ARTIFICIAL INTELLIGENCE AND GENDER EQUALITY: A SYSTEMATIC MAPPING STUDY

J. David Patón-Romero

Department of Computer Science, Norwegian University of Science and Technology (NTNU) Sem Sælands Vei 7, 7034 Trondheim, Norway

> Ricardo Vinuesa FLOW, Engineering Mechanics, KTH Royal Institute of Technology SE-100 44 Stockholm, Sweden

Letizia Jaccheri Department of Computer Science, Norwegian University of Science and Technology (NTNU) Sem Sælands Vei 7, 7034 Trondheim, Norway

> Maria Teresa Baldassarre Department of Informatics, University of Bari "Aldo Moro" (UniBa) Via Edoardo Orabona, 4, 70126 Bari, Italy

ABSTRACT

Sustainability is not only understood as a manner to safeguard the environment, but also to fight against injustices and inequalities that exist on social and economic level. One of the biggest challenges that exists in social sustainability is to achieve gender equality, as defended by the Sustainable Development Goal (SDG) 5 of the 2030 Agenda. But this is a complex challenge and must be addressed from different spheres and fields of knowledge. Artificial Intelligence (AI) has proven to be an essential asset in the development of new and innovative technologies. Its development, adoption, and constant use by a growing part of the world's population demonstrates the social impact it entails and the importance of also becoming an asset for social sustainability and, in this case especially, for gender equality.

That is why this study aims to collect the current knowledge about the fields of AI and gender equality, through the development of a Systematic Mapping Study (SMS) that identifies the most significant advances in this regard, as well as the main gaps that must be covered. The results and findings obtained in this work show the novelty of joint analysis of both areas, as well as increasing attention they have received in recent years. Likewise, they also demonstrate the need to address specific and urgent issues within gender equality, both in the field of AI and caused by its development.

KEYWORDS

Artificial Intelligence; Gender Equality; Social Sustainability; Systematic Mapping Study.

1. INTRODUCTION

Sustainability has become a key actor for the development and advancement of civilization. However, many times it is only interpreted as an environmental characteristic and other perspectives such as social and economic sustainability are ignored (Purvis et al., 2019). These three perspectives go hand in hand and must be addressed together, since, for example, it is not possible to aim to achieve sustainable development by and for the environment if changes are not conducted in society itself to support it.

According to the findings identified by (Harari, 2018), sustainability and Artificial Intelligence (AI) are two of the biggest challenges faced by humanity. From Information Technology (IT), AI has become one of the most relevant and innovative fields (Carleton et al., 2020; Menzies, 2019). The unstoppable progress that AI is experiencing demonstrates the importance of pursuing AI applications that can help to achieve a sustainable development and use in this regard (Nishant et al., 2020; Vinuesa et al., 2020). Thus, it is essential to relate both fields and work to achieve sustainability in and by AI. However, the focus has always been on the relationship between AI and technology in general with energy/climate neutrality (i.e., environmental sustainability) and marginally on its relationship with social and economic sustainability (Becker, 2015). Some of these aspects are discussed by (Vinuesa & Sirmacek, 2021).

Thus, this study aims to shed light into the issue and show the current relationship between AI and social sustainability, focusing on gender equality, which is one of the Sustainable Development Goals (SDGs) established by the 2030 Agenda (United Nation, 2015) which is having more focus and relevance regarding social sustainability (Rosa, 2017).

Current research shows that women are underrepresented in technology research, practice, and education (Albusays et al., 2021). Likewise, gender imbalance in technology has been seen as harming the economy, as highlighted by the *European Commission* when identifying an annual productivity loss of 16 billion Euro for the European economy (European Commission, 2018). In the same way, the OECD (*Organization for Economic Co-operation and Development*) states that "greater inclusion of women in the digital economy and increased diversity bring value, both social and economic" (OECD, 2018).

Therefore, the present literature analysis through a Systematic Mapping Study in the areas of AI and gender equality will be useful, since it will allow knowing the latest knowledge and establishing the pillars that will guide the development of new and innovative research and ideas in this regard. Thanks to a greater understanding of the interplay between AI and gender equality, it will it be possible to understand the changes and challenges that exist towards achieving a sustainable development through the SDGs.

