Remote Sensing of Precipitation Using Reflected GNSS Signals: Response Analysis of Polarimetric Observations

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Abstract—For the first time, rain effects on the polarimetric 1 observations of the global navigation satellite system reflec-2 tometry (GNSS-R) are investigated. The physical feasibility of 3 tracking the modifications in the surface roughness by rain splash 4 and the surface salinity by the accumulation of freshwater is 5 theoretically discussed. An empirical analysis is carried out using 6 measurements of a coastal GNSS-R station with two side-looking antennas in right- and left-handed circular polarizations (RHCP 8 and LHCP). Discernible drops in RHCP and LHCP powers are 9 observed during rain over a calm sea. The power drop becomes 10 larger at higher elevation angles. The average LHCP power drops 11 12 by \approx 5 dB at an elevation angle of 45°. The amplitude of the correlation sum shows a dampening, responding to rain rate 13 systematically. The LHCP observations show higher sensitivity to 14 rainfall compared to RHCP observations. The retrieved standard 15 deviation of surface heights shows a steady increase in the rain 16 rate. The derived surface salinity shows a decrease rains 17 higher than 10 mm/h. This study confirms the potential under 18 environmental conditions of the GNSS-R ground-based station, 19 e.g., with salinity mostly lower than 30 psu, over a calm sea, 20 being a starting point for future investigations. 21

Index Terms—GNSS-reflectometry (GNSS-R), polarimetric
 observations, rain, sea surface salinity (SSS), surface-roughening.

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I. INTRODUCTION

THE dependence of human beings on precipitation as a freshwater resource is clear. This component of the water cycle plays a key role in the economy and sustainable developments. Scarce or extreme rainfalls can lead to droughts

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or floods, threatening lives and properties. Besides, global warming is expected to change the extreme precipitation patterns in terms of magnitude and frequency [1]. Monitoring global precipitation events can assist scientists to better recognize the climate change patterns.

In situ measurements and weather radars are the traditional 34 methods to measure precipitation. They provide the required 35 precipitation information for regional-scale studies, where a 36 dense network of these instruments is established. These 37 techniques are not able to capture the global precipitation due 38 to the coverage limitations, especially over oceans and regions 39 with underdeveloped infrastructures. To this end, meteorologi-40 cal satellites are used that are mainly equipped with advanced 41 microwave and infrared instruments. The tropical rainfall 42 measuring mission (TRMM) was one of the key satellites that 43 contributed to improving our knowledge on the distribution 44 and variability of precipitation within the tropics, operating 45 from 1997 to 2015 [2]. The global precipitation measure-46 ment (GPM) mission refers to a network of next-generation 47 satellites providing precipitation information [3]. The GPM 48 core observatory was launched in 2014 carrying a microwave 49 radiometer. The GPM centers bring precipitation observations 50 from the operational satellites together and provide the stan-51 dard data products. In [4], a review of the precipitation data 52 sources and instruments is provided. 53

The exploitation of reflected global navigation satellite 54 system (GNSS) signals from the Earth's surface, so-called 55 GNSS-reflectometry (GNSS-R), has emerged as a powerful 56 technique to obtain a variety of geophysical parameters and 57 surface properties, see, e.g., [5]. The GNSS-R technique 58 is a multistatic radar method using existing signals from 59 numerous GNSS satellites as the transmitters. The small and 60 cost-effective receivers can be implemented at ground-based 61 stations or onboard different air/spaceborne platforms such as 62 satellites. The cyclone GNSS (CYGNSS) is a constellation of 63 eight low Earth orbiting (LEO) microsatellites, launched in 64 December 2016, fully dedicated to GNSS-R [6]. Given the 65 proven capabilities of GNSS-R to obtain a variety of surface 66 and atmospheric parameters, precipitation monitoring can be a 67 novel application of GNSS-R, which still needs investigations 68 to enhance the knowledge on interactions between rain and the 69 air-sea interface and consequently their impact on GNSS-R 70 measurements. Due to the novelty of the technique, a very 71 limited number of studies discuss the rain effects on GNSS-R 72 observations, and this process is not yet well understood. 73

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The precipitation signature was firstly seen in the measure-74 ments of the TechDemoSat-1 (TDS-1) satellite [7]. Based on 75 a recent scattering model proposed in [8], the authors char-76 acterized the rain splash, the altered ocean roughness by the 77 raindrops impinging on the surface, as a possible phenomenon 78 reducing the received signal power at low wind speeds. This 79 was also in agreement with the study discussing the roughness 80 change as an increment of the sea surface slopes variance 81 [9]. Later, a similar signature was also reported in CYGNSS 82 measurements with the same explanation for the rain effects 83 at low wind speeds [10]. At high winds, an underestimation of 84 wind speed using CYGNSS measurements is reported which 85 could be potentially due to the damping effect of rain on larger 86 scale surface waves [11]. In a simulation study, it is shown 87 that the atmospheric attenuation by raindrops is insignificant 88 in space-borne GNSS-R L-band measurements [12]. This type 89 of effect is expected to be even smaller in magnitude in 90 ground-based GNSS-R observations (such as those in this 91 study) due to the much shorter distance between the specular 92 point and the receiver. 93

