

1 InSAR measured permafrost degradation of palsa peatlands 2 in northern Sweden

3 **Authors:** Samuel Valman^{1,2*}, Matthias B. Siewert^{3*}, Doreen Boyd², Martha Ledger^{4,5}, David
4 Gee⁶, Betsabe de la Barreda-Bautista^{4,2}, Andrew Sowter⁶, Sofie Sjögersten⁴

5 **Affiliations:**

6 ¹ Nottingham Geospatial Institute, University of Nottingham, Nottingham NG7 2TU, UK

7 ² School of Geography, University of Nottingham, University Park, NG7 2RD, Nottingham, UK

8 ³ Department of Ecology and Environmental Sciences, Umeå University, Umeå, Sweden

9 ⁴ School of Biosciences, University of Nottingham, Sutton Bonington Campus, College Road, LE12 5RD,
10 Loughborough, UK

11 ⁵ School of Biological Sciences, Kadoorie Biological Sciences Building, University of Hong Kong, Pok Fu Lam
12 Road, Hong Kong

13 ⁶ TerraMotion, Ingenuity Centre, Triumph Rd, Nottingham NG7 2TU

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15 *These authors contributed equally to this work.

16 Corresponding author: Sofie Sjögersten sofie.sjogersten@nottingham.ac.uk

17 **Abstract.** Climate warming is degrading palsa peatlands across the circumpolar permafrost region. Permafrost
18 degradation may lead to ecosystem collapse and potentially strong climate feedbacks, as this ecosystem is an
19 important carbon store and can transition to being a strong methane emitter. Landscape level measurement of
20 permafrost degradation is needed to monitor this impact of warming. Surface subsidence is a useful metric of
21 change and can be monitored using InSAR satellite technology. We combined InSAR data, processed using the
22 ASPIS algorithm to monitor ground motion between 2017 and 2021, with [airborne](#) optical and LiDAR data to
23 investigate the rate of subsidence across palsa peatlands in northern Sweden. We show that 55% ~~of the area~~ of
24 Sweden's eight largest palsa peatlands ~~are~~ currently subsiding, which can be attributed to these permafrost
25 landforms and their degradation. The most rapid degradation occurring in the largest palsa complexes in the
26 most northern part of the region of study, also corresponding to the areas with the highest % palsa cover within
27 the overall mapped wetland area. Further, higher degradation rates were found in areas where winter
28 precipitation has increased substantially. The roughness index calculated from a LiDAR-derived DEM, used as
29 a proxy for degradation, increases alongside subsidence rates and may be used as a complementary proxy for
30 palsa degradation. We show that combining datasets captured using remote sensing enables regional-scale
31 estimation of ongoing permafrost degradation, an important step to-wards estimating the future impact of
32 climate change on permafrost-dependent ecosystems.

33

34 **Keywords:** Permafrost, subsidence, Arctic, InSAR, palsa, peatlands

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37 2.01.0 Introduction

38 Permafrost regions are critical components in the climate system, due to their essential carbon (C) storage
39 service (Harris et al., 2022). The circumpolar permafrost region in particular, stores around 1300±200 Pg of
40 organic C, corresponding to around 50% of the global terrestrial C pool (Hugelius et al., 2020; Köchy et al.,
41 2015). It covers around 21 million km² or 22% of the Northern Hemisphere's landscapes-exposed land surface
42 (Obu, 2021). Northern peatlands themselves store an estimated 415±150 Pg of C in an area covering around 3.7
43 million km² of which around 1.7 million km² is permafrost ~~substantially within overlapping with~~ the circumpolar
44 permafrost region in discontinuous and sporadic permafrost zones (Hugelius et al., 2020). Permafrost in these
45 peatlands raises the surface above the water table forming so-called palsa (pl. palsas) or, in extended form, peat
46 plateaus (Seppälä, 2011). These account for substantial areas of global permafrost, including in northern
47 Fennoscandia (Ballantyne ~~C. K.~~, 2018; Gislén et al., 2017; Tarnocai et al., 2009), ~~for example, in~~ northern
48 Sweden, 137 km² of the ~~these~~ palsa ~~peatland has have~~ been recorded from field reports reported (Backe, 2014).
49 Climate warming, and the associated alteration in the precipitation regime, is increasingly recognized to be a
50 particular threat to permafrost (Biskaborn et al., 2019), with the subarctic Fennoscandian permafrost region, and
51 the palsa ~~peatlands~~ within, particularly vulnerable (Christiansen et al., 2010; Farbroten et al., 2013).