The rest of this study is organized as follows: Section 2 contains the background about 2030 Agenda, gender equality, and AI; Section 3 presents the research methodology followed to analyze the state of the art in the fields of gender equality and AI; Section 4 shows the results obtained from the analysis performed; Section 5 discusses the findings, limitations, and implications that have been reached; finally, Section 6 contains the conclusions and lines for future work in this regard. Likewise, Appendix A includes the list of primary studies selected during the analysis of the state of the art; and Appendix B shows the answers to the established research questions from each of these primary studies.

2. BACKGROUND

2.1. 2030 Agenda & Sustainable Development Goal 5

The 2030 Agenda (United Nation, 2015) is an initiative promoted and agreed upon by the 193 Member States of the United Nations (UN) with the aim to achieve the so-called Sustainable Development Goals (SDGs). This includes a total of 17 Goals and 169 Targets that address the three pillars of sustainability (environmental, social, and economic) (Purvis et al., 2019), including areas such as climate change, economic inequality, innovation, natural resources consumption, peace, and justice, among other priorities. Likewise, for each of the Targets there is also a set of indicators that make it possible to measure the progress made in this regard (United Nations, 2017).

Among these SDGs, this study aims to focus on Goal 5 (Gender Equality), one of the Goals belonging to the field of social sustainability. The main purpose of this Goal is to *"achieve gender equality and empower all women and girls"*, for which it establishes 9 Targets (United Nation, 2015).

It is important to highlight that the 2030 Agenda identifies two types of Targets within the SDGs: 1) "outcome" Targets (i.e., circumstances to be attained), labeled by numbers; and 2) "means of implementation" Targets, labeled by lower case letters.

2.2. Gender Equality & Artificial Intelligence

In order to achieve gender equality, as defended by SDG 5 of the 2030 Agenda, actions must be conducted in all areas of knowledge. AI has proven to be an increasingly important actor in the development of new and innovative systems used by all levels of society (Lu et al., 2018). That is why it is vital that the entire life cycle of these systems is committed to achieving a better society and, therefore, gender equality must play an important role in this regard.

In general terms, it could be said that the main objective of AI in social sustainability is "the study and practice of design, build and use of AI systems with a positive impact on the society". However, when it comes to relating the terms of gender equality and AI, there is no clear criterion or definitions per se. To establish this relationship, the definitions and same logic as that used for the terms *Green by IT* and *Green in IT* (idea proposed in (Erdélyi, 2013)) will be followed, which defend sustainability in and by IT. Thus, in gender equality and AI we are faced with two perspectives:

- *Gender by AI*: in which AI provides the necessary tools to achieve gender equality through different contexts (i.e., AI as an enabler).
- *Gender in AI*: in which AI itself produces a negative impact on gender equality (e.g., lack of balance during the development of a system) and, therefore, said impact must be reduced (i.e., AI as a producer).

3. RESEARCH METHODOLOGY

A Systematic Mapping Study (SMS) is a research method used to collect, analyze, and categorize existing information from a specific context. In the specific case of this study, the guidelines established by (Kitchenham, 2007) have been followed, adopting also the lessons learned for the data extraction and analysis identified by (Brereton et al., 2007), and considering examples of application of SMSs in Software Engineering such as (Petersen et al., 2008). Thus, the characteristics established during the planning stage are shown below, as well as how the execution stage was conducted.

3.1. Planning Stage

3.1.1.Research Questions

The main goal of this study is to inspect the current state and existing relationship between the fields of artificial intelligence and gender equality. In this way, it is intended to collect and categorize all the information in this regard and identify the gaps that exist in order to develop new research proposals. To do this, the research questions (RQs) shown in Table 1 have been established.

Table 1. Research questions

Research question	Motivation
RQ1. What kind of studies exist on AI and gender equality?	Determine the type, number of publications, and trend over
	recent years in relation to AI and gender equality.
RQ2. What gender equality Targets are addressed in and by	Determine what gender equality Targets are addressed in/by
AI?	AI to identify possible opportunities and threats.
RQ3. What kind of AI proposals exist to address gender	Determine the AI proposals that exist to address gender
equality?	equality to identify trends and possible gaps in or by AI.