The previous studies have investigated the rain impact on 94 GNSS-R measurements obtained by spaceborne receivers with 95 nadir-looking left-handed circular polarization (LHCP) anten-96 nas. The impact on the scattered GNSS signals in right-handed 97 circular polarization (RHCP) is still unknown. The depolariza-98 tion effect by the flattening of the heavy precipitation is shown 99 in polarimetric GNSS Radio Occultation measurements [13], 100 but this type of effect in polarimetric GNSS-R data is not yet 101 studied. 102

The objective of this study is not only to investigate the rain 103 effects on polarimetric observations but also to further charac-104 terize the geophysical signatures and the interactions between 105 rain and sea surface. Due to the lack of spaceborne polarimet-106 ric measurements, this study focuses on a ground-based exper-107 iment that can be potentially scaled up to spaceborne investi-108 gations in future. Section II discusses the physical theory and 109 additionally, based on simulations, describes how the sea sur-110 face salinity (SSS) change due to rainfall could be potentially 111 detectable using polarimetric observations. Section III explains 112 the used data set for the empirical study. Section IV reports 113 on the analysis, and finally Section V summarizes the results 114 and gives the concluding remarks. 115

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II. PHYSICAL THEORY

117 A. Forward Models

The GNSS signals are originally transmitted in RHCP. 118 Reaching the sea surface, part of the signals are bounced off 119 in LHCP, while the rest keep their RHCP. The polarization 120 ratio depends on the reflection geometry and sea surface 121 permittivity. At elevation angles larger than the Brewster angle, 122 the LHCP signal is dominant. If the reflection is recorded at 123 an elevation angle lower than the Brewster angle, the majority 124 of the signal remains at RHCP. Forward models, being valid 125 only for ground-based observations, describe reflected RHCP 126 and LHCP powers, $P_{\rm RHCP}^{\rm r}$ and $P_{\rm LHCP}^{\rm r}$, respectively, [14] 127

$$P_{\rm RHCP}^r = G_{\rm RHCP}^{\rm ref} |\Re_{\rm RR}| W^2 L^2 P_0 \tag{1}$$

$$P_{\rm LHCP}^r = G_{\rm LHCP}^{\rm ref} |\Re_{\rm RL}| W^2 L^2 P_0 \tag{2}$$

where $G_{\rm RHCP}$ and $G_{\rm LHCP}$ are the RHCP and LHCP antenna 130 gains, P_0 is the incoming reference power at the receiver 131 position, and W and L are the power loss due to insufficient 132 delay-Doppler tracking of the reflected signal and surface 133 roughness. Finally, \Re is the polarization-dependent Fresnel 134 reflection coefficient. Analogous to the polarization of incom-135 ing direct signals, there are two Fresnel coefficients. Co-polar 136 coefficients, \Re_{RR} and \Re_{LL} , and cross-polar coefficients, \Re_{RL} 137 and \mathfrak{R}_{LR} , can be obtained from the complex dielectric permit-138 tivity of sea surface water ϵ , and local elevation angle θ as 139 follows [15]: 140

$$\mathfrak{R}_{\mathrm{RR}} = \mathfrak{R}_{\mathrm{LL}} = \frac{1}{2}(\mathfrak{R}_{\mathrm{VV}} + \mathfrak{R}_{\mathrm{HH}}) \tag{3} \quad {}^{14'}$$

$$\mathfrak{R}_{\mathrm{RL}} = \mathfrak{R}_{\mathrm{LR}} = \frac{1}{2} (\mathfrak{R}_{\mathrm{VV}} - \mathfrak{R}_{\mathrm{HH}}). \tag{4}$$

In the above equations, \Re_{VV} and \Re_{HH} , polarization components parallel and perpendicular to the incidence plane, read 143

$$\Re_{\rm VV} = \frac{\epsilon \sin \theta - \sqrt{\epsilon - \cos^2 \theta}}{\epsilon \sin \theta + \sqrt{\epsilon - \cos^2 \theta}} \tag{5}$$

$$\mathfrak{R}_{\rm HH} = \frac{\sin\theta - \sqrt{\epsilon - \cos^2\theta}}{\sin\theta + \sqrt{\epsilon - \cos^2\theta}}.$$
(6) 146

Since P_0 is here unknown, we introduce the power ratios $P_{\rm RHCP}$ and $P_{\rm LHCP}$ as the observables 148

$$P_{\rm RHCP} = P_{\rm RHCP}^r / P^d$$
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$$= \left(G_{\rm RHCP} / G^d \right) |\Re_{\rm RR}|^2 L^2 \tag{7}$$
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$$LHCP = P_{LHCP}/P$$

$$= (G_{LHCP}/G^d) |\Re_{RL}|^2 L^2$$
(8) 152
(8)

where P_0 is canceled out being divided by the power of direct signal $P^d = G_d P_0$.

