global Analysis of the Impacts of Urbanization on Bird and Plant Diversity Reveals Key Anthropogenic Drivers." - *Proceedings of the Royal Society of London B: Biological Sciences* 281(1780).
- August, Tom et al. 2015. "Emerging Technologies for Biological Recording." - *Biological Journal of the Linnean Society* 115(3): 731–49.
- Ball-Damerow, Joan E. et al. 2019. "Research Applications of Primary Biodiversity Databases in the Digital Age." - *PLoS ONE* 14(9): 1–26.
- Beck, Jan et al. 2012. "What's on the Horizon for Macroecology?" - *Ecography* 35(8): 673–83.
- Beninde, Joscha et al. 2015. "Biodiversity in Cities Needs Space: A Meta-analysis of Factors Determining Intra-urban Biodiversity Variation." - *Ecology Letters* 18(6): 581–92.
- Billerman, S.M. et al. 2020. "Birds of the World."
- Boakes, Elizabeth H. et al. 2016. "Patterns of Contribution to Citizen Science Biodiversity Projects Increase Understanding of Volunteers' Recording Behaviour." - *Scientific Reports* 6: 1–11.
- Brondízio, Eduardo S. et al. 2019. *Global Assessment Report on Biodiversity and Ecosystem Services of the Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services*. IPBES secretariat, Bonn, Germany.
- CBD. 2012. "Cities and Biodiversity Outlook: A Global Assessment of the Links between Urbanization, Biodiversity, and Ecosystem Services." - Executive Summary: 64.
- . 2020. "Update of the Zero Draft of the Post-2020 Global Biodiversity Framework." - Preparations for the Post-2020 Biodiversity Framework.
- . 2021. "Meeting Documents, Fifteenth Meeting of the Conference of the Parties to the Convention on Biological Diversity." <https://www.cbd.int/meetings/COP-15> (January 8, 2021).
- Choi, Bernard C. K., and Anita W. P. Pak. 2006. "Multidisciplinarity, Interdisciplinarity and Transdisciplinarity in Health Research, Services, Education and Policy: 1. Definitions, Objectives, and Evidence of Effectiveness." - *Clinical and Investigative Medicine* 29(6): 351–64.
- Cincotta, R P et al. 2000. "Human Population and Biodiversity Hotspots." - *Nature* 404(April): 990–92.
- Concepción, Elena D. et al. 2016. "Impacts of Urban Sprawl on Species Richness of Plants, Butterflies, Gastropods and Birds: Not Only Built-up Area Matters." - *Urban Ecosystems* 19(1): 225–42.
- Copernicus. 2021. "CORINE Land Cover." <https://land.copernicus.eu/pan-european/corine-land-cover> (January 10, 2021).

- Darwin, C. 1859. *On the Origin of Species by Means of Natural Selection, or, the Preservation of Favoured Races in the Struggle for Life*. London, United Kingdom: J. Murray.
- DKNVS. 1760. “Det Kongelige Norske Videnskabers Selskab.” <https://www.dknvs.no/> (January 8, 2021).
- Dornelas, Maria et al. 2013. “Quantifying Temporal Change in Biodiversity: Challenges and Opportunities.” - *Proceedings of the Royal Society B: Biological Sciences* 280(1750).
- . 2018. “BioTIME: A Database of Biodiversity Time Series for the Anthropocene.” - *Global Ecology and Biogeography* 27(7): 760–86.
- Elmqvist, Thomas et al. 2013. *Urbanization, Biodiversity and Ecosystem Services: Challenges and Opportunities: A Global Assessment*. eds. Thomas Elmqvist et al. Springer.
- . 2018. “Ecosystem Services Provided by Urban Green Infrastructure.” In *Routledge Handbook of Ecosystem Services*, , 452–68.
- European Space Agency. 2021. “Sentinel Online.” <https://sentinel.esa.int/web/sentinel/home> (January 10, 2021).
- Folke, Carl et al. 2016. “Social-Ecological Resilience and Biosphere-Based Sustainability Science.” - *Ecology and Society* 21(3).
- Forman, Richard T. T. 2014a. “Framework: Foundations.” In *Urban Ecology: Science of Cities*, Cambridge University Press, 1–11.
- . 2014b. “Urban Air.” In *Urban Ecology: Science of Cities*, Cambridge: Cambridge University Press, 125–48.
- . 2014c. “Urban Soil and Chemicals.” In *Urban Ecology: Science of Cities*, Cambridge: Cambridge University Press, 91–124.
- Gaertner, Mirijam et al. 2017. “Non-Native Species in Urban Environments: Patterns, Processes, Impacts and Challenges.” - *Biological Invasions* 19(12): 3461–69.
- Gaiji, Samy et al. 2013. “Content Assessment of the Primary Biodiversity Data Published through GBIF Network: Status, Challenges and Potentials.” - *Biodiversity Informatics* 8(2): 94–172.
- Gaston, Kevin J. 2000. “Global Patterns in Biodiversity.” - *Nature* 405(6783): 220–27.
- . 2005. “Biodiversity and Extinction: Species and People.” - *Progress in physical geography* 29(2): 239–47.
- GBIF.org. 2019. “GBIF Home Page.” <https://www.gbif.org/> (November 19, 2019).
- Geschke, Andrew et al. 2018. “Compact Cities or Sprawling Suburbs? Optimising the Distribution of People in Cities to Maximise Species Diversity.” - *Journal of Applied Ecology* 55(5): 2320–31.
- Grimm, Nancy B. et al. 2000. “Integrated Approaches to Long-Term Studies of Urban Ecological Systems.” - *BioScience* 50(7): 123–41.
- van Ham, Chantal et al. 2013. *Invasive Alien Species: The Urban Dimension, Case Studies on Strengthening Local Action in Europe*. Brussels, Belgium.
- IPBES. 2018. *The Regional Assessment Report on Biodiversity and Ecosystem Services for Europe and Central Asia*. eds. M. Rounsevell et al. Bonn, Germany: Secretariat of the Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services.
- . 2019. *Summary for Policymakers of the Global Assessment Report on Biodiversity and Ecosystem Services of the Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services*. eds. S. Díaz et al. Bonn, Germany: Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services (IPBES).
- IUCN. 2000. *Guidelines for the Prevention of Biodiversity Loss Caused by Alien Invasive Species*. Gland, Switzerland.

- Ives, Christopher D. et al. 2016. "Cities Are Hotspots for Threatened Species." - *Global Ecology and Biogeography* 25(1): 117–26.
- Jahn, Thomas et al. 2012. "Transdisciplinarity: Between Mainstreaming and Marginalization." - *Ecological Economics* 79: 1–10.
- Kattge, J. et al. 2011. "TRY – a Global Database of Plant Traits." - *Global Change Biology* 17: 2905–35.
- Kaye, Jason P. et al. 2006. "A Distinct Urban Biogeochemistry?" - *Trends in Ecology and Evolution* 21(4): 192–99.
- Kowarik, Ingo. 2011. "Novel Urban Ecosystems, Biodiversity, and Conservation." - *Environmental Pollution* 159(8): 1974–83.
- Kühn, Ingolf et al. 2004. "The Flora of German Cities Is Naturally Species Rich." - *Evolutionary Ecology Research* 6: 749–64.
- Living Norway Ecological Data Network. 2020. "Living Norway Ecological Data Network." <https://livingnorway.no/> (December 9, 2020).
- Luck, Gary W. 2007. "A Review of the Relationships between Human Population Density and Biodiversity." - *Biological Reviews* 82(4): 607–45.
- MacLean, Sarah A. et al. 2018. "A Century of Climate and Land-Use Change Cause Species Turnover without Loss of Beta Diversity in California's Central Valley." - *Global Change Biology* 24(12): 5882–94.
- Magurran, Anne E. et al. 2019. "Temporal β Diversity—A Macroecological Perspective." - *Global Ecology and Biogeography* 28(12): 1949–60.
- Matthies, Sarah A. et al. 2017. "Determinants of Species Richness within and across Taxonomic Groups in Urban Green Spaces." - *Urban Ecosystems* 20(4): 897–909.
- Max-Neef, Manfred A. 2005. "Foundations of Transdisciplinarity." - *Ecological Economics* 53(1): 5–16.
- McDonnell, Mark J. 2011. "The History of Urban Ecology - An Ecologist's Perspective." In *Urban Ecology. Patterns, Processes and Applications*, eds. Jari Niemelä et al. Oxford University Press, 5–14.
- McDonnell, Mark J., and Amy K. Hahs. 2008. "The Use of Gradient Analysis Studies in Advancing Our Understanding of the Ecology of Urbanizing Landscapes: Current Status and Future Directions." - *Landscape Ecology* 23(10): 1143–55.
- . 2013. "The Future of Urban Biodiversity Research: Moving beyond the 'Low-Hanging Fruit.'" - *Urban Ecosystems* 16(3): 397–409.
- McKinney, Michael L. 2006. "Urbanization as a Major Cause of Biotic Homogenization." - *Biological Conservation* 127(3): 247–60.
- . 2008. "Effects of Urbanization on Species Richness: A Review of Plants and Animals." - *Urban Ecosystems* 11(2): 161–76.
- Melliger, Ramona Laila et al. 2018. "Diverse Effects of Degree of Urbanisation and Forest Size on Species Richness and Functional Diversity of Plants, and Ground Surface-Active Ants and Spiders." - *PLoS ONE* 13(6): 1–24.
- Miller-Rushing, Abraham et al. 2012. "The History of Public Participation in Ecological Research." - *Frontiers in Ecology and the Environment* 10(6): 285–90.
- Münkemüller, Tamara et al. 2014. "Scale Decisions Can Reverse Conclusions on Community Assembly Processes." - *Global Ecology and Biogeography* 23(6): 620–32.
- Murayama, Yuji, and Manjula Ranagalage. 2020. "Sentinel-2 Data for Land Cover / Use Mapping: A Review." - *Remote Sensing* 2291(12): 14.
- Myhrvold, Nathan P. et al. 2015. "An Amniote Life-History Database to Perform Comparative

- Analyses with Birds, Mammals, and Reptiles.” - *Ecology* 96(11): 3109–3109.
- National Aeronautics and Space Administration. 2021. “Landsat.” <https://landsat.gsfc.nasa.gov/> (January 10, 2021).
- Newbold, Tim. 2010. “Applications and Limitations of Museum Data for Conservation and Ecology, with Particular Attention to Species Distribution Models.” - *Progress in Physical Geography* 34(1): 3–22.
- NINA. 2021. “Norsk Institutt for Naturforskning.” <https://www.nina.no/> (January 8, 2021).
- Norwegian Biodiversity Information Centre. 2021. “Norwegian Biodiversity Information Centre.” <https://www.biodiversity.no/> (January 8, 2021).
- Norwegian Institute of Bioeconomy Research. 2018. “AR5.” - Norwegian Institute of Bioeconomy Research. <https://www.nibio.no/tema/jord/arealressurser/arealressurskart-ar5/> (May 20, 2018).
- . 2019a. “AR50.” <https://nibio.no/tema/jord/arealressurser/ar50> (November 23, 2019).
- . 2019b. “DMK.” - Norwegian Institute of Bioeconomy Research. <https://www.nibio.no/tjenester/nedlasting-av-kartdata/dokumentasjon/dmk> (October 29, 2019).
- Oke, T. R. 1988. “The Urban Energy Balance.” - *Progress in Physical Geography* 12(4): 471–508.
- Padayachee, Ashlyn L. et al. 2017. “How Do Invasive Species Travel to and through Urban Environments?” - *Biological Invasions* 19(12): 3557–70.
- Pautasso, Marco. 2007. “Scale Dependence of the Correlation between Human Population Presence and Vertebrate and Plant Species Richness.” - *Ecology Letters* 10(1): 16–24.
- Pellissier, Vincent et al. 2008. “Relationships between Soil Seed Bank, Vegetation and Soil Fertility along an Urbanisation Gradient.” - *Applied Vegetation Science* 11(3): 325–34.
- Pickett, S. T.A. et al. 2008. “Urban Ecological Systems: Linking Terrestrial Ecological, Physical, and Socioeconomic Components of Metropolitan Areas.” - *Urban Ecology: An International Perspective on the Interaction Between Humans and Nature*: 99–122.
- Powney, Gary D., and Nick J.B. Isaac. 2015. “Beyond Maps: A Review of the Applications of Biological Records.” - *Biological Journal of the Linnean Society* 115(3): 532–42.
- Ramalho, Cristina E, and Richard J Hobbs. 2012. “Time for a Change: Dynamic Urban Ecology.” - *Trends in Ecology & Evolution* 27(3): 179–88.
- Randin, Christophe F. et al. 2020. “Monitoring Biodiversity in the Anthropocene Using Remote Sensing in Species Distribution Models.” - *Remote Sensing of Environment* 239(January): 111626.
- Rivkin, L. Ruth et al. 2019. “A Roadmap for Urban Evolutionary Ecology.” - *Evolutionary Applications* 12(3): 384–98.
- Robinson, Orin J. et al. 2018. “Correcting for Bias in Distribution Modelling for Rare Species Using Citizen Science Data.” - *Diversity and Distributions* 24(4): 460–72.
- Salomon Cavin, Joëlle, and Christian A. Kull. 2017. “Invasion Ecology Goes to Town: From Disdain to Sympathy.” - *Biological Invasions* 19(12): 3471–87.
- Skarpaas, Olav, and Bjørn Økland. 2009. “Timber Import and the Risk of Forest Pest Introductions.” - *Journal of Applied Ecology* 46(1): 55–63.
- Speed, James D M et al. 2018. “Contrasting Spatial, Temporal and Environmental Patterns in Observation and Specimen Based Species Occurrence Data.” - *PLoS Biology* 13(4): 1–17.
- Statistics Norway. 2020. “Statistisk Sentralbyrå.” <https://www.ssb.no/> (January 6, 2020).

- Sukopp, Herbert. 2008. "On the Early History of Urban Ecology in Europe." In *Urban Ecology. An International Perspective on the Interaction Between Humans and Nature*, eds. John M. Marzluff et al. Springer.
- Sushinsky, Jessica R et al. 2013. "How Should We Grow Cities to Minimize Their Biodiversity Impacts?" - *Global Change Biology* 19(2): 401–10.
- Tiago, Patrícia et al. 2017. "Spatial Distribution of Citizen Science Casuistic Observations for Different Taxonomic Groups." - *Scientific Reports* 7(1): 1–9.
- Townsend Peterson, A. et al. 2018. "Data Leakage and Loss in Biodiversity Informatics." - *Biodiversity Data Journal* 6.
- Trondheim Municipality. 2020. "Trondheim Kommune." <https://www.trondheim.kommune.no/> (April 9, 2020).
- Troudet, Julien et al. 2017. "Taxonomic Bias in Biodiversity Data and Societal Preferences." - *Scientific Reports* 7(1): 1–14.
- Turrini, Tabea, and Eva Knop. 2015. "A Landscape Ecology Approach Identifies Important Drivers of Urban Biodiversity." - *Global Change Biology* 21(4): 1652–67.
- Uchida, Kenta et al. 2020. "Opinion: Urban Biodiversity and the Importance of Scale." - *Trends in Ecology and Evolution* 36(2): 123–31.
- United Nations. 2018. "World Urbanization Prospects : The 2018 Revision." - *World Urbanization Prospects: The 2018 Revision*: 1–2.
- . 2020a. *The Sustainable Development Goals Report*.
- United Nations, Department of Economic and Social Affairs - Division for Sustainable Development. 2020b. "Sustainable Development Goals." <http://www.un.org/sustainabledevelopment/sustainable-development-goals/> (October 28, 2020).
- von Wehrden, Henrik et al. 2019. "Interdisciplinary and Transdisciplinary Research: Finding the Common Ground of Multi-Faceted Concepts." - *Sustainability Science* 14(3): 875–88.
- Whittaker, R. H. 1972. "Evolution and Measurement of Species Diversity." - *Taxon* 21(2): 213–51.
- Wilkinson, Mark D. et al. 2016. "Comment: The FAIR Guiding Principles for Scientific Data Management and Stewardship." - *Scientific Data* 3: 1–9.
- Wilman, H. et al. 2014. "EltonTraits 1.0: Species-Level Foraging Attributes of the World's Birds and Mammals." - *Ecology* 95(October 2013): 2027.

Paper I



Photo: T.K. Petersen

1 **Species data for understanding biodiversity dynamics: The What,**
2 **Where and When of species occurrence data collection.**

3 Tanja K. Petersen*,^{1,2}, James D. M. Speed¹, Vidar Grøtan², Gunnar Austrheim¹

4

5 ¹Department of Natural History, NTNU University Museum, Norwegian University of Sci-
6 ence and Technology, Trondheim, Norway

7 ²Department of Biology, Centre for Biodiversity Dynamics, Norwegian University of Sci-
8 ence and Technology, Trondheim, Norway

9

10 ***Keywords:***

11 Alien species; Citizen science; GBIF; Land cover; Museum collections; Sampling bias;
12 Threatened species

13

14 **Abstract**

15 1. The availability and quantity of observational species occurrence records have
16 greatly increased due to technological advancements and the rise of online portals,
17 such as the Global Biodiversity Information Facility (GBIF), coalescing occurrence
18 records from multiple datasets. It is well-established that such records are biased in
19 time, space and taxonomy, but whether these datasets differ in relation to origin has
20 not been assessed. If biases are specific to different types of datasets, and the rela-
21 tive contribution from these datasets have changed over time, these shifting biases
22 would have implications for interpretations of results, and consequentially for man-
23 agement and conservation measures.

- 24 2. We examined observational GBIF records from Norway to test potential differ-
25 ences in taxonomic-, time-, and land-cover biases between ten different datasets,
26 with a focus on red-listed and non-native species.
- 27 3. The datasets differ in their taxonomic coverage, with datasets dominated by citizen
28 scientist recorders focusing greatly on birds. The number of records has increased
29 over time, in particular, citizen science datasets have had a sharp increase in recent
30 years.
- 31 4. The different datasets (including division of the datasets by conservation status)
32 showed differences in geographical coverage. Anthropogenic land-covers have
33 more records than would be expected by chance in the majority of cases. Remote
34 areas have fewer records than would be expected, underlining the prevalence of a
35 roadside bias.
- 36 5. Accounting for biases in opportunistic species occurrence records need to be a dy-
37 namic- rather than static process, as the taxonomic- and geographical biases have
38 changed over time, and differ between datasets, depending on origin and inherent
39 characteristics. Data-collection programmes should be designed to counteract the
40 biases of the specific datasets, and methods to account for the biases in existing
41 data should be developed. When utilising compiled, open-source data, care must be
42 taken to ensure complementarity between the datasets, both regarding time and
43 space. Incorporating strengths and accounting for biases between datasets can
44 strengthen the integration between species occurrence records with different ori-
45 gins for science-policy impact and management.

46

47 **Introduction**

48 The amount and availability of data on species occurrences have increased tremendously in
49 recent years (Gaiji *et al.*, 2013), as have their use in applied conservation and biodiversity

50 management (Powney and Isaac, 2015). Registering species occurrences have become far
51 easier than in the early days of biogeographical surveys due to technological advancements,
52 and can be done with the help of volunteer amateurs (“citizen scientists”) (Boakes *et al.*,
53 2016). Online portals, e.g. the Global Biodiversity Information Facility (GBIF) (GBIF.org,
54 2019a), have further increased the public availability and interest (Amano *et al.*, 2016).
55 These portals gather data from various sources, ranging from digitised natural history col-
56 lections to observations made by citizen scientists. Thus, these records are a mixture of
57 data on preserved specimens and observational records from both structured surveys and
58 opportunistic sightings (Speed *et al.*, 2018). Volunteers participating in citizen science pro-
59 grams (or autonomously reporting species occurrences) likely have different motivations
60 for reporting than do institutional recorders registering species according to a specified
61 aim, covering both intrinsic and extrinsic factors. For participants in citizen science pro-
62 grams, the most important motivational factors have been reported as a personal connec-
63 tion, interest and concern for nature, a wish to contribute to science and (biodiversity- and
64 nature-) conservation, and the value/usefulness of their contributions (Ganzevoort *et al.*,
65 2017; Tiago *et al.*, 2017; Larson *et al.*, 2020).

66 These mixed datasets thus suffer from various biases and errors due to their diverse origins
67 and underlying motivations (Newbold, 2010). A frequently recognised bias for occurrence
68 records is the “roadside” bias; observations are reported more frequently short distances
69 from roads and paths, due to easier accessibility (Kadmon *et al.*, 2004; Tye *et al.*, 2017). The
70 term can be expanded to include areas near densely populated areas (Luck, 2007; Robinson
71 *et al.*, 2018). Concern has been raised repeatedly over this bias, especially if sampled areas
72 cover significantly different environmental conditions than do un-sampled areas (Phillips *et*
73 *al.*, 2009; Bystriakova *et al.*, 2012; Speed *et al.*, 2018). This potentially leads to faulty conclu-
74 sions regarding biodiversity patterns (Kramer-Schadt *et al.*, 2013). More importantly, if such
75 potential biases are not similar among data providers (e.g., datasets mainly consisting of
76 purely opportunistic citizen science records vs. datasets from structured, targeted institu-
77 tional surveys), conclusions can differ depending on the proportional contribution from

78 the different datatypes (Tye *et al.*, 2017). Even more so, if this relative contribution from
79 various types of datasets has changed over time.

80 In terms of biodiversity management, attention is frequently focused on specific taxonomic
81 groups or on species of conservation concern (e.g. red-listed and alien species). However,
82 different data providers might prioritise differently regarding taxonomic groups, and spe-
83 cies' management status (red-listed vs. alien). Citizen scientists can be biased towards char-
84 ismatic, easily recognisable taxa (Amano *et al.*, 2016), and have a greater incentive to report
85 red-listed and rare species (Tulloch, Mustin, *et al.*, 2013). Speed *et al.* (2018) showed that
86 observational plant records and preserved specimens have different biases regarding taxo-
87 nomic coverage, time and space, and hypothesised that these differences can be translated
88 somewhat to whether the data originate from structured surveys or opportunistic records,
89 thus illustrating some of the potential issues with these mixed datasets. Note however, the
90 distinction between observation- and specimen records is not equivalent to the distinction
91 between citizen science- and institutional records; vegetation plot data will be registered as
92 observations and some specimens in natural history collections are supplied by citizens
93 (Miller-Rushing *et al.*, 2012; NTNU University Museum, 2018). Geldmann *et al.* (2016)
94 showed that spatial bias in citizen science records, depended on the sampling scheme, dis-
95 tance to roads, and the human population density.

96

97 Understanding spatio-temporal dynamics of biodiversity is paramount to achieve sustaina-
98 ble management of biodiversity issues, e.g. red-listed- and alien species; for example, there
99 is a general lack of understanding on how land-use, a main but complex driver, affects bio-
100 diversity change, as detailed data on species occurrences associated with different land-use
101 types often are limited. Fine-grain data on species distributions and -associations from lo-
102 cal- to global spatial scales, and over long time periods are required – a task virtually im-
103 possible to achieve through targeted surveys alone (Bonney *et al.*, 2009; Dickinson *et al.*,
104 2010; Theobald *et al.*, 2015). Opportunistic citizen science records are frequently used as a
105 data source for e.g. Species Distribution Modelling (SDM) (Jetz *et al.*, 2012; Beck *et al.*,

106 2014), which can be used in decision-making for managing red-listed- and alien species
107 (Thuiller *et al.*, 2005; Guisan *et al.*, 2013; Syfert *et al.*, 2014). As these models are sensitive to
108 bias in the data (Yañez-Arenas *et al.*, 2014), methods to account for varying forms of bias
109 in SDM's are still being explored (e.g. Kramer-Schadt *et al.* (2013), Dorazio (2014) and
110 Robinson *et al.* (2018)). A general caveat of using GBIF records in SDM is that only few of
111 datasets report species absences, thus requiring the use of presence-only modelling.

112 If the inherent biases differ markedly between datasets collected through institutional sur-
113 veys, as citizen science, or as a mixture of the two, and the proportional contribution from
114 these groups has changed over time, this raises the additional issue of how to deal with
115 shifting biases, rather than simply static ones.

116

117 To our knowledge, limited attention has been given to whether taxonomic-, temporal- and
118 geographical sampling biases are similar for datasets with varying origins (i.e. predomi-
119 nantly from citizen science programs or institutionally organised surveys), and whether
120 these different datasets complement or amplify each other's biases. The same holds for
121 records of conservation concern within these datasets (but see Beck *et al.* (2014) for a com-
122 parison of GBIF- original source data, Tye *et al.* (2017) for comparison of SDMs based on
123 citizen science- or institutional observation records, Troudet *et al.* (2017) for an assessment
124 of taxonomic bias in GBIF records, and Speed *et al.* (2018) for comparison of spatial-, en-
125 vironmental-, temporal-, and taxonomic coverage of observational records vs. preserved
126 specimens). Awareness of such differences can elucidate how such mixed datasets should
127 be utilised in the future to ensure complementarity, and what biases to account for. Specifi-
128 cally, it will provide guidance to (1) what geographical areas, taxonomic- and conservation
129 groups should be targeted to balance sampling effort (and by whom); (2) whether certain
130 datasets (with specified origins and characteristics) are representative of all collected data,
131 and if not: (3) how to ensure complementarity between datasets to obtain maximum cover-
132 age.

133 In this study, we aim to test the ten datasets with the most records within the study region
134 from GBIF, detailing their differences and biases in taxonomy, time and land-cover associ-
135 ations, and relating these to the various origins and characteristics of the datasets. The da-
136 taset range from “pure” opportunistic citizen science records, to observations from struc-
137 tured, targeted surveys by scientific institutions. To relate the results to biodiversity man-
138 agement, focus will be put on red-listed and alien species.

139 We hypothesise the following:

140 **H1:** The distribution of records between the three main kingdoms (H1a) and alien- vs. red-
141 listed species (H1b) differ between the datasets; also within the datasets not explicitly fo-
142 cussing on a particular taxonomic group.

143 **H2:** There has been an increase in the number of records over time, primarily reflecting an
144 increase in the activity of Citizen Scientists.

145 **H3:** The different datasets will be unevenly distributed among different land-cover types,
146 with areas heavily influenced by humans (e.g. urban areas and agricultural land; areas classi-
147 fied as “Developed area” and “Cultivated” in Table S.1 (Fig. 1, Fig. S.1)) sampled more
148 than would be expected by random chance; this oversampling is expected to be greater for
149 datasets primarily consisting of Citizen Science records than for more targeted datasets.

150

151 **Materials and methods**

152 **Land-cover and species occurrence records**

153 The study was limited to Norway (Fig. 1). This is a well-surveyed region regarding species
154 occurrence records in GBIF (Chandler *et al.*, 2017), covering great variation in land-cover,
155 climate, human population density, and with detailed land-cover data available (Statistics
156 Norway, 2020).

157 Land-cover was based on the Norwegian AR50 maps from NIBIO (Norwegian Institute
158 of Bioeconomy Research, 2019), downloaded through Geonorge (2019). Land-cover is cat-
159 egorised based on land- and tree cover type, timber productivity and soil condition (Sup-
160 porting Information S.1, Table S.1, Fig. S.1). Areas smaller than 1.5 ha are not visible in the
161 dataset. The AR50 data were last updated in year 2016.

162

163 All georeferenced records of all taxa (regardless of taxonomic level) within Norway were
164 downloaded from GBIF on 19.11.2019 (GBIF.org, 2019b). The full dataset consisted of 31
165 091 434 species occurrence records. Of these, 23 586 634 belonged to the kingdom Ani-
166 malia, 1 275 533 belonged to Fungi, 5 872 214 belonged to Plantae, 283 924 belonged to
167 Archaea, Bacteria, Chromista or Protozoa, 46 were viruses, and 73 083 had no reported
168 kingdom or were *incertae sedis*. The records ranged temporally from 1686 to 2019.

169 The following criteria were used for improving the dataset quality and -comparability: (1)
170 Records with the occurrence status “absent” were removed, as very few of the dominant
171 datasets included information on absences. Thus, including absence records would reduce
172 the comparability of the datasets; (2) records with no registered species-level information
173 were removed to standardise the taxonomic resolution of the datasets; (3) potential dupli-
174 cate records for species, date, basis of record, coordinates and coordinate uncertainty were
175 removed, as there is no guarantee that the same records have not been registered multiple
176 times by different data providers; (4) records from later than 31.12.2018 were removed,
177 thus only including full sampling years. This was done in consideration of the temporal
178 analyses; (5) only records classified as “HUMAN_OBSERVATION” were retained; as the distri-
179 bution of data types differed greatly between datasets, only comparing data within a single
180 basis of record increased the comparability among datasets. Only records from the king-
181 doms Animalia, Plantae and Fungi were retained. For the comparison of different datasets,
182 the analyses were limited to datasets including >50,000 records. The final dataset for anal-
183 yses consisted of 10 datasets holding a total of 7,560,590 records (Table 1, see Supporting

184 Information S.2, Table S.2 for detailed descriptions of the individual datasets). Most spe-
185 cies were only observed sporadically (Supporting Information S.3, Figs. S.2-S.3). The ten
186 datasets were not evenly distributed across Norway, neither individually nor in unison.
187 However, as part of the aim of the study was to assess skews in geographical distribution,
188 this was not considered an issue.

189 The datasets included in the analyses differ in origin and in several characteristics, including
190 (but not limited to) taxonomic focus, methodology, number- and skill level of the report-
191 ers. Two of the datasets can be regarded as “pure” citizen science datasets (NBIC_{CS}: Citi-
192 zen science species observations from the Species Observation Service in Norway (The
193 Norwegian Biodiversity Information Centre and Hoem, 2020b). eBird: citizen science rec-
194 ords of birds (Levatch and Padilla, 2019)). Five datasets originated from museums and/or
195 universities (KMN: vascular plant registrations from the Agder Museum of Natural History
196 and Botanical Garden (Åsen, 2019). NTNU: vascular plant registrations according to
197 standardised cross-lists (NTNU University Museum, 2020). UiO_{Lichen}: lichen registrations
198 from the University of Oslo Natural History Museum (Natural History Museum, 2020).
199 UiO_{Plant Obs}: vascular plant registrations (observational records) (Natural History Museum,
200 2019b). UiO_{Plant Notes}: vascular plant registrations (field notes) (Natural History Museum,
201 2019a)), and can be regarded to cover somewhat structured surveys and observations by
202 institutional recorders. Two datasets stemmed from a private consultant and –organisation
203 (Jordal: consultant within biology and management (Jordal, 2019) and BioFokus: non-
204 profit organisation providing survey information (Blindheim, 2020)), which both provide
205 biodiversity survey information for decision makers, and can thus be regarded as mainly
206 structured surveys and observations done by institutional recorders. Likewise, the final da-
207 taset (NBIC_{Other}) included datasets and databases from providers not hosting their own
208 GBIF Integrated Publishing Toolkit (IPT) services, such as the Norwegian Environmental
209 Agency – these are likewise regarded as mainly structured, institutional surveys. Data from
210 NBIC are quality controlled internally: the data owner is responsible for the quality of the

211 data. Dubious records are validated by experts, and the data owner is asked to provide evi-
212 dence (e.g. photos) of the record. If these cannot be provided, the record is deleted
213 (Norwegian Biodiversity Information Centre, 2018; Norwegian Species Observation
214 Service, 2020).

215

216 The latest Norwegian Red List of Species was finalised in 2015, 10 years after the first na-
217 tional assessment. The list includes species evaluated as being at risk of extinction in Nor-
218 way, if conditions remain unchanged. The classification follows the same criteria as the
219 IUCN Red List (Henriksen and Hilmo, 2015). In total, $\approx 4,500$ species are currently red-
220 listed; of these are $\approx 2,550$ animals (mainly invertebrates), ≈ 750 plants and $\approx 1,200$ fungi.
221 The first version of the Alien Species List was compiled in 2007 (Gederaas *et al.*, 2012), and
222 the latest version was refined and published in 2018 (Sandvik *et al.*, 2017; The Norwegian
223 Biodiversity Information Centre, 2018b). In total, $\approx 3,000$ species are listed as alien to
224 mainland Norway, $\approx 1,500$ of these have a risk assessment. Of these, ≈ 390 are animals,
225 ≈ 990 are plants and ≈ 100 are fungi. As per the guidelines published by the Norwegian Bi-
226 odiversity Information Centre (Sandvik *et al.*, 2017), we here use the term “alien species”
227 rather than the frequently used “invasive species”. “Alien” refers to “(...) a species introduced
228 outside its natural past or present distribution.” (IUCN, 2020). The term “invasive” suggest inva-
229 sion potential and negative ecological effects, which is not necessarily the case for all alien
230 species. To avoid subjective decisions as to which alien species to classify as “invasive”, all
231 species classified as “alien” on the Alien Species List (Gederaas *et al.*, 2012) were included,
232 and the term “alien” was used rather than “invasive”.

233 Species names of the GBIF records were matched with the Norwegian Red List, and the
234 Norwegian Alien Species List, using synonyms from the GBIF backbone taxonomy, using
235 the package `'rgbif'` (Chamberlain and Boettiger, 2017). Species within the Red List cat-
236 egories “Regionally Extinct”, “Critically Endangered”, “Endangered”, “Vulnerable”, “Near
237 Threatened” and “Data Deficient” are classified as “red-listed”. As the majority of “Data

238 Deficient” species are potentially threatened (Bland *et al.*, 2015), and old records are in-
239 cluded in the analyses, inclusion of the remaining Red List categories is warranted. Species
240 alien to Svalbard, but native to mainland Norway were not listed as alien, neither were alien
241 species which have not yet established, but are evaluated to have the potential to do so
242 within 50 years (The Norwegian Biodiversity Information Centre, 2018).

243 Maps and occurrence records were transformed to the geodetic coordinate reference sys-
244 tem WGS84/UTM zone 32 (epsg: 32632).

245 **Statistical analyses**

246 Taxonomic differences within and between datasets were examined using χ^2 - tests (base
247 package: `'stats'`), testing the null hypothesis of equal distribution of the kingdoms be-
248 tween and within the datasets. Likewise, the distribution of red-listed and alien species be-
249 tween the datasets was tested with a χ^2 -test.

250 To test for temporal patterns in the data, a Mann-Kendall test for a monotonic trend was
251 applied (package: `'trend'` (Pohlert, 2020)). The median sampling year of the datasets
252 were compared with a Kruskal-Wallis-test, followed by a posthoc pairwise Dunn-test with
253 Bonferroni-correction for multiple comparisons (packages: `'stats'` and `'FSA'`
254 (Ogle *et al.*, 2020)).

255 For examining geographical biases, the data were further reduced to match the timeframe
256 of the land-cover data. Only data from year 2004 to (and including) year 2018 were used.
257 Changes in land-cover are assumed to be minimal within this 15-year span. The remaining
258 5,622,260 records were overlaid on the AR50 map (package: `'sp'` (Pebesma and Bivand,
259 2005)). The null hypotheses was that the species occurrence records are randomly distrib-
260 uted across Norway, and the number of records is a function of the area of each land-
261 cover type.

262 5,622,260 points were randomly overlaid on the map 100 times, giving ranges of expected
263 number of points associated with each land-cover type. Dataset names and conservation

264 status (“red-listed”, and “alien”) were assigned randomly to the points in the same propor-
265 tions as in the original data. Generalised Linear Models (Poisson error distribution,
266 “identity” link function) (base package: ‘stats’) were fitted to the number of rec-
267 ords predicted by area of each land-cover type for the simulated data, providing the null
268 models; one separate model for each of the combinations of dataset and conservation sta-
269 tus. Sampling bias was concluded if the observed number fell outside the 0.95 confidence
270 interval of the model. To compare the extent of sampling bias for the different groups, the
271 absolute- and relative residuals were calculated as:

$$272 \text{ Absolute residual} = \text{No. records}_{\text{observed}} - \text{No. records}_{\text{predicted}}$$

273 and

$$274 \text{ Relative residual} = \frac{\text{Absolute residual}}{\text{mean}(\text{No. records}_{\text{observed}}, \text{No. records}_{\text{predicted}})}$$

275 To evaluate the differences in biodiversity patterns obtained using occurrence records from
276 the different datasets, or all in combination, individual-based species accumulation curves
277 were made for each dataset \times conservation status group, and the asymptotic species rich-
278 ness calculated (package: ‘iNEXT’ (Hsieh *et al.*, 2020)).

279

280 All data preparation and analyses were performed in R, version 3.6.1 (R Core Team, 2020).
281 Maps were made in ArcMap version 10.6 (ESRI, 2018).