3.1.2.Search Strategy

As a strategy for search the relevant studies and information, the *Scopus* database will be used. To this end, we decided to conduct a general search and a search for each of the Targets identified by the SDG 5 of the 2030 Agenda (United Nation, 2015) (i.e., 10 searches). In this way, specific terms of each Target can be addressed in more detail and the identification of studies in this regard is facilitated. Thus, Table 2 shows the search strings that will be used. As can be seen, these search strings are divided into two main parts (the two contexts within the scope of this study).

Table 2. Searc	ch strings
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Scope	Search string
General	("Artificial Intelligence" OR AI) AND (Gender OR "Women rights" OR "Social sustainability" OR "SDG 5")
Target 5.1	("Artificial Intelligence" OR AI) AND ((Women OR Girls OR Gender) AND Discrimination)
Target 5.2	("Artificial Intelligence" OR AI) AND ((Women OR Girls OR Gender) AND (Violence OR Exploitation OR
	Trafficking))
Target 5.3	("Artificial Intelligence" OR AI) AND (((Women OR Girls OR Gender) AND "Harmful practices") OR ((Child
	OR Early OR Forced) AND Marriage) OR "Genital mutilation")

Scope	Search string
Target 5.4	("Artificial Intelligence" OR AI) AND ("Care work" OR "Domestic work" OR "Social protection policies" OR
	"Shared responsibility")
Target 5.5	("Artificial Intelligence" OR AI) AND ((Women OR Girls OR Gender) AND ("Equal opportunities" OR
	Participation OR Leadership))
Target 5.6	("Artificial Intelligence" OR AI) AND ((Sexual OR Reproductive) AND (Health OR Rights))
Target 5.a	("Artificial Intelligence" OR AI) AND ((Women OR Girls OR Gender) AND Equal* AND Rights)
Target 5.b	("Artificial Intelligence" OR AI) AND ((Women OR Girls) AND Technology)
Target 5.c	("Artificial Intelligence" OR AI) AND ((Women OR Girls OR Gender) AND (Equal* OR Empower*))

These search strings will be applied to the title, abstract and keywords of the studies. Likewise, publications from 2010 and onwards will be considered, since it has been during the last decade when, mainly, the area of gender equality has had its momentum.

3.1.3. Selection Criteria

All the documents and information collected through the searches will be analyzed considering the title, abstract, and keywords of each one. This will determine which studies will be included for a more detailed analysis. To do this, on the one hand, those studies that meet the following inclusion criteria will be considered for further analysis:

- **I1.** Studies in English dealing with AI and gender equality.
- **I2.** Studies published between 2010 and 2021 in journals, conferences, and/or workshops, with peer review process.

On the other hand, the studies that meet any of the following exclusion criteria will be automatically discarded:

- E1. Discussion or opinion studies, as well as those that are only available as abstract or presentation.
- E2. Duplicate studies (in which case will be considered the most complete and recent).
- E3. Studies whose main contribution is not related to AI and gender equality, or where AI and gender equality are not related to each other.

In the same way, the snowballing effect (Wohlin, 2014) will be followed, so the documents referenced in the considered studies will also be evaluated for their possible inclusion.

3.1.4. Quality Assessment Criteria

One of the most critical points to obtain representative and relevant results and references for future research is the quality assessment of the studies. To do this, the following issues have been established that will be analyzed following a scoring system of three values (-1, 0, +1), generating a quality result for each study between -4 and +4:

- a. The study presents a detailed description and guidance on how AI can contribute to gender equality. *Yes* (+1); *Partially* (0); *No* (-1).
- b. The study validates the proposal or idea that it defends. *Empirically validated* (+1); *Theoretically validated* (0); *Not validated* (-1).
- c. The study has been published in a relevant journal¹/conference². *High ranking* (+1); *Medium ranking* (0); *Low ranking or not indexed* (-1).
- d. The study has been cited by other authors in publications. More than five cites (+1); Between one and four cites or recently published in 2021 (0); Not cited (-1).