B. Salinity Change

As (5) and (6) imply, Fresnel coefficients are dependent 156 on the relative permittivity of the seawater and the elevation 157 angle of the signals. The permittivity of seawater is in turn 158 controlled by the SSS and sea surface temperature (SST). 159 The permittivity value can be obtained using models relying 160 on the L-band measurements [16]-[18]. These models are 161 used in ocean state retrievals using GNSS-R measurements 162 [19]. Fig. 1 shows how SSS and SST changes can alter 163 permittivity value based on simulations using the model pro-164 posed in [18]. As shown, at SSS values lower than almost 165 30 psu, which is the SSS range in the area studied in the following sections, the SSS plays a more significant role in 167 controlling the permittivity value. The calculations using the 168 Klein-Swift dielectric constant model confirm that the sensi-169 tivity of emissivity, and therefore reflectivity, to SST at L-band 170 is insignificant also at higher SSS values and SSTs lower than 171 15 °C [20]. 172

Conditions of SSS, considered in the following simulations, apply for the Western Baltic Sea where long-term GNSS-R measurements were conducted at the Onsala Observatory, discussed in the following sections. The historical observations of a station about 29-km away, between 2001 and 2009, show that there are large SSS variations around 25 psu remaining below 30 most of a year.

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Fig. 1. (a) Real and (b) imaginary parts of dielectric permittivity of sea water versus SSS and SST.

Rain creates a fresher layer of water accumulating on the 180 sea surface. This causes a rapid change in the SSS. The depth 181 of the freshwater layer evolves, increasing or decreasing as a 182 function of rain accumulation rate and sea state mixing the 183 water. The mixing rate is in turn dependent on surface wind 184 speed. So, more significant SSS changes are expected over a 185 calm ocean, i.e., at low wind speeds. This has been also the 186 condition providing the means to detect roughness change by 187 rainfall, as discussed in [7]. 188

Using a forward rain impact model (RIM), the SSS change is predictable as follows [21]:

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$$S_{\text{RIM}} = S_0 \left[\left(\prod_{i=1}^{n} \left[1 + \frac{R_{1i}}{\sqrt{k_z * t_i}} e^{-z^2/(4k_z t_i)} \right] \right) \right]^{-1} \\ * \left[1 + \frac{R_2}{\sqrt{k_z * t}} e^{-z^2/(4k_z t)} \right]^{-1}$$
(9)

where S_{RIM} is the new salinity after precipitation, z is the 193 depth in meters, n is the total number of rain events, S_0 is the 194 initial salinity in psu, t is time in seconds, k_z is the vertical 195 eddy diffusivity coefficient, R_{1i} and R_2 are the rain surface 196 impulse function, the rain accumulation integrated over the 197 characteristic mixing depth, for each event in meters. It is 198 shown that the skin depth at the L-band frequency (1.4 GHz) 199 and a fixed water temperature (20 °C) and salinity (34 psu) 200 is 0.955 cm [22]. By definition, the skin depth is the medium 201 thickness through which the electric field amplitude of the 202 propagating electromagnetic wave falls to 1/e, i.e., 37%, of its 203 original value. Accordingly, the L-band penetration depth is 204



Fig. 2. SSS versus rain rate at different duration lengths (D) and $S_0 = 25$ psu.

not more than 1 cm, and we consider z = 0.005 m in this ²⁰⁵ study. ²⁰⁶

Fig. 2 shows the SSS change due to rain events at different 207 rates and duration (1, 6, 12, 18, and 24 h). For instance, 208 the blue curve shows a 1-h rainfall at a constant rate of 209 10 mm/h drops the SSS from 25 psu to \approx 24 psu. Following 210 that, Fig. 3 shows the expected power levels and their ratio in 211 these rain event scenarios. As demonstrated, altered SSS by 212 rain could affect the observations at significant rain rates with 213 a long enough duration. Besides, the power change for both 214 RHCP and LHCP reflected GNSS signals and their ratio at 215 different SSS values are illustrated in Fig. 4. As shown, sig-216 nificant drops of SSS are distinguishable in the power changes. 217 This phenomenon increases the power of the LHCP signals, 218 whereas, it decreases that scattered in RHCP. The discrepancy 219 in the LHCP to RHCP power ratio will be therefore more 220 significant leading to a more detectable SSS signature. This 221 can be an advantage of polarimetric GNSS-R observations. 222

The RIM used here considers the surface salinity profile 223 as a function of depth and time. Wind speed is not included 224 and the model is assumed to describe the near-surface salinity 225 profile at low wind speeds (0-3 m/s). The analysis in the 226 following sections is conducted at a calm sea state. This 227 is the condition in which the model is valid. Nevertheless, 228 we should consider a level of uncertainty at different wind 229 speeds. Besides, the wind might have a weaker effect in 230 the coastal areas due to the land sheltering and the lim-231 ited fetch [23], leading to a faster accumulation of the 232 freshwater. 233