282

283 **Results**

284 **Taxonomic differences**

285 The number of records from each dataset differed ($\chi^2 = 26\,019\,773$, $df = 9$, $p\text{-value} < 0.001$)
286 with the vast majority of the records belonging to the NBIC_{CS} dataset, followed by the Ui-
287 O_{Plant Notes} (see Table 1 for description of dataset names). The kingdoms were not equally

288 distributed between and within the datasets ($\chi^2 = 3\,813\,957$, $df = 18$, $p\text{-value} < 0.001$). Obvi-
289 ously, the datasets with a specified taxonomic scope were dominated by records belonging
290 to the particular kingdom, but the datasets including several kingdoms differed as well; the
291 BioFokus- and NBIC_{CS} datasets had an overabundance of animals and fungi, whereas the
292 NBIC_{other} dataset only had an overabundance of animal records. The Jordal dataset had an
293 overabundance of plants and fungi (Fig. 2). Within the animal kingdom, birds was the most
294 frequently recorded class, followed by insects and mammals overall. For the multi-taxa da-
295 taset, the distribution within the animal kingdom differed: the BioFokus datasets held
296 most records of insects, followed by birds and mammals, the Jordal dataset was dominated
297 by birds, followed by insects and bivalves, and the NBIC_{CS}- and NBIC_{other} datasets were
298 dominated by records of birds, followed by insects and mammals (Fig. S.4). When account-
299 ing for the different sample sizes, the distribution of red-listed- and alien species differed
300 between the datasets, with the BioFokus-, eBird-, NBIC_{CS}-, NBIC_{other}-, and UiOLichen holding
301 more red-listed-, and the KMN-, Jordal-, NTNU-, UiO_{Plant Notes}- and UiO_{Plant Obs} datasets
302 holding more alien species than what would be expected by random ($\chi^2 = 104\,807$, $df = 9$,
303 $p\text{-value} < 0.001$) (Fig. 2b).

304 **Temporal differences**

305 The Mann-Kendall test detected a tendency in the overall dataset; the number of records
306 had increased over time ($\tau = 16.732$, $n = 200$, $p\text{-value} < 0.001$) (Fig. 3a). Median year dif-
307 fered for all datasets, (medians: KMN = 1986, BioFokus = 2011, eBird = 2015, Jordal =
308 2007, NBIC_{CS} = 2014, NBIC_{other} = 2014, NTNU = 1985, UiOLichen = 2000, UiO_{Plant Notes} =
309 1961, UiO_{Plant Obs} = 2009, Kruskal-Wallis = 496.44, $df = 9$, $p\text{-value} = < 0.001$. $p\text{-value} < 0.001$
310 for all pairwise comparisons) (Fig. 3b).

311 **Geographic differences**

312 The simulated numbers of records within the groups (conservation status \times dataset) were
313 predicted by the area of the specified land cover type (Table 2, Fig. 4).

314 Each land-cover type was relatively over- or under-sampled for different datasets (the ob-
315 served number of records fell outside of the 0.95 Confidence Interval of models based on
316 the simulated data), except for Snow/ice, which was under-sampled by all datasets. The re-
317 sults are summarised in Table 3, and the full table can be seen in the Supporting Infor-
318 mation S.6.

319 Models and results regarding datasets (regardless of conservation status) can be seen in the
320 Supporting Information (Supporting Information S.5).

321 Comparing the absolute residuals between predicted and observed number of records
322 within each land-cover type, the largest numerical discrepancies were seen for open firm
323 ground, developed areas and cultivated land (Fig. 5a). However, comparing the relative re-
324 siduals (disregarding un-mapped areas and snow/ice), only alien records associated with
325 open firm ground showed a consistent pattern between datasets (under-sampling) (Fig. 5b).

326 **Asymptotic species richness**

327 The asymptotic species richness differed for most of the datasets (Supporting Information
328 S.7). For both red-listed- and alien species, only the estimates for the NBIC_{CS} datasets
329 (NBIC_{CS} red-listed = 2 412 (C.I. = 2 333 – 2 513), NBIC_{CS} alien = 867 (C.I. = 833 - 920))
330 overlapped with the estimates for all datasets combined (Combined red-listed = 2 550 (C.I.
331 = 2 469 – 2 654), Combined alien = 861 (C.I. = 836 - 902)).

332

333 **Discussion**

334 Various forms of biases have been shown for the increasing amount of species data availa-
335 ble from open databases, such as GBIF. However the potential taxonomic-, temporal-, and
336 geographical biases differ between datasets according to the origin and characteristics of
337 the datasets, and how these different datasets might complement each other, have not been
338 addressed. Additionally, whether these biases extend to red-listed and alien species remain
339 un-investigated. We found that multi-taxa datasets from GBIF are biased towards different

340 kingdoms (supporting H1a). More records of red-listed species are registered than alien
341 species; (supporting H1b). When categorising the records according to datasets and conser-
342 vation status, the geographical biases differ between the datasets, with a few general pat-
343 terns. Anthropogenic land-covers are generally oversampled (with a few exceptions),
344 whereas less directly human-affected- and/or remote areas are under-sampled (somewhat
345 supporting H3).

346 **Differences in taxonomic groups and conservation status between datasets**

347 The taxonomic bias within and between the datasets differ markedly, both in the sense that
348 several of the datasets are concerned with a single taxonomic group, and in that the multi-
349 taxa datasets are skewed towards a single group. The datasets originating from museums all
350 focus on plants (except for UiO_{Lichen}; lichens are here classified as fungi). These patterns
351 are reflected when comparing the multi-taxa datasets: the two datasets from the Norwegian
352 Biodiversity Information Centre (NBIC) are both dominated by animal records, whereas
353 the BioFokus and Jordal are both dominated by plants. Interestingly, only two out of the
354 ten datasets can be regarded as citizen science, but yet they make up the bulk of the rec-
355 ords. The dominance of birds within these datasets reflect the long-term popularity of or-
356 nithology (Devictor *et al.*, 2010), the incentive for people to report on charismatic, recog-
357 nisable species, and that many citizen science programmes have focused on birds (Tulloch,
358 Mustin, *et al.*, 2013). However, if the datasets dominated by citizen science records are not
359 considered, the avian dominance is much less pronounced. This echoes the taxonomic dif-
360 ferences observed by Troudet *et al.* (2017) and Speed *et al.* (2018). Theobald *et al.* (2015)
361 found the taxonomic bias in citizen science- and institutional datasets to be consistent;
362 however, they did see an overweight of respectively birds and plants in the two groups.
363 This underlines the careful considerations which much be taken eventually when using citi-
364 zen science in multi-taxa analyses – nevertheless, within popular taxa, citizen science rec-
365 ords can be a useful supplement to institutional observations, as this allows for otherwise
366 impossible sample sizes (Tulloch, Possingham, *et al.*, 2013; Powney and Isaac, 2015). Citi-
367 zen science data on popular taxa have proven useful for discovering population trends,

368 conservation and management (e.g. for birds: Lehtikoinen *et al.* (2019) and examples in
369 Sullivan *et al.* (2009)).

370 The datasets with more alien- than red-listed records were all datasets focused on vascular
371 plants; for all other datasets, more red-listed- than alien records were registered. This illus-
372 trates that most species on the Alien Species List are plants (The Norwegian Biodiversity
373 Information Centre, 2018). The dominance of red-listed- compared to alien species among
374 half of the datasets, in particular in the datasets dominated by citizen science records
375 (NBICCs and eBird) points to a greater interest for rarities among citizen scientists and a
376 potential lack of interest or knowledge regarding alien species. Among the other datasets,
377 the difference can be due to a traditionally larger focus on red-listed species, or that red-
378 listed species are likely registered as observations (i.e. not destructively sampled) (NTNU
379 University Museum, 2018), whereas alien species are potentially sampled as specimens to
380 ensure validation later. The numerical difference between the conservation status groups
381 can nevertheless be an artefact of the number of species in either status group: approxi-
382 mately three times as many species are on the Norwegian Red List compared to the Alien
383 Species List (The Norwegian Biodiversity Information Centre, 2015, 2018).

384 **Geographical biases**

385 The most anthropogenic land-cover types have higher numbers of records than what
386 would be expected for most, but not all groups. Developed areas were oversampled overall
387 in all but three datasets (KMN, NTNU and Jordal); when focussing on either red-listed- or
388 alien records, the same pattern emerges, with the exception of the Jordal dataset being
389 oversampled and the UiO_{Plant Obs} being under-sampled for red-listed species. This pattern
390 likely has multiple underlying causes: despite a general omission of cities in ecological his-
391 tory (reviewed by Salomon Cavin and Kull (2017)), the last decades have seen increased fo-
392 cus on urban ecology, especially on cities as centres of spread for alien species (Gaertner *et*
393 *al.*, 2017). This has likely amplified the oversampling of alien species in urban areas. The

394 oversampling of red-listed species are likely a combined effect of roadside bias and inter-
395 est/prestige, as the oversampling is particularly large for datasets dominated by citizen sci-
396 ence records.

397 Agricultural areas are similarly oversampled for most groups. This again reflects the road-
398 side bias, as agricultural areas are generally found near developed areas (Fig. S.1), and thus
399 have high accessibility. Grazing land is particularly oversampled, reflecting how such areas
400 are regarded as of conservation concern, thus warranting attention from different recorders
401 (Pärtel *et al.*, 2005).

402 The picture is highly nuanced regarding the different forest categories. The cases of over-
403 sampling may reflect that sampling tends to be done where high species richness is ex-
404 pected *a priori* (Boakes *et al.*, 2016), the high amount of woodland in Norway (>35%), and
405 the high species richness of forests ($\approx 60\%$ of Norwegian species are associated with wood-
406 lands). The highest number and concentration of red-listed species are found in coniferous
407 woodlands and broad-leaved deciduous woodland, respectively (Gjerde *et al.*, 2010), which
408 is somewhat seen in the positive residuals of red-listed records from most datasets. Some
409 of the datasets hold fewer red-listed records than expected for coniferous- (KMN, eBird,
410 Jordal NBIC_{CS} (red-listed), and UiO_{Plant Obs}) and deciduous (eBird, NBIC_{CS} (red-listed),
411 NTNU (red-listed), UiO_{Lichen}, and UiO_{Plant Obs}) forests. This discrepancy presumably stems
412 from the taxonomical difference between the datasets: red-listed woodland species in Nor-
413 way are mainly fungi, insects and lichens (Gjerde *et al.*, 2010; Henriksen and Hilmo, 2015),
414 and the number of red-listed plants outnumber red-listed animals; according to the
415 Norwegian Biodiversity Information Centre (2015), only 14 out of 82 red-listed birds are
416 associated with forests. Both datasets mainly collected by citizen scientists are heavily dom-
417 inated by (or exclusively consists of) birds, which are easier to observe in open areas. Un-
418 classified forests have fewer records than predicted for almost all datasets, except
419 NBIC_{Other} and UiO_{Plant Obs}, reflecting that this forest type is found in more remote, inacces-
420 sible areas; these two datasets have likely targeted such areas specifically. The land-covers

421 with fewer records than predicted for most of the datasets are characterised by being lo-
422 cated in more remote and/or inaccessible areas, less directly affected by humans:
423 snow/ice-covered areas, mires and open firm ground. In some instances, this reflects genu-
424 ine low species richness and –abundance (discussed below), as is likely the case for
425 “snow/ice” (having the largest relative residuals) and the most alpine cases of “open firm
426 ground”. However, some areas are likely under-sampled due to inaccessibility (e.g. mires),
427 genuine difference in spatial- and taxonomic focus and interest of the datasets.

428 The discrepancies between predicted and observed number of records should be inter-
429 preted with caution. Some land-cover types are naturally more species poor than others. It
430 can thus be expected that a lower number of records should be reported, than would be
431 expected solely from area. This is the case of alpine areas; it is estimated that only $\approx 14\%$ of
432 the native vascular plants of Norway occur in mountains (Austrheim *et al.*, 2010). Alpine
433 areas are here found within the land-cover types “Open firm ground” and “Snow/ice”,
434 both of which have fewer records than predicted by the null models. Consequently, parts
435 of the differences between observations and predictions can be attributed to the null mod-
436 els not taking intrinsic differences in species richness and abundances into account. Never-
437 theless, as we were not modelling species richness, but number of records (a proxy of sam-
438 pling effort), the main signals are mainly attributable to differences in sampling effort.

439 **Dataset complementarity**

440 The general quality of the data found in open databases, such as GBIF is a point worth
441 general discussion. Various opinions on the matter exist (Newbold, 2010; Gaiji *et al.*, 2013;
442 Powney and Isaac, 2015). The biases shown underlines how the individual datasets stored
443 in GBIF are not all compiled with the intention of covering all taxa, periods or habitat.
444 Thus, indiscriminately using such compiled datasets without accounting for the differences
445 in sampling effort (whether this is spatial, temporal or taxonomic) will inevitably lead to
446 flawed results. The differences in both taxonomic- and geographic focus of different da-
447 tasetes from open databases shown in this study raise the question on how to compile such

448 datasets to ensure optimal coverage, and whether datasets with certain origin and charac-
449 teristics are complementary. If multi-taxa management decisions are to be made based on
450 analyses including e.g. GBIF data, several considerations must be taken into account.

451 1. Regarding taxonomic complementarity, it is clear that careful examination of the in-
452 cluded datasets is necessary, as indiscriminate data use will result in taxonomic imbalances.

453 2. Likewise, as the temporal coverage of the datasets is highly variable, timespan of individ-
454 ual datasets should be considered in relation to the questions asked.

455 3. Considering the geographical dissimilarities between the datasets, it is evident that if con-
456 clusions regarding the importance of different land-cover types for species of conservation
457 concern are drawn upon analyses of single datasets, contrasting results will follow.

458 4. The geographic coverage of the single datasets used in analyses should be investigated to
459 ensure that certain areas are not over- or under-represented.

460 The overarching theme of these points can be summarised as not to assume a greater qual-
461 ity and validity of the available data than what is warranted. Care must be taken as to not
462 stretch the conclusions based on single datasets further than the extents of the individual
463 datasets, geographically or taxonomically.

464 **Integrating multiple datasets for understanding and managing biodiversity**

465 Data availability thus remains the main challenge for understanding biodiversity patterns,
466 and ultimately for how we manage biodiversity (Magurran *et al.*, 2019). This study has ex-
467 amined how different datasets, with different origins and characteristics, can complement
468 each other in filling data availability gaps, specifically the gaps for three kingdoms (animals,
469 plants and fungi), red-listed and alien species, and their distributions across land-covers and
470 time.

471 Despite the emerging paradigm of data reuse and sharing among scientists, lack of data
472 publishing is still an issue; only 10% of biocollections are estimated to be digitally available,
473 including data used prior to recent changes in data publishing policies provided by funding

474 agencies and journals (Ball-Damerow *et al.*, 2019). Traditionally, most collected data have
475 been stored locally, and data not directly used in publications have remained unused and
476 potentially forgotten with time (Osawa, 2019). This also leaves the worst case scenario that
477 not all parts of datasets are published. Likewise, standardisation of biodiversity data among
478 data providers is important to ensure interoperability (Poisot *et al.*, 2019). An attempt at
479 this is to use the Darwin Core Archive format adopted by GBIF (Wieczorek *et al.*, 2012;
480 Osawa, 2019). Despite these efforts, substantial quantities of primary biodiversity data (and
481 metadata) remain undiscovered (Chavan and Penev, 2011). This leaves a gap in the founda-
482 tion of biodiversity research. In the light of the results presented here, if the lack of data
483 sharing is uneven among datasets with different origins, the gap is even more severe.

484 Open source, compiled biodiversity data have potential to be used for biodiversity model-
485 ling, if spatially biased sampling effort can be corrected for (Higa *et al.*, 2015). Unfortu-
486 nately, a recent review found that only 69% of the examined papers addressed some aspect
487 of data quality (Ball-Damerow *et al.*, 2019). Our results caution that careful considerations
488 of the data used in such studies are needed; as the contribution from different datasets
489 have changed over time, so has the geographical bias. Therefore, accounting for bias
490 should be a dynamic process, dependent on timespan of the included data and the data
491 contributors. If observational datasets of mixed origins are used indiscriminately, the re-
492 ported spatio-temporal patterns could merely reflect spatio-temporal shifts in bias. Future
493 surveys and citizen science programmes should aim to include otherwise neglected taxo-
494 nomic groups, especially in under-sampled land-cover types, such as remote mountainous
495 areas. In particular, non-avian animals are underrepresented compared to their actual abun-
496 dance, and open firm ground and mires should be investigated more closely. Citizen sci-
497 ence programmes focussing on non-avian taxa should be designed, learning from the suc-
498 cess of previous programmes for e.g. birds (Sullivan *et al.*, 2009), butterflies (Butterfly
499 Conservation, 2020), and bumblebees (Bumblebee Conservation Trust, 2019), and use
500 their established frameworks. Both citizen scientists and institutional recorders should be

501 encouraged to record observations in secluded areas, and to include observations of “less
502 prestigious” species.

503 The quality of data from respectively institutional recorders and citizen scientists will vary
504 immensely depending on methods and organism group. Whereas trained professionals
505 likely exhibit greater skills regarding some of the more challenging groups, this is not nec-
506 essarily the case for all taxa. If quality can be ensured, citizen scientists can provide other-
507 wise impossible amounts of data to facilitate science-policy impact of the sustainable biodi-
508 versity management. This study has shown the different biases from different datasets, and
509 illustrate some of the challenges with accounting for all of them in a single study.

510

511 **Authors’ contributions**

512 TKP, GA, JDMS and VG conceived the idea and designed the methodology; TKP re-
513 trieved and analysed the data; TKP wrote the first draft of the manuscript. All authors con-
514 tributed critically to the drafts and gave final approval for publication.

515

516 **Data availability statement**

517 All relevant data are available from public repository (GBIF Occurrence Download –
518 19.11.2019, doi: 10.15468/dl.dmdxne) (GBIF.org, 2019b).

519 Land-cover data are available through Kartkatalogen (Geonorge, 2019), and was down-
520 loaded on 23.11.2019.

521 All R code written to perform the data download- and analyses can be viewed and down-
522 loaded in a public repository: <https://doi.org/10.5281/zenodo.4455460>.

523

524 **References**

- 525 Amano, T., Lamming, J. D. L. and Sutherland, W. J. (2016) ‘Spatial Gaps in Global
526 Biodiversity Information and the Role of Citizen Science’, *BioScience*, 66(5), pp. 393–
527 400. doi: 10.1093/biosci/biw022.
- 528 Åsen, P. (2019) *Vascular Plants, Field notes, Agder naturmuseum (KMN). Version 1.160. Agder*
529 *Museum of Natural History and Botanical Garden. Occurrence dataset.*
530 <https://doi.org/10.15468/gja4jo>, Accessed: 15 May 2020.
- 531 Austrheim, G., Bråthen, K. A., Ims, R. A., Mysterud, A. and Ødegaard, F. (2010) ‘Alpine
532 environment’, in Kålås, J. A., Henriksen, S., Skjelseth, S., and Viken, Å. E. (eds)
533 *Environmental conditions and impacts for Red List species. Norway: Norwegian*
534 *Biodiversity Information Centre*, pp. 107–117.
- 535 Ball-Damerow, J. E., Brenskelle, L., Barve, N., Soltis, P. S., Sierwald, P., Bieler, R.,
536 LaFrance, R., *et al.* (2019) ‘Research applications of primary biodiversity databases
537 in the digital age’, *PLoS ONE*, 14(9), pp. 1–26. doi: 10.1371/journal.pone.0215794.
- 538 Beck, J., Böller, M., Erhardt, A. and Schwanghart, W. (2014) ‘Spatial bias in the GBIF
539 database and its effect on modeling species’ geographic distributions’, *Ecological*
540 *Informatics*, 19, pp. 10–15. doi: 10.1016/j.ecoinf.2013.11.002.
- 541 Bland, L. M., Collen, B., Orme, C. D. L. and Bielby, J. (2015) ‘Predicting the conservation
542 status of data-deficient species’, *Conservation Biology*, 29(1), pp. 250–259. doi:
543 10.1111/cobi.12372.
- 544 Blindheim, T. (2020) *BioFokus. Version 1.1384. Occurrence dataset.*
545 <https://doi.org/10.15468/jxbhqx>, Accessed: 15 May 2020.
- 546 Boakes, E. H., Gliozzo, G., Seymour, V., Harvey, M., Smith, C., Roy, D. B. and Haklay,
547 M. (2016) ‘Patterns of contribution to citizen science biodiversity projects increase
548 understanding of volunteers’ recording behaviour’, *Scientific Reports*, 6, pp. 1–11. doi:
549 10.1038/srep33051.
- 550 Bonney, R., Cooper, C. B., Dickinson, J., Kelling, S., Phillips, T., Rosenberg, K. V. and
551 Shirk, J. (2009) ‘Citizen Science: A Developing Tool for Expanding Science

552 Knowledge and Scientific Literacy’, *BioScience*, 59(11), pp. 977–984. doi:
553 10.1525/bio.2009.59.11.9.

554 Bumblebee Conservation Trust (2019) *Bumblebee Conservation Trust*.
555 <https://www.bumblebeeconservation.org/>, Accessed: 11 February 2020.

556 Butterfly Conservation (2020) *Butterfly Conservation*. <https://butterfly-conservation.org/>,
557 Accessed: 11 February 2020.

558 Bystriakova, N., Peregrym, M., Erkens, R. H. J., Bezsmertna, O. and Schneider, H. (2012)
559 ‘Sampling bias in geographic and environmental space and its effect on the
560 predictive power of species distribution models’, *Systematics and Biodiversity*, 10, pp.
561 305–315. doi: 10.1080/14772000.2012.705357.

562 Chamberlain, S. and Boettiger, C. (2017) ‘R Python, and Ruby clients for GBIF species
563 occurrence data.’

564 Chandler, M., See, L., Copas, K., Bonde, A. M. Z., López, B. C., Danielsen, F., Legind, J.
565 K., *et al.* (2017) ‘Contribution of citizen science towards international biodiversity
566 monitoring’, *Biological Conservation*, 213, pp. 280–294. doi:
567 10.1016/j.biocon.2016.09.004.

568 Chavan, V. and Penev, L. (2011) ‘The data paper: a mechanism to incentivize data
569 publishing in biodiversity science.’, *BMC bioinformatics*, 12. doi: 10.1186/1471-2105-
570 12-S15-S2.

571 Devictor, V., Whittaker, R. J. and Beltrame, C. (2010) ‘Beyond scarcity: Citizen science
572 programmes as useful tools for conservation biogeography’, *Diversity and
573 Distributions*, 16(3), pp. 354–362. doi: 10.1111/j.1472-4642.2009.00615.x.

574 Dickinson, J. L., Zuckerberg, B. and Bonter, D. N. (2010) ‘Citizen Science as an
575 Ecological Research Tool: Challenges and Benefits’, *Annual Review of Ecology,
576 Evolution, and Systematics*, 41(1), pp. 149–172. doi: 10.1146/annurev-ecolsys-102209-
577 144636.

578 Dorazio, R. M. (2014) ‘Accounting for imperfect detection and survey bias in statistical
579 analysis of presence-only data’, *Global Ecology and Biogeography*, 23(12), pp. 1472–
580 1484. doi: 10.1111/geb.12216.

581 ESRI (2018) 'ArcGIS Desktop'. Redlands: Environmental Systems Research Institute.
582 Gaertner, M., Wilson, J. R. U., Cadotte, M. W., MacIvor, J. S., Zenni, R. D. and
583 Richardson, D. M. (2017) 'Non-native species in urban environments: patterns,
584 processes, impacts and challenges', *Biological Invasions*, 19(12), pp. 3461–3469. doi:
585 10.1007/s10530-017-1598-7.

586 Gaiji, S., Chavan, V., Ariño, A. H., Otegui, J., Hobern, D., Sood, R. and Robles, E. (2013)
587 'Content assessment of the primary biodiversity data published through GBIF
588 network: Status, challenges and potentials', *Biodiversity Informatics*, 8(2), pp. 94–172.
589 doi: 10.17161/bi.v8i2.4124.

590 Ganzevoort, W., van den Born, R. J. G., Halffman, W. and Turnhout, S. (2017) 'Sharing
591 biodiversity data: citizen scientists' concerns and motivations', *Biodiversity and
592 Conservation*. Springer Netherlands, 26(12), pp. 2821–2837. doi: 10.1007/s10531-
593 017-1391-z.

594 GBIF.org (2019a) *GBIF Home Page*. <https://www.gbif.org/>, Accessed: 19 November
595 2019.

596 GBIF.org (2019b) 'GBIF Occurrence Download (19 November 2019). Accessed from R
597 via rgbiP'. doi: <https://doi.org/10.15468/dl.dmdxne>.

598 Gederaas, L., Moen, T. L., Skjelseth, S. and Larsen, L.-K. (2012) *Alien species in Norway—
599 with the Norwegian Black List 2012*. Norway: Norwegian Biodiversity Information
600 Centre.

601 Geldmann, J., Heilmann-Clausen, J., Holm, T. E., Levinsky, I., Markussen, B., Olsen, K.,
602 Rahbek, C., *et al.* (2016) 'What determines spatial bias in citizen science? Exploring
603 four recording schemes with different proficiency requirements', *Diversity and
604 Distributions*, 22(11), pp. 1139–1149. doi: 10.1111/ddi.12477.

605 Geonorge (2019) *Kartkatalogen*. [https://kartkatalog.geonorge.no/metadata/4bc2d1e0-
606 f693-4bf2-820d-c11830d849a3](https://kartkatalog.geonorge.no/metadata/4bc2d1e0-f693-4bf2-820d-c11830d849a3), Accessed: 23 November 2019.

607 Gjerde, I., Brandrud, T. E., Ohlson, M. and Ødegaard, F. (2010) 'Woodland', in Kålås, J.
608 A., Henriksen, S., Skjelseth, S., and Viken, Å. E. (eds) *Environmental conditions and
609 impacts for Red List species*. Norway: Norwegian Biodiversity Information Centre, pp.

610 67–78.

611 Guisan, A., Tingley, R., Baumgartner, J. B., Naujokaitis-Lewis, I., Sutcliffe, P. R., Tulloch,
612 A. I. T., Regan, T. J., *et al.* (2013) ‘Predicting species distributions for conservation
613 decisions’, *Ecology Letters*, 16, pp. 1424–1435. doi: 10.1111/ele.12189.

614 Henriksen, S. and Hilmo, O. (2015) *Norwegian Red List of species — methods and results,*
615 *Norwegian Red List of Species.*

616 Higa, M., Yamaura, Y., Koizumi, I., Yabuhara, Y., Senzaki, M. and Ono, S. (2015)
617 ‘Mapping large-scale bird distributions using occupancy models and citizen data
618 with spatially biased sampling effort’, *Diversity and Distributions*, 21(1), pp. 46–54. doi:
619 10.1111/ddi.12255.

620 Hsieh, T. C., Ma, K. H. and Chao, A. (2020) ‘iNEXT: iNterpolation and EXTrapolation
621 for species diversity. R package version 2.0.20’.

622 IUCN (2020) *Invasive species.* [https://www.iucn.org/theme/species/our-work/invasive-](https://www.iucn.org/theme/species/our-work/invasive-species)
623 [species](https://www.iucn.org/theme/species/our-work/invasive-species), Accessed: 6 January 2021.

624 Jetz, W., McPherson, J. M. and Guralnick, R. P. (2012) ‘Integrating biodiversity
625 distribution knowledge: Toward a global map of life’, *Trends in Ecology and Evolution*,
626 27(3), pp. 151–159. doi: 10.1016/j.tree.2011.09.007.

627 Jordal, J. B. (2019) *Jordal. Version 1.91. Biolog J.B. Jordal AS. Occurrence dataset.*
628 <https://doi.org/10.15468/wqsad9>, Accessed: 15 May 2020.

629 Kadmon, R., Farber, O. and Danin, A. (2004) ‘Effect of roadside bias on the accuracy of
630 predictive maps produced by bioclimatic models’, *Ecological Applications*, 14(2), pp.
631 401–413. doi: 10.1890/02-5364.

632 Kramer-Schadt, S., Niedballa, J., Pilgrim, J. D., Schröder, B., Lindenborn, J., Reinfelder,
633 V., Stillfried, M., *et al.* (2013) ‘The importance of correcting for sampling bias in
634 MaxEnt species distribution models’, *Diversity and Distributions*, 19(11), pp. 1366–
635 1379. doi: 10.1111/ddi.12096.

636 Larson, L. R., Cooper, C. B., Futch, S., Singh, D., Shipley, N. J., Dale, K., LeBaron, G. S.,
637 *et al.* (2020) ‘The diverse motivations of citizen scientists: Does conservation
638 emphasis grow as volunteer participation progresses?’, *Biological Conservation.*

639 Elsevier, 242(January), p. 108428. doi: 10.1016/j.biocon.2020.108428.

640 Lehtikoinen, A., Brotons, L., Calladine, J., Campedelli, T., Escandell, V., Flousek, J.,
641 Grueneberg, C., *et al.* (2019) ‘Declining population trends of European mountain
642 birds’, *Global Change Biology*, 25(2), pp. 577–588. doi: 10.1111/gcb.14522.

643 Levatich, T. and Padilla, F. (2019) *EOD - eBird Observation Dataset. Cornell Lab of*
644 *Ornithology. Occurrence dataset.* <https://doi.org/10.15468/aomfnb>, Accessed: 15 May
645 2020.

646 Luck, G. W. (2007) ‘A review of the relationships between human population density and
647 biodiversity’, *Biological Reviews*, 82(4), pp. 607–645. doi: 10.1111/j.1469-
648 185X.2007.00028.x.

649 Magurran, A. E., Dornelas, M., Moyes, F. and Henderson, P. A. (2019) ‘Temporal β
650 diversity—A macroecological perspective’, *Global Ecology and Biogeography*, 28(12),
651 pp. 1949–1960. doi: 10.1111/geb.13026.

652 Miller-Rushing, A., Primack, R. and Bonney, R. (2012) ‘The history of public
653 participation in ecological research’, *Frontiers in Ecology and the Environment*, 10(6), pp.
654 285–290. doi: 10.1890/110278.

655 Natural History Museum, U. of O. (2019a) *Vascular Plants, Field notes, Oslo (O). Version*
656 *1.186. Occurrence dataset.* <https://doi.org/10.15468/w8gru5>, Accessed: 15 May 2020.

657 Natural History Museum, U. of O. (2019b) *Vascular Plants, Observations, Oslo (O). Version*
658 *1.181. Occurrence dataset.* <https://doi.org/10.15468/tvnjk7>, Accessed: 15 May 2020.

659 Natural History Museum, U. of O. (2020) *Lichen field notes, Oslo (O). Version 1.180.*
660 *Occurrence dataset.* <https://doi.org/10.15468/zrxfcu>, Accessed: 15 May 2020.

661 Newbold, T. (2010) ‘Applications and limitations of museum data for conservation and
662 ecology, with particular attention to species distribution models’, *Progress in Physical*
663 *Geography*, 34(1), pp. 3–22. doi: 10.1177/0309133309355630.

664 Norwegian Biodiversity Information Centre (2018) *Kvalitetsikring og foredling av dataene.*
665 [https://artsdatabanken.no/Pages/233743/Kvalitetssikring_og_foredling_av_datae](https://artsdatabanken.no/Pages/233743/Kvalitetssikring_og_foredling_av_dataene)
666 [ne](https://artsdatabanken.no/Pages/233743/Kvalitetssikring_og_foredling_av_dataene), Accessed: 29 June 2020.

667 Norwegian Institute of Bioeconomy Research (2019) *AR50.*

668 <https://nibio.no/tema/jord/arealressurser/ar50>, Accessed: 23 November 2019.

669 Norwegian Species Observation Service (2020) *Håndtering av avvikende rapporter i*
670 *Artsobservasjoner*. <https://www.artsobservasjoner.no/Home/DeviatingReports>,
671 Accessed: 29 June 2020.

672 NTNU University Museum (2018) *Samlingsplan 2018 - 2025, Samlingsplan NTNU*
673 *Vitenskapsmuseet*. Trondheim, Norway.

674 NTNU University Museum (2020) *Vascular plant field notes, NTNU University Museum.*
675 *Version 1.97. Sampling event dataset*. <https://doi.org/10.15468/kkb2x0>, Accessed: 15
676 May 2020.

677 Ogle, D. H., Wheeler, P. and Dinno, A. (2020) ‘Fisheries Stock Analysis. R package
678 version 0.8.30’.

679 Osawa, T. (2019) ‘Perspectives on biodiversity informatics for ecology’, *Ecological Research*,
680 34(4), pp. 446–456. doi: 10.1111/1440-1703.12023.

681 Pärtel, M., Bruun, H. H. and Sammuli, M. (2005) ‘Biodiversity in temperate European
682 grasslands: origin and conservation.’, *Grassland Science in Europe*, 10, pp. 1–14.

683 Pebesma, E. and Bivand, R. (2005) ‘Classes and methods for spatial data in R’, *R News*,
684 5(2).

685 Phillips, S. J., Dudík, M., Dudík, D., Elith, J., Graham, C. H., Lehmann, A., Leathwick,
686 J., *et al.* (2009) ‘Sample selection bias and presence-only distribution models:
687 implications for background and pseudo-absence data’, *Ecological Applications*, 19(1),
688 pp. 181–197.

689 Pohlert, T. (2020) ‘trend: Non-Parametric Trend Tests and Change-Point Detection. R
690 package version 1.1.2’.

691 Poisot, T., Bruneau, A., Gonzalez, A., Gravel, D. and Peres-Neto, P. (2019) ‘Ecological
692 Data Should Not Be So Hard to Find and Reuse’, *Trends in Ecology and Evolution*,
693 34(6), pp. 494–496. doi: 10.1016/j.tree.2019.04.005.

694 Powney, G. D. and Isaac, N. J. B. (2015) ‘Beyond maps: A review of the applications of
695 biological records’, *Biological Journal of the Linnean Society*, 115(3), pp. 532–542. doi:
696 10.1111/bij.12517.

- 697 R Core Team (2020) ‘R: A language and environment for statistical computing’. Vienna,
698 Austria: R Foundation for Statistical Computing.
- 699 Robinson, O. J., Ruiz-Gutierrez, V. and Fink, D. (2018) ‘Correcting for bias in
700 distribution modelling for rare species using citizen science data’, *Diversity and*
701 *Distributions*, 24(4), pp. 460–472. doi: 10.1111/ddi.12698.
- 702 Salomon Cavin, J. and Kull, C. A. (2017) ‘Invasion ecology goes to town: from disdain to
703 sympathy’, *Biological Invasions*, 19(12), pp. 3471–3487. doi: 10.1007/s10530-017-
704 1588-9.
- 705 Sandvik, H., Gederaas, L. and Hilmo, O. (2017) *Guidelines for the Generic Ecological Impact*
706 *Assessment of Alien Species*. version 3. Trondheim: Norwegian Biodiversity
707 Information Centre.
- 708 Speed, J. D. M., Bendiksby, M., Finstad, A. G., Hassel, K., Kolstad, L. and Prestø, T.
709 (2018) ‘Contrasting spatial , temporal and environmental patterns in observation
710 and specimen based species occurrence data’, *PLoS Biology*, 13(4), pp. 1–17. doi:
711 10.1371/journal.pone.0196417.
- 712 Statistics Norway (2020) *Statistisk Sentralbyrå*. <https://www.ssb.no/>, Accessed: 6 January
713 2020.
- 714 Sullivan, B. L., Wood, C. L., Iliff, M. J., Bonney, R. E., Fink, D. and Kelling, S. (2009)
715 ‘eBird: A citizen-based bird observation network in the biological sciences’,
716 *Biological Conservation*, 142(10), pp. 2282–2292. doi: 10.1016/j.biocon.2009.05.006.
- 717 Syfert, M. M., Joppa, L., Smith, M., Coomes, D., Bachman, S. and Brummitt, N. (2014)
718 ‘Using species distribution models to inform IUCN Red List assessments’, *Biological*
719 *Conservation*, 177, pp. 174–184.
- 720 The Norwegian Biodiversity Information Centre (2015) *Norwegian Red List for species*.
721 [https://www.biodiversity.no/Pages/135380/Norwegian_Red_List_for_Species?Ke](https://www.biodiversity.no/Pages/135380/Norwegian_Red_List_for_Species?Key=14)
722 [y=14](https://www.biodiversity.no/Pages/135380/Norwegian_Red_List_for_Species?Key=14), Accessed: 30 December 2019.
- 723 The Norwegian Biodiversity Information Centre (2018) *The Alien Species List of Norway*.
724 <https://www.biodiversity.no/alien-species>, Accessed: 30 December 2019.
- 725 The Norwegian Biodiversity Information Centre and Hoem, S. (2020a) *Norwegian*

726 *Biodiversity Information Centre - Other datasets. Version 13.132. Occurrence dataset.*
727 <https://doi.org/10.15468/tm56sc>, Accessed: 15 May 2020.