3.1.5.Data Extraction

A series of answers have been established for each of the research questions (as shown in Table 3). In this way, the same data extraction criteria will be applied to all studies, facilitating their analysis and categorization.

Table 3. Classification schema

Research question	Answers						
RQ1. What kind of studies exist on AI and gender equality?*	a. State of the art analysis	c. Validation					
	b. Proposal	d. Others					

¹ Following the Journal Citation Reports (JCR): <u>https://jcr.clarivate.com/</u>

² Following the GII-GRIN-SCIE Conference Rating: <u>https://scie.lcc.uma.es/</u>

Research question	Answers		
RQ2. What gender equality Targets are addressed in and by	a. Target 5.1	d. Target 5.4	g. Target 5.a
AI?**	b. Target 5.2	e. Target 5.5	h. Target 5.b
	c. Target 5.3	f. Target 5.6	i. Target 5.c
RQ3. What kind of AI proposals exist to address gender	a. Gender by AI		
equality?	b. Gender in AI		

*The answers to RQ1 follow the idea of the example of (Petersen et al., 2008).

**The answers to RQ2 have their origin in the Targets of the SDG 5 from the 2030 Agenda (United Nation, 2015).

3.1.6.Synthesis Methods

A both quantitative and qualitative synthesis of data will be conducted related to the answers to the research questions and the quality evaluations performed, respectively. These syntheses will be represented by tables and/or graphs with the results in a matter of numbers and/or percentages, as well as bubble plots to analyze how the research questions are related through their answers.

3.2. Execution Stage

In order to apply the protocol established during the planning stage, three main phases have been followed during the execution stage:

- **First phase.** Based on the identification of potential studies. To do this, first, after performing the 10 searches applying the search strings (cf. Table 2) on the *Scopus* database, 3,558 studies were obtained. Then, the selection criteria were applied to these studies, considering the abstract of each one, and 169 potential studies were obtained.
- Second phase. Oriented to the identification of primary studies, by means of which the selection criteria were applied again, but this time on the complete content of each of the 169 potential studies. As a result, 29 primary studies were obtained.
- **Third phase.** This last phase is dedicated to the compilation of results, for which the characterization of the primary studies was performed through the answers to the research questions and the obtaining of the main findings, as well as the quality assessment of said studies.

4. **RESULTS**

The general results obtained after the execution of the SMS are shown below, answering each of the established research questions. It is important to highlight that, related to these results, Appendix A includes the list of references of primary studies, while Appendix B contains a summary table with the mapping of the answers to the research questions of each of these primary studies.

4.1. RQ1. What Kind of Studies Exist on AI and Gender Equality?

The main objective of this RQ is based on identifying the type of studies that currently exist in AI and gender equality. In this way, it is possible to determine which are the most relevant studies when, e.g., conducting a new proposal in this regard or considering a specific proposal that is validated for the application of a case.

Our results (represented in Figure 1) show that about 24% of the studies found (S05, S08, S09, S11, S17, S18, and S21) are based on or contain some analysis of the state of the art on the field that concerns us.

Likewise, 69% of the studies (20 in total) deal with specific proposals to address gender equality from some point in and by AI. However, of all these proposals only 11 are validated by some empirical case (S01, S02, S04, S10, S13, S16, S22, S23, S27, S28, and S29).

It is also important to highlight 2 studies (S03 and S06), which are based on experiments aimed at assessing the impact of gender in and by AI.

Finally, related to this research question, it is also important to analyze the evolution of the studies over the last few years. Figure 2 shows how this progression has been, through which it can be seen that mainly in the last 2 years there has been a boom in publications in this regard. This is because it is in recent years when more efforts are beginning to be made to achieve the objectives of the 2030 Agenda (United Nation, 2015).



Figure 1. Results for the RQ1 (percentage of studies in each of the four categories)



4.2. RQ2. What Gender Equality Targets are Addressed in and by AI?

This research question is the main one of the present study, since its objective is to analyze and map the studies according to the Target(s) of the SDG 5 from the 2030 Agenda (United Nation, 2015) that address. In this way, it is possible to identify which are the Targets that are usually dealt with in and by AI, as well as the possible Targets that are not yet covered and need development in this regard.