C. Roughness Change

When a raindrop impinges on the sea surface, a cavity with a crown is firstly created which will change to a vertical stalk at the center of rings of gravity-capillary waves. These waves propagate outwards, known as "ring waves" [24]. These small-scale waves alter the surface roughness which could affect the scatterometric measurements. 240

Numerous laboratory experiments are investigating the altered surface waves by rain. Based on them, theoretical models have been proposed to describe the generated waves and the impact on the surface, see, e.g., [25]–[28]. Although the ring waves intensify the centimeter-scale roughness, rain could also attenuate ocean surface gravity waves [27]. 240





Fig. 3. Power of (a) reflected RCHP, (b) RHCP, and (c) signals and their ratio, and SST = 10 °C vs. rain rate at different rainfall duration lengths D, $\theta = 30^{\circ}$, and $S_0 = 25$ psu.

The developed theories based on laboratory experiments do 247 not yet sufficiently answer the questions on the mechanism 248 of rain impact on ocean surface waves in real environments 249 [29]. The simulations based on GNSS signal scattering models 250 and the log Gaussian spectrum of rain-generated waves [28], 251 do not show an exact match with the rain impacts observed in 252 the empirical measurements [7]; however, the authors did not 253 exclude the possibility of other effects existence such as swell 254 and downdraft. In a recent study, field observations are used 255 to describe rain impact on surface roughness [29]. Therein, 256 more significant rain-enhanced gravity-capillary waves, with 257 wavelengths smaller than $\lambda = 56$ mm are reported. These 258 waves control the intensity of forward GNSS scattering in the 259 regime of weak diffuse scattering, i.e., at low wind speed, 260 which is the environmental condition for observing altered 261 roughness effects in GNSS-R observations. 262

Further studies are required to characterize the rain-ocean 263 interactions and consequently the effects on signal forward 264

Power of (a) reflected RCHP, (b) RHCP, and (c) signals and their Fig. 4. ratio versus elevation angle θ at different SSS and SST = 10 °C.

scattering patterns. We should therefore admit the effect of rain 265 on the ocean surface is one of the least understood processes. 266 We will discuss the result of this study on the surface change 267 more in an empirical sense trying to enhance the knowledge 268 on this process in a top-down approach. 269

In this study, the polarization-independent power loss due 270 to the surface roughness is considered as [14]

$$L = \exp\left[(-1/2)\left(4\pi^2/\lambda^2\right)\sigma^2\sin^2\theta\right]$$
 (10) 272

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where σ is the standard deviation of sea surface height and λ is 273 the wavelength of the GNSS signal. It should be noted that the 274 power loss model described here applies to coherent scattering 275 and does not account for possible polarization-dependent sig-276 natures of the roughness. This means the reflected signals can 277 exhibit some degrees of ellipticity originated from the structure 278 of the sea surface roughness. For instance, the horizontal 279 parallel crests can act as an oriented structure affecting the 280 horizontal component of the signals. Such roughness-induced 281 polarimetric effects need to be further investigated in future 282 studies. 283



Fig. 5. Eastward view of the GFZ GNSS-R station at (a) OSO, (b) zenith looking, RHCP, and side looking, RHCP and LHCP, antennas, and (c) sea targeted at the antenna boresight.

III. EXPERIMENTAL DATA

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A GFZ coastal GNSS-R station at Onsala Space Observa-285 tory (OSO) (57.393°N; 11.914°E) in Sweden is considered 286 in this study for investigating the effect of precipitation on 287 GNSS-R polarimetric observations. The data set covers a 288 period of one year from January to December 2016. The 289 station is equipped with a GNSS occultation, reflectometry, 290 and scatterometry (GORS) receiver [30] with two reflectome-291 try antennas in different polarization, i.e., RHCP and LHCP. 292 The antennas are tilted $\approx 98^{\circ}$ with respect to the zenith. The 293 boresight of the reflectometry antennas is set at an azimuth 294 angle of $\approx 150^{\circ}$ to capture sea surface reflections with the 295 highest gain values. The station environment and antennas are 296 shown in Fig. 5. 297

The gain pattern of the antennas for incoming signals as a 298 function of satellite elevation and azimuth angles is shown 299 in Fig. 6. The experiment setup assigns high gains to the 300 signals at grazing elevation angles, i.e., signals from satellites 301



Fig. 6. Gain pattern of the reflectometry antennas as a function of satellite elevation and azimuth angles.



Fig. 7. Spatial extent of the specular points over the sea.

at 0° to about 40° elevation angles. The signals from the 302 satellites with azimuth angles close to the antennas' boresight, 303 i.e., 150°, are recorded at higher gains. The gain pattern is 304 assumed to be equal for both the RHCP and LHCP sea-looking 305 antennas. 306

The receiver uses designated channels to track the signals of the up-looking antenna (master channel) and side-looking 308 antennas (slave channels), as previously described by [31]. In-phase and quadrature samples (I/Q) of the respective chan-310 nels are recorded. Fig. 8 shows an example of the receiver output from the GPS satellite PRN 7 signals captured by the 312 sea-looking antennas.