728 The Norwegian Biodiversity Information Centre and Hoem, S. (2020b) *Norwegian Species*
729 *Observation Service. Version 1.82. Occurrence dataset.* The Norwegian Biodiversity
730 Information Centre (NBIC). <https://doi.org/10.15468/zjhzl>, Accessed: 15 May
731 2020.

732 Theobald, E. J., Ettinger, A. K., Burgess, H. K., DeBey, L. B., Schmidt, N. R., Froehlich,
733 H. E., Wagner, C., *et al.* (2015) ‘Global change and local solutions: Tapping the
734 unrealized potential of citizen science for biodiversity research’, *Biological*
735 *Conservation*, 181, pp. 236–244. doi: 10.1016/j.biocon.2014.10.021.

736 Thuiller, W., Richardson, D. M., Pyšek, P., Midgley, G. F., Hughes, G. O. and Rouget, M.
737 (2005) ‘Niche-based modelling as a tool for predicting the risk of alien plant
738 invasions at a global scale’, *Global Change Biology*, 11, pp. 2234–2250. doi:
739 10.1111/j.1365-2486.2005.01018.x.

740 Tiago, P., Gouveia, M. J., Capinha, C., Santos-Reis, M. and Pereira, H. M. (2017) ‘The
741 influence of motivational factors on the frequency of participation in citizen science
742 activities’, *Nature Conservation*, 18, pp. 61–78. doi:
743 10.3897/natureconservation.18.13429.

744 Troudet, J., Grandcolas, P., Blin, A., Vignes-Lebbe, R. and Legendre, F. (2017)
745 ‘Taxonomic bias in biodiversity data and societal preferences’, *Scientific Reports*, 7(1),
746 pp. 1–14. doi: 10.1038/s41598-017-09084-6.

747 Tulloch, A. I. T., Mustin, K., Possingham, H. P., Szabo, J. K. and Wilson, K. A. (2013)
748 ‘To boldly go where no volunteer has gone before: Predicting volunteer activity to
749 prioritize surveys at the landscape scale’, *Diversity and Distributions*, 19(4), pp. 465–
750 480. doi: 10.1111/j.1472-4642.2012.00947.x.

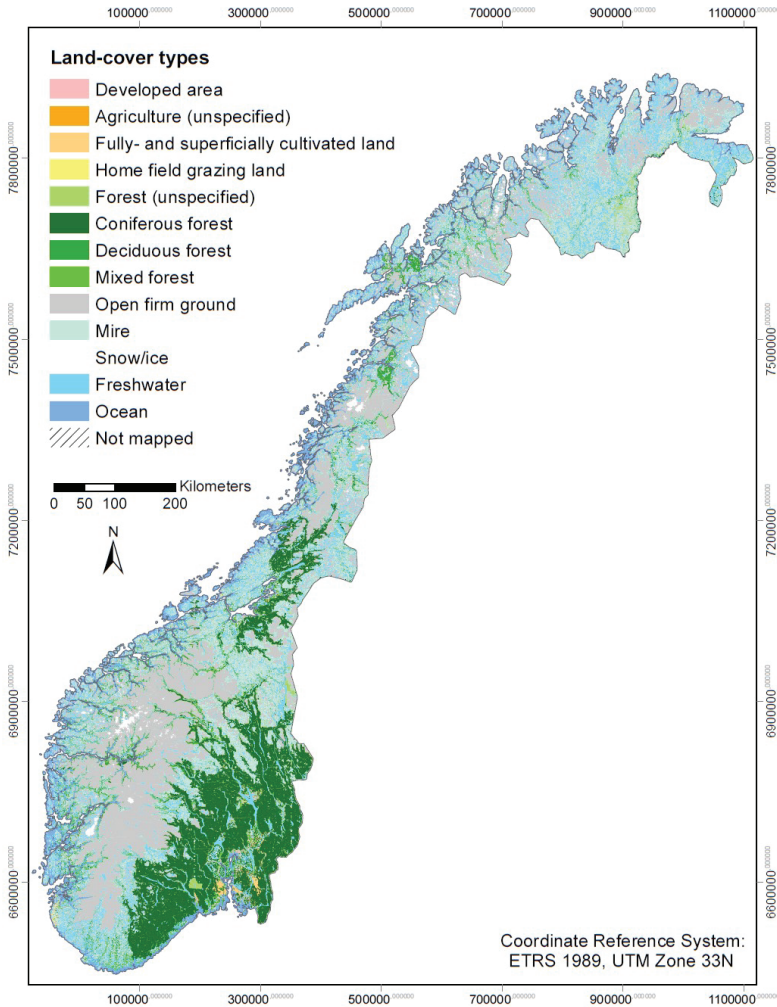
751 Tulloch, A. I. T., Possingham, H. P., Joseph, L. N., Szabo, J. and Martin, T. G. (2013)
752 ‘Realising the full potential of citizen science monitoring programs’, *Biological*
753 *Conservation*, 165, pp. 128–138. doi: 10.1016/j.biocon.2013.05.025.

754 Tye, C. A., McCleery, R. A., Fletcher, R. J., Greene, D. U. and Butryn, R. S. (2017)

755 'Evaluating citizen vs. professional data for modelling distributions of a rare
756 squirrel', *Journal of Applied Ecology*, 54(2), pp. 628–637. doi: 10.1111/1365-
757 2664.12682.

758 Wieczorek, J., Bloom, D., Guralnick, R., Blum, S., Döring, M., Giovanni, R., Robertson,
759 T., *et al.* (2012) 'Darwin core: An evolving community-developed biodiversity data
760 standard', *PLoS ONE*, 7(1). doi: 10.1371/journal.pone.0029715.

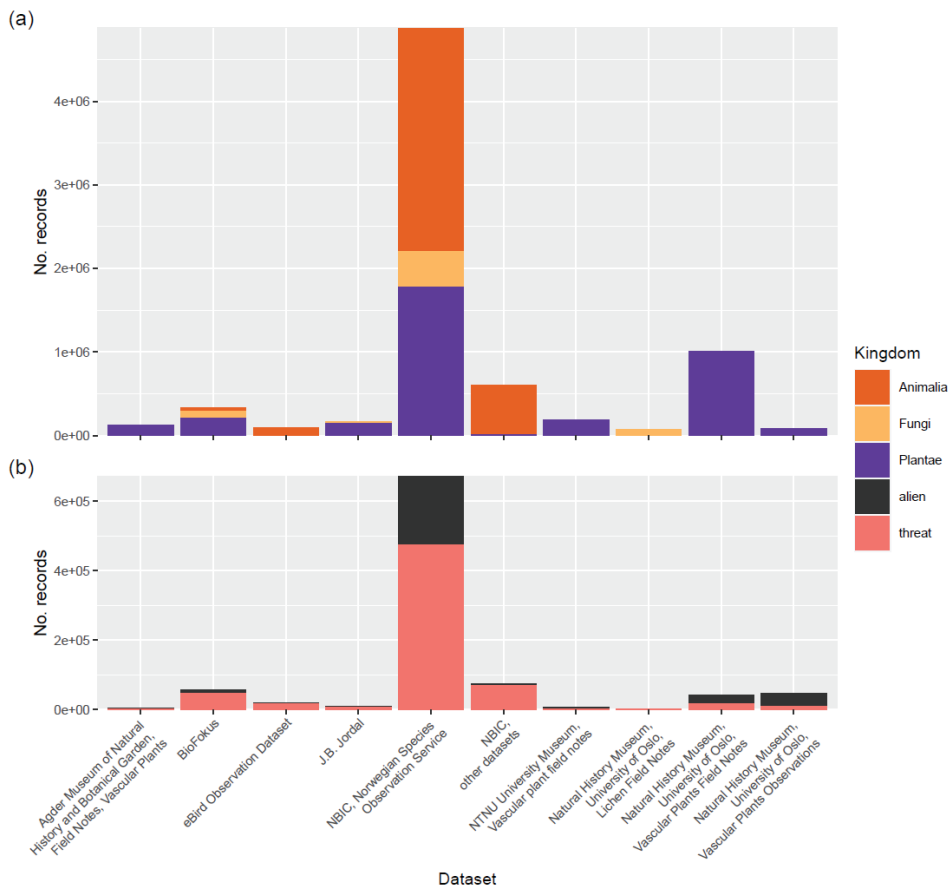
761 Yañez-Arenas, C., Guevara, R., Martínez-Meyer, E., Mandujano, S. and Lobo, J. M.
762 (2014) 'Predicting species' abundances from occurrence data: Effects of sample size
763 and bias', *Ecological Modelling*, 294, pp. 36–41. doi: 10.1016/j.ecolmodel.2014.09.014.
764
765



767

768 Figure 1. Map of Norway. Detailed maps of the individual land-cover types are shown in
769 the Supporting Information S.1, Fig. S.1.

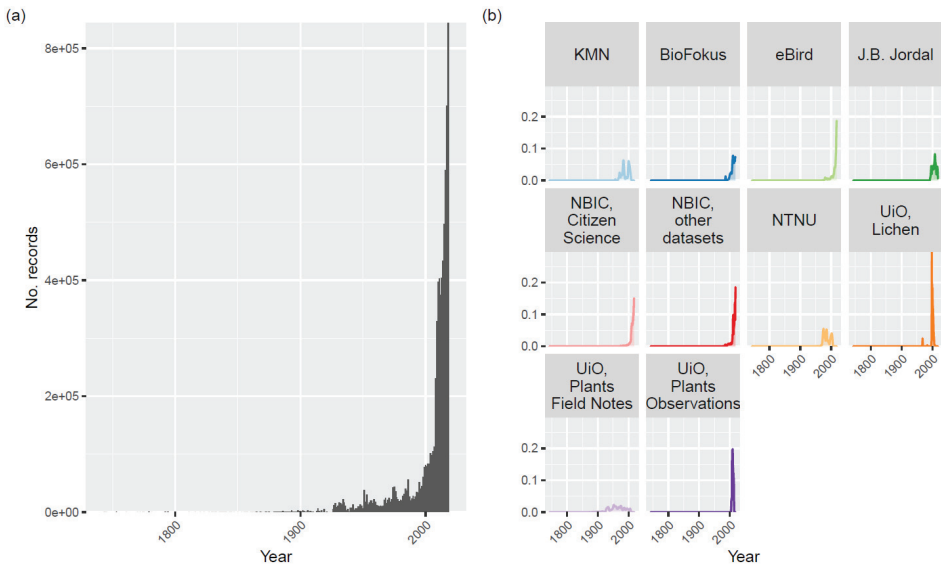
770



771

772 Figure 2. Number of records within each of the datasets used in the analyses. (a) Number
 773 of records from the included kingdoms in each dataset; (b) number of red-listed- or alien
 774 species records in each dataset. Note the differences in y-axis values due to species neither
 775 on the Red List nor the Alien Species List included in (b).

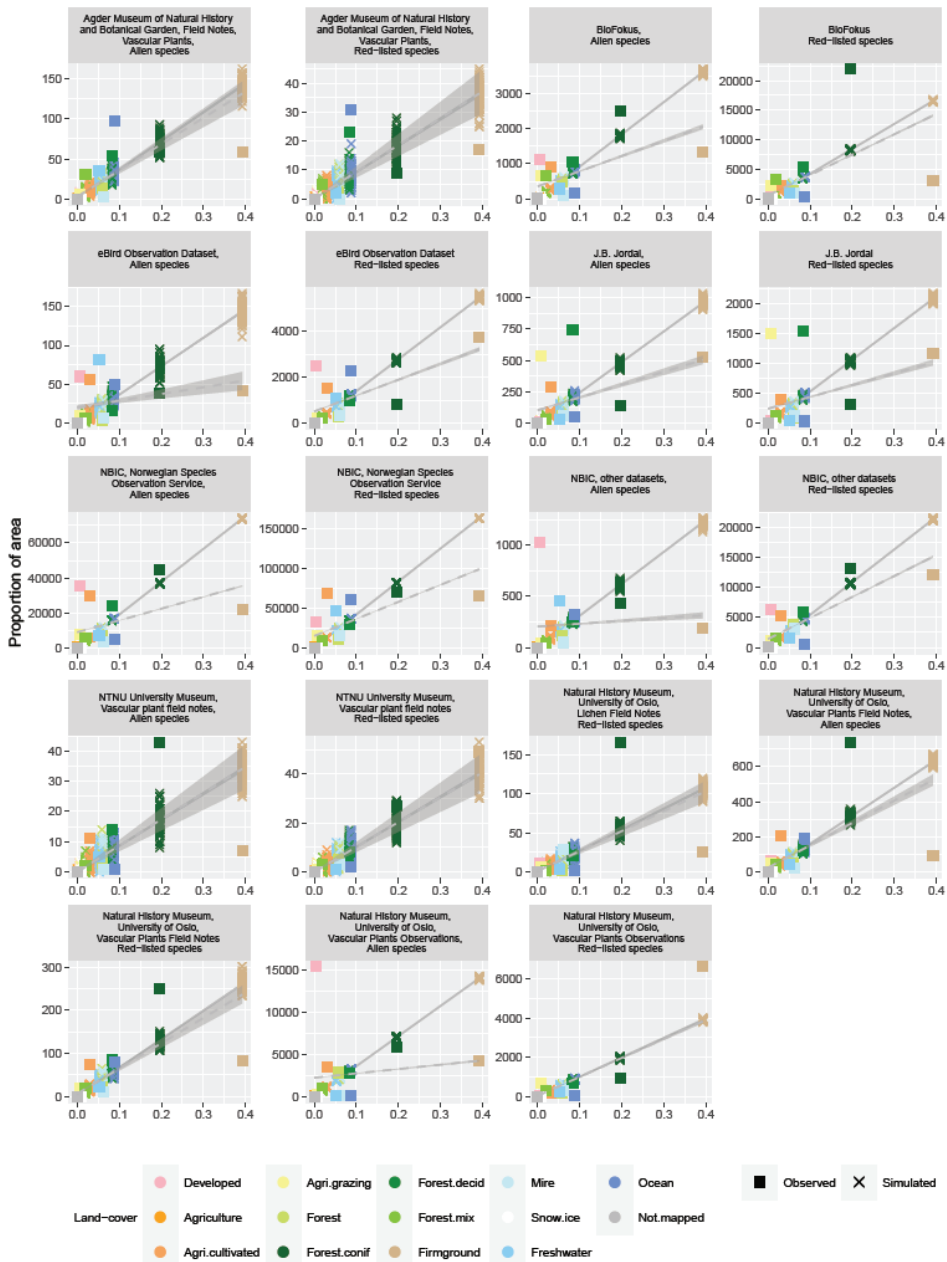
776



777

778 Figure 3. (a) Number of GBIF records across years in total. (b) Density plots of the number of records, divided by datasets. Note that the y-axis in (b) indicate proportion rather
779 than absolute number. Acronyms refers to the datasets described in Table 1.
780

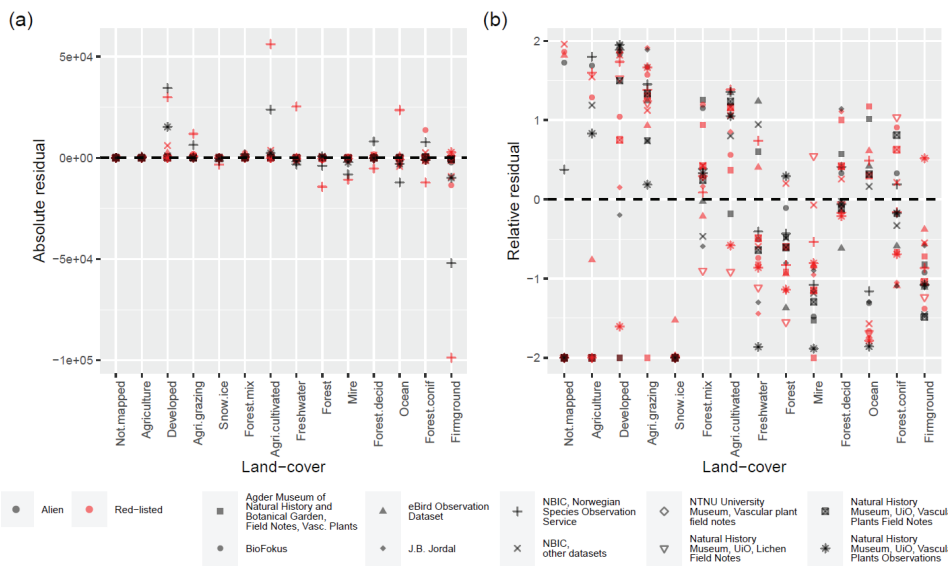
781



783 Figure 4. Null models (GLM) of the number of records as a function of area (proportion
 784 of total area within Norway) for the simulated data (crosses), vs. the observed number of
 785 records for each land-cover type (squares). Solid lines indicate model predictions; gray ribbons
 786 indicate the 0.95 confidence interval. Dashed lines indicate regressions similar to the
 787 null-models fitted through the observed values.

788

789



790

791 Figure 5. Differences between observed number of records within each land-cover type,
 792 and the number of records predicted by area. (a) Absolute residuals

793 ($No. records_{observed} - No. records_{predicted}$); (b) relative residuals

794 ($\frac{Absolute\ residual}{mean(No. records_{observed}, No. records_{predicted})}$). Colours indicate conservation status,

795 shapes indicate dataset. The land-cover types are ordered increasingly with respect to area.

796

797 Table 1. Datasets included in the analyses. The datasets are displayed in descending order according to the total no. records.
 798 The description is simplified from the description and methodology as presented on the GBIF webpage. More detailed de-
 799 scriptions can be found in Supporting Information S.2., Table S.2.

Dataset Name (abbreviation)	Publisher (reference)	# Animals			Description
		# Plants			
		# Fungi			
Norwegian Species Observation Service, (NBIC _{CS})	Norwegian Biodiversity Information Centre (The Norwegian Biodiversity Information Centre and Hoem, 2020b)	2 678 373			Citizen science species observations (Artsobservasjonet.no)
		1 776 878			
		422 930			
Vascular Plants, Field notes, Oslo (UIO _{Plant Notes})	Natural History Museum, University of Oslo (Natural History Museum, 2019a)	0			Vascular Plants, Field notes from Norway
		1 006 937			
		0			
Norwegian Biodiversity Information Centre - Other datasets (NBIC _{Other})	Norwegian Biodiversity Information Centre (The Norwegian Biodiversity Information Centre and Hoem, 2020a)	585 898			Other data providers sharing occurrence data via the Norwegian Species Map Service
		17288			
		289			
BioFokus (BioFokus)		39 179			
		212 265			

	BioFokus (Blindheim, 2020)	82724	A non-profit organisation providing biodiversity survey information on biological diversity to decision makers and the public in Norway.
Vascular plant field notes, NTNU University Museum (NTNU)	NTNU University Museum (NTNU University Museum, 2020)	0 188 910 0	Vascular plant field notes, Norway, using standardised cross-lists. The dataset is a collection of observations made during various research projects at the NTNU University Museum
Jordal (Jordal)	Biolog J.B. Jordal AS (Jordal, 2019)	282 150 411 21 856	John Bjarne Jordal, sole proprietor. Consultant within biology and nature management)
Vascular Plants, Field notes, Agder naturmuseum (KMN) (KMN)	Agder Museum of Natural History and Botanical Garden (Åsen, 2019)	0 125 115 0	Vascular Plants, Field notes from Norway
EOD - eBird Observation Dataset (eBird)	Cornell Lab of Ornithology (Levatch and Padilla, 2019)	93 109 0 0	eBird: a collective enterprise taking a novel approach to citizen science by developing cooperative partnerships among experts in a wide range of fields
		0	Vascular Plants, Observations, Oslo

Vascular Plants, Observations, Oslo (UO _{Plant} Obs)	Natural History Museum, University of Oslo (Natural History Museum, 2019b)	82 634	
		0	
Lichen field notes, Oslo (UO _{Lichen})	Natural History Museum, University of Oslo (Natural History Museum, 2020)	0	Lichens, Field notes from Norway
		0	
		75 512	

800

801

XXXIX

802 Table 2. Model output. Simulated occurrence data randomly distributed across the AR50
803 map; conservation status and dataset name assigned in the same proportions as for the
804 GBIF data (100 repetitions). Generalised linear models (Poisson error distribution,
805 “identity”-link function) of the simulated data were fitted, predicting number of rec-
806 ords falling within each land-cover by the area of the respective land-cover type. P-values
807 below 0.05 are highlighted in bold text. Acronyms refers to the datasets described in Table
808 1.

(a) Red-listed species occurrence records				
	Estimate	Std. error	z-value	p-value
<i>Dataset: KMN</i>				
Intercept	3.337e-04	3.618e-06	-92.25	<0.001
Proportion of total area	9.218e+01	9.601e-01	96.00	<0.001
<i>Dataset: BioFokus</i>				
Intercept	-4.525e-02	3.253e-02	-1.39	0.164
Proportion of total area	4.223e+04	2.056e+01	2054.50e	<0.001
<i>Dataset: eBird</i>				
Intercept	-2.288e-03	2.173e-02	-0.105	0.916
Proportion of total area	1.397e+04	1.182e+01	1181.673	<0.001
<i>Dataset: Jordal</i>				
Intercept	1.002e-02	1.681e-02	0.596	0.551
Proportion of total area	5.324e+03	7.300e+00	729.297	<0.001
<i>Dataset: NBIC_{CS}</i>				
Intercept	-8.217e-02	1.181e-01	0.696	0.486
Proportion of total area	4.165e+05	6.456e+01	645.341	<0.001
<i>Dataset: NBIC_{Other}</i>				
Intercept	-1.640e-03	4.370e-02	-0.038	0.97
Proportion of total area	5.451e+04	2.335e+01	2333.829	<0.001
<i>Dataset: NTNU</i>				
Intercept	-3.694e-04	3.680e-06	-100.4	<0.001
Proportion of total area	1.020e+02	1.010e+00	101.0	<0.001

<i>Dataset: UiOLichen</i>				
Intercept	-9.760e-04	6.637e-06	-147.1	<0.001
Proportion of total area	2.696e+02	1.642e+00	164.2	<0.001
<i>Dataset: UiOPlant Notes</i>				
Intercept	-2.406e-03	2.899e-05	-83.01	<0.001
Proportion of total area	6.647e+02	2.578e+00	257.81	<0.001
<i>Dataset: UiOPlant Obs</i>				
Intercept	2.763e-02	2.474e-02	1.117	0.264
Proportion of total area	9.981e+03	9.996e+00	998,450	<0.001

(b) Alien species occurrence records

	Estimate	Std. error	z-value	p-value
<i>Dataset: KMN</i>				
Intercept	-1.310e-03	7.510e-05	-17.45	<0.001
Proportion of total area	3.620e+02	1.903e+00	190.27	<0.001
<i>Dataset: BioFokus</i>				
Intercept	4.047e-02	2.651e-02	1.527	0.127
Proportion of total area	9.240e+03	9.619e+00	960.567	<0.001
<i>Dataset: eBird</i>				
Intercept	1.471e-02	1.137e-02	1.294	0.196
Proportion of total area	3.658e+02	1.919	190.657	<0.001
<i>Dataset: Jordal</i>				
Intercept	2.351e-02	1.726e-02	1.362	0.173
Proportion of total area	2.442e03	4.948e+00	493.657	<0.001
<i>Dataset: NBIC_{CS}</i>				
Intercept	5.979e-04	8.174e-02	0.007	0.994
Proportion of total area	1.889e+05	4.347e+01	4344.328	<0.001
<i>Dataset: NBIC_{Other}</i>				
Intercept	8.834e-03	1.390e-02	0636	0.525
Proportion of total area	3.120e+03	5.598e+00	558.283	<0.001
<i>Dataset: NTNU</i>				
Intercept	-3.128e-04	2.901e-05	-10.78	<0.001

Proportion of total area	8.640e+01	9.296e-01	92.95	<0.001
---------------------------------	-----------	-----------	-------	------------------

Dataset: $\widehat{U_i O}_{Plant\ Notes}$

Intercept	-5.791e-03	2.618e-05	-221.2	<0.001
------------------	------------	-----------	--------	------------------

Proportion of total area	1.600e+03	4.000e+00	399.9	<0.001
---------------------------------	-----------	-----------	-------	------------------

Dataset: $\widehat{U_i O}_{Plant\ Obs}$

Intercept	1.108e-02	3.710e-02	0.299	0.765
------------------	-----------	-----------	-------	-------

Proportion of total area	3.595e+04	1.897e+01	1895.303	<0.001
---------------------------------	-----------	-----------	----------	------------------

809

810

811

812 Table 3. Over- vs. under-sampled land-cover types for each dataset. A summary of which land-cover types has either more or
813 fewer observed records than expected by the Generalised Linear Models summarised in Table 2. ↑ indicates more records
814 than expected, ↓ indicates fewer records than expected. “n.s.” indicates that the observed number of records fell within the
815 0.95 C.I. of the model predictions. See Supporting Information, Table S.6 for detailed numbers. Acronyms refers to the da-
816 taset described in Table 1.

	KMN	Bio-Fokus	eBird	Jordal	NBICs	NBIC_{Other}	NTNU	UioLichen	UioPlant	UioPlant
	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Red-listed</i> <i>Alien</i>	<i>Obs</i> <i>Red-listed</i> <i>Alien</i>
Developed area	↓ ↑	↑	↑	↑	↑	↑	↓	↑	↑	↓
Agriculture (unsp.)	↓	↑	↑	↓	↑	↑	↓	↑	↑	↓
Cultivated land	↑	↑	↑	↓	↑	↑	↓	↑	↑	↓
Home fields grazing land	↓	↑	↑	↑	↑	↑	↑	↑	↑	↑
Forest (unsp.)	↓	↓	↓	↓	↓	↓	↓	↓	↓	↓
Coniferous forest	↓	↑	↓	↓	↓	↓	↓	↓	↓	↓

Supplementary material

Tanja K. Petersen, James D. M. Speed, Vidar Grøtan, Gunnar Austrheim

Species data for understanding biodiversity dynamics: The What, Where and When of species occurrence data collection.

S.1. Land-cover categories in AR50

Table S.1. Land-cover categories in the AR50 maps for Norway used in the analyses. The descriptions of the land-cover types are based on the descriptions from the AR50 product specification (Norwegian Institute of Bioeconomy Research, 2016)¹. “Percentage of total area” describes the percentage of the entire AR50 map covered by the respective land-cover type.

AR50 category	Name	Description	Area (km ²)	Percentage of total area (%)
10	Developed area	Area which is used for traffic or houses	1 945.5	0.55
20	Agriculture	Fully cultivated land, superficially cultivated land, and home fields grazing land; exact use not specified	56.9	0.016
24	Fully- and superficially cultivated land	Fully cultivated land be used for agriculture or meadows, and be renewed by ploughing.	10 612.6	3.0

¹ Norwegian Institute of Bioeconomy Research. 2016. 50 *Produktspesifikasjon: Arealressurskart*.

		Superficially cultivated land is mostly evened out on the surface, allowing for mechanical harvesting		
25	Home fields grazing land	Home fields which can be used for grazing, but cannot be mechanically harvested. At least 50% of the area is covered by grasses	2 249.5	0.6
30	Forest	Forest type unregistered	20 844.7	5.9
31	Coniferous forest	At least 50% of the area is covered by conifers	69 424.4	19.5
32	Deciduous forest	Less than 20% of the forested area is covered by conifers	29 440.8	8.3
33	Mixed forest	Between 20 and 50% of the forested area is covered by conifers	6 956.7	2.0
50	Open firm ground	Firm ground which is not farmland, forest, developed or used for communications purposes	139 314.5	39.2
60	Mire	Area with marsh vegetation and a peat layer at least 30 cm thick	21 833.9	6.1
70	Snow/ice	Mix of snow and ice which does not melt during the summer	2 831.1	0.80
81	Freshwater	Lakes and rivers	18 511.4	5.2
82	Ocean	Ocean	31 150.1	8.8
99	Not mapped	Area of unknown character	1.3	0.00036

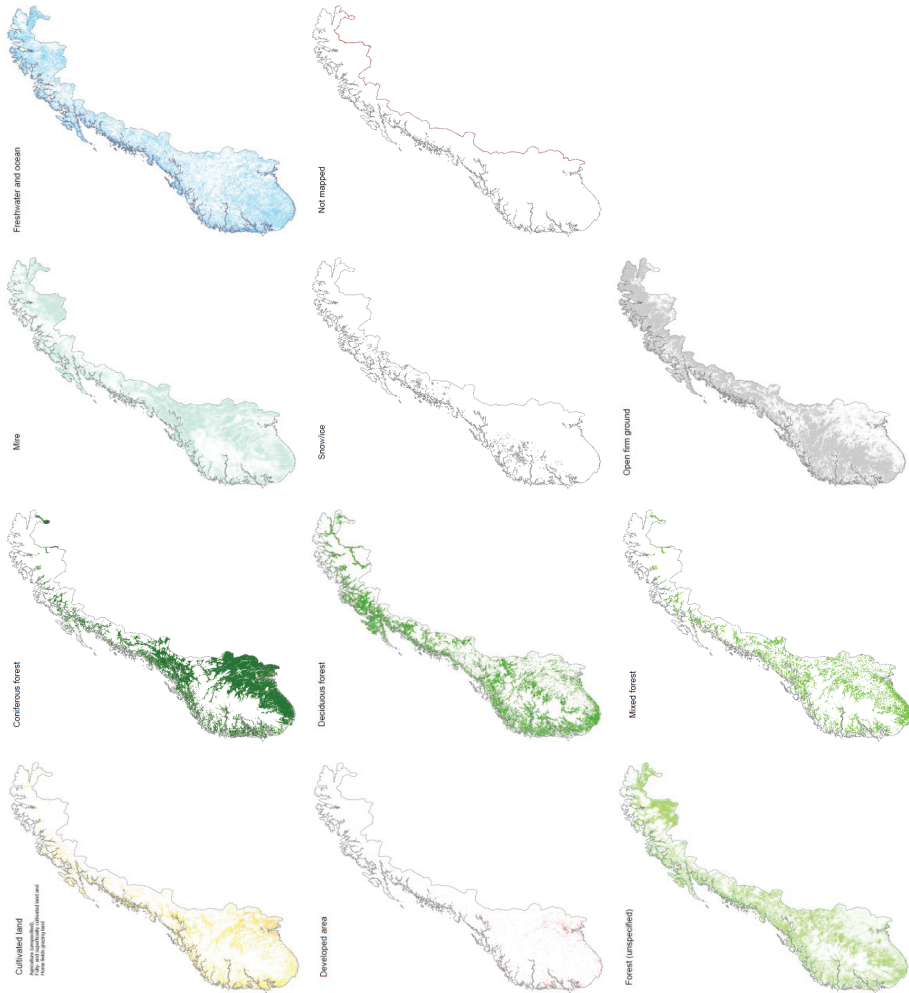


Figure S.1 Detailed maps of the individual land-cover categories included in the analyses. Scale, geographic scope and colour legend are similar to Fig. 1, except snow/ice (light grey) and unmapped areas (red), which were changed here to increase visibility.

S.2. Datasets included in analyses

Table S.2 Datasets included in the analyses. The datasets are displayed in descending order according to the total no. records.

The description is modified from the description and methodology as presented on the linked GBIF webpage.

Dataset Name (abbreviation)	Citation	Publisher	# Animals	# Plants	Description
Norwegian Species Observation Service (NBICs)	The Norwegian Biodiversity Information Centre, Hoem S (2020). Norwegian Species Observation Service. Version 1.82. The Norwegian Biodiversity Information Centre (NBIC). Occurrence dataset https://doi.org/10.15468/zjpbzel accessed via GBIF.org on 2020-05-15	Norwegian Biodiversity Information Centre	2 678 373	1 776 878	Citizen science species observations from the Species Observation Service in Norway (Artsobservasjonser.no)
Vascular Plants, Field notes, Oslo (O) (UfO _{Plant Notes})	Natural History Museum, University of Oslo (2019). Vascular Plants, Field notes, Oslo (O). Version 1.186. Occurrence dataset https://doi.org/10.15468/w8gru5 accessed via GBIF.org on 2020-05-15.	Natural History Museum, University of Oslo	0	1 006 937	Vascular Plants, Field notes from Norway
Norwegian Biodiversity Information Centre, Hoem S (2020).				0	

Norwegian Biodiversity Information Centre - Other datasets (NBIC _{other})	Norwegian Biodiversity Information Centre - Other datasets. Version 1.3.132. The Norwegian Biodiversity Information Centre (NBIC). Occurrence dataset https://doi.org/10.15468/tm56s accessed via GBIF.org on 2020-05-15	Norwegian Biodiversity Information Centre	17 288 289	Other data providers sharing occurrence data via the Norwegian Species Map Service, allowing NBIC to share their data as they do not provide their own IPT service. This includes datasets/databases from the following providers: The Norwegian Environment Agency (http://www.miljodirektoratet.no/); <i>Predator database, water species database, naturdatabase</i> and <i>salmon registry</i> .
BioFokus (BioFokus)	Blindheim T (2020). BioFokus. Version 1.1384. BioFokus. Occurrence dataset https://doi.org/10.15468/jxbhqx accessed via GBIF.org on 2020-05-15	BioFokus	39 179 212 265 82 724	BioFokus is a non-profit organization for providing biodiversity survey information on biological diversity to decision makers and the public in Norway. BioFokus focuses on field work to secure updated and relevant data on botany, zoology, ecology, and survey for respective areas. BioFokus employees cover a wide range of species-groups and have broad understanding of ecology in the respective nature types, including

forest, cultural landscapes and fresh water

Vascular plant field notes, NTNU University Museum (NTNU)	NTNU University Museum (2020). Vascular plant field notes, NTNU University Museum. Version 1.97. Sampling event dataset https://doi.org/10.15468/jxbhqx accessed via GBIF.org on 2020-05-15	NTNU University Museum	0	188 910	0	Vascular plant field notes, Norway. Observation of vascular plants from more several thousand localities using standardised cross-lists. The dataset is a collection of observations made during various research project, including surveys and shorter field trips, at the NTNU University Museum from the first part of the 1900s and onwards. Included are also a few observations made by skilled amateurs. The dataset contains observation from several thousand localities throughout Norway. The observations were recorded in the field using standardised cross-list forms covering most vascular plants to be found in Norway.
Jordal (Jordal)	Jordal J B (2019). Jordal. Version 1.91. Biolog J.B. Jordal AS. Occurrence dataset https://doi.org/10.15468/wqsad accessed via GBIF.org on 2020-05-15	Biolog J.B. Jordal AS	282	150 411	21 856	John Bjarne Jordal, sole proprietor. Consultant within biology and nature management (http://www.jbjordal.no/)
Vascular Plants, Field notes, Agder naturmuseum (KMIN)	Åsen P (2019). Vascular Plants, Field notes, Agder naturmuseum (KMIN). Version 1.160. Agder Museum of Natural History and Botanical Garden. Occurrence	Agder Museum of Natural History and Botanical Garden	0	125 115	0	Vascular Plants, Field notes from Norway

dataset https://doi.org/10.15468/gja4jo accessed via GBIF.org on 2020-05-15	Levatich T, Padilla F (2019). EOD - eBird Observation Dataset. Cornell Lab of Ornithology. Occurrence dataset https://doi.org/10.15468/aom-fnb accessed via GBIF.org on 2020-05-15	Cornell Lab of Ornithology	93 109 0 0	eBird is a collective enterprise that takes a novel approach to citizen science by developing cooperative partnerships among experts in a wide range of fields: population ecologists, conservation biologists, quantitative ecologists, statisticians, computer scientists, GIS and informatics specialists, application developers, and data administrators
Vascular Plants, Observations, Oslo (O) (UIO _{plant.Obs})	Natural History Museum, University of Oslo (2019). Vascular Plants, Observations, Oslo (O). Version 1.181. Occurrence dataset https://doi.org/10.15468/tvnjk7 accessed via GBIF.org on 2020-05-15	Natural History Museum, University of Oslo	0 82 634 0	Vascular Plants, Observations, Oslo (O)
Lichen field notes, Oslo (O) (UIO _{lichen})	Natural History Museum, University of Oslo (2020). Lichen field notes, Oslo (O). Version 1.180. Occurrence dataset https://doi.org/10.15468/zrfxcu accessed via GBIF.org on 2020-05-15	Natural History Museum, University of Oslo	0 0 75512	Lichens, Field notes from Norway

S.3 Taxonomical differences in GBIF records

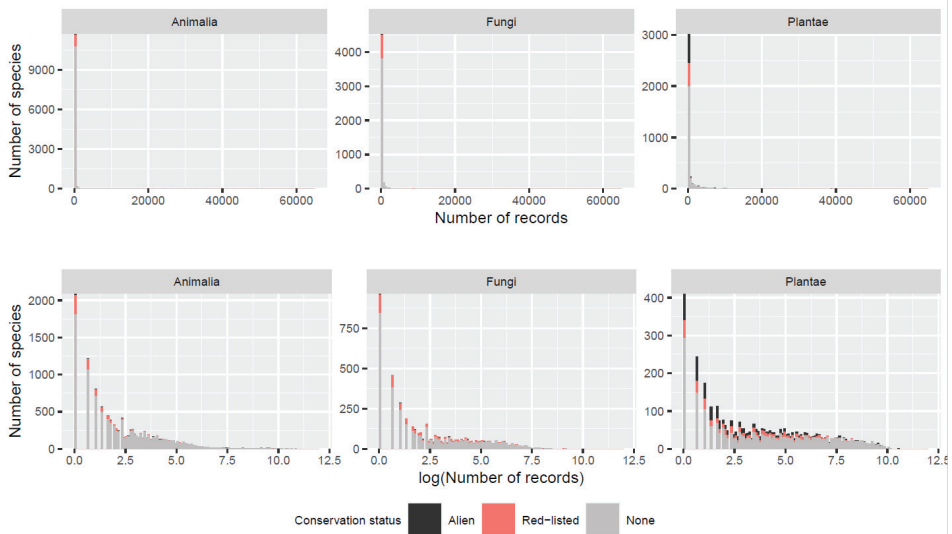


Figure S.2. Number of species by number of records in the analysed GBIF records within each kingdom. The number of species by total number of records (top panels), and by $\log(\text{number of records})$ (bottom panels). Conservation status is indicated by colour.