From the results obtained (represented in Figure 3), there are only studies that address 5 of the 9 Targets established. Targets 5.3, 5.4, 5.5, and 5.a are not covered in any of the studies, so there is no evidence on possible developments in AI that address gender equality in the specific contexts of these Targets (it will be discussed later in detail).

Regarding the Targets that are covered, Target 5.1 is the one with the most development in the AI area, since 52% of the studies (15 in total) base their objective on addressing the context of this one. Likewise, following a decreasing order, Target 5.2 is found in 34% of the studies (10), Target 5.6 in 17% (5), Target 5.b in 17% (5), and Target 5.c in 7% (2).



Figure 3. Results for the RQ2 (percentage of studies addressing each of the Targets within SDG 5)

4.3. RQ3. What Kind of AI Proposals Exist to Address Gender Equality?

The last of the research questions is focused on knowing the studies that are oriented both to achieve gender equality in different areas through the use of AI (*Gender by AI*) and to implement measures that help achieve better gender equality issues in AI itself (*Gender in AI*). In this regard, the results show equality in both perspectives, since 18 studies deal with *Gender by AI* and 13 with *Gender in AI*.

First, the studies S01, S02, S03, S04, S05, S06, S07, S10, S12, S13, S14, S15, S16, S23, S24, S25, S26, and S29 include some proposal or characteristics related to the context of *Gender by AI*. Second, *Gender in AI* is addressed in the studies S03, S05, S08, S09, S11, S17, S18, S19, S20, S21, S22, S27, and S28. And, finally, it should be noted that the studies S03 and S05 consider both perspectives.

Thus, 62% of the studies are framed in the perspective of Green by AI, while 45% deal with Green in AI.

5. DISCUSSION

5.1. Principal Findings

The main objective of this study is based on knowing the state of the art and the current relationship between gender equality and AI. In this way, it is intended to identify important aspects and gaps that help to develop new innovative ideas in this regard. After analyzing the results, the following observations can be made:

- Focus on *Gender by AI*. Although there is a high level of equality in the number of studies that deal with *Gender by AI* and *Gender in AI*, in recent years the main focus has been the development of studies related to *Gender by AI*. This could be due to the greater ease of developing a specific technology to address an aspect related to gender equality (such as, e.g., an AI device that detects dangerous situations for women) than, e.g., change the business/management processes that guide organizations when developing new AI proposals so that they follow a set of best practices that respect gender equality. Undoubtedly, the latter is more complicated, because it is not only necessary to understand both fields to develop useful and applicable best practices for most contexts, but also a high number of practical cases and the involvement of external actors who allow validation of these practices are needed. Therefore, the fact that the focus is currently on *Gender by AI* is an issue that can generate a lot of controversy, since the "cart is being put before the horse" and the question arises as to whether AI proposals that help gender equality can actually be developed, when in AI itself and in all the processes that surround it (i.e., the basis) there is no such equality.
- Inequality when dealing with the Targets. It is a normal result that there is a difference in the number of studies when addressing different contexts, but, in this case, the difference is quite large. To improve understanding, we can talk about three groups:
 - Advanced development: the Target 5.1 is the only one in this group, since it is the most addressed by the studies found. This is because it is a fairly general Target whose objective is *"end all forms* of discrimination against all women and girls everywhere" (United Nation, 2015). In general terms, the main purpose of SDG 5 of the 2030 Agenda is the one that defines this Target and that is why most studies tend to focus on it. However, we must not forget that there are other Targets with more specific purposes and that they must be addressed urgently.
 - Medium development: the Targets 5.2, 5.6, 5.b, and 5.c are found in this group. It is always relevant to find evidence that supports, in this case, the specific context of each of the Targets. However, the evidence is quite scarce, and it is necessary to continue developing new ideas, as well as improving the current ones. From a practical point of view, following the evidence found, these Targets can be addressed in a simple way in and by AI. For example, Target 5.2 aims to "eliminate all forms of violence against all women and girls in the public and private spheres, including trafficking and sexual and other types of exploitation" (United Nation, 2015), which has been shown to be easily addressed through specific AI proposals that analyze various situations to identify potential dangers affecting the integrity of women and girls. We are on the right path, but we must not get lost and continue dedicating efforts to develop proposals in and by AI in these Targets.
 - Null development: the Targets 5.3, 5.4, 5.5, and 5.a are not addressed by any study. This may be due to the difficulty of identifying an idea in and/or by the AI that effectively and efficiently addresses the specific contexts of these Targets. For example, regarding the Target 5.3 ("eliminate all harmful practices, such as child, early and forced marriage and female genital mutilation" (United Nation, 2015)), it is difficult to understand that AI can do something about it, since it deals with behaviors rooted in certain cultures and the application field of AI may not be sufficient to adequately address such a cultural change. However, the Target 5.a ("undertake reforms to give women equal rights to economic resources, as well as access to ownership and control over land and other forms of property, financial services, inheritance and natural resources, in accordance with national laws" (United Nation, 2015)) can be applied in the field of AI through, e.g., systems that analyze the profiles of certain candidates for obtaining economic resources of different kinds, without any type of bias related to the gender of each person. For this reason, all the Targets of the SDG 5 from the 2030 Agenda are equally important, but it is necessary and transcendental to guide and begin to dedicate efforts to develop these Targets that have not yet been explored, in order to propose ideas in and by AI that allow progress in gender equality in this regard.