The antennas, here, have a small baseline (≈ 20 cm) with 314 respect to each other and are mounted with a height of about 3-315 m above the reflecting sea surface. For the given geometry the 316 differential delay between direct and reflected signals cannot 317 be resolved in the code delay domain. Multipath patterns of 318 the direct and the reflected signals occur in the receiver output 319 as previously explained, for example, by Larson et al. [32]. 320 As shown in Fig. 8, a low-frequency pattern is evident in the 321 interferometric pattern, which is attributed to the direct signal. 322 A separation algorithm is applied to find the low-frequency 323 pattern of the direct signal and the higher-frequency pattern of 324 the reflected signal, as described by [33]. To be more specific, 325 a first-order polynomial is fit to the low-frequency variations of 326 I/Q correlation sum in each segment. Having the contribution 327

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Fig. 8. Reflectometry observations of the GPS satellite PRN 7 on October 15, 2016, obtained from the Onsala station using two sea-looking antennas with right- and left-handed circular polarizations (RHCP and LHCP). The receiver output, i.e., the correlation sums, at in-phase and quadrature channels are shown as solid and dotted lines, respectively.

of the direct signal from this fitting, the I/Q contributions of the
 reflected signal is determined. For the here given geometry the
 separation algorithm yields estimates of the direct and reflected
 signal power every ten minutes.

The experiment uses wind, tide gauge, and precipitation measurements as the ancillary information for the analysis. The wind and sea-level measurements are obtained from the nearest meteorological and tide gauge stations, respectively. For the precipitation estimates, we use the GPM, half-hourly $0.1^{\circ} \times 0.1^{\circ}$ Version 06B Level 3 IMERG final run product.

The GNSS-R station can measure reflected signals at a 338 sampling rate of 0.1 Hz from GPS satellites with elevation 339 angles ranging from 0° to 50°. The spatial coverage of the 340 specular points is shown in Fig. 7. During the one-year obser-341 vation period, from January to December 2016, 175178 mea-342 surements are recorded, from which 26413 data are captured 343 during rainfall, i.e., at rain rates higher than 0 mm/h. The 344 maximum and average recorded rain rates are 23.04 and 345 0.09 mm/h, respectively. 346

IV. ANALYSIS

Fig. 9 visualizes the power in both LHCP and RHCP versus 348 the elevation angle along with simulated measurements using 349 (1) and (2), derived from the entire data set. There is a general 350 agreement between the observed and simulated measurements; 351 however, the discrepancy of $P_{\rm RHCP}$ is larger compared to 352 P_{LHCP} . Especially, there is a mismatch between the RHCP 353 simulated and observed measurements at low elevation angles 354 $(\leq 10^{\circ}).$ 355

To investigate rain effects, the data are limited to wind 356 speeds lower than 5 m/s as it is the necessary condition for 357 detecting rain splash. Additionally, only the data at winds 358 blowing from the land side, i.e., with azimuths between 0° and 359 150°, are considered. The location of the station and the con-360 sidered azimuth range is shown in Fig. 10. In this condition, 361 the wind-wave generation is limited by the short fetch. We use 362 363 this sheltering effect of the coastline on the nearshore waves to further exclude wind-associated effects in the measurements. 364 Then, this coastal experiment better provides the environmen-365 tal conditions for tracking the rain splash effects compared 366 to those using spaceborne measurements. Hoseini et al. [23] 367 reported on the insignificant correlation between the offshore 368 or land breeze and sea surface roughness at this GNSS-R 369 station owing to the limited fetch. Furthermore, no meaningful 370 correlation between precipitation and wind speed could be 371 identified in our statistical analysis. The correlation coefficient 372 between precipitation and wind speed is 0.1. 373



Fig. 9. Power of reflected (a) RCHP and (b) LHCP signals versus elevation angle θ along with simulated observations at $\sigma = 0.03$ m in red (entire data set). The SST values are obtained from a nearby meteorological station. The SSS for each observation is estimated as an average value based on historical records of between 2001 and 2009 at a station located about 29-km away.

Fig. 11 demonstrates the power of the reflected signal in 374 both polarizations, in different cases, during rainfalls and 375 rain-free measurements. A significant discrepancy between the 376 two cases is shown which is larger at higher elevation angles. 377 Besides, the simulations show the power behavior at different 378 surface states. The similar patterns between the simulated 379 and observed behavior show a possibility that the observed 380 effects appear due to the rain splash altering the surface state, 381 as discussed in [7]. In that spaceborne analysis, studying 382 the intensity of the effect at different incidence angles was 383 postponed to a future work due to the uncertainty of mea-384 surements. Here, the dependence of the rain effect on the 385 reflection geometry is evident. We cannot exclude that the 386 power drop in the empirical data might be intensified by other 387 types of effects such as swell and downdraft which can be 388 potentially the factors causing larger standard deviations dur-389 ing rainfall in Fig. 11. According to the figure, a higher level 390 of sensitivity to rain in LHCP measurements is seen compared 391



Fig. 10. GNSS-R station location and the wind azimuth range condition $(0^{\circ}-150^{\circ})$ in the analysis.

to those RHCP observations. Besides, the probability density
function (PDF) of the power measurements, derived as the
kernel density estimation (KDE) is given in Fig. 12. It indicates
that rain has distorted the PDF in both sets of measurements,
dampening the peak and increasing the probability at lower
values of power.