52 Climatic models project unsuitable conditions for permafrost within the coming century, with the most pessimistic
53 estimates projecting unsuitability even sooner - by 2040 in Fennoscandia (Chadburn et al., 2017; Fewster et al.,
54 2022; Könönen et al., 2022; Stefan et al., 2006). As palsa ~~peatlands~~ are often found in the sporadic or
55 discontinuous permafrost zone (Zuidhoff & Kolstrup, 2000), they are particularly sensitive to climate warming
56 and any resultant permafrost thaw and disappearance. Their sensitivity mainly results from the alterations in the
57 thermal insulation effect of peat deposits and snow as the climate changes (Seppälä, 2011; Smith & Riseborough,
58 1996). Specifically, organic peat has a high thermal conductivity when wet and frozen, but low conductivity when
59 dry and thawed. Snow has a highly insulating effect on ground temperature. Thus, extended periods of air-
60 temperatures below 0°C and thin snow cover in winter are beneficial to maintain or grow the perennial frozen
61 ~~permafrost~~ core of palsas and peat plateaus. Low summer precipitation, which reduces the thermal conductivity
62 of peat, also helps to preserve the frozen cores in palsa. In contrast, increased snowfall has been linked to
63 permafrost degradation as it increases winter insulation. Further, high summer precipitation leads to higher
64 thermal conductivity of peat, and combined with warm summer temperatures, can degrade permafrost by
65 increasing permafrost temperatures and subsequent thawing of the frozen peat core of palsas. The strong insulating
66 properties of peat allow the occurrence of permafrost at the southern extent of the northern permafrost region and
67 valley bottoms in areas otherwise too warm for permafrost (Johansson et al., 2013; Seppälä, 2011; Smith &
68 Riseborough, 1996).

69 Warming of the permafrost in palsa ~~peatlands~~ typically leads at the surface, to top-down thaw, (i.e. thickening of
70 the active layer), and eventual subsidence of the surface, as well as lateral thaw, sometimes called abrupt thaw
71 or thermokarst, which occurs at the margin of peat plateaux and palsa edges (Seppälä, 2011; Smith &
72 Riseborough, 1996; Zuidhoff, 2002). This is often associated with water-logged conditions and, as a result,
73 increased methane (CH₄) emissions (Glagolev et al., 2011; Hugelius et al., 2020; Matthews et al., 1997;
74 Miglovetz et al., 2021; Schuur et al., 2009; Turetsky et al., 2020; Varner et al., 2022), which is a central theme
75 for permafrost research (Sjöberg *et al.*, 2020). A subsequent impact of this permafrost degradation is an
76 alteration in vegetation cover, its hydrology, and human use of the landscape (e.g., infrastructure and reindeer
77 husbandry)(Markkula et al., 2019; Ramage et al., 2021). Given the potentially large impacts of permafrost thaw
78 on the global climate, ecosystem function and human activity, quantification and monitoring of the subsidence
79 in peat deposits affected by permafrost thaw and degradation, as well as an understanding of their sensitivity to
80 changing climatic parameters, is urgently required (IPCC, 2021).