- Low number of practical cases. When developing a proposal, it is essential to conduct practical cases that validate it and demonstrate its applicability, quality, effectiveness, and efficiency when addressing the objective for which it was developed. However, of the 20 proposals identified through the SMS, only 11 (i.e., 55%) have been validated. This supposes a too low number of validated proposals, since all or the vast majority should have been applied in some practical case, showing that they are complete and serious proposals. For this reason, it is extremely important that any development conducted in and/or by AI to, in this case, address some Target(s) of SDG 5 from the 2030 Agenda, be accompanied by a practical application and validation.
- Lack of joint development of both fields. Although there is evidence on the development of ideas in and by AI to address the specific context of gender equality, there is very little. This, together with the analysis of the progression of publications over the last few years, demonstrates the novelty of this field and the growing interest in conducting new research in this regard. It is very important to closely follow this progression, as well as analyze the new studies that arise and develop new ideas that contribute to this research field so important and necessary to achieve gender equality, improving the field of social sustainability and complying with the provisions of the 2030 Agenda.

5.2. Limitations

Although an attempt has been made to design this study to avoid or mitigate the possible limitations (such as the case of performing a general search and 9 searches for each of the Targets, with the aim to find studies with very specific terminology for certain contexts), there are always limitations that can affect when identifying and analyzing results and findings.

It should not be forgotten that the present analysis comes from the perspective of the authors and may not be interpreted in the same way by other researchers or professionals in the field. Likewise, it is possible that certain literature on the field has been overlooked, or even that some more recent evidence on the studies found has not yet been published at the time of the SMS execution. Consequently, to mitigate the risks in this regard, several authors analyzed and interpreted the data and results obtained here, contributed to the final consensus, thus reducing the bias among each other.

5.3. Implications

The development of this study has a high implication and significance for the fields in which it is found. As has been shown, there are few studies that put the fields of AI and gender equality in common. Thus, thanks to this study, not only the state of the art in this regard is identified, but also the gaps and possible lines of research that improve existing studies or that address new and innovative ideas not considered until now.

In section "5.1 Principal Findings" a discussion has been conducted in which different lines of future work/research are identified. These can be used by researchers who are in the fields of AI, IT in general, gender equality and social sustainability. Therefore, this study is a necessary starting point and the demonstration of the importance of the fields that concern us, which will attract new researchers and professionals to the development of new proposals with the goal of achieving gender equality in and by AI.

6. CONCLUSIONS AND FUTURE WORK

The increasing use of technology and AI by a wide range of people around the world shows that they must be driven by and for the whole of society, avoiding gender, culture, religion, and other kinds of discrimination. However, women and other vulnerable and discriminated minorities are underrepresented in this regard and the progress to get around this situation is slow and scant (Adams & Khomh, 2020; Albusays et al., 2021).