Fig. 13 shows the RHCP and LHCP reflection amplitudes 398 at two different precipitation conditions during the setting 399 period of the GPS satellite PRN 7. The shown exemplary 400 cases do not necessarily meet the aforementioned conditions 401 on the wind speed and its azimuth. Hence, wind speed and 402 direction profiles are additionally given. The left column in the 403 figure shows the I/Q components of the reflected signal during 404 a rain-free period. As expected, a general trend of decreasing 405 RHCP amplitude, and on the contrary, increasing LHCP 406 amplitude over the satellite elevation angle is evident, which is 407 due to the reflection geometry. Both cases are associated with 408 offshore winds, so we expect to have a minimized wind-driven 409 roughness effect. Fig. 13(a) on the left column reports on 410 slightly higher wind speed. As a result, an insignificantly 411 rougher sea surface with more power loss could be anticipated 412 for the left column. The right column shows the case with 413 the same GPS satellite and reflection geometry but during a 414 rain-affected period. 415

Although the wind conditions might imply larger amplitude in rain-affected case, the RHCP and LHCP correlation sums in Fig. 13(d) and (f), exhibit smaller amplitudes leading to lower power ratios in Fig. 13(h). This power loss in the measurements of both polarizations compared to those in 13(g) can be explained with the rainfall. The striking fact 421 is the systematic response to the rain rate in Fig. 13(h). 422 There is an increase in the rain rate, from almost 0.5 to 423 3.6 mm/h, at elevation angles between 17° and 26°. The 424 amplitudes shrink within this range along with a decrease in 425 power ratios. The rain rate decrease also stimulates a relative 426 increase in both of the power ratios when the satellite elevation 427 angle passes 26°. However, the power ratios at elevation 428 angles from 26° to 50° with a rain rate of 1.1 mm/h are 429 slightly lower compared to the power ratios in Fig. 13(g). The 430 amplitude change due to rain is more prominent in the LHCP 431 observations compared to those in RHCP. 432

Considering the SSS change as the additional rain effect, we develop an inversion algorithm to estimate the standard deviation of surface heights and the SSS. $P_{\text{RHCP}}(\sigma, S)$ and $P_{\text{LHCP}}(\sigma, S)$ are considered as the observable parameters, see (7) and (8). The values of σ and S are sought which minimizes the cost function

$$\delta(\sigma, S) = \frac{1}{n} \sum_{i=1}^{n} \left| P_{\text{RHCP},i}^{o} - P_{\text{RHCP},i}^{s}(\sigma, S) \right|$$
⁴³⁹

$$+\frac{1}{n}\sum_{i=1}^{n} |P_{\text{LHCP},i}^{o} - P_{\text{LHCP},i}^{s}(\sigma, S)| \quad (11) \quad 440$$

where the superscripts o and s indicate the observed or simu-441 lated parameters, respectively. We assume that the variations of 442 S and σ within 3 h are insignificant. So, these parameters are 443 retrieved as the average value for every three-hour time span 444 and *n* is the number of observations residing in each temporal 445 bin. The conditions on the wind speed and its direction are 446 applied here. Fig. 14 shows the obtained values for both 447 parameters as a function of the rain rate. 448

According to Fig. 14, most of the data have a σ value 449 of \approx 3.5 cm. A steady increase in the average σ over the 450 rain rate is observed with the correlation coefficient of 0.96. 451 The derived SSS values show no significant trend at rain 452 rates lower than 10 mm/h. Most of the data points, in a 453 rain-free condition, reside at the SSS of ≈ 26 psu. A slight 454 downward trend in the SSS at low rain rates, between 0 455 and 2.5 mm/h, is observed, decreasing the average SSS to 456 \approx 24 psu. Nevertheless, at higher rain rates up to 10 mm/h 457 no significant change is observed in the average SSS value, 458 which could statistically approve the SSS change due to rain. 459 However, at rain rates larger than 10 mm/h a significant drop 460 in the SSS values is observed. 461

The altered surface state and the effects on the power ratios 462 shown here are much larger than the theoretical study on 463 the surface effect of rain based on laboratory measurements, 464 e.g., [28]. There is also a quantitative mismatch on the 465 observed effects in TDS-1 and CYGNSS measurements with 466 the theoretical studies [7], [10]. This inconsistency could be, 467 at least partially, associated with the environmental differences 468 between laboratory and real sea conditions. In a laboratory, 469 the theoretical knowledge is derived based on observing wave 470 parameters in a water tank influenced by artificially-generated 471 rain and wind. Laxague and Zappa [29] discussed that the 472 variation in such conditions is spatial, and not temporal. This 473 is a considerable difference in a real sea, where rain may 474



Fig. 11. (a) RHCP and (c) LHCP power ratios and in different cases, during rain events, at rates higher than 0.2 mm/h, and at no rain along with model-simulated (b) RHCP and (d) LHCP power ratios at different standard deviations of surface heights σ . Average and maximum rain rates of the data during rainfall are 1.5 and 23.0 mm/h, respectively.