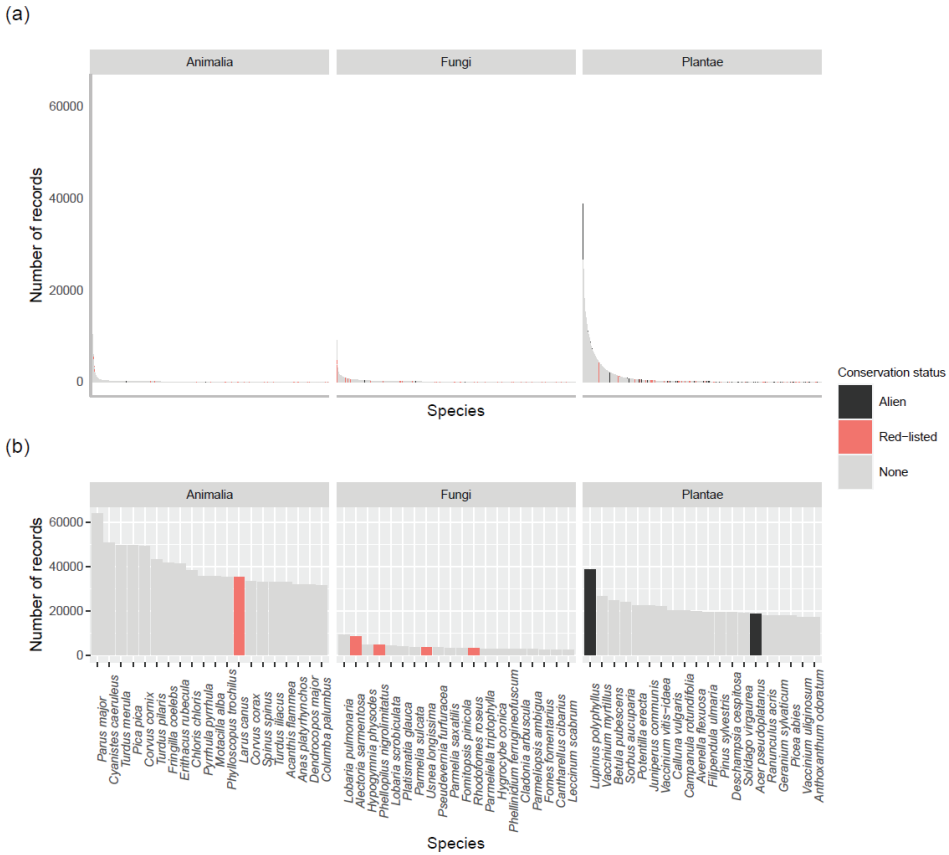


Figure S.3. Number of records for each species within the kingdoms. (a) For all species (species names not indicated), and (b) the 20 most frequent species within each of the kingdoms. Conservation status is indicated by color.

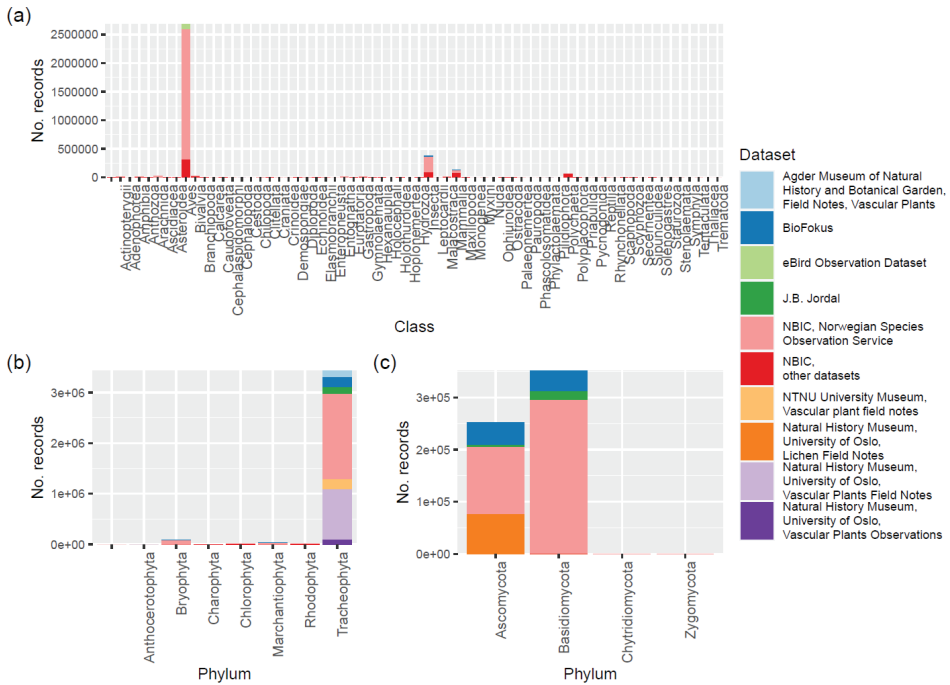


Figure S.4. Number of records at lower taxonomic levels within the kingdoms (a) Animalia, (b) Plantae and (c) Fungi.

S.4 Interannual variation in GBIF records for the two recorder groups

To compare the within-year variation between the datasets, the proportion of records registered in each month was calculated (separately for each dataset and each kingdom to account for the uneven sampling sizes). Treating “month” as numerical values, the median months of sampling were compared with a Kruskal-Wallis test, and posthoc pairwise comparisons were performed with a Dunn Test with Bonferroni correction for multiple comparisons. The median sampling month differed between datasets for all taxonomic groups (Animals: Kruskal-Wallis $\chi^2 = 53.245$, $df = 4$, $p\text{-value} < 0.001$; Plants: Kruskal-Wallis $\chi^2 = 23.718$, $df = 7$, $p\text{-value} = 0.001$; Fungi: Kruskal-Wallis $\chi^2 = 46.051$, $df = 4$, $p\text{-value} < 0.001$; All: Kruskal-Wallis $\chi^2 = 46.247$, $df = 9$, $p\text{-value} < 0.001$). All pairwise comparisons were significantly different (adjusted $p\text{-value} < 0.001$), except for the Jordal and BioFokus datasets for animal records.

Table S.3. Interannual temporal variation between datasets. Mean and median sampling month, including standard deviation and interquartile range.

	Dataset	Mean	S.D.	Median	IQR
Animals	BioFokus	6.88	1.70	7	2
	eBird	6.15	2.32	6	2
	Jordal	6.66	2.13	7	3
	NBIC _{CS}	6.03	2.63	6	4
	NBIC _{Other}	4.20	3.50	3	6
Plants	KMN	7.13	1.03	7	2
	BioFokus	7.54	1.42	8	3
	Jordal	7.85	1.38	8	2

	NBIC _{CS}	6.91	1.61	7	2
	NBIC _{Other}	7.57	1.44	8	2
	N'TNU	7.42	1.06	7	1
	UiO _{Plant Notes}	7.22	0.932	7	1
	UiO _{Plant Nores}	7.22	1.44	7	2
Fungi	BioFokus	8.16	1.81	9	2
	Jordal	8.64	1.42	9	1
	NBIC _{CS}	7.57	2.60	8	3
	NBIC _{Other}	6.56	2.68	7	4
	UiO _{Lichen}	6.92	2.34	7	4
All	KMN	7.13	1.03	7	2
	BioFokus	7.61	1.60	8	3
	eBird	6.15	2.32	6	2
	Jordal	7.95	1.41	8	2
	NBIC _{CS}	6.48	2.37	7	3
	NBIC _{Other}	4.30	3.50	3	6
	N'TNU	7.42	1.06	7	1
	UiO _{Lichen}	6.92	2.34	7	4
	UiO _{Plant Notes}	7.22	0.93	7	1
	UiO _{Plant Nores}	7.22	1.44	7	2

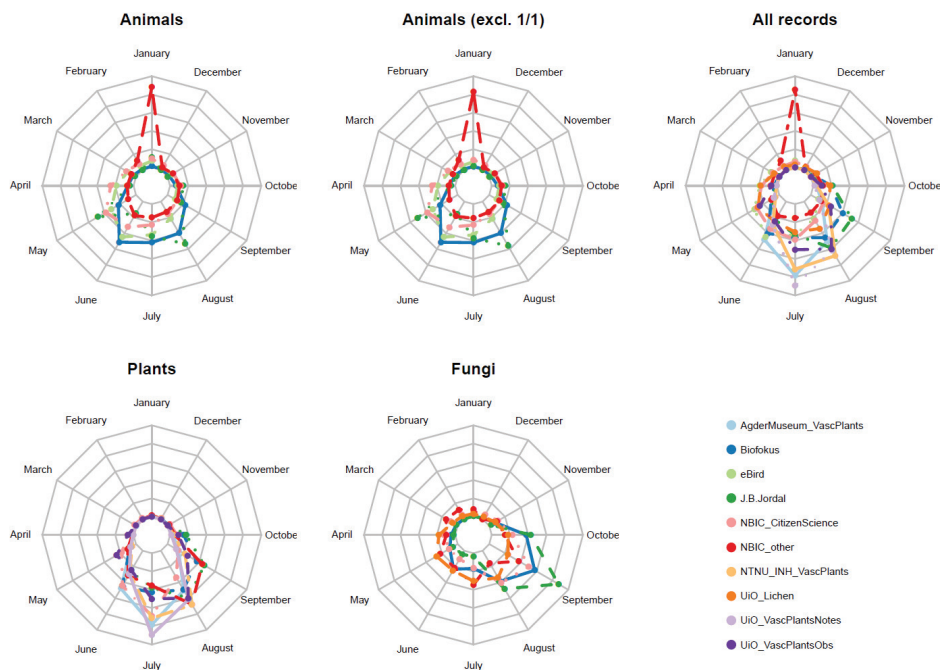


Figure S.5. Proportion of records in total and within each of the three kingdoms registered each month. Dataset is indicated by colour. Records without information of month of recording are not included. “Animals (excl. 1/1)” is excluding animal records registered to January 1st.

S.5 Separate models for individual datasets

For the simulated data, the number of records within each land-cover type is predicted by the area of the specified land cover type (Table S.2).

Only open firm ground, mires and snow/ice showed consistent patterns across all datasets; all had fewer records that would be expected by the area. Developed areas were under-

sampled in the KMN-, NTNU-, and Jordal datasets, and oversampled for all others. Unspecified agricultural areas were under-sampled (i.e. had no records) in the KMN-, NTNU-, UiOLichen- and UiOPlant Notes datasets, but were oversampled for all others. Cultivated areas and grazing land were under-sampled in the KMN- and NTNU dataset, respectively, otherwise oversampled in all cases. Coniferous forest was oversampled in the NBICCS-, BioFokus-, NTNU-, UiOLichen- and UiOPlant Notes datasets, under-sampled for all others. Deciduous- and mixed forest were both under-sampled in the NBICother dataset, otherwise consistently oversampled, whereas unspecified forest was oversampled in the NTNU- and UiOPlant Obs datasets, otherwise consistently under-sampled. Freshwater was oversampled in the NBICother-, NBICCS-, eBird- and KMN datasets, otherwise under-sampled. Ocean was oversampled in the UiOPlant Notes-, eBird-, KMN- and NBICother datasets, otherwise under-sampled. Unmapped areas were oversampled in the UiOPlant Obs-, NBICother-, eBird- and BioFokus datasets, and under-sampled for the NBICCS-, KMN-, Jordal- and UiOPlant Notes datasets, and the number of records from the UiOLichen. NTNU datasets fell within the 0.95 CI.

Comparing the absolute residuals between predicted and observed number of records within each land-cover type, the largest numerical discrepancies were seen for open firm ground, developed area and cultivated land (Figure S.6-S.9). When comparing the relative residuals, no other consistent patterns are seen (Figure S.9).

Table S.4. Generalised linear models (Poisson error distribution, “identity”-link function) of the simulated, random data, predicting the number of records for each land-cover type as a function of the area of the specified land-cover type in Norway.

	Estimate	Std. error	z value	p value
KMN				
Intercept	4.675e-02	3.010e-02	1.553	0.12
Proportion of area	1.326e+04	1.152e+01	1150.920	<0.001
BioFokus				
Intercept	1.200e-02	1.001e-01	0.12	0.905
Proportion of area	2.800e+05	5.293e+01	5289.24	<0.001
eBird				
Intercept	9.353e-03	5.294e-02	0.177	0.86
Proportion of area	7.675e+04	2.771e+01	2769.340	<0.001
Jordal				
Intercept	-6.343e-02	5.745e-02	-1.104	0.27
Proportion of area	1.105e+05	3.326e+01	3323.834	<0.001
NBIC_{Cs}				
Intercept	4.839e-01	4.015e-01	1.205	0.228
Proportion of area	4.429e+06	2.105e+02	21038.426	<0.001
NBIC_{Other}				
Intercept	-5.170e-02	1.346e-01	-0.384	0.701
Proportion of area	5.266e+05	7.259e+01	7254.215	<0.001
NTNU				
Intercept	-1.192e-02	1.412e-02	-0.844	0.398

Proportion of area	8.879e+03	9.425e+00	942.060	<0.001
UiO_{Lichen}				
Intercept	1.164e-02	1.697e-02	0.686	0.493
Proportion of area	5.041e+03	7.104e+00	709.638	<0.001
UiO_{Plant Notes}				
Intercept	1.287e-03	3.977e-02	0.032	0.974
Proportion of area	4.439e+04	2.108e+01	2106.276	<0.001
UiO_{Plant Obs}				
Intercept	6.962e-02	5.961e-02	1.168	0.243
Proportion of area	8.186e+04	2.862e+01	2859.838	<0.001

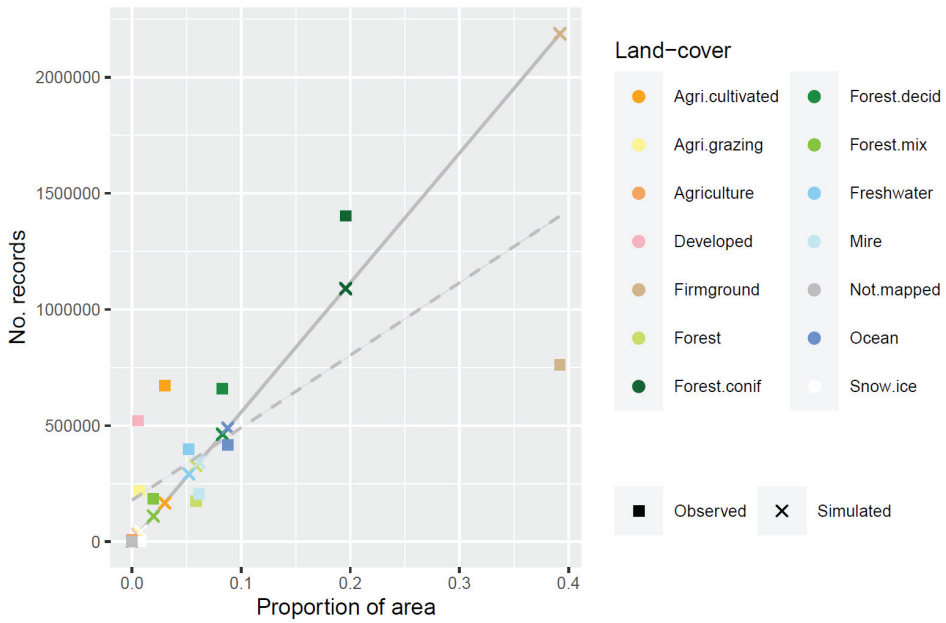


Figure S.6. Null model (GLM) of the total number of records as a function of area (proportion of the total area within the study area) for the random, simulated data (crosses), vs. the observed number of records for each land-cover type (squares), not including information on dataset or conservation status. The solid line indicate the predicted values by the null-model, gray ribbons indicate the 0.95 C.I..

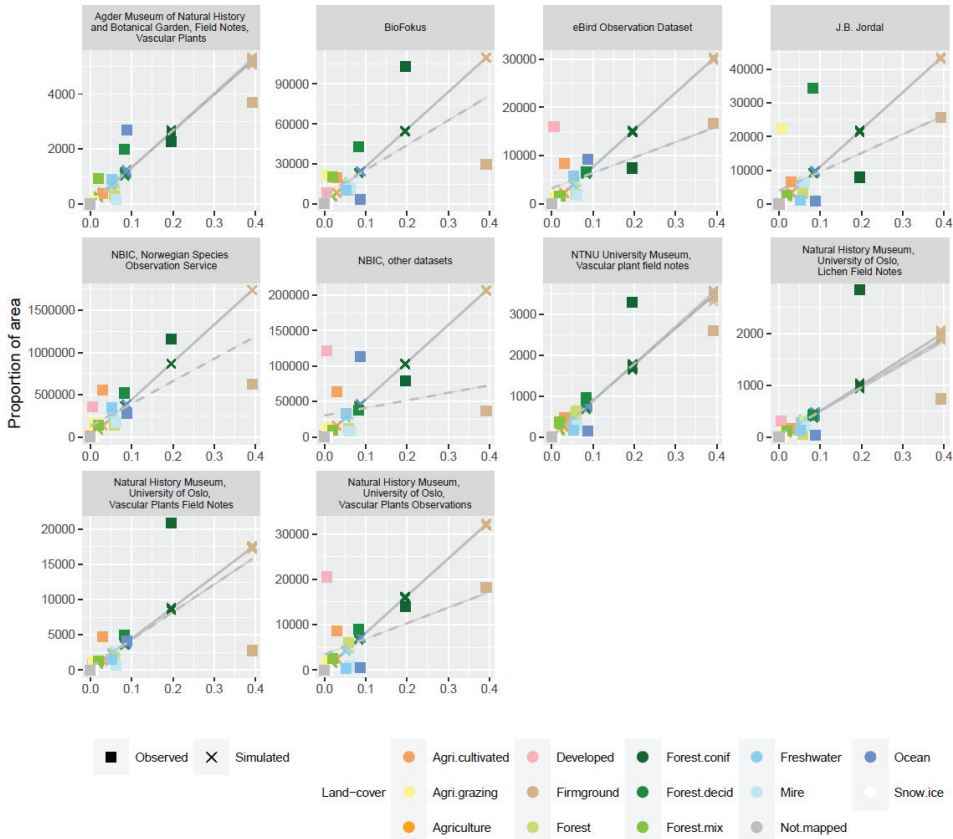


Figure S.7. Null models (GLM) of the number of records as a function of area (proportion of the total area within the study area) for the random, simulated data (crosses), vs. the observed number of records for each land-cover type (squares). Solid lines indicate the predicted values by the null-model, gray ribbons indicate the 0.95 C.I.

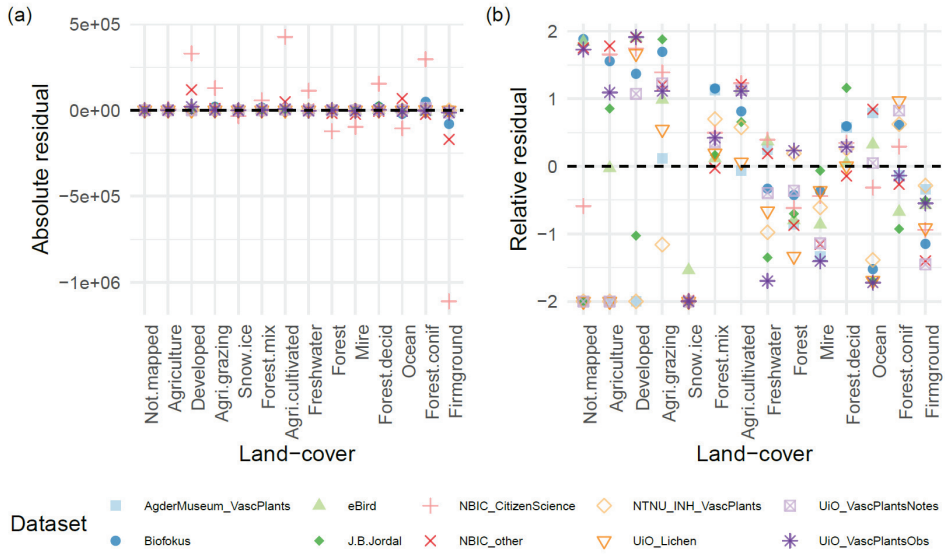


Figure S.9. (a) Absolute and (b) relative residuals between observed number of species occurrence records, and the predicted number of records, based on the area of each land-cover type. Colours/shapes indicate dataset.

Table S.5. Observed number of records for each land-cover type, the number of records predicted for each land-cover type based on area, and the absolute and relative difference between the two. Negative residuals are highlighted in gray. * indicates that the observed number of records falls within the 0.95 confidence interval.

Land-cover	Dataset	Observed	Predicted	Absolute residual	Relative residual
Developed area	KMN	0.00	72.83	-72.83	-2.00
	BioFokus	8202.00	1536.21	6665.79	1.37
	eBird	16080.00	421.14	15658.86	1.90
	Jordal	195.00	606.52	-411.52	-1.03
	NBIC _{CS}	353357.00	24305.34	329051.66	1.74
	NBIC _{Other}	121287.00	2889.51	118397.49	1.91
	NTNU	0.00	48.71	-48.71	-2.00
	UiO _{Lichen}	305.00	27.67	277.33	1.67
	UiO _{Plant Notes}	807.00	243.61	563.39	1.07
	UiO _{Plant Obs}	20516.00	449.23	20066.77	1.91
Agriculture (unsp.)	KMN	0.00	2.18	-2.18	-2.00
	BioFokus	360.00	44.94	315.06	1.56
	eBird	12.00	12.32	-0.32	-0.03
	Jordal	44.00	17.68	26.32	0.85
	NBIC _{CS}	7543.00	711.24	6831.76	1.66
	NBIC _{Other}	1452.00	84.45	1367.55	1.78
	NTNU	0.00	1.41	-1.41	-2.00
	UiO _{Lichen}	0.00	0.82	-0.82	-2.00
	UiO _{Plant Notes}	0.00	7.13	-7.13	-2.00
	UiO _{Plant Obs}	45.00	13.20	31.80	1.09
Cultivated land	KMN	371.00	396.85	-25.85	-0.07

	BioFokus	19845.00	8375.70	11469.30	0.81
	eBird	8373.00	2296.12	6076.88	1.14
	Jordal	6529.00	3307.18	3221.82	0.66
	NBIC _{CS}	557986.00	132515.66	425470.34	1.23
	NBIC _{Other}	64208.00	15754.44	48453.56	1.21
	NTNU	481.00	265.62	215.38	0.58
	UiO _{Lichen}	160.00	150.84	9.16	0.06
	UiO _{Plant Notes}	4763.00	1328.20	3434.80	1.13
	UiO _{Plant Obs}	8603.00	2449.00	6154.00	1.11
Home fields grazing land	KMN	95.00	84.30	10.70	0.12
	BioFokus	21515.00	1778.38	19736.62	1.69
	eBird	1434.00	487.53	946.47	0.99
	Jordal	22524.00	702.15	21821.85	1.88
	NBIC _{CS}	156314.00	28136.80	128177.20	1.39
	NBIC _{Other}	13422.00	3345.02	10076.98	1.20
	NTNU	15.00	56.39	-41.39	-1.16
	UiO _{Lichen}	56.00	32.04	23.96	0.54
	UiO _{Plant Notes}	1181.00	282.01	898.99	1.23
	UiO _{Plant Obs}	1824.00	520.04	1303.96	1.11
Forest	KMN	310.00	776.68	-466.68	-0.86
	BioFokus	10656.00	16392.88	-5736.88	-0.42
	eBird	1929.00	4493.95	-2564.95	-0.80
	Jordal	3109.00	6472.87	-3363.87	-0.70
	NBIC _{CS}	136870.00	259358.86	-122488.86	-0.62
	NBIC _{Other}	12076.00	30834.60	-18758.60	-0.87
	NTNU	629.00	519.89	109.11	0.19
	UiO _{Lichen}	59.00	295.21	-236.21	-1.33
	UiO _{Plant Notes}	1804.00	2599.55	-795.55	-0.36
	UiO _{Plant Obs}	6057.00	4793.12	1263.88	0.23
Coniferous forest	KMN	2260.00	2595.81	-335.81	-0.14

	BioFokus	103112.00	54790.45	48321.55	0.61
	eBird	7443.00	15020.26	-7577.26	-0.67
	Jordal	7934.00	21634.64	-13700.64	-0.93
	NBIC _{CS}	1162695.00	866862.89	295832.11	0.29
	NBIC _{Other}	78797.00	103059.66	-24262.66	-0.27
	NTNU	3309.00	1737.67	1571.33	0.62
	UiO _{Lichen}	2841.00	986.65	1854.35	0.97
	UiO _{Plant Notes}	20885.00	8688.57	12196.43	0.82
	UiO _{Plant Obs}	13922.00	16020.04	-2098.04	-0.14
Deciduous forest	KMN	1985.00	1099.58	885.42	0.57
	BioFokus	42799.00	23208.55	19590.45	0.59
	eBird	6656.00	6362.40	293.60	0.05
	Jordal	34497.00	9164.12	25332.88	1.16
	NBIC _{CS}	520766.00	367192.44	153573.56	0.35
	NBIC _{Other}	37835.00	43654.74	-5819.74	-0.14
	NTNU	956.00	736.05	219.95	0.26
	UiO _{Lichen}	420.00	417.94	2.06	0.00
	UiO _{Plant Notes}	4925.00	3680.37	1244.63	0.29
	UiO _{Plant Obs}	9061.00	6785.93	2275.07	0.29
Mixed forest	KMN	936.00	260.18	675.82	1.13
	BioFokus	20395.00	5490.77	14904.23	1.15
	eBird	1693.00	1505.25	187.75	0.12
	Jordal	2560.00	2168.03	391.97	0.17
	NBIC _{CS}	144012.00	86872.08	57139.92	0.49
	NBIC _{Other}	10060.00	10327.96	-267.96	-0.03
	NTNU	361.00	174.13	186.87	0.70
	UiO _{Lichen}	120.00	98.89	21.11	0.19
	UiO _{Plant Notes}	1247.00	870.72	376.28	0.36
	UiO _{Plant Obs}	2466.00	1605.49	860.51	0.42
Open firm ground	KMN	3684.00	5199.70	-1515.70	-0.34

	BioFokus	29709.00	109752.61	-80043.61	-1.15
	eBird	16658.00	30087.59	-13429.59	-0.57
	Jordal	25778.00	43337.14	-17559.14	-0.51
	NBIC _{CS}	624927.00	1736442.04	-1111515.04	-0.94
	NBIC _{Other}	36376.00	206442.37	-170066.37	-1.40
	NTNU	2610.00	3480.81	-870.81	-0.29
	UiO _{Lichen}	738.00	1976.39	-1238.39	-0.91
	UiO _{Plant Notes}	2764.00	17404.37	-14640.37	-1.45
	UiO _{Plant Obs}	18263.00	32090.22	-13827.22	-0.55
Mire	KMN	163.00	814.94	-651.94	-1.33
	BioFokus	11842.00	17200.45	-5358.45	-0.37
	eBird	1874.00	4715.34	-2841.34	-0.86
	Jordal	6367.00	6791.75	-424.75	-0.06
	NBIC _{CS}	174161.00	272135.72	-97974.72	-0.44
	NBIC _{Other}	8667.00	32353.62	-23686.62	-1.15
	NTNU	291.00	545.50	-254.50	-0.61
	UiO _{Lichen}	215.00	309.75	-94.75	-0.36
	UiO _{Plant Notes}	747.00	2727.61	-1980.61	-1.14
	UiO _{Plant Obs}	880.00	5029.24	-4149.24	-1.40
Snow/ice	KMN	0.00	105.89	-105.89	-2.00
	BioFokus	5.00	2234.19	-2229.19	-1.99
	eBird	80.00	612.49	-532.49	-1.54
	Jordal	0.00	882.13	-882.13	-2.00
	NBIC _{CS}	93.00	35348.36	-35255.36	-1.99
	NBIC _{Other}	16.00	4202.39	-4186.39	-1.98
	NTNU	0.00	70.85	-70.85	-2.00
	UiO _{Lichen}	0.00	40.24	-40.24	-2.00
	UiO _{Plant Notes}	0.00	354.29	-354.29	-2.00
	UiO _{Plant Obs}	0.00	653.31	-653.31	-2.00
Freshwater	KMN	889.00	691.48	197.52	0.25
	BioFokus	10443.00	14594.59	-4151.59	-0.33
	eBird	5807.00	4000.97	1806.03	0.37

	Jordal	1118.00	5762.79	-4644.79	-1.35
	NBIC _{CS}	344041.00	230907.42	113133.58	0.39
	NBIC _{Other}	33102.00	27452.06	5649.94	0.19
	NTNU	159.00	462.86	-303.86	-0.98
	UiO _{Lichen}	132.00	262.82	-130.82	-0.66
	UiO _{Plant Notes}	1552.00	2314.38	-762.38	-0.39
	UiO _{Plant Obs}	349.00	4267.32	-3918.32	-1.70
Ocean	KMN	2680.00	1163.40	1516.60	0.79
	BioFokus	3335.00	24555.65	-21220.65	-1.52
	eBird	9327.00	6731.69	2595.31	0.32
	Jordal	797.00	9696.04	-8899.04	-1.70
	NBIC _{CS}	282800.00	388505.50	-105705.50	-0.31
	NBIC _{Other}	113587.00	46188.61	67398.39	0.84
	NTNU	141.00	778.77	-637.77	-1.39
	UiO _{Lichen}	37.00	442.20	-405.20	-1.69
	UiO _{Plant Notes}	4085.00	3893.99	191.01	0.05
	UiO _{Plant Obs}	537.00	7179.80	-6642.80	-1.72
Not mapped	KMN	0.00	0.09	-0.09	-2.00
	BioFokus	34.00	1.03	32.97	1.88
	eBird	7.00	0.29	6.71	1.84
	Jordal	0.00	0.34	-0.34	-2.00
	NBIC _{CS}	9.00	16.52	-7.52	-0.59
	NBIC _{Other}	28.00	1.85	26.15	1.75
	NTNU	0.00	0.02	-0.02	-2.00*
	UiO _{Lichen}	0.00	0.03	-0.03	-2.00*
	UiO _{Plant Notes}	0.00	0.16	-0.16	-2.00
	UiO _{Plant Obs}	5.00	0.37	4.63	1.73

S.6 Observed and predicted number of records

Table S.6. Observed- and predicted number of records for each land-cover type, and absolute and relative difference between the two. Negative residuals are highlighted in gray. * indicates that the observed number of records falls within the 0.95 confidence interval.