That is why this study is focused on analyzing the state of the art in the fields of AI and gender equality. On the one hand, AI is becoming a fundamental field for the development of new and innovative technologies, so it is vital that it represents a positive asset for sustainability (Harari, 2018; Nishant, 2020). On the other hand, gender equality must be addressed in all fields without exception, and, above all, it needs a boost with new ideas and proposals in the field of technology and AI (European Commission, 2018; OECD, 2018).

Through the results presented here, not only the current status in this regard has been identified, but also a series of problems and gaps that must be addressed. The novelty of this work has been demonstrated, due to

the small number of studies in this area, as well as the large increase in studies and the growing importance that these fields are taking in recent years. Therefore, it is necessary to continue with this momentum and address the gaps that exist by developing proposals and empirical validations that cover the different specific contexts of the Targets identified by the SDG 5 of the 2030 Agenda (United Nation, 2015) both in and by AI.

Thus, as future work, we are conducting new studies on gender equality in different areas related to technology, such as IT processes and entrepreneurship in the IT sector, in order to identify different points of view and links that help together to develop new and better proposals to address gender equality in this regard. Likewise, we also intend to develop a framework of best practices that establish the bases for the development, validation, evaluation, and improvement of proposals for both *Gender by AI* and *Gender in AI*. In this way, we want to facilitate and promote these fields both at the research and professional level in organizations.

A society unable to change will not generate any progress. Let us be the change our society needs, promoting new and inclusive ideas for all humankind.

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APPENDIX A. PRIMARY STUDIES SELECTED

Table 4. Primary studies selected

ID	Reference
S01	Hossain, N., Ovi, J. H., Tasnim, S., Islam, N., & Zishan, S. R., 2021. Design and development of Wearable multisensory
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- S11 Wellner, G. & Rothman, T., 2020. Feminist AI: Can We Expect Our AI Systems to Become Feminist?. Philosophy & Technology, 33(2), pp. 191-205.
- S12 Hernández-Álvarez, M. & Granizo, S. L., 2020. Detection of Human Trafficking Ads in Twitter Using Natural Language Processing and Image Processing. Proc. of Advances in Artificial Intelligence, Software and Systems Engineering at the 11th Int. Conference on Applied Human Factors and Ergonomics (AHFE 2020). New York, NJ, USA, pp. 77-83.
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- S28 Vachovsky, M. E., Wu, G., Chaturapruek, S., Russakovsky, O., Sommer, R., & Fei-Fei, L., 2016. Toward More Gender Diversity in CS through an Artificial Intelligence Summer Program for High School Girls. *Proc. of 47th ACM Technical Symposium on Computing Science Education (SIGCSE '16)*. Memphis, TN, USA, pp. 303-308.
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APPENDIX B. PRIMARY STUDIES MAPPING

ID	RQ1		RQ2										RÇ	23	Quality	
ID	а	b	с	d	а	b	с	d	e	f	g	h	i	а	b	score
S01		Х	Х			Х								Х		+1
S02		Х	Х		Х							Х		Х		+3
S03				Х	Х	Х						Х		Х	Х	+4
S04		Х	Х			Х								Х		+2
S05	Х				Х							Х	Х	Х	Х	+1
S06				Х	Х									Х		+3
S07		Х								Х				Х		+2
S08	Х				Х					Х					Х	+3
S09	Х				Х							Х			Х	+1
S10		Х	Х		Х									Х		+3
S11	Х				Х										Х	+2
S12		Х				Х								Х		0
S13		Х	Х			Х								Х		+3
S14		Х				Х								Х		0
S15		Х				Х								Х		+1
S16		Х	Х							Х				Х		+2
S17	Х				Х										Х	+3
S18	Х				Х								Х		Х	+3
S19		Х			Х										Х	+3
S20		Х			Х										Х	+1
S21	Х				Х										Х	+1
S22		Х	Х		Х										Х	+3
S23		Х	Х							Х				Х		+3
S24		Х				Х								Х		+3
S25		Х				Х								Х		+3
S26		Х				Х								Х		+3
S27		Х	Х		Х										Х	+4
S28		Х	Х									Х			Х	+4
S29		Х	Х							Х				Х		+4

Table 5. Primary studies mapping