Fig. 12. PDF of (a) RCHP and (b) LHCP power in cases during rainfall and at no rain, derived as KDE.

affect the surface state rapidly, but environmental conditions
do not vary spatially at a significant level for long periods.
As a result, the laboratory-derived models are not satisfactory
to characterize the problem of rain splash impacts on the sea

surface waves and the change in the GNSS forward scattering 479 patterns. In situ measurements of the impact of rain show a 480 strong increase in the roughness in the scale of the gravity 481 capillary waves with a wavelength between 5.4 and 56 mm 482 [29]. These waves control the intensity of the reflection when 483 the coherent component is considerable, i.e., when the wind 484 speed is low and the Rayleigh parameter is large enough, 485 see [7]. This is the condition applied to the analysis here. 486

The discussions in Section II theoretically prove the potential of simultaneous retrieval of SSS and surface state using polarimetric observations. The average values in Fig. 14 demonstrate the rain-associated systematic behaviors of these parameters. However, considerable uncertainties in σ and the SSS change are also seen which could be associated with different factors.

We cannot exclude non-rain environmental changes, such 494 as swell and downdraft, which could still affect the retrievals 495 and increase the uncertainty. We are not yet able to include 496 all possible factors in the analysis since there is no readily 497 available data set and method for parameterizing these com-498 plex processes. Additionally, we have visualized the impacts 499 as a function of rain rate; however, the raindrop size and 500 rainfall duration can impact the intensity of effects. It is 501 known that larger raindrops can modify surface roughness 502 more significantly, see, e.g., [34]. The raindrop size also 503 controls the terminal velocity and, consequently, the down-504 draft intensity, see [10] and references therein. At low 505 wind speeds such as those in the analysis here, small 506 wave damping in the low-frequency band of wave spectrum 507 might happen whose rate could increase as rain duration 508 prolongs [35]. 509



Fig. 13. Exemplary case of the rain impact on reflectometry observations from GPS satellite PRN 7. The graphs on the left side are related to the observations over a rain-free period. Right panels are associated with a time span that includes precipitation records based on the GPMs data. The reflectometry observations on both sides are selected to be over the same elevation angles of the satellite (a) and (b) wind speed and direction, (c) and (d) receiver output (correlation sums), quadrature in dotted and in-phase in solid line, from the RHCP antenna, (e) and (f) receiver output (correlation sums), quadrature in dotted and in-phase in solid line, from the reflection power to direct signal power estimated from the RHCP and LHCP correlation sums against rain rate.

Besides, technical limitations should be considered. Here, 510 the measurement noise might degrade the retrieval accuracy. 511 A noise power analysis is carried out in [23] using the 512 same data set. Noise power is calculated as the variance of 513 quadrature components of the zenith-looking antenna over a 514 temporal segment. Accordingly, the average noise power is in 515 general between -1 and 0 dB, depending on the elevation 516 angle. However, the standard deviation of the noise, varying 517 between 2.5 and 3.5 dB, depending again on the elevation 518 angle, is large enough to affect the retrieval. 519

Finally, we rely here on physical models by calculating 520 the difference between observed and modeled powers in 521 (11). Any inaccuracies in the models could also affect the 522 retrieval efficiency. According to (10), the surface roughness 523 is supposed to have a polarization-independent behavior. The 524 empirical data demonstrate inconsistencies since the RHCP 525 observations show a lower level of sensitivity to the roughness 526 change compared to LHCP observations. The existence of 527 this dependence is shown in Figs. 11 and 13. Fig. 15 shows 528 the discrepancies between the modeled and observed LHCP 529 to RHCP power ratio. According to the theoretical model, 530 the ratio is supposed to be roughness-free, but here the rough-531 ness effect is evident, which is larger in magnitude at high 532 elevation angles. The model has not been able to sufficiently 533 describe the polarization-dependent results found, e.g., by 534

[23] and [33] neither. Besides, the challenges on modeling and correcting the roughness effects to retrieve soil moisture is discussed by [36]. There is still a need and room for investigations on this topic.