Land-cover	Dataset	Conservation status	Observed	Predicted	Absolute residual	Relative residual
Developed area	KMN	Red-listed	0	0.51	-0.51	-2.00
		Alien	0	1.99	-1.99	-2.00
	BioFokus	Red-listed	736	231.69	504.31	1.04
		Alien	1122	50.74	1071.26	1.83
	eBird	Red-listed	2490	76.67	2413.33	1.88
		Alien	60	2.02	57.98	1.87
	Jordal	Red-listed	34	29.23	4.77	0.15
		Alien	11	13.43	-2.43	-0.20
	NBIC _{CS}	Red-listed	32192	2285.22	29906.78	1.73
		Alien	35475	1036.34	34438.66	1.89
	NBIC _{Other}	Red-listed	6355	299.08	6055.92	1.82
		Alien	1027	17.13	1009.87	1.93
	NTNU	Red-listed	0	0.56	-0.56	-2.00
		Alien	0	0.47	-0.47	-2.00
	UiO _{Lichen}	Red-listed	11	1.48	9.52	1.53
	UiO _{Plant Notes}	Red-listed	8	3.65	4.35	0.75
		Alien	61	8.77	52.23	1.50
	UiO _{Plant Obs}	Red-listed	6	54.79	-48.79	-1.61
		Alien	15501	197.27	15303.73	1.95
	Agriculture (unsp.)	KMN	Red-listed	0	0.01	-0.01
Alien			0	0.06	-0.06	-2.00
BioFokus		Red-listed	31	6.73	24.27	1.29
		Alien	18	1.52	16.48	1.69
eBird		Red-listed	1	2.24	-1.24	-0.77

		Alien	0	0.07	-0.07	-2.00
	Jordal	Red-listed	0	0.86	-0.86	-2.00
		Alien	0	0.42	-0.42	-2.00
	NBIC _{CS}	Red-listed	600	66.75	533.25	1.60
		Alien	575	30.31	544.69	1.80
	NBIC _{Other}	Red-listed	68	8.74	59.26	1.54
		Alien	2	0.51	1.49	1.19
	NTNU	Red-listed	0	0.02	-0.02	-2.00
		Alien	0	0.01	-0.01	-2.00
	UiO _{Lichen}	Red-listed	0	0.04	-0.04	-2.00
	UiO _{Plant Notes}	Red-listed	0	0.10	-0.10	-2.00
		Alien	0	0.25	-0.25	-2.00
	UiO _{Plant Obs}	Red-listed	0	1.63	-1.63	-2.00
		Alien	14	5.78	8.22	0.83
	KMN	Red-listed	4	2.76	1.24	0.37
		Alien	9	10.83	-1.83	-0.18
	BioFokus	Red-listed	2248	1263.42	984.58	0.56
		Alien	904	276.48	627.52	1.06
	eBird	Red-listed	1539	418.03	1120.97	1.15
		Alien	56	10.96	45.04	1.35
	Jordal	Red-listed	396	159.30	236.70	0.85
		Alien	289	73.10	215.90	1.19
Cultivated land	NBIC _{CS}	Red-listed	68657	12459.86	56197.14	1.39
		Alien	29489	5650.36	23838.64	1.36
	NBIC _{Other}	Red-listed	5301	1630.67	3670.33	1.06
		Alien	218	93.37	124.63	0.80
	NTNU	Red-listed	4	3.05	0.95	0.27
		Alien	11	2.58	8.42	1.24
	UiO _{Lichen}	Red-listed	3	8.06	-5.06	-0.92
	UiO _{Plant Notes}	Red-listed	75	19.88	55.12	1.16
		Alien	203	47.85	155.15	1.24
	UiO _{Plant Obs}	Red-listed	164	298.63	-134.63	-0.58

		Alien	3453	1075.51	2377.49	1.05
	KMN	Red-listed	0	0.59	-0.59	-2.00
		Alien	5	2.30	2.70	0.74
	BioFokus	Red-listed	2236	268.22	1967.78	1.57
		Alien	654	58.74	595.26	1.67
	eBird	Red-listed	243	88.76	154.24	0.93
		Alien	10	2.34	7.66	1.24
	Jordal	Red-listed	1502	33.83	1468.17	1.91
		Alien	540	15.54	524.46	1.89
Home fields graz- ing land	NBIC _{CS}	Red-listed	14502	2645.48	11856.52	1.38
		Alien	7549	1199.72	6349.28	1.45
	NBIC _{Other}	Red-listed	1232	346.23	885.77	1.12
		Alien	43	19.83	23.17	0.74
	NTNU	Red-listed	1	0.65	0.35	0.43
		Alien	0	0.55	-0.55	-2.00
	UiO _{Lichen}	Red-listed	7	1.71	5.29	1.21
	UiO _{Plant Notes}	Red-listed	20	4.22	15.78	1.30
		Alien	51	10.16	40.84	1.34
	UiO _{Plant Obs}	Red-listed	693	63.43	629.57	1.66
Alien		275	228.37	46.63	0.19	
	KMN	Red-listed	2	5.40	-3.40	-0.92
		Alien	13	21.20	-8.20	-0.48
	BioFokus	Red-listed	1311	2472.81	-1161.81	-0.61
		Alien	485	541.09	-56.09	-0.11
	eBird	Red-listed	294	818.18	-524.18	-0.94
		Alien	4	21.43	-17.43	-1.37
Forest (unsp.)	Jordal	Red-listed	84	311.77	-227.77	-1.15
		Alien	61	143.04	-82.04	-0.80
	NBIC _{CS}	Red-listed	10097	24386.49	-14289.49	-0.83
		Alien	7119	11058.88	-3939.88	-0.43
	NBIC _{Other}	Red-listed	3903	3191.55	711.45	0.20
		Alien	111	182.73	-71.73	-0.49

	NTNU	Red-listed	5	5.97	-0.97	-0.18
		Alien	3	5.06	-2.06	-0.51
	UiOLichen	Red-listed	2	15.78	-13.78	-1.55
	UiOPlant Notes	Red-listed	21	38.92	-17.92	-0.60
		Alien	50	93.66	-43.66	-0.61
	UiOPlant Obs	Red-listed	160	584.44	-424.44	-1.14
		Alien	2825	2104.98	720.02	0.29
	KMN	Red-listed	9	18.04	-9.04	-0.67
		Alien	59	70.86	-11.86	-0.18
	BioFokus	Red-listed	21994	8265.05	13728.95	0.91
		Alien	2519	1808.42	710.58	0.33
	eBird	Red-listed	804	2734.62	-1930.62	-1.09
		Alien	39	71.61	-32.61	-0.59
	Jordal	Red-listed	324	1042.00	-718.00	-1.05
		Alien	140	478.04	-338.04	-1.09
Coniferous forest	NBIC _{CS}	Red-listed	69319	81508.00	-12189.00	-0.16
		Alien	44651	36962.46	7688.54	0.19
	NBIC _{Other}	Red-listed	13206	10667.22	2538.78	0.21
		Alien	437	610.72	-173.72	-0.33
	NTNU	Red-listed	27	19.97	7.03	0.30
		Alien	43	16.91	26.09	0.87
	UiOLichen	Red-listed	166	52.76	113.24	1.04
	UiOPlant Notes	Red-listed	249	130.09	118.91	0.63
		Alien	740	313.05	426.95	0.81
		UiOPlant Obs	Red-listed	948	1953.34	-1005.34
		Alien	5905	7035.53	-1130.53	-0.17
	KMN	Red-listed	23	7.64	15.36	1.00
		Alien	54	30.01	23.99	0.57
Deciduous forest	BioFokus	Red-listed	5306	3500.94	1805.06	0.41
		Alien	1068	766.05	301.95	0.33
	eBird	Red-listed	995	1158.35	-163.35	-0.15
		Alien	16	30.34	-14.34	-0.62

	Jordal	Red-listed	1538	441.38	1096.62	1.11
		Alien	741	202.51	538.49	1.14
	NBIC _{CS}	Red-listed	29165	34525.71	-5360.71	-0.17
		Alien	23683	15656.83	8026.17	0.41
	NBIC _{Other}	Red-listed	5845	4518.50	1326.50	0.26
		Alien	252	258.70	-6.70	-0.03
	NTNU	Red-listed	8	8.46	-0.46	-0.06
		Alien	14	7.16	6.84	0.65
	UiO _{Lichen}	Red-listed	21	22.35	-1.35	-0.06
	UiO _{Plant Notes}	Red-listed	84	55.10	28.90	0.42
		Alien	118	132.60	-14.60	-0.12
	UiO _{Plant Obs}	Red-listed	668	827.43	-159.43	-0.21
		Alien	2798	2980.17	-182.17	-0.06
Mixed forest	KMN	5	5	1.81	3.19	0.94
		31	31	7.10	23.90	1.25
	BioFokus	3283	3283	828.23	2454.77	1.19
		672	672	181.27	490.73	1.15
	eBird	221	221	274.04	-53.04	-0.21
		7	7	7.19	-0.19	-0.03
	Jordal	123	123	104.43	18.57	0.16
		26	26	47.93	-21.93	-0.59
	NBIC _{CS}	8896	8896	8168.15	727.85	0.09
		5462	5462	3704.15	1757.85	0.38
	NBIC _{Other}	1646	1646	1069.00	577.00	0.43
		38	38	61.21	-23.21	-0.47
	NTNU	3	3	2.00	1.00	0.40
		2	2	1.69	0.31	0.17
	UiO _{Lichen}	2	2	5.29	-3.29	-0.90
	UiO _{Plant Notes}	20	20	13.03	6.97	0.42
		40	40	31.37	8.63	0.24
	UiO _{Plant Obs}	263	263	195.78	67.22	0.29
		985	985	705.07	279.93	0.33

Open firm ground	KMN	Red-listed	17	36.14	-19.14	-0.72
		Alien	59	141.93	-82.93	-0.83
	BioFokus	Red-listed	3036	16556.05	-13520.05	-1.38
		Alien	1333	3622.48	-2289.48	-0.92
	eBird	Red-listed	3732	5477.82	-1745.82	-0.38
		Alien	41	143.43	-102.43	-1.11
	Jordal	Red-listed	1173	2087.25	-914.25	-0.56
		Alien	525	957.56	-432.56	-0.58
	NBIC _{CS}	Red-listed	64506	163271.52	-98765.52	-0.87
		Alien	22008	74040.76	-52032.76	-1.08
	NBIC _{Other}	Red-listed	12146	21367.86	-9221.86	-0.55
		Alien	193	1223.34	-1030.34	-1.45
	NTNU	Red-listed	49	40.00	9.00	0.20
		Alien	7	33.87	-26.87	-1.31
	UiO _{Lichen}	Red-listed	25	105.69	-80.69	-1.23
	UiO _{Plant Notes}	Red-listed	82	260.58	-178.58	-1.04
		Alien	93	627.09	-534.09	-1.48
	UiO _{Plant Obs}	Red-listed	6643	3912.78	2730.22	0.52
		Alien	4239	14093.09	-9854.09	-1.08
	Mire	KMN	Red-listed	0	5.66	-5.66
Alien			3	22.24	-19.24	-1.52
BioFokus		Red-listed	1039	2594.63	-1555.63	-0.86
		Alien	85	567.75	-482.75	-1.48
eBird		Red-listed	351	858.48	-507.48	-0.84
		Alien	6	22.49	-16.49	-1.16
Jordal		Red-listed	116	327.12	-211.12	-0.95
		Alien	57	150.09	-93.09	-0.90
NBIC _{CS}		Red-listed	14758	25587.85	-10829.85	-0.54
		Alien	3470	11603.67	-8133.67	-1.08
NBIC _{Other}		Red-listed	3117	3348.77	-231.77	-0.07
		Alien	49	191.73	-142.73	-1.19
NTNU		Red-listed	3	6.27	-3.27	-0.71

		Alien	1	5.31	-4.31	-1.37
	UiO _{Lichen}	Red-listed	29	16.56	12.44	0.55
	UiO _{Plant Notes}	Red-listed	11	40.84	-29.84	-1.15
		Alien	21	98.27	-77.27	-1.30
	UiO _{Plant Obs}	Red-listed	261	613.23	-352.23	-0.81
		Alien	66	2208.68	-2142.68	-1.88
Snow/ice	KMN	Red-listed	0	0.74	-0.74	-2.00
		Alien	0	2.89	-2.89	-2.00
	BioFokus	Red-listed	2	336.98	-334.98	-1.98
		Alien	0	73.78	-73.78	-2.00
	eBird	Red-listed	15	111.51	-96.51	-1.53
		Alien	0	2.93	-2.93	-2.00
	Jordal	Red-listed	0	42.50	-42.50	-2.00
		Alien	0	19.52	-19.52	-2.00
	NBIC _{CS}	Red-listed	14	3323.56	-3309.56	-1.98
		Alien	0	1507.21	-1507.21	-2.00
	NBIC _{Other}	Red-listed	1	434.97	-433.97	-1.99
		Alien	0	24.91	-24.91	-2.00
	NTNU	Red-listed	0	0.81	-0.81	-2.00
		Alien	0	0.69	-0.69	-2.00
	UiO _{Lichen}	Red-listed	0	2.15	-2.15	-2.00
	UiO _{Plant Notes}	Red-listed	0	5.30	-5.30	-2.00
		Alien	0	12.76	-12.76	-2.00
	UiO _{Plant Obs}	Red-listed	0	79.68	-79.68	-2.00
		Alien	0	286.90	-286.90	-2.00
	Freshwater	KMN	Red-listed	2	4.80	-2.80
		Alien	35	18.87	16.13	0.60
BioFokus		Red-listed	1012	2201.54	-1189.54	-0.74
		Alien	287	481.74	-194.74	-0.51
eBird		Red-listed	1097	728.42	368.58	0.40
		Alien	81	19.09	61.91	1.24
Jordal		Red-listed	45	277.57	-232.57	-1.44

	Alien	27	127.35	-100.35	-1.30
NBIC _{CS}	Red-listed	47065	21711.30	25353.70	0.74
	Alien	6518	9845.72	-3327.72	-0.41
NBIC _{Other}	Red-listed	1537	2841.43	-1304.43	-0.60
	Alien	454	162.68	291.32	0.94
NTNU	Red-listed	1	5.32	-4.32	-1.37
	Alien	5	4.50	0.50	0.10
UiO _{Lichen}	Red-listed	4	14.05	-10.05	-1.11
UiO _{Plant Notes}	Red-listed	21	34.65	-13.65	-0.49
	Alien	43	83.38	-40.38	-0.64
UiO _{Plant Obs}	Red-listed	207	520.33	-313.33	-0.86
	Alien	67	1874.07	-1807.07	-1.86
KMN	Red-listed	31	8.08	22.92	1.17
	Alien	97	31.75	65.25	1.01
BioFokus	Red-listed	340	3704.15	-3364.15	-1.66
	Alien	169	810.51	-641.51	-1.31
eBird	Red-listed	2303	1225.58	1077.42	0.61
	Alien	49	32.10	16.90	0.42
Jordal	Red-listed	33	467.00	-434.00	-1.74
	Alien	46	214.26	-168.26	-1.29
NBIC _{CS}	Red-listed	60109	36529.70	23579.30	0.49
	Alien	4406	16565.61	-12159.61	-1.16
NBIC _{Other}	Red-listed	576	4780.77	-4204.77	-1.57
	Alien	322	273.71	48.29	0.16
NTNU	Red-listed	2	8.95	-6.95	-1.27
	Alien	1	7.58	-6.58	-1.53
UiO _{Lichen}	Red-listed	2	23.65	-21.65	-1.69
UiO _{Plant Notes}	Red-listed	79	58.30	20.70	0.30
	Alien	193	140.30	52.70	0.32
UiO _{Plant Obs}	Red-listed	50	875.45	-825.45	-1.78
	Alien	116	3153.15	-3037.15	-1.86
Ocean	KMN Red-listed	0	0.00	0.00	-2.00*

		Alien	0	0.00	0.00	-2.00*
	BioFokus	Red-listed	3	0.11	2.89	1.86
		Alien	1	0.07	0.93	1.72
	eBird	Red-listed	1	0.05	0.95	1.82
		Alien	0	0.02	-0.02	-2.00*
	Jordal	Red-listed	0	0.03	-0.03	-2.00*
		Alien	0	0.03	-0.03	-2.00*
	NBIC _{CS}	Red-listed	0	1.43	-1.43	-2.00
		Alien	1	0.68	0.32	0.37
Not mapped	NBIC _{Other}	Red-listed	19	0.20	18.80	1.96
		Alien	0	0.02	-0.02	-2.00*
	NTNU	Red-listed	0	0.00	0.00	-2.00*
		Alien	0	0.00	0.00	-2.00*
	UiO _{Lichen}	Red-listed	0	0.00	0.00	-2.00*
	UiO _{Plant Notes}	Red-listed	0	0.00	0.00	-2.00*
		Alien	0	0.00	0.00	-2.00*
	UiO _{Plant Obs}	Red-listed	0	0.06	-0.06	-2.00
		Alien	0	0.14	-0.14	-2.00

S.7. Effects of recorder group on species diversity estimates

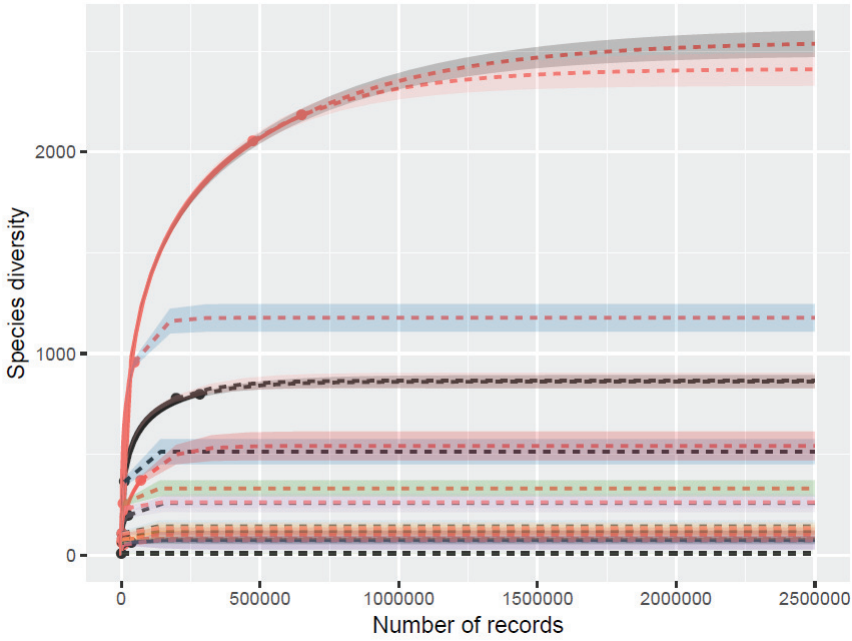


Figure S.10. Species accumulation curves based on individual based rarefaction. Conservation status is indicated by line colour, dataset is indicated by ribbon colour (see Fig. S.4 for colour legend). Solid lines reflect the interpolated species richness, dashed lines reflect the extrapolated species richness, filled circles show the observed species richness. Ribbons illustrate the 0.95 confidence interval.

Table S.7. Observed and estimated, asymptotic species richness for the four record groups, and the combined citizen science- and professional data.

Dataset	Observed species richness	Estimated species richness	S.E.	Lower 0.95 CI	Upper 0.95 CI
KMN, red-listed	63	76.486	8.310	67.436	104
KMN, alien	110	144.021	15.705	124.375	190.513
BioFokus, red-listed	955	1 177.722	40.849	1 110.932	1 273.121
BioFokus, alien	364	513.619	34.510	459.762	597.766
eBird, red-listed	78	78.000	0.440	78.000	79.080
eBird, alien	8	8.000	0.548	8.000	9.479
Jordal, red-listed	257	331.408	25.670	295.564	400.568
Jordal, alien	61	73.497	8.455	64.751	102.628
NBIC_{CS}, red-listed	2 055	2 412.084	45.691	2 333.161	2 513.104
NBIC_{CS}, alien	777	866.706	21.783	833.116	920.404
NBIC_{Other}, red-listed	372	542.664	41.081	479.183	643.744
NBIC_{Other}, alien	103	119.527	9.006	109.084	147.897
NTNU, red-listed	106	135.380	16.595	116.474	188.410
NTNU, alien	81	117.952	17.838	96.069	171.618
UiO_{Lichen}, red-listed	78	99.990	12.405	85.849	139.607
UiO_{Plant Notes}, red-listed	233	263.419	13.711	246.094	303.666
UiO_{Plant Notes}, alien	198	259.359	24.309	227.031	327.685
UiO_{Plant Obs}, red-listed	55	75.248	20.185	58.975	158.151
UiO_{Plant Obs}, alien	63	75.500	8.457	66.752	104.637
Combined, red-listed	2 182	2 550.166	46.987	2 468.978	2 654.322
Combined, alien	798	860.969	16.400	836.112	902.037



Paper II

Photo: T.K. Petersen



OPEN

Urban aliens and threatened near-naturals: Land-cover affects the species richness of alien- and threatened species in an urban-rural setting

Tanja K. Petersen^{1,2} [✉], James D. M. Speed¹ , Vidar Grøtøen² & Gunnar Austrheim¹

Urbanisation has strong effects on biodiversity patterns, but impacts vary among species groups and across spatial scales. From a local biodiversity management perspective, a more general understanding of species richness across taxonomic groups is required. This study aims to investigate how fine-scale land-cover variables influence species richness patterns of locally threatened and alien species. The study was performed in Trondheim, Norway, covering a steep urbanisation gradient. Spatially correlated Generalised Linear Mixed Effects Models predicting the number of all-, threatened-and alien species by taxon, habitat, habitat heterogeneity and mean aspect within 500 m × 500 m grid cells were constructed. The habitat categories were based on detailed land-cover maps. The highest number of threatened species was found in habitats relatively less affected by humans, whereas the number of alien species were only dependent on taxonomic group and spatial correlation. It is shown that land-cover variables within an administrative border can be used to make predictions on species richness within overarching species groups. Recommendations to biodiversity management agencies are to ensure protection of natural habitats to favour locally threatened species, and closely monitor urban areas to mitigate the introduction and spread of alien species.

The majority of the world's population now live in cities, and urbanisation is predicted to increase further in the future^{1,2}. Cities are frequently located in biodiversity hotspots, and increases in urban areas inevitably will happen at the cost of other habitats important for biodiversity^{3–6}. This calls for studies detailing how to manage biodiversity efficiently and properly in urban areas.

Various effects of urbanisation on biodiversity have been suggested and reported, depending on the exact variables in question, and the trends differ among taxa^{5,7,8}. For example, urbanisation can be a homogenising force on multiple spatial scales, impoverishing the local native species pool^{7,9,10}, while supplying alien species^{11,12}. Thus, alpha diversity might increase, despite larger-scale beta diversity decrease^{13,14}. The positive correlation between plant species richness and urbanisation can be caused by high habitat heterogeneity in urban areas, due to patches of remnant (semi-)natural habitats, allowing species with different requirements to persist¹⁵. Other reasons can be the introduction of alien plant species, e.g. for ornamental purposes, and a natural high productivity independent of human settlement^{7,11,16}.

Cadotte *et al.* (2017)¹⁷ reported that alien species richness generally increases with urbanisation. In contrast, other studies have linked urban areas with relatively high numbers of native and/or threatened species (see e.g. Kühn and Klotz (2006)¹⁴, Kowarik (2011)⁵ and references, and Ives *et al.* (2016)¹⁸).

Studies of biodiversity in relation to land use, urbanisation and human disturbance have been carried out on vastly different spatial scales (meters to >1000 km) (e.g. Blair (1996)¹³, Kühn and Klotz (2006)¹⁴, Pautasso (2007)¹⁹, Ahrne *et al.* (2009)²⁰ and Bertolino *et al.* (2015)²¹). Studies on a fine spatial scale, including a broad urbanisation gradient (ranging from industrialised to natural areas) are largely lacking however (but see Turrini

¹Department of Natural History, NTNU University Museum, Norwegian University of Science and Technology (NTNU), Erling Skakkes gt 47b, Trondheim, Norway. ²Centre for Biodiversity Dynamics, Department of Biology, NTNU, NO-7491, Trondheim, Norway. [✉]e-mail: tanja.k.petersen@ntnu.no

and Knop (2015)²³ and Concepción *et al.* (2016)²³²⁰. If the results of biodiversity research are to be used by local management, it is crucial that these results are obtained and delivered on a relatively small spatial scale, appropriate for potential management intervention. Hence, this study is performed on a spatial scale in close accordance with the spatial scale of urban planning and management.

For conservation purposes, focus is frequently placed on restricted groups, such as protection of native, threatened species or mitigation of alien species (see e.g. the Norwegian 'Natural diversity' law²⁴, the EU Habitats Directive²⁵, and the EU Regulation on Invasive Alien Species²⁶). In particular, the similarities and differences in variables determining their distributions are of interest, e.g. alien species being favoured by urbanisation, whereas native species can be threatened by such^{27,28}. Knowledge of how broad land-cover variables affect the distribution and richness of these groups could help guide decisions on city development and biodiversity management on municipality level. Threatened and alien species have rarely been investigated simultaneously (but see Deuschewitz *et al.* (2003)²⁹, Kühn *et al.* (2004)¹⁶ and Matthies *et al.* (2017)³⁰).

The aim of this study is to investigate which general, fine-scale land-cover variables influence species richness patterns of all species, as well as specifically rare and non-native species. As a case study system, we use a northern European municipality with a strong urban to rural gradient. We predict that:

Urban areas are predicted to have higher levels of alien species richness compared to non-urban areas, as cities function as introduction sites for (plant) species associated with gardens. Similarly, key pathways for introduction of alien species are through trade and traffic, which are more prevalent in urban areas than outside^{31,17,31,32}.

Urban areas are predicted to have relatively low levels of threatened species richness due to the high level of disturbance in urban areas^{7,8}. However, naturally high levels of biodiversity or suitable microhabitats within the urban matrix can potentially lead to the opposite pattern^{5,16,18}. Urban areas are viewed here as areas dominated by build-up area and immediately surrounding areas.

Forests are predicted to have high levels of both threatened- and alien species richness, as approx. 48% of the Norwegian Red-listed species are generally affiliated with forests, while several alien tree species have been planted for forestry purposes throughout Norway^{33–35}. The associations between species richness and forest cover might depend on more fine-scale forest composition and structure, out of scope of this study.

Coastal areas are ecotones, and are thus expected to host a high number of species³⁶.

Open areas with sparse vegetation (or otherwise disturbed habitat) are predicted to have high levels of alien species richness, as these are able to exploit disturbed habitat^{5,17,27}; e.g. alien plant species with a ruderal life strategy³⁷.

Habitat heterogeneity affect the richness of both groups positively, as more diverse habitat within an area provide resources for different requirements^{4,29,30,36,38–41}.

Topography: North-facing slopes are expected to be negatively correlated with overall species richness, as plants are negatively affected by a lack of light and lower temperatures⁴².

Materials and Methods

Study area. The study was carried out within Trondheim Municipality (Norway) administrative borders, around 63.42°N, 10.38°E (Fig. 1a,b). It is a southern-boreal⁴³, coastal municipality with an area of 342 km², a population of approximately 190,000 people⁴⁴, and annual mean temperature and precipitation are approximately 5 °C and 887 mm⁴⁵. The municipality holds a steep urbanisation gradient: from the city centre and industrial areas, through rural areas including agricultural areas and commercial forests, and to nature reserves and areas managed for biodiversity conservation. The municipality covers highly different nature types, including both coastline, subalpine areas and limnic systems, and thus has a high potential for varied biological communities and high levels of biodiversity⁴⁴. Trondheim municipality is fairly well-sampled with regards to species occurrence records, e.g. due to the presence and activity of the University Museum.

Data retrieval and data cleaning. *Land cover data.* Land cover was based on the Norwegian AR5 maps (Land Resource map 1:5000) from NIBIO⁴⁶. Shapefiles of the land cover maps were provided by the Trondheim Municipality in April 2018. The AR5 maps are both continually and periodically updated, and provides the most complete data on national land resources⁴⁷. Land cover is categorised based on land cover type, tree cover type, timber productivity and soil condition, giving 66 functionally unique categories within Trondheim municipality (hereafter called "land cover types") (Supplementary Material 1, Table S.1). The map was overlaid by a 500 m × 500 m grid.

Updates of the AR5 maps are mainly done if the categorical classification of an area changes, and the responsible authorities are notified of this change⁴⁶. Consequently, "unannounced" changes are not reflected in the data. As the land cover data was matched with GBIF records from 2013–2018, changes within this period are not taken into account.

Aspect of the terrain was retrieved from a Digital Terrain Model raster with a resolution of 25 m × 25 m. The circular aspect (unit: degrees) was transformed to a "northness"-measure by $Northness = \cos(\text{Aspect}^\circ)$, hence fitting a scale of -1 to 1 (in this definition: -1 = south-facing, 1 = north-facing). The values were rescaled to a gradient from 0 to 1 to match the scale of other included variables. All flat areas were given NA-values. For each grid cell, northness was calculated as the mean of all raster cells within the overlaid grid cell.

GBIF occurrence records. Large amounts of data on species occurrences are available from online databases, such as the Global Biodiversity Information Facility (GBIF)⁴⁸, and the Norwegian Biodiversity Information Centre⁴⁹.

All occurrence records from a bounding box around Trondheim Municipality (the exact municipality border was too detailed to include in the process) were downloaded from GBIF on 06/03-2018⁵⁰ (864,715 records in

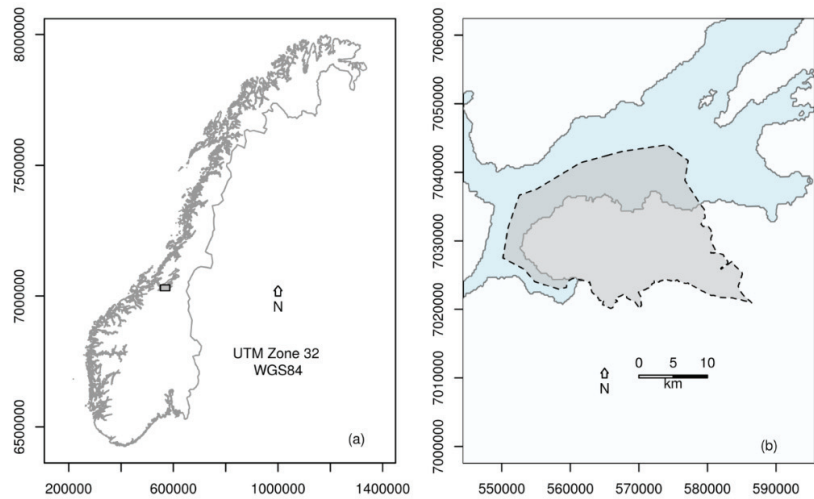


Figure 1. Location of study system. (a) Map of Norway, position of Trondheim Municipality indicated with a grey box, (b) Trondheim Municipality, municipality border indicated with dashed line and grey colour. The figure was made in R, version 3.6.1³⁵.

total, giving 9,117 unique species names; 48,468 records not identified to species level), and subsequently spatially filtered according to the municipality border. The data was additionally filtered according to the following criteria, reducing the number of records to 251,803 across 3,097 unique species names:

1. Records containing a full species name for comparability with the threatened- and alien species lists.
2. Coordinate uncertainty of ≤ 354 m, ($1/2$ length of the diagonal of $500\text{ m} \times 500\text{ m}$ grid cells).
3. Records made between January 1st 2013 and March 6th 2018 to ensure compatibility with the used land cover maps, and a negligible amount of land cover change.

Of these records, 94.9% were within the kingdom Animalia (91.4% of the total data set were birds), 3.7% within Plantae, and 1.3% within Fungi. 0.65‰ (163 records) were from outside these kingdoms (Supplementary material 2, Table S.2). The data set was divided into threatened- and alien species (only including animals, plants and fungi).

The “threatened” status was defined based on one or more assessments of the national Norwegian Red List from 2006, 2010 and 2015, provided by The Norwegian Biodiversity Information Centre. See Supplementary material 3 for detailed description of inclusion details.

The “alien” status was based on the Alien Species List (v. 2012³¹) from The Norwegian Biodiversity Information Centre. Only species alien in mainland Norway were retained (excluding species alien only to Svalbard). All alien species were included, regardless of risk category. Discrepancies in nomenclature between GBIF records and species lists were resolved using the Taxonomic Name Resolver (function “tnrs” from the taxize-package³²). Only terrestrial and limnic species were included in the data sets. All species classified as marine by The Norwegian Biodiversity Information Centre were manually excluded from the lists (excluding birds; all bird species in the data set were regarded as terrestrial).

32,585 records (121 unique species names) could be categorised as threatened (99.3% animals, 98.9% birds, 0.3% plants and 0.4% fungi), and 3,447 (177 species) as alien (64.6% animals, 63.4% birds, 34.3% plants and 1.1% fungi) (Supplementary material 2, Table S.2). The risk of species mis-identification is considered negligible, as the majority of records are associated with organisations deemed reliable regarding species identification (e.g. the Norwegian Ornithological Society, the Norwegian Botanical Society and the NTNU University Museum herbarium). Furthermore, as individual species are not analysed, it is unlikely that single erroneous records will affect the aggregated species pool.

The number of threatened- and alien species, and the overall species richness, registered within each grid cell was calculated, and divided into five taxonomic groups: birds, non-avian animals, plants, fungi and other taxa. “Other taxa” was excluded from further analyses due to a low amount of data.

Preparation of variables. *Land cover variables.* To reduce the number of land cover types while avoiding subjectively defining categories, hierarchical cluster analysis was used to identify grid cells with similar composition, creating a limited number of clustered land cover type categories. All grid cells within the administrative border of the Trondheim municipality were used for the cluster analysis, including cells only partially within the

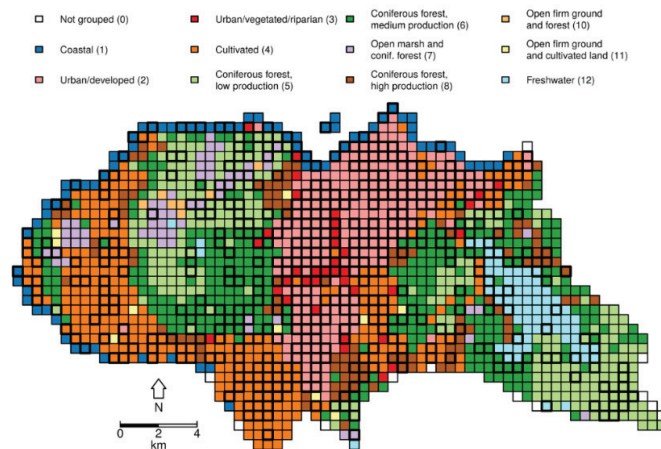


Figure 2. Trondheim Municipality coloured by habitat. Colour definitions shown in the legend. Numbers and names refer to cluster number and the given habitat name. Grid cells used for modelling are indicated with a black border. The figure was made in R, version 3.6.1³⁵.

municipality border, including only the within-municipality area. Marine grid cells (entirely covered by ocean) were not included, resulting in 1509 grid cells in total.

The cluster analysis was done using the function “hclust” on a dissimilarity matrix based on the AR5 land cover in each grid cell, using “Complete linkage” as the clustering method, and a Bray-Curtis dissimilarity matrix of the individual grid cells (function “vegdist”, package “vegan”³³). Cut-off value was set at $height=0.99$ (referring to the height of the cluster-tree, where $height=1$ indicates no clustering, and $height=0$ each individual branch (grid cell) being an autonomous cluster), resulting in 17 clusters in total, of which 6 included ≤ 3 grid cells. The clusters including ≤ 3 grid cells were mainly found on the municipality border. These were excluded from further analysis. Each cluster will hereafter be referred to as a “habitat”.

The habitats were named according to the (on average) dominating land cover types within the cells (Fig. 2, Supplementary Material 1, Fig. S.1). The number of grid cells per habitat was median 76.5 (interquartile range 30–242.25, Table 1). The most frequent habitat within the municipality was *Cultivated*, followed by *Coniferous forest* and *Urban/developed areas*.

Outliers in the number of records or number of species (evaluated separately for each taxon level) were excluded based on Tukey’s method (0.75 quantile + $1.5 \cdot IQR$). Subsequently, the habitats *Open firm ground and forest* and *Open firm ground and cultivated land* were excluded from the analyses, as only one and two grid cells remained, respectively. 485 grid cells were included in the subsequent analyses (Fig. 2).

Habitat heterogeneity was calculated for each grid cell as the Simpson’s Diversity Index. The index is calculated as $1-D$, where D is $D = \sum \left(\frac{n}{N} \right)^2$, and n is the total area of a particular land cover type, N is the total area of the grid cell. The index ranges between zero (completely homogeneous land cover) and one (infinite heterogeneity in land cover; a hypothetical value). The grid cells included in the analyses ranged between 0.012 and 0.884.

Statistical analyses. Generalised Linear Mixed Effects Models (GLMM) were constructed, predicting the threatened- (Poisson error distribution), alien- (Poisson error distribution), and overall (negative binomial error distribution due to overdispersion) number of species in each grid cell by habitat, habitat heterogeneity, northness, including an interaction with taxon (birds, non-avian animals, plants, fungi and other taxa) for all variables. Total number of records within grid cells were used as offset to account for differences in sampling effort. To account for spatial autocorrelation ($Moran's I > 1$ in exploratory Generalised Linear Models), a Matérn correlation function was used as a random effect (package *spaMM*⁵⁴). The models were fitted using Maximum Likelihood. Model selection was performed as stepwise backwards selection, based on AIC on the full models of the form: $No. \text{ species} = \text{habitat} * \text{taxon} + \text{habitat heterogeneity} * \text{taxon} + \text{northness} * \text{taxon} + \text{Matern}(1 | \text{longitude} + \text{latitude})$. The models were subsequently used to predict species richness across all grid cells within the Trondheim municipality, using 100 records as an offset.

All data preparation, analyses and figures were made in R, version 3.6.1³⁵.

Results

Different models proved to be optimal for the three species groups (all-, threatened- and alien species). For overall species richness, all predictors and interaction terms were retained, whereas threatened species richness was predicted by habitat, northness and taxon. Alien species richness was only predicted by taxon (Tables 2–4, Figs. 3–6). The spatial correlation parameters for total- ($\nu=0.460$, $\rho=0.001$), threatened- ($\nu=2.254$, $\rho=0.005$), and

Habitat no.	Name	No. (total)	No. (models)
0	Not grouped	12	0
1	Coastal	79	26
2	Urban/developed	249	142
3	Urban/vegetated/riparian	36	15
4	Cultivated	319	122
5	Coniferous forest, low production	240	51
6	Coniferous forest, medium production	315	68
7	Open marsh and coniferous forest	59	15
8	Coniferous forest, high production	109	28
10	Open firm ground and forest	7	0
11	Open firm ground and cultivated land	10	0
12	Freshwater	74	18
	Sum	1509	485

Table 1. Distribution of grid cells among habitats. The grid cells in the Not grouped-habitat include six clusters containing ≤ 3 grid cells. The number of grid cells used for modelling were the ones fulfilling the criteria listed in the methods. All grid cells were used for the predictions, except for habitat 10 and 11, as no grid cells from these habitats were included in the model building, thus having values undefined for the parameter.

alien ($v=0.759$, $\rho=0.002$) species richness indicate that all model predictions are spatially correlated (Tables 2–4). Plotting the estimated correlations against distances indicate total species richness having the farthest-reaching correlations, followed by alien and threatened species richness (Supplementary material 4, Fig. S.2).