81 The degradation of the permafrost of palsa ~~peatlands~~ has been observed right across the circumpolar
82 permafrost region in a number of studies, including in northern Scandinavia (Åkerman & Johansson, 2008; de la
83 Barreda-Bautista et al., 2022; Luoto & Seppälä, 2003; Olvmo et al., 2020; Sannel et al., 2016; Varner et al.,

84 2022); Russia (Glagolev et al., 2011; Miglovets et al., 2021; van Huissteden et al., 2021); the USA (Douglas et
85 al., 2021; Douglas et al., 2015; Sannel, 2020) and Canada (Mamet et al., 2017; Sannel & Kuhry, 2011; Short et
86 al., 2014; Vallée & Payette, 2007). Although rapid degradation in response to short term climatic events has
87 been observed, typically permafrost degradation has been investigated via long-term monitoring at decadal
88 timescales in response to changes in temperature and precipitation conditions (Åkerman & Johansson, 2008; de
89 la Barreda-Bautista et al., 2022; Olvmo et al., 2020; Sannel et al., 2016). These longer-term studies have shown
90 strong relationships between permafrost degradation and summer temperatures, length of the thaw period,
91 winter precipitation and snow depth (Smith et al., 2022). These types of analyses are very useful for quantifying
92 how much of the landscape has already transitioned and understanding the climate change drivers behind these
93 changes, but they do not capture the initial stages of permafrost degradation in palsas ~~peatlands~~ and the lower
94 rates of subsidence that have yet to result in observable changes in the vegetation or thermokarst formation. The
95 latter is crucial to understand the ongoing response of palsas ~~peatlands~~ to climate warming and to predict when
96 pulses of greenhouse gases to the atmosphere and other impacts (e.g., on infrastructure) are likely to occur.
97 Thus, we need approaches that detect early signs of degradation at landscape scales, with repeated observations,
98 ~~are urgently required.~~

99 Due to the vast extent and remoteness of permafrost areas there is no current complete annual degradation rate
100 measurements. So, we looked to satellite remote sensing to underpin the measurement and monitoring
101 assessment of permafrost peatlands, their degradation and resultant climate impacts (~~Armstrong-McKay et al.,~~
102 ~~2022;~~ Hugelius et al., 2020; ~~Obu, 2021;~~ ~~Schuur et al., 2015;~~ Swingedouw et al., 2020). Optical remote sensing
103 approaches can be augmented with RaDAR remote sensing methods, including InSAR, to capture the early
104 response of permafrost to warming, ~~since~~ these methods can detect vertical land surface motion at millimetre
105 precision across a range of natural landscapes, with greater confidence in the direction of surface motion than
106 the absolute magnitude (Alshammari et al., 2020; Alshammari et al., 2018; ~~Bartsch et al., 2016;~~ de la Barreda-
107 Bautista et al., 2022; Short et al., 2014; van Huissteden et al., 2021) ~~(Bartsch et al., 2016)~~. The regular sampling
108 frequency, insensitivity to cloud and, in the case of Sentinel-1, low cost, means InSAR from Sentinel-1 should
109 be well suited to measure and monitor ongoing changes in permafrost affected by climate change. Further,
110 Sentinel-1 for InSAR is effective at both local and regional scales - the 20m × 20m spatial resolution enables
111 measurement of surface motion within local sites (de la Barreda-Bautista et al., 2022), and can do so over entire
112 and complex landscapes, such as the circumpolar permafrost region (Reinosch et al., 2020).

113 The overall aim of this study was to carry out a regional-scale analysis of permafrost degradation across the
114 palsas ~~peatlands~~ of northern Sweden, principally using Sentinel-1 InSAR-derived subsidence as an indication of
115 degradation. Pertinent to this is that any InSAR-detected changes can be associated with known and delineated
116 targets in the wider landscape. Furthermore, it is also important to understand any within-site dynamics of
117 permafrost degradation. This paper therefore has specific objectives to: (i) measure the subsidence rate between
118 2017-2021 of all major palsa peatlands-complexes in the northern Sweden region; (ii) determine in which palsa
119 peatlands-complexes subsidence is greatest, and (iii) assess if the spatial patterns of degradation can be linked to
120 climatic variables and properties of the different sites across the region. To achieve these objectives, we
121 combined large-scale regional analysis with higher resolution site-specific analysis of patterns in subsidence,
122 using a combination of datasets - satellite (Sentinel-1) InSAR; occupied airborne optical and LiDAR data; and
123 snow depth, precipitation, and temperature time-series from meteorological stations across the region.