This study has been focused on rain-caused modifications 539 of reflecting surface properties. Intense rainfalls can addition-540 ally induce depolarization effects on the signals propagating 541 through the atmosphere in grazing angle geometry due to the 542 asymmetry between the horizontal and vertical dimensions 543 of the big droplets. This effect has been studied in Radio 544 Occultation [13], [37] and ground-based direct measurements 545 [38]. The change in the ellipticity of the incoming circularly 546 polarized signals could in turn affect the power estimations at 547 intense rains and low elevation angles. In contrast, the studies 548 here show that the observed signatures are more intense at high 549 elevation angles and also noticeable at low rain rates. Although 550 this implies that the ellipticity change of the incoming signal 551 could not be the dominant type of effect causing the signatures 552 observed here, we encourage characterizing this type of effect 553 in future studies. There are still important questions on how 554 thereby the GNSS-R observations are affected requiring decent 555 investigations using optimized setups and analysis constraints 556 for this purpose. Such an investigation is beyond scope of this 557 article. The potential consequences in polarimetric GNSS-R 558 measurements could be carried out in future studies with long 559

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Fig. 14. Obtained values of (a) standard deviation of sea surface heights σ and (b) SSS along with simulated SSS at different rain duration D versus rain rate. The average values and standard deviations are shown in red.



LHCP to RHCP power ratio versus elevation angle along with Fig. 15. simulations shown in red. The SST values are obtained from a nearby meteorological station. The SSS for each observation is estimated as an average value based on historical records of between 2001 and 2009 at a station located about 29-km away

enough data sets especially with dense measurements at high 560 rain rates. 561

V. SUMMARY AND DISCUSSION

We investigated the response of polarimetric GNSS-R 563 observations to precipitation for the first time. A discussion 564 was made on the theoretical potential and the necessity of 565 dual-polarization measurements for tracking the rain-caused 566 SSS change in addition to sea surface modifications. The 567 combination of the RHCP and LHCP observations amplifies 568 the left signatures in the power ratios by the SSS change. Using 569 a **RIM**, the change in the signal power due to SSS change is 570 simulated which shows a considerable impact at high rain rates 571 with long enough duration. 572

An analysis is carried out using measurements of a 573 coastal GNSS-R station with two (RHCP and LHCP) 574

side-looking antennas. The rain effects result in the discrep-575 ancy of the average power ratios in cases during rainfall and 576 at no rain, which is proportional to the elevation angles. The 577 geometry-dependence of the discrepancy magnitude implies 578 a higher probability that the observed effects are mainly 579 associated with the roughness change. 580

The left signature in both LHCP and RHCP power is large 581 enough to be distinguishable in the measurements. The average 582 LHCP power at an elevation angle of 45° drops by ≈ 5 dB; 583 however, a significant increase in the standard deviation of the 584 LHCP power is observed. In general, the LHCP observations 585 enjoy a higher level of sensitivity to rain events, compared 586 to RHCP measurements. This fact is further approved by 587 investigating the amplitude of correlation sums in exemplary 588 cases. The amplitude of the LHCP correlation sum shows more 589 evident dampening by rainfall. This is a piece of evidence 590 that the roughness loss is polarization-dependent, contrary to 591 what the theoretical model implies. As a result, the models, 592 especially those describing the power loss by the surface 593 roughness, could be subjected to refinement in future studies. 594

An algorithm retrieving the SSS and surface state using the 595 polarimetric measurements is proposed. The retrieved standard 596 deviation of surface heights and the SSS also confirms that the 597 general significant modification of the surface state, whereas, 598 the SSS shows no meaningful behavior responding to rainfalls 599 with rates below 10 mm/h. The retrieved SSS demonstrates 600 a drop at higher rain rates. However, the data in this range 601 are too sparse and studies with a substantially higher number 602 of observations at extreme rain rates are encouraged and still 603 required. 604

The reported roughness change by rain splash in this study is 605 substantially larger than the modifications described by models derived based on laboratory measurements. We discussed that this can be due to the inefficiency of these models in describing the altered roughness in a real sea. The results show an agreement with recent in situ measurements in the 610 real environment. This also explains the quantitative mismatch 611 between the effects seen in the spaceborne measurements and 612 model-based simulations in the previous studies. 613

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None of the data sets used so far for the investigation of rain 614 sensitivity are optimized for this purpose. Future investigations 615 could be carried out on data sets from GNSS-R altimetry 616 instruments due to the similarities of required environmental 617 conditions, i.e., coherent reflections from calm surfaces. How-618 ever, there is a dissimilarity in the desired reflection geometry. 619 We are here interested in reflections at high elevation angles 620 rather than those at grazing angles, considering the significant 621 role of roughness change. The reflecting sea in the area of 622 this study has smaller SSS values compared to global oceans 623 remaining below 30 psu most of the time. In future studies 624 on environments with higher SSS, SST change due to cold 625 rainfalls can also be a significant research question over 626 oceans with SSTs above 15 °C. The change in the ellipticity 627 of the incoming circularly polarized signals propagating the 628 rainy atmosphere deserves attention in future studies. With 629 the derived knowledge in this study and previous investiga-630 tions, we encourage future experiments fully dedicated to this 631 topic. 632

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