Total species richness. For overall species richness, northness had a negative effect on species richness, the magnitude varying with taxon (Table 2, Fig. 4): non-avian animals responded most strongly to northness, followed by fungi, plants and birds. The response to habitat heterogeneity varied by taxon: plants and birds responding positively to increasing levels of habitat heterogeneity, fungi and non-avian animals having a negative response (Fig. 5). Similarly, the response to habitat differed among taxa, all other variables being held constant at mean values: fewest birds are predicted in *Open marsh and coniferous forest* followed by *Coastal*, *Freshwater*, *Coniferous forest; high production* and *Urban/developed*. The highest number was predicted for *Cultivated*, followed by *Urban/vegetated/riparian*, *Coniferous forest; low production* and *medium production*. However, 0.95 C.I. overlapped for all habitats. For non-avian animals, *Cultivated* and *Urban/developed* had lower predicted species richness compared to *Open marsh and coniferous forest*, and *Cultivated* was lower than *Coniferous forest; low production* and *medium production* as well. All other 0.95 C.I. overlapped. The highest number of fungi species was predicted for *Open marsh and coniferous forest*, 0.95 C.I. only overlapping with *Coniferous forest; high production*. The lowest number was predicted for *Coastal*, 0.95 C.I. overlapping with *Freshwater*, *Cultivated*, *Urban/vegetated/riparian* and *Urban/developed*. The lowest number of plants was predicted for *Freshwater*, followed by *Urban/developed* (0.95 overlapping with *Urban/vegetated/riparian*, *Cultivated* and *Coniferous forest; high production*). The highest number was predicted for *Open marsh and coniferous forest*, 0.95 C.I. overlapping with *Coastal*, *Coniferous forest; medium production*, *low production* and *Coniferous forest; high production* (Fig. 6).

Threatened species richness. For threatened species, increasing values of northness increase the predicted number of species (Table 3, Fig. 4). The highest species richness values are found for birds, followed by non-avian animals, fungi and plants. However, 0.95 C.I. overlap for all taxa except for birds and plants in *Urban/developed* and *Cultivated* areas. The highest numbers of species are found in *Open marsh and coniferous forest*, followed by *Cultivated*, *Coastal*, *Freshwater*, *Urban/developed*, *Coniferous forest; low production*, *high production*, *Urban/vegetated/riparian*, and *Coniferous forest; medium production*. However, all 0.95 C.I. overlap (Fig. 6).

Alien species richness. For alien species, only taxon was retained as a predictor; the highest number of species predicted for plants, followed by non-avian animals, fungi and birds. However, the 0.95 C.I. overlapped for all taxa except birds and plants (Table 4, Fig. 6).

Discussion

Urban areas are often found to have high levels of biodiversity, but little is known on how fine-scale land use is structuring species diversity in cities. We used species occurrence records from GBIF and official land cover classifications to determine how habitat affects total species richness, and the number of threatened and alien species. We did so by constructing spatially correlated Generalised Linear Mixed Effects Models based on habitat, habitat heterogeneity, aspect and taxonomic group within 500 m \times 500 m grid cells across the Trondheim municipality, selecting the best models based on Δ AIC. The best models varied for overall-, threatened and alien

<i>Fixed effects (family: negative binomial, shape = 0.789)</i>			
Marginal AIC: 8014.325	Estimate	Cond.SE	t-value
(Intercept)	-0.979	0.347	-2.819
Urban/developed	0.198	0.288	0.689
Urban/vegetated/riparian	0.295	0.428	0.689
Cultivated	0.339	0.297	1.142
Coniferous forest, low production	0.273	0.323	0.845
Coniferous forest, medium production	0.265	0.314	0.841
Open marsh and coniferous forest	-0.172	0.433	-0.396
Coniferous forest, high production	0.192	0.367	0.525
Freshwater	0.148	0.399	0.371
Plantae	-1.998	0.503	-3.975
Animal	-1.660	0.545	-3.047
Fungi	-4.834	0.938	-5.156
Habitat heterogeneity	0.007	0.326	0.021
Northness	-0.056	0.287	-0.194
Urban/developed: Plantae	-1.418	0.402	-3.525
Urban/vegetated/riparian: Plantae	-1.378	0.612	-2.253
Cultivated: Plantae	-1.063	0.421	-2.526
Coniferous forest, low production: Plantae	-0.621	0.458	-1.355
Coniferous forest, medium production: Plantae	-0.534	0.443	-1.205
Open marsh and coniferous forest: Plantae	0.648	0.599	1.082
Coniferous forest, high production: Plantae	-0.685	0.521	-1.314
Freshwater: Plantae	-4.267	0.930	-4.589
Urban/developed: Animal	-0.956	0.434	-2.206
Urban/vegetated/riparian: Animal	-0.634	0.663	-0.956
Cultivated: Animal	-1.719	0.468	-3.677
Coniferous forest, low production: Animal	-0.551	0.501	-1.099
Coniferous forest, medium production: Animal	-0.628	0.485	-1.294
Open marsh and coniferous forest: Animal	0.612	0.654	0.935
Coniferous forest, high production: Animal	-0.860	0.581	-1.480
Freshwater: Animal	-1.281	0.655	-1.954
Urban/developed: Fungi	1.684	0.862	1.952
Urban/vegetated/riparian: Fungi	1.569	1.052	1.492
Cultivated: Fungi	1.142	0.883	1.294
Coniferous forest, low production: Fungi	2.126	0.906	2.345
Coniferous forest, medium production: Fungi	2.071	0.894	2.316
Open marsh and coniferous forest: Fungi	4.054	0.988	4.103
Coniferous forest, high production: Fungi	2.513	0.945	2.659
Freshwater: Fungi	0.215	1.160	0.185
Plantae: Habitat heterogeneity	2.002	0.481	4.165
Animal: Habitat heterogeneity	-0.022	0.528	-0.041
Fungi: Habitat heterogeneity	-0.367	0.604	-0.608
Plantae: Northness	-0.084	0.398	-0.211
Animal: Northness	-0.681	0.453	-1.503
Fungi: Northness	-0.423	0.501	-0.844
<i>Random effects (family: Gaussian)</i>			
Correlation parameters		Variance parameters	
ν	ρ	λ	
0.460	0.00123	0.118	

Table 2. Model output, total species richness. Model output from the spatially correlated GLMM of total species richness. The model was constructed with a negative binomial error structure. The factor levels *Coastal* and *Aves* are used as intercepts, thus categorical predictor values are relative to these.

species richness, with total species richness depending on all predictors and their interaction with taxon, whereas threatened species richness depended on habitat, aspect and taxon, and alien species richness only depended on taxon. The relationship between species richness in general are highly complex and dependent on multiple factors and interactions (Table 2, Figs. 3–6). Threatened, native species are associated with non-anthropogenic habitats (Table 3, Figs. 4 and 6), whereas alien species are mainly affected by spatial correlations on the investigated spatial

Fixed effects (family: Poisson)			
Marginal AIC: 1400.967	Estimate	Cond.SE	t-value
(Intercept)	-2.982	0.248	-12.022
Urban/developed	-0.420	0.195	-2.151
Urban/vegetated/riparian	-0.681	0.322	-2.114
Cultivated	0.003	0.204	0.015
Coniferous forest, low production	-0.506	0.314	-1.611
Coniferous forest, medium production	-0.753	0.272	-2.764
Open marsh and coniferous forest	0.436	0.431	1.013
Coniferous forest, high production	-0.533	0.311	-1.714
Freshwater	-0.333	0.325	-1.026
Plantae	-0.987	0.320	-3.089
Animal	-0.259	0.283	-0.912
Fungi	-0.353	0.411	-0.858
Northness	0.384	0.234	1.642
Random effects (family: Gaussian)			
Correlation parameters		Variance parameters	
ν	ρ	λ	
2.254	0.00511	0.1984	

Table 3. Model output from the spatially correlated GLMM of threatened species richness. The model was constructed with a Poisson error structure. The factor levels *Coastal* and *Aves* are used as intercepts, thus categorical predictor values are relative to these.

scale (Table 4). The key findings of this study advance our understanding of the field by confirming the association of threatened, native species with more natural habitats, and the potential for establishment of alien species across all habitats on a management-relevant spatial scale.

The retention of all predictors and interactions in the model of overall species richness illustrate the complex relationships between environmental variables and different taxonomic groups. Nevertheless, the overall negative effect of northness reflect the low species richness of north-facing slopes, compared to south-facing ones⁴² (Fig. 4). The different taxa responded differently to increasing habitat heterogeneity, the only unidirectional response being for plants (positive) (Fig. 5). This supports the results of Matthies *et al.* (2017)³⁰ and Beninde *et al.* (2015)⁵⁶, in which respectively habitat heterogeneity and habitat richness were positively associated with species richness in urban areas. However, other studies have found a positive correlation for restricted taxonomic groups, such as arthropods^{39,40}, birds and mammals³⁰, which was not observed here.

Non-surprisingly, the different taxa responded differently to various habitats (Fig. 6). Interestingly, whereas plants, fungi and non-avian animals generally responded negatively to urban areas (differences not necessarily significantly different from other habitats however), the effect was less pronounced for birds. This could reflect their high mobility, and potential for an “urban adapter/exploiter”-status of some bird species⁵³. In contrast, the habitat with the highest predicted number of both plant-, non-avian animal-, and fungi species, had the lowest predicted number of bird species.

Threatened species richness generally responded positively to increasing northness, in contrast to what would be expected (Fig. 4). This could potentially be an artefact of the habitat associations; *Coastal* areas had higher northness values ($mean = 0.758$, $S.E. = 0.022$ compared to the overall $mean = 5.35$, $S.E. = 0.005$). The effect of taxon reflects the differences in the number of species within each taxonomic group being classified as threatened; (50 bird species, 26 plant species, 12 non-avian animal species, 33 fungi species included in the study). For all taxa, the lowest species numbers are predicted for all variants of coniferous forest, contrary to the initial expectations, and urban areas (Fig. 6). The negative effect of urban areas on threatened species richness mirrors the findings of Aronson *et al.* (2014)⁹, and emphasises how vulnerable native species are not pre-adapted to the changed environments of the city. Contrary to expectations, the effect of the various forest habitats on threatened species is lower than for most other habitats (Table 3). The low number of threatened species in forests can be due to the lack of sampling, showing a spatial bias in the data rather than an effect. This should however be accounted for by using the number of records as an offset in the models. Rather, large parts of the forested areas in Trondheim are strongly affected by previous afforestation for timber production, where mainly coniferous species (both native and alien) were planted⁵³. These forests might not provide the needed conditions for native species⁵⁷. Plantations and secondary vegetation have been shown to harbour fewer species than primary forests^{58,59}. The lack of association between forested areas and threatened species calls for a nuanced perspective on what forest types constitute suitable habitats for species of interest, as indicated by Ingram *et al.* (2015)⁵⁸ and Horák *et al.* (2019)⁵⁹. The highest species numbers are predicted for *Open marsh and coniferous forest* and *Coastal areas* (Fig. 6); the former is likely the habitat category reflecting the lowest human impact. The high number of threatened species in coastal habitats can likely be ascribed to these habitats being ecotones, providing varied habitat conditions. Ecotones have

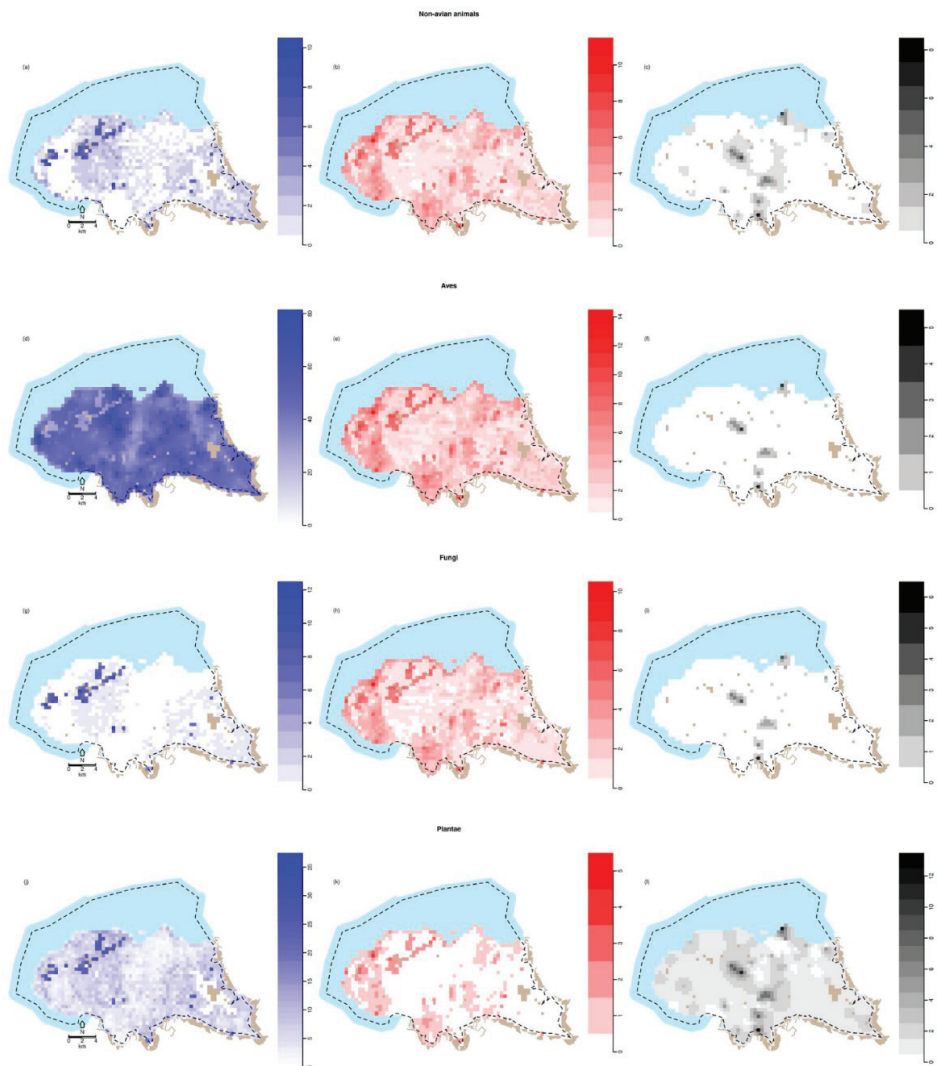


Figure 3. Maps of predicted species richness. Predicted number of species for each taxonomic- and species group given the realised habitat, habitat heterogeneity and northness. All predictions were made using 100 records (i.e. sampling events) as the offset. (a) Non-avian animals in total; (b) Threatened non-avian animals; (c) Alien non-avian animals; (d) Birds in total; (e) Threatened birds; (f) Alien birds; (g) Fungi in total; (h) Threatened fungi; (i) Alien fungi; (j) Plants in total; (k) Threatened plants; (l) Alien plants. The figure was made in R, version 3.6.1⁵⁵.

been suggested to have an increased species richness³⁶. Lloyd *et al.* (2000)³⁸ found ecotonal species to mainly be natives, which is supported by the findings here.

Interestingly, in the model of alien species richness, only taxon was retained as a significant predictor, reflecting the differences in the number of species within each taxonomic group being classified as alien (5 bird species, 156 plant species, 10 non-avian animal species, 6 fungi species included in the study). The lack of response to either of the other investigated variables stands in stark contrast to the expectations and previous findings, but can be attributed to alien species often being generalist opportunists; the spatial scale investigated does not reflect

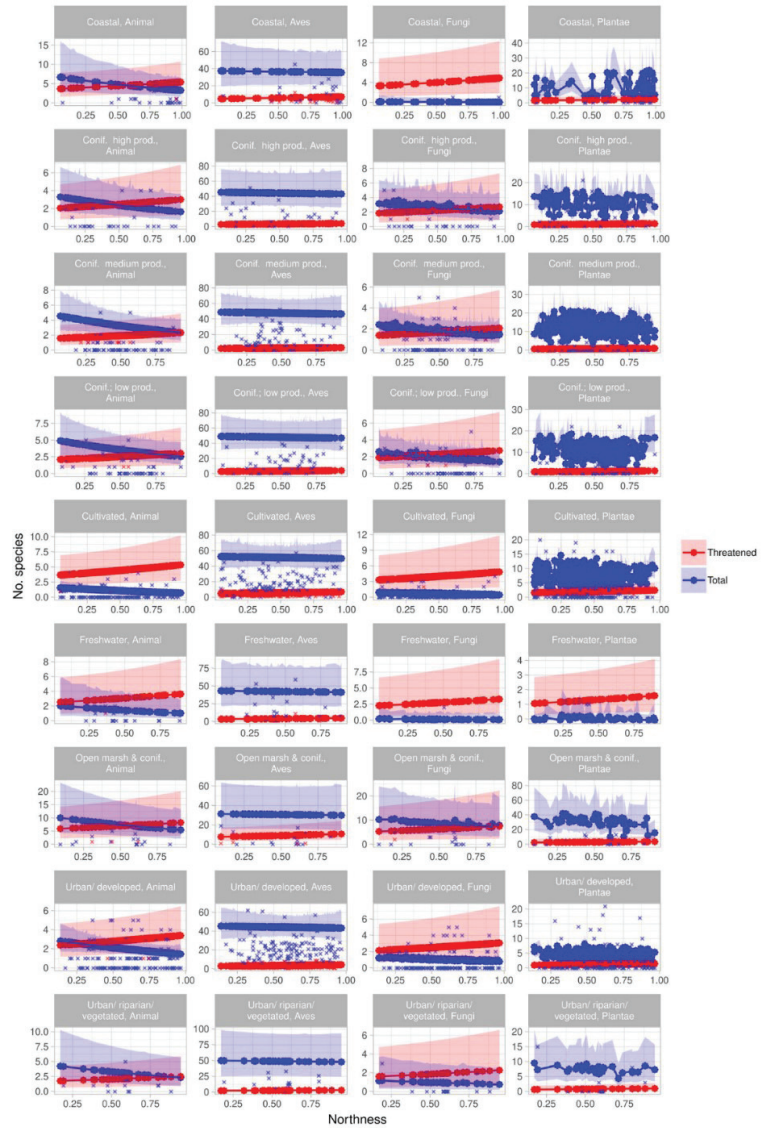


Figure 4. Effect of northness. Effect of northness on predicted species richness across taxa and within habitats. Crosses indicate observed values (incl. spatial effects and variations in all predictors), filled circles are the predictions (spatial effects removed, and all other predictors set to their mean values), and ribbons indicate 0.95 confidence intervals around the predictions. Note the different y-axes. The figure was made in R, version 3.6.1⁵⁵.

the fine-scale conditions affecting the individual species. This result highlights that on this spatial scale, all parts of the municipality are open for potential invasion by alien species. Given the spatial correlations (Supplementary material 4, Fig. S.2), it is evident that founder events and subsequent spread of alien species are of crucial importance; on the investigated scale, even more important than the configuration of environment. As many alien species are introduced through urban areas mainly due trade and traffic^{12,15,31}, emphasis must be put on the

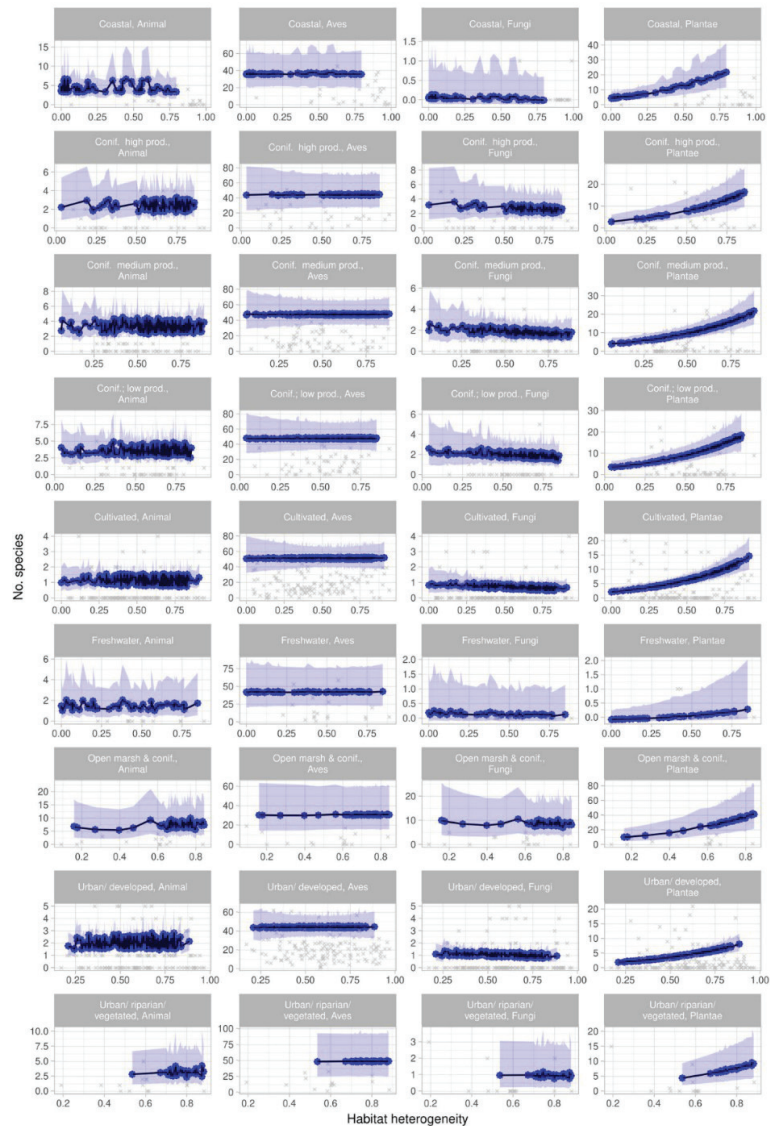


Figure 5. Effect of habitat heterogeneity. Effect of habitat heterogeneity on predicted species richness across taxa and within habitats. Crosses indicate observed values (incl. spatial effects and variations in all predictors), filled circles are the predictions (spatial effects removed, and all other predictors set to their mean values), and ribbons indicate 0.95 confidence intervals around the predictions. Note the different y-axes. The figure was made in R, version 3.6.1⁵⁵.

importance of avoiding unintentional introduction of potential invasive species. As an example, the review by Kowarik (2011)⁵ found cities to be hotspots of alien plant species. In addition, port cities have been suggested as even greater hotspots of introductions, leaving Trondheim even more vulnerable^{60,61}.

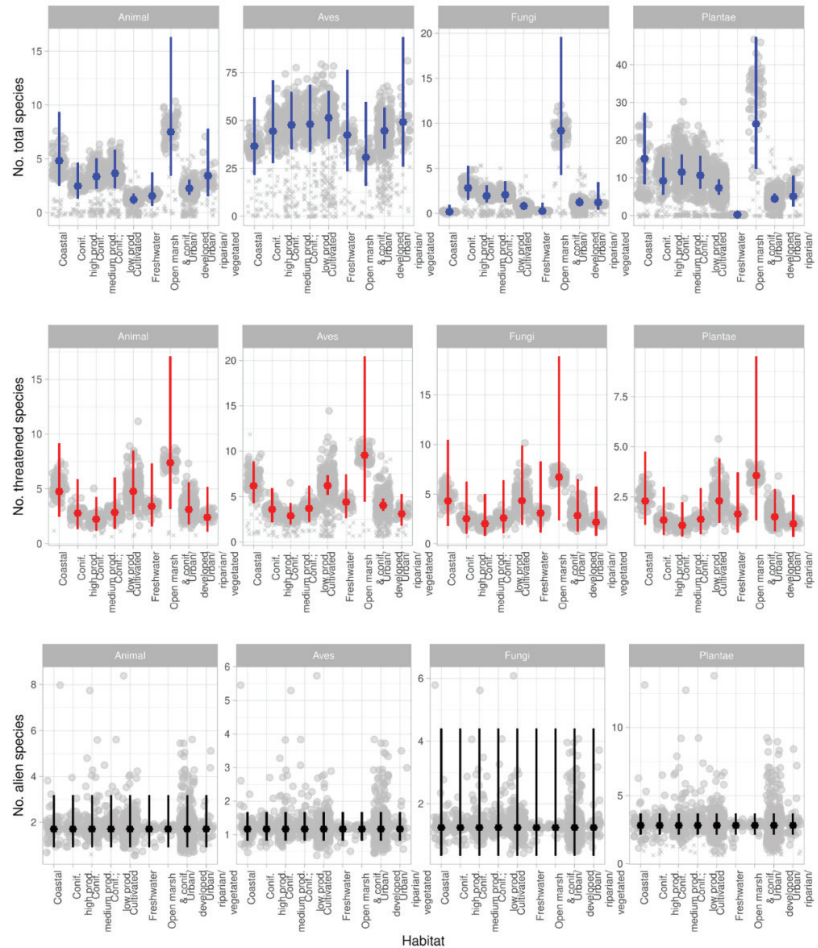


Figure 6. Effects of habitat and taxonomic group. Effect of habitat on predicted species richness across taxa and within habitats. Crosses indicate observed values (incl. spatial effects and variations in all predictors), filled grey circles are the predictions (incl. spatial effects and variations in all predictors), coloured circles indicate mean predicted values (spatial effects removed), and all other predictors set to their mean values, and lines indicate the 0.95 confidence interval of the prediction. Note the different y-axes. The figure was made in R, version 3.6.1⁵⁵.

As the explanatory variables used in these models are “indirect” (*sensu* Guisan and Zimmermann (2000)⁶²), the habitats are proxies for underlying environmental (direct) drivers. Therefore, a direct extrapolation to other geographical areas should be cautious⁶². However, the general methods are highly applicable elsewhere.

Of the 1,509 grid terrestrial cells, 485 qualified for analyses; species occurrence data was sparse in the rest. Those used in the analyses were biased towards urban areas (Table 1), supporting the general trend in citizen science data; concentrated around inhabited or areas otherwise accessible to the public^{63,64}. For example, areas within Trondheim municipality relatively far from human activity are under-sampled, with two habitats not being represented in the analyses at all (*Open firm ground and forest – and cultivated land*). This bias was accounted for in the models, but the differentiated sampling effort nevertheless leaves varying degrees of uncertainty for each habitat and taxon. The sample sizes differed among species groups, with many more observations of threatened than alien species. The differences thus might reflect sampling strategy rather than reality.

As the models are by nature rather crude, they inevitably lack predictor variables, which could have increased model accuracy. However, including highly detailed variables was not the aim of this study. Since

<i>Fixed effects (family: Poisson)</i>			
Marginal AIC: 712.727	Estimate	Cond.SE	t-value
(Intercept)	-4.441	0.80	-24.715
Plantae	0.878	0.167	5.254
Animal	0.390	0.327	1.164
Fungi	0.059	0.645	0.092
<i>Random effects (family: Gaussian)</i>			
Correlation parameters		Variance parameters	
ν	ρ	λ	
0.759	0.00178	0.597	

Table 4. Model output from the spatially correlated GLMM of alien species richness. The model was constructed with a Poisson error structure. The factor level *Aves* is used as intercept, thus categorical predictor values are relative to this.

the data set included a wide array of species, these will not respond in similar ways to variation in the included variables, or to missing variables⁶⁵. The more species included in the models, the more opposing mechanisms are attempted to be fitted within a single modelling framework, giving a poorer fit, compared to models with a narrower scope.

The number of GBIF records have increased in recent years (see Speed *et al.* (2018)⁶⁶). Of all species recorded in Trondheim, approximately 1/3 have been recorded within the municipality from year 2013 to 2018. Of the 6,020 species from the downloaded data set not included in the analyses, 33.9% (2,039) have only been recorded once, and 85.5% (5,150) have been recorded <10 times. Most of these infrequent species are insects. This taxonomic skew is likely due to this species group being poorly sampled or requiring expert knowledge to identify to species level.

Different correlations with environmental variables are expected at different spatial scales for different organisms^{19,23,66}. Taxa and species with opposing responses to the included variables could mask each other, thus not revealing the underlying mechanisms⁶⁶. Simultaneously, the mechanisms underlying species distributions vary with spatial scale, not necessarily in the same direction for different taxa^{19,23,67}. As multiple different taxonomic groups were included in this study, the used spatial scale is potentially inappropriate for all taxa.

According to Pautasso (2007)¹⁹, a negative correlation between urbanisation and species richness is expected when the study grain is smaller than 1 km, as in this study, but positive at larger scales. This is ascribed to the larger scale reflecting human settlement in productive areas, competing for space with other species, whereas the small scales reflect more detailed environmental- and land cover effects.

Our results indicate that if the Trondheim municipality is to be managed to favour biodiversity, favouring threatened species and excluding alien species, the following actions can be recommended:

Habitat heterogeneity on a relatively small spatial scale should be ensured, favouring overall species richness. This should however not be confused with fragmentation of natural habitat.

To favour threatened species, non-anthropogenic- and coastal areas should be monitored and protected, potentially expanding the actions to ecotones in general.

To limit the spread of alien species, initial introduction and establishment should be avoided. Thus, urban- and other anthropogenic areas should be closely monitored and managed^{12,68}.

Protection of important and heterogeneous habitat should be accounted for in unison with ensuring large habitat patches, rather than multiple smaller ones; a metastudy by Beninde *et al.* (2015)⁶⁶ showed patch area to have the largest positive effect on urban biodiversity.

Conclusions

Overall-, threatened- and alien species richness are not determined by the same land-cover variables. Total- and threatened species richness responds to both habitat and aspect, whereas alien species richness does not respond to any of the variables included in this study. The highest numbers of threatened species are associated with non-anthropogenic habitats, but in contrast to expectations, not more positively associated with forested areas than other habitats, calling for detailed investigations of the importance of forest environments for threatened species. Our finding that alien species do not respond to land-cover variables, but only spatial correlations, confirms the importance of founder events, and highlights the status of cities as gateways for alien species in general.

To mitigate the knowledge gaps from under-sampled habitats, we urge for sampling outside inhabited areas and for less investigated taxa. Using models build on administrative land cover maps and open database occurrence records can be a useful tool for local biodiversity management, by providing guidelines regarding where to aim future efforts, both regarding future conservation efforts and future investigations. Further work is however needed in dealing with the inherent biases of such databases.

In the case of Trondheim, an averaged sized Northern European city, the recommendations for biodiversity management are to ensure protection of natural habitats within a spatial resolution of 250,000 m², and to closely monitor and manage urban areas to mitigate the introduction and spread of alien species.

Data availability

All relevant data is available from public repository (GBIF Occurrence Download – March 6th 2018, <https://doi.org/10.15468/dl.ruacxc>).

Received: 13 January 2020; Accepted: 2 May 2020;

Published online: 22 May 2020

References

- Grimm, N. B. *et al.* Global change and the ecology of cities. *Science* (80-), **319**, 756–760 (2008).
- United Nations. World Urbanization Prospects: The 2018 Revision. *World Urbanization Prospect. 2018 Revis.* 1–2 (2018).
- Cincotta, R. P., Wisniewski, J. & Engelman, R. Human population and biodiversity hotspots. *Nature* **404**, 990–992 (2000).
- Araujo, M. B. The coincidence of people and biodiversity in Europe. *Glob. Ecol. Biogeogr.* **12**, 5–12 (2003).
- Kowarik, I. Novel urban ecosystems, biodiversity, and conservation. *Environ. Pollut.* **159**, 1974–1983 (2011).
- Ancillotto, L., Bosso, L., Salinas-Ramos, V. B. & Russo, D. The importance of ponds for the conservation of bats in urban landscapes. *Landscape Urban Plan.* **190**, 103607 (2019).
- McKinney, M. L. Urbanization, biodiversity, and conservation. *Bioscience* **52** (2002).
- Aronson, M. F. J. *et al.* A global analysis of the impacts of urbanization on bird and plant diversity reveals key anthropogenic drivers. *Proc. R. Soc. London B Biol. Sci.* **281** (2014).
- Alberti, M. The effects of urban patterns on ecosystem function. *Int. Reg. Sci. Rev.* **28**, 168–192 (2005).
- McKinney, M. L. Urbanization as a major cause of biotic homogenization. *Biol. Conserv.* **127**, 247–260 (2006).
- Gaston, K. J. Biodiversity and extinction: species and people. *Prog. Phys. Geogr.* **2**, 239–247 (2005).
- Padayachee, A. L. *et al.* How do invasive species travel to and through urban environments? *Biol. Invasions* **19**, 3557–3570 (2017).
- Blair, R. B. Land use and avian species diversity along an urban gradient. *Ecol. Appl.* **6**, 506–519 (1996).
- Kühn, I. & Klotz, S. Urbanization and homogenization - Comparing the floras of urban and rural areas in Germany. *Biol. Conserv.* **127**, 292–300 (2006).
- Francis, R. A. & Chadwick, M. A. Urban invasions: non-native and invasive species in cities. *Geography* **100**, 144–152 (2015).
- Kühn, I., Brandl, R. & Klotz, S. The flora of German cities is naturally species rich. *Evol. Ecol. Res.* **6**, 749–764 (2004).
- Cadotte, M. W., Yasui, S. L. E., Livingstone, S. & MacIvor, J. S. Are urban systems beneficial, detrimental, or indifferent for biological invasion? *Biol. Invasions* **19**, 3489–3503 (2017).
- Ives, C. D. *et al.* Cities are hotspots for threatened species. *Glob. Ecol. Biogeogr.* **25**, 117–126 (2016).
- Pautasso, M. Scale dependence of the correlation between human population presence and vertebrate and plant species richness. *Ecol. Lett.* **10**, 16–24 (2007).
- Ahrné, K., Bengtsson, J. & Elmqvist, T. Bumble bees (*Bombus* spp) along a gradient of increasing urbanization. *PLoS One* **4**, 1–9 (2009).
- Bertolino, S. *et al.* Spatially explicit models as tools for implementing effective management strategies for invasive alien mammals. *Mamm. Rev.* online ver. 1–13 (2020).
- Turrini, T. & Knop, E. A landscape ecology approach identifies important drivers of urban biodiversity. *Glob. Chang. Biol.* **21**, 1652–1667 (2015).
- Concepción, E. D. *et al.* Impacts of urban sprawl on species richness of plants, butterflies, gastropods and birds: not only built-up area matters. *Urban Ecosyst.* **19**, 225–242 (2016).
- Klima- og miljødepartementet. Naturmangfoldloven. <https://lovdata.no/dokument/NL/lov/%0A2009-06-19-100> (2009).
- European Commission. Council Directive 92/43/EEC of 21 May 1992 on the conservation of natural habitats and of wild fauna and flora. <http://data.europa.eu/eli/dir/1992/43/oj> (1992).
- European Commission. Regulation (EU) No 1143/2014 of the European Parliament and of the Council of 22 October 2014 on the prevention and management of the introduction and spread of invasive alien species. <http://data.europa.eu/eli/reg/2014/1143/oj> (2014).
- Polce, C., Kunin, W. E., Biesmeijer, J. C., Dauber, J. & Phillips, O. L. Alien and native plants show contrasting responses to climate and land use in Europe. *Glob. Ecol. Biogeogr.* **20**, 367–379 (2011).
- Godefroid, S. & Ricotta, C. Alien plant species do have a clear preference for different land uses within urban environments. *Urban Ecosyst.* **21**, 1189–1198 (2018).
- Deuschewitz, K., Lausch, A., Kühn, I. & Klotz, S. Native and alien plant species richness in relation to spatial heterogeneity on a regional scale in Germany. *Glob. Ecol. Biogeogr.* **12**, 299–311 (2003).
- Matthies, S. A., Rüter, S., Schaarschmidt, F. & Prasse, R. Determinants of species richness within and across taxonomic groups in urban green spaces. *Urban Ecosyst.* **20**, 897–909 (2017).
- Pyšek, P., Prach, K. & Mandák, B. Invasions of alien plants into habitats of Central European landscape: an historical pattern. In *Plant Invasions: Ecological Mechanisms and Human Responses* (eds. Starfinger, U., Edwards, K., Kowarik, I. & Williamson, M.) 23–32 (1998).
- Genovesi, P., Scalera, R. & van Ham, C. *Invasive alien species: the urban dimension: case studies on strengthening local action in Europe*. <https://portals.iucn.org/library/node/29131> (2013).
- Prestø, T. Bymarka. In *Bli med ut!* (eds. Fremstad, E. & Dolmen, D.) vol. 4, 5–11 (NTNU University Museum, Department of Natural History. <https://www.ntnu.no/museum/bli-med-ut>, 2005).
- Prestø, T., Hagen, D. & Vange, V. Sembrafuru Pinus cembra invaderer bynært kulturlandskap. Eksempel fra Ladehalvøya, Trondheim. *Bjøttia* **1**, 1–68 (2013).
- Henriksen, S. & Hilmo, O. *Norwegian Red List of species — methods and results. Norwegian Red List of Species* (2015).
- Walker, S. *et al.* Properties of ecotones: Evidence from five ecotones objectively determined from a coastal vegetation gradient. *J. Veg. Sci.* **14**, 579–590 (2003).
- Prinzig, A., Durka, W., Klotz, S. & Brandl, R. Which species become aliens? *Evol. Ecol.* **4**, 385–405 (2002).
- Lloyd, K. M. *et al.* Evidence on ecotone concepts from switch, environmental and anthropogenic ecotones. *J. Veg. Sci.* **11**, 903–910 (2000).
- Maskell, L. C., Bullock, J. M., Smart, S. M., Thompson, K. & Hulme, P. E. The distribution and habitat associations of non-native plant species in urban riparian habitats. *J. Veg. Sci.* **17**, 499–508 (2006).
- Sattler, T., Duelli, P., Obrist, M. K., Arlettaz, R. & Moretti, M. Response of arthropod species richness and functional groups to urban habitat structure and management. *Landscape Ecol.* **25**, 941–954 (2010).
- Aronson, M. F. J. *et al.* Biodiversity in the city: key challenges for urban green space management. *Front. Ecol. Environ.* **15**, 189–196 (2017).
- Holland, P. G. & Steyn, D. G. Vegetational responses to latitudinal variations in slope angle and aspect. *J. Biogeogr.* **2**, 179 (1975).
- Moen, A. *Vegetation. Vegetasjon* (Norwegian Mapping Authority, 1999).
- Trondheim Kommune. Trondheim Kommune. <https://www.trondheim.kommune.no/> (2020).
- Statistics Norway. Statistisk sentralbyrå. <https://www.ssb.no/> (2018).

46. Norwegian Institute of Bioeconomy Research. AR5. *Norwegian Institute of Bioeconomy Research*, <https://www.nibio.no/tema/jord/arealressurser/arealressurskart-ar5/> (2018).
47. Kartverket.SOSI, <https://kartverket.no/geodataarbeid/Standarder/SOSI/> (2019).
48. Global Biodiversity Information Facility. Global Biodiversity Information Facility, <https://www.gbif.org/> (2018).
49. Norwegian Biodiversity Information Centre. Artsdatabanken, <https://www.biodiversity.no/> (2018).
50. GBIForg. GBIF Occurrence Download (06 March 2018). Accessed from R via rgbif, <https://doi.org/10.15468/dl.ruaccx> (2018).
51. Gederas, L., Moen, T. L., Skjølseth, S. & Larsen, L.-K. *Alien species in Norway—with the Norwegian Black List 2012*. (Norwegian Biodiversity Information Centre, 2012).
52. Chamberlain, S. A. & Szöcs, E. taxize: taxonomic search and retrieval in R. *F1000Research* **2**, (2013).
53. Oksanen, J. *et al.* Package 'vegan' - Community Ecology Package. *Rpackage version 2.5-6* (2019).
54. Rousset, F. & Perdy, J.-B. Testing environmental and genetic effects in the presence of spatial autocorrelation. *Ecography (Cop.)*, **37**, 781–790 (2014).
55. R Core Team. *R: A Language and Environment for Statistical Computing*. (R Foundation for Statistical Computing, 2019).
56. Beninde, J., Veith, M. & Hochkirch, A. Biodiversity in cities needs space: a meta-analysis of factors determining intra-urban biodiversity variation. *Ecol. Lett.* **18**, 581–592 (2015).
57. Brockerhoff, E. G., Jactel, H., Parrotta, J. A., Quine, C. P. & Sayer, J. Plantation forests and biodiversity: Oxymoron or opportunity? *Biodivers. Conserv.* **17**, 925–951 (2008).
58. Ingram, D. J. *et al.* Global effects of land use on local terrestrial biodiversity. *Nature* **520**, 45–50 (2015).
59. Horák, J. *et al.* Green desert? Biodiversity patterns in forest plantations. *For. Ecol. Manage.* **433**, 343–348 (2019).
60. Tordini, E. *et al.* Diversity patterns of alien and native plant species in Trieste port area: exploring the role of urban habitats in biodiversity conservation. *Urban Ecosyst.* **20**, 1151–1160 (2017).
61. Early, R. *et al.* Global threats from invasive alien species in the twenty-first century and national response capacities. *Nat. Commun.* **7** (2016).
62. Guisan, A. & Zimmermann, N. E. Predictive habitat distribution models in ecology. *Ecol. Modell.* **135**, 147–186 (2000).
63. Graham, C. H., Ferrier, S., Huettman, F., Moritz, C. & Peterson, A. T. New developments in museum-based informatics and applications in biodiversity analysis. *Trends Ecol. Evol.* **19**, 497–503 (2004).
64. Speed, J. D. M. *et al.* Contrasting spatial, temporal and environmental patterns in observation and specimen based species occurrence data. *PLoS Biol.* **13**, 1–17 (2018).
65. Wood, J. R. *et al.* No single driver of biodiversity: Divergent responses of multiple taxa across land use types. *Ecosphere* **8** (2017).
66. Gaston, K. J. Global patterns in biodiversity. *Nature* **405**, 220–227 (2000).
67. Borgström, S. T., Elmqvist, T., Angelstam, P. & Alfsen-Norodom, C. Scale mismatches in management of urban landscapes. *Ecol. Soc.* **11** (2006).
68. Gaertner, M. *et al.* Non-native species in urban environments: patterns, processes, impacts and challenges. *Biol. Invasions* **19**, 3461–3469 (2017).

Acknowledgements

We thank the personnel of Trondheim Municipality for providing data and descriptions of land cover, and Marc Daverdin for provision and assistance with the used Digital Terrain Model.

Author contributions

T.K.P., G.A., J.D.M.S. and V.G. conceived the idea and designed the methodology; T.K.P. retrieved and analysed the data; T.K.P. wrote the first draft of the manuscript. All authors contributed critically to the drafts and gave final approval for publication.

Competing interests

The authors declare no competing interests.

Additional information

Supplementary information is available for this paper at <https://doi.org/10.1038/s41598-020-65459-2>.

Correspondence and requests for materials should be addressed to T.K.P.

Reprints and permissions information is available at www.nature.com/reprints.

Publisher's note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.



Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit <http://creativecommons.org/licenses/by/4.0/>.

© The Author(s) 2020

Supplementary material

Petersen, T. K. et al. (2020). Urban aliens and threatened near-naturals : Land-cover affects the species richness of alien- and threatened species in an urban- rural setting. *Scientific Reports*, *10* (8513), 1–14.

1. Land cover characteristic of the habitats

The characteristic AR5 land cover of each habitat are evaluated based on the mean area of each land cover (m²) within the grid cells assigned to the respective habitats. An overview of the characterising land covers can be seen in Fig. S1. The dominating land cover has been determining for the used name for each of the habitats.

The official AR5 land covers falling under the labels defined here are as shown in Table S.1 (Ahlstrøm et al. 2014)¹.

The cluster analysis was based on the land cover within the classified grid cell. The analysis was also done by incorporating the land cover within the focal grid cell and the eight first order neighbour-cells, to smooth out the categorisation. The overall patterns were similar to what was found for the “strict” analysis, however a large degree of the more fine scale variation in land cover between categories was lost. Thus, further analyses were performed using habitat categories based only on the focal grid cell.

¹ Ahlstrøm, A. P., Bjørklund, K. and Frydenlund, J. (2014) *AR5 klassifikasjonssystem. Klassifikasjon av arealressurser*.

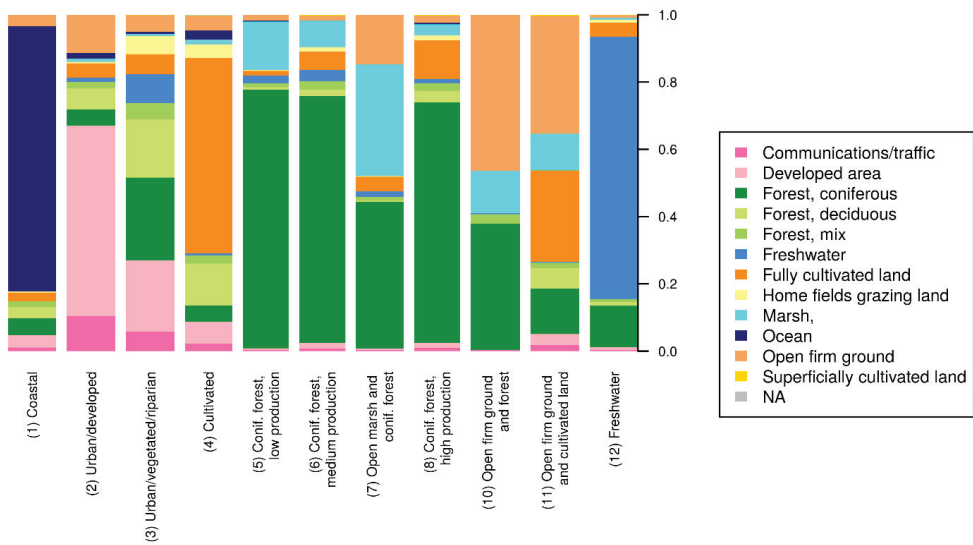


Figure S.1. Mean proportion of grid cell within each habitat covered by the respective AR5 land covers.

Table S.1 Labels of AR5 land cover types in Fig. S.1 and the included sub-classes. Only sub-classes occurring within the study area are included. Categories are translated from Ahlström et al. (2014).

Label	Area type	Tree cover	Productivity	Soil condition
Communications/traffic	Communications/traffic	-	-	-
Developed areas	Developed areas	-	-	-
Forest, coniferous	Forest	Coniferous	Impediment	Bedrock
				Shallow soil
				Soil
				Organic soil

					Boulder
					Shallow soil
			Low		Soil
					Organic soil
					Shallow soil
			Medium		Soil
					Organic soil
					Shallow soil
			High		Soil
					Organic soil
			Very high		Soil
					Bedrock
					Shallow soil
			Impediment		Soil
					Organic soil
					Soil
			Low		Organic soil
					Shallow soil
			Medium		Soil
					Organic soil
					Soil
			High		Organic soil
					Shallow soil
					Soil
			Impediment		Organic soil
					Shallow soil
			Low		Soil

				Organic soil
				Shallow soil
			Medium	Soil
				Organic soil
			High	Soil
Freshwater	Freshwater	-	-	-
Fully cultivated land	Fully cultivated land	-	-	Soil
				Organic soil
		Deciduous	-	Soil
Home fields grazing land	Home fields grazing land	-	-	Shallow soil
				Soil
				Organic soil
		Open	Impediment	-
			Impediment	-
		Coniferous	Low	-
			Medium	-
Marsh	Marsh		High	-
			Impediment	-
		Deciduous	Medium	-
			Impediment	-
		Mix	Low	-
Ocean	Ocean	-	-	-
				Artificial surface
				Bedrock
Open firm ground	Open firm ground	-	Impediment	Boulder
				Shallow soil
				Soil

					Medium	Soil
					High	Soil
Superficially cultivated land	Superficially cultivated land	-	-	-	-	Shallow soil Soil
NA		-	-	-	-	-

2. Distribution of data in taxonomic groups

Table S.2. Distribution of records between taxonomic groups. The distribution of the data from GBIF after the initial data cleaning procedure (described in the method section “*GBIF occurrence records*”), among taxonomic groups, including both number of records and number of registered species names. Capital letters indicate that subgroups are shown below in the table. Bold text indicate kingdom-level. "OTHER" indicate records/species not included in the animal-, plant- or fungi kingdom.

		All species	Threatened species	Alien species
TOTAL	Records	251,803	32,585	3,447
	Species	3,097	121	177
ANIMALIA	Records	239,038	32,351	2,226
	Species	1,353	62	15
Aves	Records	230,161	32,228	2,186
	Species	222	50	5
Mammalia	Records	729	97	7
	Species	28	4	1
Reptilia and Amphibia	Records	74	12	0
	Species	6	1	0
Arthropoda	Records	6,098	13	8

	Species	844	6	6
PLANTAE	Records	9,233	102	1,182
	Species	952	26	156
Bryophyta	Records	516	11	0
	Species	181	4	0
TRACHEOPHYTA	Records	8,481	91	1,182
	Species	702	22	156
Pinopsida	Records	357	0	124
	Species	23	0	15
Magnoliopida	Records	6,369	56	989
	Species	510	20	129
FUNGI	Records	3,369	132	39
	Species	763	33	6
OTHER	Records	163	-	-
	Species	29	-	-

3. Criteria for inclusion of threatened species

Description of the sorting of the Red List from the Norwegian Biodiversity Information Centre. The categories from 2006, 2010 and 2015 for each evaluated species evaluated were compared:

The official Norwegian Red Lists from 2006, 2010 and 2015, including the notes on the evaluations, were provided by the Norwegian Biodiversity Information Centre. The three lists provided the basis for the modified version of the Norwegian Red List used in this study. The used abbreviations are in line with the official IUCN categories: *DD* = Data Deficient, *LC* = Least Concern, *NT* = Near Threatened, *VU* = Vulnerable, *EN* = Endangered, *CR* = Critically Endangered, *RE* = Regionally Extinct, *NE* = Not Evaluated, *NA* = Not Available.

If species have not been evaluated in any year (categorised as either NA or NE), they are immediately discarded.

Similarly, if species were only evaluated as LC, or as any combination of LC and non-evaluated, they are discarded.

If species have previously been evaluated to DD (on the threatened part of the Red List), but was evaluated as LC in the latest version (2015), they were discarded.

Species listed as LC in the previous two assessments were discarded, regardless of their category in 2006.

All species evaluated to be in the threatened categories (RE, CR, EN, VU and NT) in 2015 were included in the used Red List. Species listed as DD were evaluated separately (see the further description).

All species always within the threatened categories were included in the list (incl. combinations with NA, NE and DD - thus, species never categorised as LC).

All species evaluated as data deficient in all years (DD) were included in the final version of the Red List. Similarly, all species listed as DD once, with any combination of NE and NA, were included.

All species listed as regionally extinct (RE) at any point, were included in the list.

For all species listed as LC, NA, NE or DD in the latest assessment (2015), but previously listed as any of the threatened categories (RE, CR, EN, VU, NT and DD), the notes on the reasoning for down-grading of the respective species were assessed and evaluated. Generally, species where there was great uncertainty regarding the actual current status of the species, was included in the list. Otherwise, the species was discarded.

Description/reasoning for all of the individually evaluated species can be presented upon requests.

4. Spatial correlation

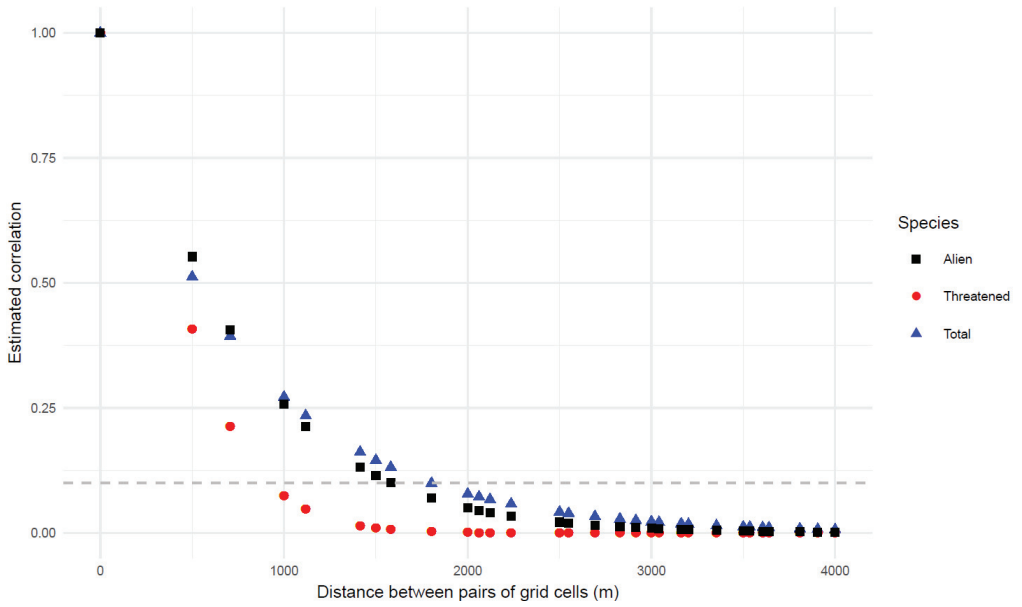


Figure S.2. Spatial correlation of models.

Estimated correlation of grid cells as a function of distance between the grid cells included in the models. Correlation- and distances are based on the spatial correlation parameters ν and ρ , as described in the model outputs (table 2-4, main text).

Paper III



Photo: T.K. Petersen

Paper 3: Petersen, T.K., Speed, J. D. M., Grøtan, V., Austrheim, G. Competitors and ruderals go to town: Plant community composition and function along an urbanization gradient. *Nordic Journal of Botany*

This article is awaiting publication and is therefore not included



Paper IV

Photo: T.K. Petersen

Paper 4: Petersen, T.K., Speed, J. D. M., Grøtan, V., Frøyen, Y.K., Austrheim, G.
Urbanisation and land-cover change affect the functional turnover of bird communities but not the extent of species composition change. Submitted manuscript to Journal of Urban Ecology.

This article is awaiting publication and is therefore not included

Doctoral theses in Biology
Norwegian University of Science and Technology
Department of Biology

Year	Name	Degree	Title
1974	Tor-Henning Iversen	Dr. philos Botany	The roles of statholiths, auxin transport, and auxin metabolism in root gravitropism
1978	Tore Slagsvold	Dr. philos Zoology	Breeding events of birds in relation to spring temperature and environmental phenology
1978	Egil Sakshaug	Dr. philos Botany	The influence of environmental factors on the chemical composition of cultivated and natural populations of marine phytoplankton
1980	Arnfinn Langeland	Dr. philos Zoology	Interaction between fish and zooplankton populations and their effects on the material utilization in a freshwater lake
1980	Helge Reinertsen	Dr. philos Botany	The effect of lake fertilization on the dynamics and stability of a limnetic ecosystem with special reference to the phytoplankton
1982	Gunn Mari Olsen	Dr. scient Botany	Gravitropism in roots of <i>Pisum sativum</i> and <i>Arabidopsis thaliana</i>
1982	Dag Dolmen	Dr. philos Zoology	Life aspects of two sympatric species of newts (<i>Triturus</i> , <i>Amphibia</i>) in Norway, with special emphasis on their ecological niche segregation
1984	Eivin Røskaft	Dr. philos Zoology	Sociobiological studies of the rook <i>Corvus frugilegus</i>
1984	Anne Margrethe Cameron	Dr. scient Botany	Effects of alcohol inhalation on levels of circulating testosterone, follicle stimulating hormone and luteinizing hormone in male mature rats
1984	Asbjørn Magne Nilssen	Dr. scient Botany	Alveolar macrophages from expectorates – Biological monitoring of workers exposed to occupational air pollution. An evaluation of the AM-test
1985	Jarle Mork	Dr. philos Zoology	Biochemical genetic studies in fish
1985	John Solem	Dr. philos Zoology	Taxonomy, distribution and ecology of caddisflies (<i>Trichoptera</i>) in the Dovrefjell mountains
1985	Randi E. Reinertsen	Dr. philos Zoology	Energy strategies in the cold: Metabolic and thermoregulatory adaptations in small northern birds

1986	Bernt-Erik Sæther	Dr. philos Zoology	Ecological and evolutionary basis for variation in reproductive traits of some vertebrates: A comparative approach
1986	Torleif Holthe	Dr. philos Zoology	Evolution, systematics, nomenclature, and zoogeography in the polychaete orders <i>Oweniimorpha</i> and <i>Terebellomorpha</i> , with special reference to the Arctic and Scandinavian fauna
1987	Helene Lampe	Dr. scient Zoology	The function of bird song in mate attraction and territorial defence, and the importance of song repertoires
1987	Olav Hogstad	Dr. philos Zoology	Winter survival strategies of the Willow tit <i>Parus montanus</i>
1987	Jarle Inge Holten	Dr. philos Botany	Autecological investigations along a coast-inland transect at Nord-Møre, Central Norway
1987	Rita Kumar	Dr. scient Botany	Somaclonal variation in plants regenerated from cell cultures of <i>Nicotiana sanderae</i> and <i>Chrysanthemum morifolium</i>
1987	Bjørn Åge Tømmerås	Dr. scient Zoology	Olfaction in bark beetle communities: Interspecific interactions in regulation of colonization density, predator - prey relationship and host attraction
1988	Hans Christian Pedersen	Dr. philos Zoology	Reproductive behaviour in willow ptarmigan with special emphasis on territoriality and parental care
1988	Tor G. Heggberget	Dr. philos Zoology	Reproduction in Atlantic Salmon (<i>Salmo salar</i>): Aspects of spawning, incubation, early life history and population structure
1988	Marianne V. Nielsen	Dr. scient Zoology	The effects of selected environmental factors on carbon allocation/growth of larval and juvenile mussels (<i>Mytilus edulis</i>)
1988	Ole Kristian Berg	Dr. scient Zoology	The formation of landlocked Atlantic salmon (<i>Salmo salar</i> L.)
1989	John W. Jensen	Dr. philos Zoology	Crustacean plankton and fish during the first decade of the manmade Nesjø reservoir, with special emphasis on the effects of gill nets and salmonid growth
1989	Helga J. Vivås	Dr. scient Zoology	Theoretical models of activity pattern and optimal foraging: Predictions for the Moose <i>Alces alces</i>
1989	Reidar Andersen	Dr. scient Zoology	Interactions between a generalist herbivore, the moose <i>Alces alces</i> , and its winter food resources: a study of behavioural variation
1989	Kurt Ingar Draget	Dr. scient Botany	Alginate gel media for plant tissue culture

1990	Bengt Finstad	Dr. scient Zoology	Osmotic and ionic regulation in Atlantic salmon, rainbow trout and Arctic charr: Effect of temperature, salinity and season
1990	Hege Johannesen	Dr. scient Zoology	Respiration and temperature regulation in birds with special emphasis on the oxygen extraction by the lung
1990	Åse Krøkje	Dr. scient Botany	The mutagenic load from air pollution at two workplaces with PAH-exposure measured with Ames Salmonella/microsome test
1990	Arne Johan Jensen	Dr. philos Zoology	Effects of water temperature on early life history, juvenile growth and prespawning migrations of Atlantic salmon (<i>Salmo salar</i>) and brown trout (<i>Salmo trutta</i>): A summary of studies in Norwegian streams
1990	Tor Jørgen Almaas	Dr. scient Zoology	Pheromone reception in moths: Response characteristics of olfactory receptor neurons to intra- and interspecific chemical cues
1990	Magne Husby	Dr. scient Zoology	Breeding strategies in birds: Experiments with the Magpie <i>Pica pica</i>
1991	Tor Kvam	Dr. scient Zoology	Population biology of the European lynx (<i>Lynx lynx</i>) in Norway
1991	Jan Henning L'Abée Lund	Dr. philos Zoology	Reproductive biology in freshwater fish, brown trout <i>Salmo trutta</i> and roach <i>Rutilus rutilus</i> in particular
1991	Asbjørn Moen	Dr. philos Botany	The plant cover of the boreal uplands of Central Norway. I. Vegetation ecology of Sølendet nature reserve; haymaking fens and birch woodlands
1991	Else Marie Løbersli	Dr. scient Botany	Soil acidification and metal uptake in plants
1991	Trond Nordtug	Dr. scient Zoology	Reflectometric studies of photomechanical adaptation in superposition eyes of arthropods
1991	Thyra Solem	Dr. scient Botany	Age, origin and development of blanket mires in Central Norway
1991	Odd Terje Sandlund	Dr. philos Zoology	The dynamics of habitat use in the salmonid genera <i>Coregonus</i> and <i>Salvelinus</i> : Ontogenic niche shifts and polymorphism
1991	Nina Jonsson	Dr. philos Zoology	Aspects of migration and spawning in salmonids
1991	Atle Bones	Dr. scient Botany	Compartmentation and molecular properties of thioglucoside glucohydrolase (myrosinase)
1992	Torgrim Breiehagen	Dr. scient Zoology	Mating behaviour and evolutionary aspects of the breeding system of two bird species: the Temminck's stint and the Pied flycatcher

1992	Anne Kjersti Bakken	Dr. scient Botany	The influence of photoperiod on nitrate assimilation and nitrogen status in timothy (<i>Phleum pratense</i> L.)
1992	Tycho Anker-Nilssen	Dr. scient Zoology	Food supply as a determinant of reproduction and population development in Norwegian Puffins <i>Fratercula arctica</i>
1992	Bjørn Munro Jenssen	Dr. philos Zoology	Thermoregulation in aquatic birds in air and water: With special emphasis on the effects of crude oil, chemically treated oil and cleaning on the thermal balance of ducks
1992	Arne Vollan Aarset	Dr. philos Zoology	The ecophysiology of under-ice fauna: Osmotic regulation, low temperature tolerance and metabolism in polar crustaceans.
1993	Geir Slupphaug	Dr. scient Botany	Regulation and expression of uracil-DNA glycosylase and O ⁶ -methylguanine-DNA methyltransferase in mammalian cells
1993	Tor Fredrik Næsje	Dr. scient Zoology	Habitat shifts in coregonids.
1993	Yngvar Asbjørn Olsen	Dr. scient Zoology	Cortisol dynamics in Atlantic salmon, <i>Salmo salar</i> L.: Basal and stressor-induced variations in plasma levels and some secondary effects.
1993	Bård Pedersen	Dr. scient Botany	Theoretical studies of life history evolution in modular and clonal organisms
1993	Ole Petter Thangstad	Dr. scient Botany	Molecular studies of myrosinase in Brassicaceae
1993	Thrine L. M. Heggberget	Dr. scient Zoology	Reproductive strategy and feeding ecology of the Eurasian otter <i>Lutra lutra</i> .
1993	Kjetil Bevanger	Dr. scient Zoology	Avian interactions with utility structures, a biological approach.
1993	Kåre Haugan	Dr. scient Botany	Mutations in the replication control gene trfA of the broad host-range plasmid RK2
1994	Peder Fiske	Dr. scient Zoology	Sexual selection in the lekking great snipe (<i>Gallinago media</i>): Male mating success and female behaviour at the lek
1994	Kjell Inge Reitan	Dr. scient Botany	Nutritional effects of algae in first-feeding of marine fish larvae
1994	Nils Røv	Dr. scient Zoology	Breeding distribution, population status and regulation of breeding numbers in the northeast-Atlantic Great Cormorant <i>Phalacrocorax carbo carbo</i>
1994	Annette-Susanne Hoepfner	Dr. scient Botany	Tissue culture techniques in propagation and breeding of Red Raspberry (<i>Rubus idaeus</i> L.)

1994	Inga Elise Bruteig	Dr. scient Botany	Distribution, ecology and biomonitoring studies of epiphytic lichens on conifers
1994	Geir Johnsen	Dr. scient Botany	Light harvesting and utilization in marine phytoplankton: Species-specific and photoadaptive responses
1994	Morten Bakken	Dr. scient Zoology	Infanticidal behaviour and reproductive performance in relation to competition capacity among farmed silver fox vixens, <i>Vulpes vulpes</i>
1994	Arne Moksnes	Dr. philos Zoology	Host adaptations towards brood parasitism by the Cockoo
1994	Solveig Bakken	Dr. scient Botany	Growth and nitrogen status in the moss <i>Dicranum majus</i> Sm. as influenced by nitrogen supply
1994	Torbjørn Forseth	Dr. scient Zoology	Bioenergetics in ecological and life history studies of fishes.
1995	Olav Vadstein	Dr. philos Botany	The role of heterotrophic planktonic bacteria in the cycling of phosphorus in lakes: Phosphorus requirement, competitive ability and food web interactions
1995	Hanne Christensen	Dr. scient Zoology	Determinants of Otter <i>Lutra lutra</i> distribution in Norway: Effects of harvest, polychlorinated biphenyls (PCBs), human population density and competition with mink <i>Mustela vison</i>
1995	Svein Håkon Lorentsen	Dr. scient Zoology	Reproductive effort in the Antarctic Petrel <i>Thalassoica antarctica</i> ; the effect of parental body size and condition
1995	Chris Jørgen Jensen	Dr. scient Zoology	The surface electromyographic (EMG) amplitude as an estimate of upper trapezius muscle activity
1995	Martha Kold Bakkevig	Dr. scient Zoology	The impact of clothing textiles and construction in a clothing system on thermoregulatory responses, sweat accumulation and heat transport
1995	Vidar Moen	Dr. scient Zoology	Distribution patterns and adaptations to light in newly introduced populations of <i>Mysis relicta</i> and constraints on Cladoceran and Char populations
1995	Hans Haavardsholm Blom	Dr. philos Botany	A revision of the <i>Schistidium apocarpum</i> complex in Norway and Sweden
1996	Jorun Skjærmo	Dr. scient Botany	Microbial ecology of early stages of cultivated marine fish; impact fish-bacterial interactions on growth and survival of larvae
1996	Ola Ugedal	Dr. scient Zoology	Radiocesium turnover in freshwater fishes
1996	Ingibjörg Einarisdóttir	Dr. scient Zoology	Production of Atlantic salmon (<i>Salmo salar</i>) and Arctic charr (<i>Salvelinus alpinus</i>): A study of some physiological and immunological responses to rearing routines

1996	Christina M. S. Pe-reira	Dr. scient Zoology	Glucose metabolism in salmonids: Dietary effects and hormonal regulation
1996	Jan Fredrik Børseth	Dr. scient Zoology	The sodium energy gradients in muscle cells of <i>Mytilus edulis</i> and the effects of organic xenobiotics
1996	Gunnar Henriksen	Dr. scient Zoology	Status of Grey seal <i>Halichoerus grypus</i> and Harbour seal <i>Phoca vitulina</i> in the Barents sea region
1997	Gunvor Øie	Dr. scient Botany	Eevaluation of rotifer <i>Brachionus plicatilis</i> quality in early first feeding of turbot <i>Scophthalmus maximus</i> L. larvae
1997	Håkon Holien	Dr. scient Botany	Studies of lichens in spruce forest of Central Norway. Diversity, old growth species and the relationship to site and stand parameters
1997	Ole Reitan	Dr. scient Zoology	Responses of birds to habitat disturbance due to damming
1997	Jon Arne Grøttum	Dr. scient Zoology	Physiological effects of reduced water quality on fish in aquaculture
1997	Per Gustav Thingstad	Dr. scient Zoology	Birds as indicators for studying natural and human-induced variations in the environment, with special emphasis on the suitability of the Pied Flycatcher
1997	Torgeir Nygård	Dr. scient Zoology	Temporal and spatial trends of pollutants in birds in Norway: Birds of prey and Willow Grouse used as
1997	Signe Nybø	Dr. scient Zoology	Impacts of long-range transported air pollution on birds with particular reference to the dipper <i>Cinclus cinclus</i> in southern Norway
1997	Atle Wibe	Dr. scient Zoology	Identification of conifer volatiles detected by receptor neurons in the pine weevil (<i>Hylobius abietis</i>), analysed by gas chromatography linked to electrophysiology and to mass spectrometry
1997	Rolv Lundheim	Dr. scient Zoology	Adaptive and incidental biological ice nucleators
1997	Arild Magne Landa	Dr. scient Zoology	Wolverines in Scandinavia: ecology, sheep depredation and conservation
1997	Kåre Magne Nielsen	Dr. scient Botany	An evolution of possible horizontal gene transfer from plants to soil bacteria by studies of natural transformation in <i>Acinetobacter calcoaceticus</i>
1997	Jarle Tufto	Dr. scient Zoology	Gene flow and genetic drift in geographically structured populations: Ecological, population genetic, and statistical models
1997	Trygve Hesthagen	Dr. philos Zoology	Population responses of Arctic charr (<i>Salvelinus alpinus</i> (L.)) and brown trout (<i>Salmo trutta</i> L.) to acidification in Norwegian inland waters

1997	Trygve Sigholt	Dr. philos Zoology	Control of Parr-smolt transformation and seawater tolerance in farmed Atlantic Salmon (<i>Salmo salar</i>) Effects of photoperiod, temperature, gradual seawater acclimation, NaCl and betaine in the diet
1997	Jan Østnes	Dr. scient Zoology	Cold sensation in adult and neonate birds
1998	Seethaledsumy Visvalingam	Dr. scient Botany	Influence of environmental factors on myrosinases and myrosinase-binding proteins
1998	Thor Harald Ringsby	Dr. scient Zoology	Variation in space and time: The biology of a House sparrow metapopulation
1998	Erling Johan Solberg	Dr. scient Zoology	Variation in population dynamics and life history in a Norwegian moose (<i>Alces alces</i>) population: consequences of harvesting in a variable environment
1998	Sigurd Mjøen Saastad	Dr. scient Botany	Species delimitation and phylogenetic relationships between the Sphagnum recurvum complex (Bryophyta): genetic variation and phenotypic plasticity
1998	Bjarte Mortensen	Dr. scient Botany	Metabolism of volatile organic chemicals (VOCs) in a head liver S9 vial equilibration system in vitro
1998	Gunnar Austrheim	Dr. scient Botany	Plant biodiversity and land use in subalpine grasslands. – A conservation biological approach
1998	Bente Gunnveig Berg	Dr. scient Zoology	Encoding of pheromone information in two related moth species
1999	Kristian Overskaug	Dr. scient Zoology	Behavioural and morphological characteristics in Northern Tawny Owls <i>Strix aluco</i> : An intra- and inter-specific comparative approach
1999	Hans Kristen Stenøien	Dr. scient Botany	Genetic studies of evolutionary processes in various populations of nonvascular plants (mosses, liverworts and hornworts)
1999	Trond Arnesen	Dr. scient Botany	Vegetation dynamics following trampling and burning in the outlying haylands at Sølendet, Central Norway
1999	Ingvar Stenberg	Dr. scient Zoology	Habitat selection, reproduction and survival in the White-backed Woodpecker <i>Dendrocopos leucotos</i>
1999	Stein Olle Johansen	Dr. scient Botany	A study of driftwood dispersal to the Nordic Seas by dendrochronology and wood anatomical analysis
1999	Trina Falck Galloway	Dr. scient Zoology	Muscle development and growth in early life stages of the Atlantic cod (<i>Gadus morhua</i> L.) and Halibut (<i>Hippoglossus hippoglossus</i> L.)
1999	Marianne Giæver	Dr. scient Zoology	Population genetic studies in three gadoid species: blue whiting (<i>Micromisistius poutassou</i>), haddock (<i>Melanogrammus aeglefinus</i>) and cod (<i>Gadus morhua</i>) in the North-East Atlantic

1999	Hans Martin Hanslin	Dr. scient Botany	The impact of environmental conditions of density dependent performance in the boreal forest bryophytes <i>Dicranum majus</i> , <i>Hylocomium splendens</i> , <i>Plagiochila asplenigides</i> , <i>Ptilium crista-castrensis</i> and <i>Rhytidiadelphus lokeus</i>
1999	Ingrid Bysveen Mjølnørød	Dr. scient Zoology	Aspects of population genetics, behaviour and performance of wild and farmed Atlantic salmon (<i>Salmo salar</i>) revealed by molecular genetic techniques
1999	Else Berit Skagen	Dr. scient Botany	The early regeneration process in protoplasts from <i>Brassica napus</i> hypocotyls cultivated under various g-forces
1999	Stein-Are Sæther	Dr. philos Zoology	Mate choice, competition for mates, and conflicts of interest in the Lekking Great Snipe
1999	Katrine Wangen Rustad	Dr. scient Zoology	Modulation of glutamatergic neurotransmission related to cognitive dysfunctions and Alzheimer's disease
1999	Per Terje Smiseth	Dr. scient Zoology	Social evolution in monogamous families:
1999	Gunnbjørn Bremset	Dr. scient Zoology	Young Atlantic salmon (<i>Salmo salar</i> L.) and Brown trout (<i>Salmo trutta</i> L.) inhabiting the deep pool habitat, with special reference to their habitat use, habitat preferences and competitive interactions
1999	Frode Ødegaard	Dr. scient Zoology	Host specificity as a parameter in estimates of arthropod species richness
1999	Sonja Andersen	Dr. scient Zoology	Expressional and functional analyses of human, secretory phospholipase A2
2000	Ingrid Salvesen	Dr. scient Botany	Microbial ecology in early stages of marine fish: Development and evaluation of methods for microbial management in intensive larviculture
2000	Ingar Jostein Øien	Dr. scient Zoology	The Cuckoo (<i>Cuculus canorus</i>) and its host: adaptations and counteradaptations in a coevolutionary arms race
2000	Pavlos Makridis	Dr. scient Botany	Methods for the microbial control of live food used for the rearing of marine fish larvae
2000	Sigbjørn Stokke	Dr. scient Zoology	Sexual segregation in the African elephant (<i>Loxodonta africana</i>)
2000	Odd A. Gulseth	Dr. philos Zoology	Seawater tolerance, migratory behaviour and growth of Charr, (<i>Salvelinus alpinus</i>), with emphasis on the high Arctic Dieset charr on Spitsbergen, Svalbard
2000	Pål A. Olsvik	Dr. scient Zoology	Biochemical impacts of Cd, Cu and Zn on brown trout (<i>Salmo trutta</i>) in two mining-contaminated rivers in Central Norway

2000	Sigurd Einum	Dr. scient Zoology	Maternal effects in fish: Implications for the evolution of breeding time and egg size
2001	Jan Ove Evjemo	Dr. scient Zoology	Production and nutritional adaptation of the brine shrimp <i>Artemia</i> sp. as live food organism for larvae of marine cold water fish species
2001	Olga Hilmo	Dr. scient Botany	Lichen response to environmental changes in the managed boreal forest systems
2001	Ingebrigt Uglem	Dr. scient Zoology	Male dimorphism and reproductive biology in corkwing wrasse (<i>Symphodus melops</i> L.)
2001	Bård Gunnar Stokke	Dr. scient Zoology	Coevolutionary adaptations in avian brood parasites and their hosts
2002	Ronny Aanes	Dr. scient Zoology	Spatio-temporal dynamics in Svalbard reindeer (<i>Rangifer tarandus platyrhynchus</i>)
2002	Mariann Sandsund	Dr. scient Zoology	Exercise- and cold-induced asthma. Respiratory and thermoregulatory responses
2002	Dag-Inge Øien	Dr. scient Botany	Dynamics of plant communities and populations in boreal vegetation influenced by scything at Sølendet, Central Norway
2002	Frank Rosell	Dr. scient Zoology	The function of scent marking in beaver (<i>Castor fiber</i>)
2002	Janne Østvang	Dr. scient Botany	The Role and Regulation of Phospholipase A ₂ in Monocytes During Atherosclerosis Development
2002	Terje Thun	Dr. philos Biology	Dendrochronological constructions of Norwegian conifer chronologies providing dating of historical material
2002	Birgit Hafjeld Borgen	Dr. scient Bi- ology	Functional analysis of plant idioblasts (Myrosin cells) and their role in defense, development and growth
2002	Bård Øyvind Solberg	Dr. scient Bi- ology	Effects of climatic change on the growth of dominating tree species along major environmental gradients
2002	Per Winge	Dr. scient Bi- ology	The evolution of small GTP binding proteins in cellular organisms. Studies of RAC GTPases in <i>Arabidopsis thaliana</i> and the Ral GTPase from <i>Drosophila melanogaster</i>
2002	Henrik Jensen	Dr. scient Bi- ology	Causes and consequences of individual variation in fitness-related traits in house sparrows
2003	Jens Rohloff	Dr. philos Biology	Cultivation of herbs and medicinal plants in Norway – Essential oil production and quality control
2003	Åsa Maria O. Espmark Wibe	Dr. scient Bi- ology	Behavioural effects of environmental pollution in threespine stickleback <i>Gasterosteus aculeatur</i> L.
2003	Dagmar Hagen	Dr. scient Bi- ology	Assisted recovery of disturbed arctic and alpine vegetation – an integrated approach

2003	Bjørn Dahle	Dr. scient Biology	Reproductive strategies in Scandinavian brown bears
2003	Cyril Lebogang Taolo	Dr. scient Biology	Population ecology, seasonal movement and habitat use of the African buffalo (<i>Syncerus caffer</i>) in Chobe National Park, Botswana
2003	Marit Stranden	Dr. scient Biology	Olfactory receptor neurones specified for the same odorants in three related Heliothine species (<i>Helicoverpa armigera</i> , <i>Helicoverpa assulta</i> and <i>Heliothis virescens</i>)
2003	Kristian Hassel	Dr. scient Biology	Life history characteristics and genetic variation in an expanding species, <i>Pogonatum dentatum</i>
2003	David Alexander Rae	Dr. scient Biology	Plant- and invertebrate-community responses to species interaction and microclimatic gradients in alpine and Arctic environments
2003	Åsa A Borg	Dr. scient Biology	Sex roles and reproductive behaviour in gobies and guppies: a female perspective
2003	Eldar Åsgard Bendiksen	Dr. scient Biology	Environmental effects on lipid nutrition of farmed Atlantic salmon (<i>Salmo salar</i> L.) parr and smolt
2004	Torkild Bakken	Dr. scient Biology	A revision of Nereidinae (Polychaeta, Nereididae)
2004	Ingar Pareliusson	Dr. scient Biology	Natural and Experimental Tree Establishment in a Fragmented Forest, Ambohitantely Forest Reserve, Madagascar
2004	Tore Brembu	Dr. scient Biology	Genetic, molecular and functional studies of RAC GTPases and the WAVE-like regulatory protein complex in <i>Arabidopsis thaliana</i>
2004	Liv S. Nilsen	Dr. scient Biology	Coastal heath vegetation on central Norway; recent past, present state and future possibilities
2004	Hanne T. Skiri	Dr. scient Biology	Olfactory coding and olfactory learning of plant odours in heliothine moths. An anatomical, physiological and behavioural study of three related species (<i>Heliothis virescens</i> , <i>Helicoverpa armigera</i> and <i>Helicoverpa assulta</i>)
2004	Lene Østby	Dr. scient Biology	Cytochrome P4501A (CYP1A) induction and DNA adducts as biomarkers for organic pollution in the natural environment
2004	Emmanuel J. Gerreta	Dr. philos Biology	The Importance of Water Quality and Quantity in the Tropical Ecosystems, Tanzania
2004	Linda Dalen	Dr. scient Biology	Dynamics of Mountain Birch Treelines in the Scandes Mountain Chain, and Effects of Climate Warming

2004	Lisbeth Mehli	Dr. scient Biology	Polygalacturonase-inhibiting protein (PGIP) in cultivated strawberry (<i>Fragaria x ananassa</i>): characterisation and induction of the gene following fruit infection by <i>Botrytis cinerea</i>
2004	Børge Moe	Dr. scient Biology	Energy-Allocation in Avian Nestlings Facing Short-Term Food Shortage
2005	Matilde Skogen Chauton	Dr. scient Biology	Metabolic profiling and species discrimination from High-Resolution Magic Angle Spinning NMR analysis of whole-cell samples
2005	Sten Karlsson	Dr. scient Biology	Dynamics of Genetic Polymorphisms
2005	Terje Bongard	Dr. scient Biology	Life History strategies, mate choice, and parental investment among Norwegians over a 300-year period
2005	Tonette Røsteliën	PhD Biology	Functional characterisation of olfactory receptor neurone types in heliothine moths
2005	Erlend Kristiansen	Dr. scient Biology	Studies on antifreeze proteins
2005	Eugen G. Sørmo	Dr. scient Biology	Organochlorine pollutants in grey seal (<i>Halichoerus grypus</i>) pups and their impact on plasma thyroid hormone and vitamin A concentrations
2005	Christian Westad	Dr. scient Biology	Motor control of the upper trapezius
2005	Lasse Mork Olsen	PhD Biology	Interactions between marine osmo- and phagotrophs in different physicochemical environments
2005	Åslaug Viken	PhD Biology	Implications of mate choice for the management of small populations
2005	Ariaya Hymete Sahle Dingle	PhD Biology	Investigation of the biological activities and chemical constituents of selected <i>Echinops</i> spp. growing in Ethiopia
2005	Anders Gravbrøt Finstad	PhD Biology	Salmonid fishes in a changing climate: The winter challenge
2005	Shimane Washington Makabu	PhD Biology	Interactions between woody plants, elephants and other browsers in the Chobe Riverfront, Botswana
2005	Kjartan Østbye	Dr. scient Biology	The European whitefish <i>Coregonus lavaretus</i> (L.) species complex: historical contingency and adaptive radiation
2006	Kari Mette Murvoll	PhD Biology	Levels and effects of persistent organic pollutants (POPs) in seabirds, Retinoids and α -tocopherol – potential biomarkers of POPs in birds?
2006	Ivar Herfindal	Dr. scient Biology	Life history consequences of environmental variation along ecological gradients in northern ungulates

2006	Nils Egil Tokle	PhD Biology	Are the ubiquitous marine copepods limited by food or predation? Experimental and field-based studies with main focus on <i>Calanus finmarchicus</i>
2006	Jan Ove Gjershaug	Dr. philos Biology	Taxonomy and conservation status of some booted eagles in south-east Asia
2006	Jon Kristian Skei	Dr. scient Biology	Conservation biology and acidification problems in the breeding habitat of amphibians in Norway
2006	Johanna Järnegren	PhD Biology	<i>Acesta oophaga</i> and <i>Acesta excavata</i> – a study of hidden biodiversity
2006	Bjørn Henrik Hansen	PhD Biology	Metal-mediated oxidative stress responses in brown trout (<i>Salmo trutta</i>) from mining contaminated rivers in Central Norway
2006	Vidar Grotan	PhD Biology	Temporal and spatial effects of climate fluctuations on population dynamics of vertebrates
2006	Jafari R Kideghesho	PhD Biology	Wildlife conservation and local land use conflicts in Western Serengeti Corridor, Tanzania
2006	Anna Maria Billing	PhD Biology	Reproductive decisions in the sex role reversed pipefish <i>Syngnathus typhle</i> : when and how to invest in reproduction
2006	Henrik Pärn	PhD Biology	Female ornaments and reproductive biology in the bluethroat
2006	Anders J. Fjellheim	PhD Biology	Selection and administration of probiotic bacteria to marine fish larvae
2006	P. Andreas Svensson	PhD Biology	Female coloration, egg carotenoids and reproductive success: gobies as a model system
2007	Sindre A. Pedersen	PhD Biology	Metal binding proteins and antifreeze proteins in the beetle <i>Tenebrio molitor</i> - a study on possible competition for the semi-essential amino acid cysteine
2007	Kasper Hancke	PhD Biology	Photosynthetic responses as a function of light and temperature: Field and laboratory studies on marine microalgae
2007	Tomas Holmern	PhD Biology	Bushmeat hunting in the western Serengeti: Implications for community-based conservation
2007	Kari Jørgensen	PhD Biology	Functional tracing of gustatory receptor neurons in the CNS and chemosensory learning in the moth <i>Heliothis virescens</i>
2007	Stig Ulland	PhD Biology	Functional Characterisation of Olfactory Receptor Neurons in the Cabbage Moth, (<i>Mamestra brassicae</i> L.) (Lepidoptera, Noctuidae). Gas Chromatography Linked to Single Cell Recordings and Mass Spectrometry

2007	Snorre Henriksen	PhD Biology	Spatial and temporal variation in herbivore resources at northern latitudes
2007	Roelof Frans May	PhD Biology	Spatial Ecology of Wolverines in Scandinavia
2007	Vedasto Gabriel Ndibalema	PhD Biology	Demographic variation, distribution and habitat use between wildebeest sub-populations in the Serengeti National Park, Tanzania
2007	Julius William Nyahongo	PhD Biology	Depredation of Livestock by wild Carnivores and Illegal Utilization of Natural Resources by Humans in the Western Serengeti, Tanzania
2007	Shombe Ntaruluka Hassan	PhD Biology	Effects of fire on large herbivores and their forage resources in Serengeti, Tanzania
2007	Per-Arvid Wold	PhD Biology	Functional development and response to dietary treatment in larval Atlantic cod (<i>Gadus morhua</i> L.) Focus on formulated diets and early weaning
2007	Anne Skjetne Mortensen	PhD Biology	Toxicogenomics of Aryl Hydrocarbon- and Estrogen Receptor Interactions in Fish: Mechanisms and Profiling of Gene Expression Patterns in Chemical Mixture Exposure Scenarios
2008	Brage Bremset Hansen	PhD Biology	The Svalbard reindeer (<i>Rangifer tarandus platyrhynchus</i>) and its food base: plant-herbivore interactions in a high-arctic ecosystem
2008	Jiska van Dijk	PhD Biology	Wolverine foraging strategies in a multiple-use landscape
2008	Flora John Magige	PhD Biology	The ecology and behaviour of the Masai Ostrich (<i>Struthio camelus massaicus</i>) in the Serengeti Ecosystem, Tanzania
2008	Bernt Rønning	PhD Biology	Sources of inter- and intra-individual variation in basal metabolic rate in the zebra finch, <i>Taeniopygia guttata</i>
2008	Solvi Wehn	PhD Biology	Biodiversity dynamics in semi-natural mountain landscapes - A study of consequences of changed agricultural practices in Eastern Jotunheimen
2008	Trond Moxness Kortner	PhD Biology	The Role of Androgens on previtellogenic oocyte growth in Atlantic cod (<i>Gadus morhua</i>): Identification and patterns of differentially expressed genes in relation to Stereological Evaluations
2008	Katarina Mariann Jørgensen	Dr. scient Biology	The role of platelet activating factor in activation of growth arrested keratinocytes and re-epithelialisation
2008	Tommy Jørstad	PhD Biology	Statistical Modelling of Gene Expression Data
2008	Anna Kusnierczyk	PhD Biology	<i>Arabidopsis thaliana</i> Responses to Aphid Infestation

2008	Jussi Evertsen	PhD Biology	Herbivore sacoglossans with photosynthetic chloroplasts
2008	John Eilif Hermansen	PhD Biology	Mediating ecological interests between locals and globals by means of indicators. A study attributed to the asymmetry between stakeholders of tropical forest at Mt. Kilimanjaro, Tanzania
2008	Ragnhild Lyngved	PhD Biology	Somatic embryogenesis in <i>Cyclamen persicum</i> . Biological investigations and educational aspects of cloning
2008	Line Elisabeth Sundt-Hansen	PhD Biology	Cost of rapid growth in salmonid fishes
2008	Line Johansen	PhD Biology	Exploring factors underlying fluctuations in white clover populations – clonal growth, population structure and spatial distribution
2009	Astrid Jullumstrø Feuerherm	PhD Biology	Elucidation of molecular mechanisms for pro-inflammatory phospholipase A2 in chronic disease
2009	Pål Kvello	PhD Biology	Neurons forming the network involved in gustatory coding and learning in the moth <i>Heliothis virescens</i> : Physiological and morphological characterisation, and integration into a standard brain atlas
2009	Trygve Devold Kjellsen	PhD Biology	Extreme Frost Tolerance in Boreal Conifers
2009	Johan Reinert Vikan	PhD Biology	Coevolutionary interactions between common cuckoos <i>Cuculus canorus</i> and <i>Fringilla</i> finches
2009	Zsolt Volent	PhD Biology	Remote sensing of marine environment: Applied surveillance with focus on optical properties of phytoplankton, coloured organic matter and suspended matter
2009	Lester Rocha	PhD Biology	Functional responses of perennial grasses to simulated grazing and resource availability
2009	Dennis Ikanda	PhD Biology	Dimensions of a Human-lion conflict: Ecology of human predation and persecution of African lions (<i>Panthera leo</i>) in Tanzania
2010	Huy Quang Nguyen	PhD Biology	Egg characteristics and development of larval digestive function of cobia (<i>Rachycentron canadum</i>) in response to dietary treatments - Focus on formulated diets
2010	Eli Kvingedal	PhD Biology	Intraspecific competition in stream salmonids: the impact of environment and phenotype
2010	Sverre Lundemo	PhD Biology	Molecular studies of genetic structuring and demography in <i>Arabidopsis</i> from Northern Europe

2010	Iddi Mihijai Mfunda	PhD Biology	Wildlife Conservation and People's livelihoods: Lessons Learnt and Considerations for Improvements. The Case of Serengeti Ecosystem, Tanzania
2010	Anton Tinchov Antonov	PhD Biology	Why do cuckoos lay strong-shelled eggs? Tests of the puncture resistance hypothesis
2010	Anders Lyngstad	PhD Biology	Population Ecology of <i>Eriophorum latifolium</i> , a Clonal Species in Rich Fen Vegetation
2010	Hilde Færevik	PhD Biology	Impact of protective clothing on thermal and cognitive responses
2010	Ingerid Brønne Arbo	PhD Medical technology	Nutritional lifestyle changes – effects of dietary carbohydrate restriction in healthy obese and overweight humans
2010	Yngvild Vindenes	PhD Biology	Stochastic modeling of finite populations with individual heterogeneity in vital parameters
2010	Hans-Richard Brattbakk	PhD Medical technology	The effect of macronutrient composition, insulin stimulation, and genetic variation on leukocyte gene expression and possible health benefits
2011	Geir Hysing Bolstad	PhD Biology	Evolution of Signals: Genetic Architecture, Natural Selection and Adaptive Accuracy
2011	Karen de Jong	PhD Biology	Operational sex ratio and reproductive behaviour in the two-spotted goby (<i>Gobiusculus flavescens</i>)
2011	Ann-Iren Kittang	PhD Biology	<i>Arabidopsis thaliana</i> L. adaptation mechanisms to microgravity through the EMCS MULTIGEN-2 experiment on the ISS: The science of space experiment integration and adaptation to simulated microgravity
2011	Aline Magdalena Lee	PhD Biology	Stochastic modeling of mating systems and their effect on population dynamics and genetics
2011	Christopher Gravningen Sørmo	PhD Biology	Rho GTPases in Plants: Structural analysis of ROP GTPases; genetic and functional studies of MIRO GTPases in <i>Arabidopsis thaliana</i>
2011	Grethe Robertsen	PhD Biology	Relative performance of salmonid phenotypes across environments and competitive intensities
2011	Line-Kristin Larsen	PhD Biology	Life-history trait dynamics in experimental populations of guppy (<i>Poecilia reticulata</i>): the role of breeding regime and captive environment
2011	Maxim A. K. Teichert	PhD Biology	Regulation in Atlantic salmon (<i>Salmo salar</i>): The interaction between habitat and density
2011	Torunn Beate Hancke	PhD Biology	Use of Pulse Amplitude Modulated (PAM) Fluorescence and Bio-optics for Assessing Microalgal Photosynthesis and Physiology

2011	Sajeda Begum	PhD Biology	Brood Parasitism in Asian Cuckoos: Different Aspects of Interactions between Cuckoos and their Hosts in Bangladesh
2011	Kari J. K. Attramadal	PhD Biology	Water treatment as an approach to increase microbial control in the culture of cold water marine larvae
2011	Camilla Kalvatn Egset	PhD Biology	The Evolvability of Static Allometry: A Case Study
2011	AHM Raihan Sarker	PhD Biology	Conflict over the conservation of the Asian elephant (<i>Elephas maximus</i>) in Bangladesh
2011	Gro Dehli Villanger	PhD Biology	Effects of complex organohalogen contaminant mixtures on thyroid hormone homeostasis in selected arctic marine mammals
2011	Kari Bjørneraas	PhD Biology	Spatiotemporal variation in resource utilisation by a large herbivore, the moose
2011	John Odden	PhD Biology	The ecology of a conflict: Eurasian lynx depredation on domestic sheep
2011	Simen Pedersen	PhD Biology	Effects of native and introduced cervids on small mammals and birds
2011	Mohsen Falahati-Anbaran	PhD Biology	Evolutionary consequences of seed banks and seed dispersal in <i>Arabidopsis</i>
2012	Jakob Hønborg Hansen	PhD Biology	Shift work in the offshore vessel fleet: circadian rhythms and cognitive performance
2012	Elin Noreen	PhD Biology	Consequences of diet quality and age on life-history traits in a small passerine bird
2012	Irja Ida Ratikainen	PhD Biology	Foraging in a variable world: adaptations to stochasticity
2012	Aleksander Handå	PhD Biology	Cultivation of mussels (<i>Mytilus edulis</i>): Feed requirements, storage and integration with salmon (<i>Salmo salar</i>) farming
2012	Morten Kraabøl	PhD Biology	Reproductive and migratory challenges inflicted on migrant brown trout (<i>Salmo trutta</i> L.) in a heavily modified river
2012	Jisca Huisman	PhD Biology	Gene flow and natural selection in Atlantic salmon
2012	Maria Bergvik	PhD Biology	Lipid and astaxanthin contents and biochemical post-harvest stability in <i>Calanus finmarchicus</i>
2012	Bjarte Bye Løfaldli	PhD Biology	Functional and morphological characterization of central olfactory neurons in the model insect <i>Heliothis virescens</i> .

2012	Karen Marie Hammer	PhD Biology	Acid-base regulation and metabolite responses in shallow- and deep-living marine invertebrates during environmental hypercapnia
2012	Øystein Nordrum Wiggen	PhD Biology	Optimal performance in the cold
2012	Robert Dominikus Fyumagwa	Dr. Philos Biology	Anthropogenic and natural influence on disease prevalence at the human –livestock-wildlife interface in the Serengeti ecosystem, Tanzania
2012	Jenny Bytingsvik	PhD Biology	Organohalogenated contaminants (OHCs) in polar bear mother-cub pairs from Svalbard, Norway. Maternal transfer, exposure assessment and thyroid hormone disruptive effects in polar bear cubs
2012	Christer Moe Rolandsen	PhD Biology	The ecological significance of space use and movement patterns of moose in a variable environment
2012	Erlend Kjeldsberg Hovland	PhD Biology	Bio-optics and Ecology in <i>Emiliania huxleyi</i> Blooms: Field and Remote Sensing Studies in Norwegian Waters
2012	Lise Cats Myhre	PhD Biology	Effects of the social and physical environment on mating behaviour in a marine fish
2012	Tonje Aronsen	PhD Biology	Demographic, environmental and evolutionary aspects of sexual selection
2012	Bin Liu	PhD Biology	Molecular genetic investigation of cell separation and cell death regulation in <i>Arabidopsis thaliana</i>
2013	Jørgen Rosvold	PhD Biology	Ungulates in a dynamic and increasingly human dominated landscape – A millennia-scale perspective
2013	Pankaj Barah	PhD Biology	Integrated Systems Approaches to Study Plant Stress Responses
2013	Marit Linnerud	PhD Biology	Patterns in spatial and temporal variation in population abundances of vertebrates
2013	Xinxin Wang	PhD Biology	Integrated multi-trophic aquaculture driven by nutrient wastes released from Atlantic salmon (<i>Salmo salar</i>) farming
2013	Ingrid Ertshus Mathisen	PhD Biology	Structure, dynamics, and regeneration capacity at the sub-arctic forest-tundra ecotone of northern Norway and Kola Peninsula, NW Russia
2013	Anders Foldvik	PhD Biology	Spatial distributions and productivity in salmonid populations
2013	Anna Marie Holand	PhD Biology	Statistical methods for estimating intra- and inter-population variation in genetic diversity
2013	Anna Solvang Båtnes	PhD Biology	Light in the dark – the role of irradiance in the high Arctic marine ecosystem during polar night

2013	Sebastian Wacker	PhD Biology	The dynamics of sexual selection: effects of OSR, density and resource competition in a fish
2013	Cecilie Miljeteig	PhD Biology	Phototaxis in <i>Calanus finmarchicus</i> – light sensitivity and the influence of energy reserves and oil exposure
2013	Ane Kjersti Vie	PhD Biology	Molecular and functional characterisation of the IDA family of signalling peptides in <i>Arabidopsis thaliana</i>
2013	Marianne Nymark	PhD Biology	Light responses in the marine diatom <i>Phaeodactylum tricorutum</i>
2014	Jannik Schultner	PhD Biology	Resource Allocation under Stress - Mechanisms and Strategies in a Long-Lived Bird
2014	Craig Ryan Jackson	PhD Biology	Factors influencing African wild dog (<i>Lycaon pictus</i>) habitat selection and ranging behaviour: conservation and management implications
2014	Aravind Venkatesan	PhD Biology	Application of Semantic Web Technology to establish knowledge management and discovery in the Life Sciences
2014	Kristin Collier Valle	PhD Biology	Photoacclimation mechanisms and light responses in marine micro- and macroalgae
2014	Michael Puffer	PhD Biology	Effects of rapidly fluctuating water levels on juvenile Atlantic salmon (<i>Salmo salar</i> L.)
2014	Gundula S. Bartzke	PhD Biology	Effects of power lines on moose (<i>Alces alces</i>) habitat selection, movements and feeding activity
2014	Eirin Marie Bjørkvoll	PhD Biology	Life-history variation and stochastic population dynamics in vertebrates
2014	Håkon Holand	PhD Biology	The parasite <i>Syngamus trachea</i> in a metapopulation of house sparrows
2014	Randi Magnus Sommerfelt	PhD Biology	Molecular mechanisms of inflammation – a central role for cytosolic phospholipase A2
2014	Espen Lie Dahl	PhD Biology	Population demographics in white-tailed eagle at an on-shore wind farm area in coastal Norway
2014	Anders Øverby	PhD Biology	Functional analysis of the action of plant isothiocyanates: cellular mechanisms and in vivo role in plants, and anticancer activity
2014	Kamal Prasad Acharya	PhD Biology	Invasive species: Genetics, characteristics and trait variation along a latitudinal gradient.
2014	Ida Beathe Øverjordet	PhD Biology	Element accumulation and oxidative stress variables in Arctic pelagic food chains: <i>Calanus</i> , little auks (<i>Alle alle</i>) and black-legged kittiwakes (<i>Rissa tridactyla</i>)

2014	Kristin Møller Gabrielsen	PhD Biology	Target tissue toxicity of the thyroid hormone system in two species of arctic mammals carrying high loads of organohalogen contaminants
2015	Gine Roll Skjervø	Dr. philos Biology	Testing behavioral ecology models with historical individual-based human demographic data from Norway
2015	Nils Erik Gustaf Forsberg	PhD Biology	Spatial and Temporal Genetic Structure in Landrace Cereals
2015	Leila Alipanah	PhD Biology	Integrated analyses of nitrogen and phosphorus deprivation in the diatoms <i>Phaeodactylum tricoratum</i> and <i>Seminavis robusta</i>
2015	Javad Najafi	PhD Biology	Molecular investigation of signaling components in sugar sensing and defense in <i>Arabidopsis thaliana</i>
2015	Bjørnar Sporsheim	PhD Biology	Quantitative confocal laser scanning microscopy: optimization of in vivo and in vitro analysis of intracellular transport
2015	Magni Olsen Kyrkjæide	PhD Biology	Genetic variation and structure in peatmosses (<i>Sphagnum</i>)
2015	Keshuai Li	PhD Biology	Phospholipids in Atlantic cod (<i>Gadus morhua</i> L.) larvae rearing: Incorporation of DHA in live feed and larval phospholipids and the metabolic capabilities of larvae for the de novo synthesis
2015	Ingvild Fladvad Størdal	PhD Biology	The role of the copepod <i>Calanus finmarchicus</i> in affecting the fate of marine oil spills
2016	Thomas Kvalnes	PhD Biology	Evolution by natural selection in age-structured populations in fluctuating environments
2016	Øystein Leiknes	PhD Biology	The effect of nutrition on important life-history traits in the marine copepod <i>Calanus finmarchicus</i>
2016	Johan Henrik Hårdensson Berntsen	PhD Biology	Individual variation in survival: The effect of incubation temperature on the rate of physiological ageing in a small passerine bird
2016	Marianne Opsahl Olufsen	PhD Biology	Multiple environmental stressors: Biological interactions between parameters of climate change and perfluorinated alkyl substances in fish
2016	Rebekka Varne	PhD Biology	Tracing the fate of escaped cod (<i>Gadus morhua</i> L.) in a Norwegian fjord system
2016	Anette Antonsen Fenstad	PhD Biology	Pollutant Levels, Antioxidants and Potential Genotoxic Effects in Incubating Female Common Eiders (<i>Somateria mollissima</i>)
2016	Wilfred Njama Marealle	PhD Biology	Ecology, Behaviour and Conservation Status of Masai Giraffe (<i>Giraffa camelopardalis tippelskirchi</i>) in Tanzania

2016	Ingunn Nilssen	PhD Biology	Integrated Environmental Mapping and Monitoring: A Methodological approach for end users.
2017	Konika Chawla	PhD Biology	Discovering, analysing and taking care of knowledge.
2017	Øystein Hjorthol Opedal	PhD Biology	The Evolution of Herkogamy: Pollinator Reliability, Natural Selection, and Trait Evolvability.
2017	Ane Marlene Myhre	PhD Biology	Effective size of density dependent populations in fluctuating environments
2017	Emmanuel Hosiana Masenga	PhD Biology	Behavioural Ecology of Free-ranging and Reintroduced African Wild Dog (<i>Lycaon pictus</i>) Packs in the Serengeti Ecosystem, Tanzania
2017	Xiaolong Lin	PhD Biology	Systematics and evolutionary history of <i>Tanytarsus</i> van der Wulp, 1874 (Diptera: Chironomidae)
2017	Emmanuel Clamsen Mmassy	PhD Biology	Ecology and Conservation Challenges of the Kori bustard in the Serengeti National Park
2017	Richard Daniel Ly- amuya	PhD Biology	Depredation of Livestock by Wild Carnivores in the Eastern Serengeti Ecosystem, Tanzania
2017	Katrin Hoydal	PhD Biology	Levels and endocrine disruptive effects of legacy POPs and their metabolites in long-finned pilot whales of the Faroe Islands
2017	Berit Glomstad	PhD Biology	Adsorption of phenanthrene to carbon nanotubes and its influence on phenanthrene bioavailability/toxicity in aquatic organism
2017	Øystein Nordeide Kielland	PhD Biology	Sources of variation in metabolism of an aquatic ectotherm
2017	Narjes Yousefi	PhD Biology	Genetic divergence and speciation in northern peat-mosses (<i>Sphagnum</i>)
2018	Signe Christensen- Dalgaard	PhD Biology	Drivers of seabird spatial ecology - implications for development of offshore wind-power in Norway
2018	Janos Urbancsok	PhD Biology	Endogenous biological effects induced by externally supplemented glucosinolate hydrolysis products (GHPs) on <i>Arabidopsis thaliana</i>
2018	Alice Mühlroth	PhD Biology	The influence of phosphate depletion on lipid metabolism of microalgae
2018	Franco Peniel Mbise	PhD Biology	Human-Carnivore Coexistence and Conflict in the Eastern Serengeti, Tanzania
2018	Stine Svalheim Markussen	PhD Biology	Causes and consequences of intersexual life history variation in a harvested herbivore population

2018	Mia Vedel Sørensen	PhD Biology	Carbon budget consequences of deciduous shrub expansion in alpine tundra ecosystems
2018	Hanna Maria Kauko	PhD Biology	Light response and acclimation of microalgae in a changing Arctic
2018	Erlend I. F. Fossen	PhD Biology	Trait evolvability: effects of thermal plasticity and genetic correlations among traits
2019	Peter Sjolte Ranke	PhD Biology	Demographic and genetic consequences of dispersal in house sparrows
2019	Mathilde Le Moullec	PhD Biology	Spatiotemporal variation in abundance of key tundra species: from local heterogeneity to large-scale synchrony
2019	Endre Grüner Ofstad	PhD Biology	Causes and consequences of variation in resource use and social structure in ungulates
2019	Yang Jin	PhD Biology	Development of lipid metabolism in early life stage of Atlantic salmon (<i>Salmo salar</i>)
2019	Elena Albertsen	PhD Biology	Evolution of floral traits: from ecological context to functional integration
2019	Mominul Islam Nahid	PhD Biology	Interaction between two Asian cuckoos and their hosts in Bangladesh
2019	Knut Jørgen Egelie	Phd Biology	Management of intellectual property in university-industry collaborations – public access to and control of knowledge
2019	Thomas Ray Haaland	Phd Biology	Adaptive responses to environmental stochasticity on different evolutionary time-scales
2019	Kwaslema Malle Hariohay	Phd Biology	Human wildlife interactions in the Ruaha-Rungwa Ecosystem, Central Tanzania
2019	Mari Engvig Løseth	Phd Biology	Exposure and effects of emerging and legacy organic pollutants in white-tailed eagle (<i>Haliaeetus albicilla</i>) nestlings
2019	Joseph Mbyati Mukeka	Phd Biology	Human-Wildlife Conflicts and Compensation for Losses in Kenya: Dynamics, Characteristics and Correlates
2019	Helene Løvstrand Svarva	Phd Biology	Dendroclimatology in southern Norway: tree rings, demography and climate
2019	Nathalie Briels	Phd Biology	Exposure and effects of legacy and emerging organic pollutants in developing birds – Laboratory and field studies
2019	Anders L.Kolstad	Phd Biology	Moose browsing effects on boreal production forests – implications for ecosystems and human society

2019	Bart Peeters	Phd Biology	Population dynamics under climate change and harvesting: Results from the high Arctic Svalbard reindeer
2019	Alex Kojo Datsomor	Phd Biology	The molecular basis of long chain polyunsaturated fatty acid (LC-PUFA) biosynthesis in Atlantic salmon (<i>Salmo salar L.</i>): In vivo functions, functional redundancy and transcriptional regulation of LC-PUFA biosynthetic enzymes
2020	Ingun Næve	Phd Biology	Development of non-invasive methods using ultrasound technology in monitoring of Atlantic salmon (<i>Salmo salar</i>) production and reproduction
2020	Rachael Morgan	Phd Biology	Physiological plasticity and evolution of thermal performance in zebrafish
2020	Mahsa Jalili	Phd Biology	Effects of different dietary ingredients on the immune responses and antioxidant status in Atlantic salmon (<i>Salmo salar L.</i>): possible nutrionomics approaches
2020	Haiqing Wang	Phd Biology	Utilization of the polychaete <i>Hediste diversicolor</i> (O.F. Millier, 1776) in recycling waste nutrients from land-based fish farms for value adding applications'
2020	Louis Hunninck	Phd Biology	Physiological and behavioral adaptations of impala to anthropogenic disturbances in the Serengeti ecosystems
2020	Kate Layton-Matthews	Phd Biology	Demographic consequences of rapid climate change and density dependence in migratory Arctic geese
2020	Amit Kumar Sharma	Phd Biology	Genome editing of marine algae: Technology development and use of the CRISPR/Cas9 system for studies of light harvesting complexes and regulation of phosphate homeostasis
2020	Lars Rød-Eriksen	Phd Biology	Drivers of change in meso-carnivore distributions in a northern ecosystem
2020	Lone Sunniva Jevne	Phd Biology	Development and dispersal of salmon lice (<i>Lepeophtheirus salmonis</i> Krøyer, 1837) in commercial salmon farming localities
2020	Sindre Håvarstein Eldøy	Phd Biology	The influence of physiology, life history and environmental conditions on the marine migration patterns of sea trout
2020	Vasundra Touré	Phd Biology	Improving the FAIRness of causal interactions in systems biology: data curation and standardisation to support systems modelling applications
2020	Silje Forbord	Phd Biology	Cultivation of <i>Saccharina latissima</i> (Phaeophyceae) in temperate marine waters; nitrogen uptake kinetics, growth characteristics and chemical composition

2020	Jørn Olav Løkken	Phd Biology	Change in vegetation composition and growth in the forest-tundra ecotone – effects of climate warming and herbivory
2020	Kristin Odden Ny-stuen	Phd Biology	Drivers of plant recruitment in alpine vegetation
2021	Sam Perrin	Phd Biology	Freshwater Fish Community Responses to Climate Change and Invasive Species
2021	Lara Veylit	Phd Biology	Causes and consequences of body growth variation in hunted wild boar populations
2021	Semona Issa	Phd Biology	Combined effects of environmental variation and pollution on zooplankton life history and population dynamics
2021	Monica Shilereyo	Phd Biology	Small Mammal Population Ecology and Ectoparasite Load: Assessing Impacts of Land Use and Rainfall Seasonality in the Serengeti Ecosystem, Tanzania
2021	Vanessa Bieker	Phd Biology	Using historical herbarium specimens to elucidate the evolutionary genomics of plant invasion
2021	Håkon Austad Langberg	Phd Biology	Fate and transport of forever chemicals in the aquatic environment: Partitioning and biotransformation of mixtures of Per- and Polyfluoroalkyl Substances (PFAS) from different point sources and resulting concentrations in biota

ISBN 978-82-326-5218-1 (printed ver.)
ISBN 978-82-326-5431-4 (electronic ver.)
ISSN 1503-8181 (printed ver.)
ISSN 2703-8084 (online ver.)



NTNU

Norwegian University of
Science and Technology