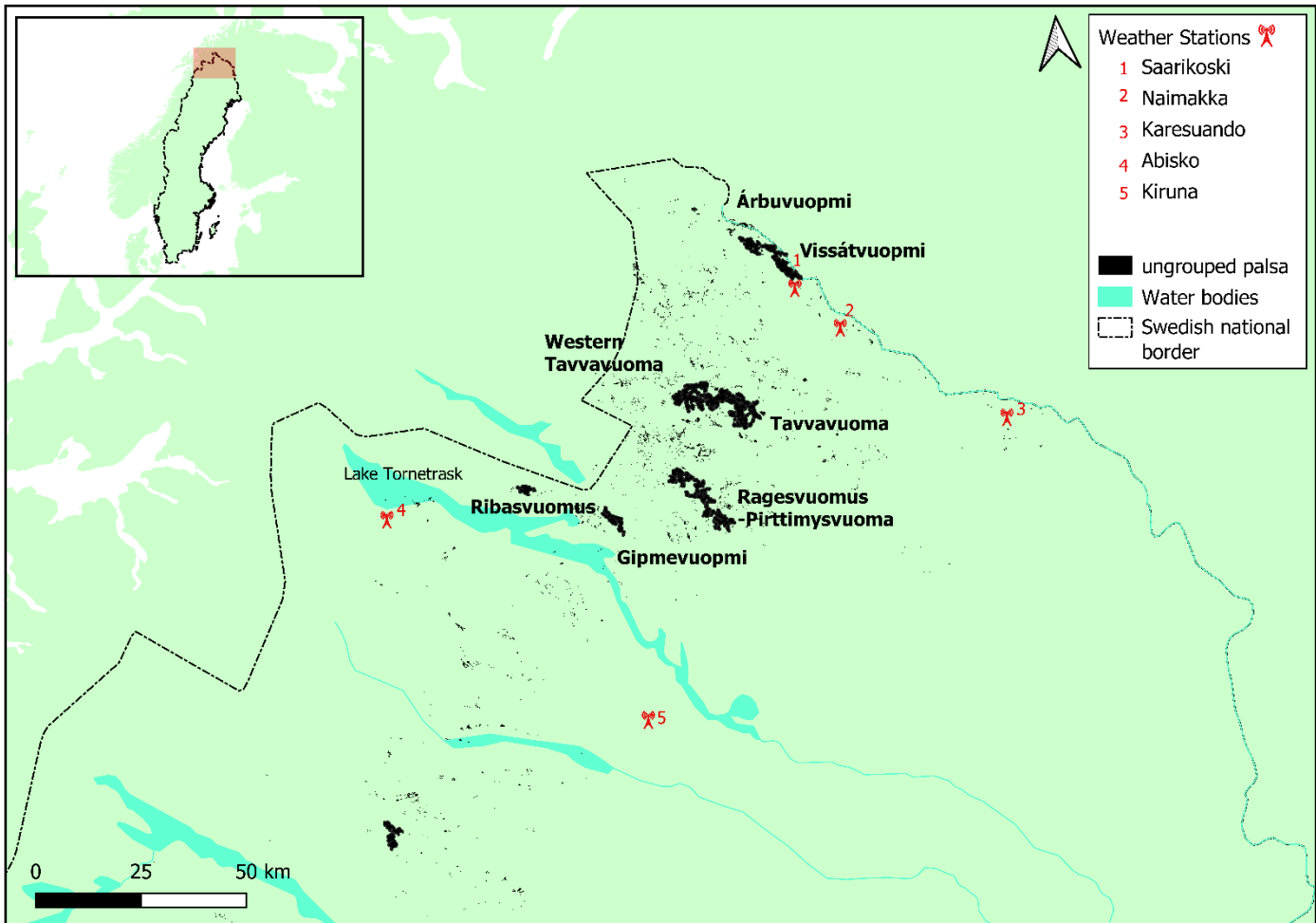
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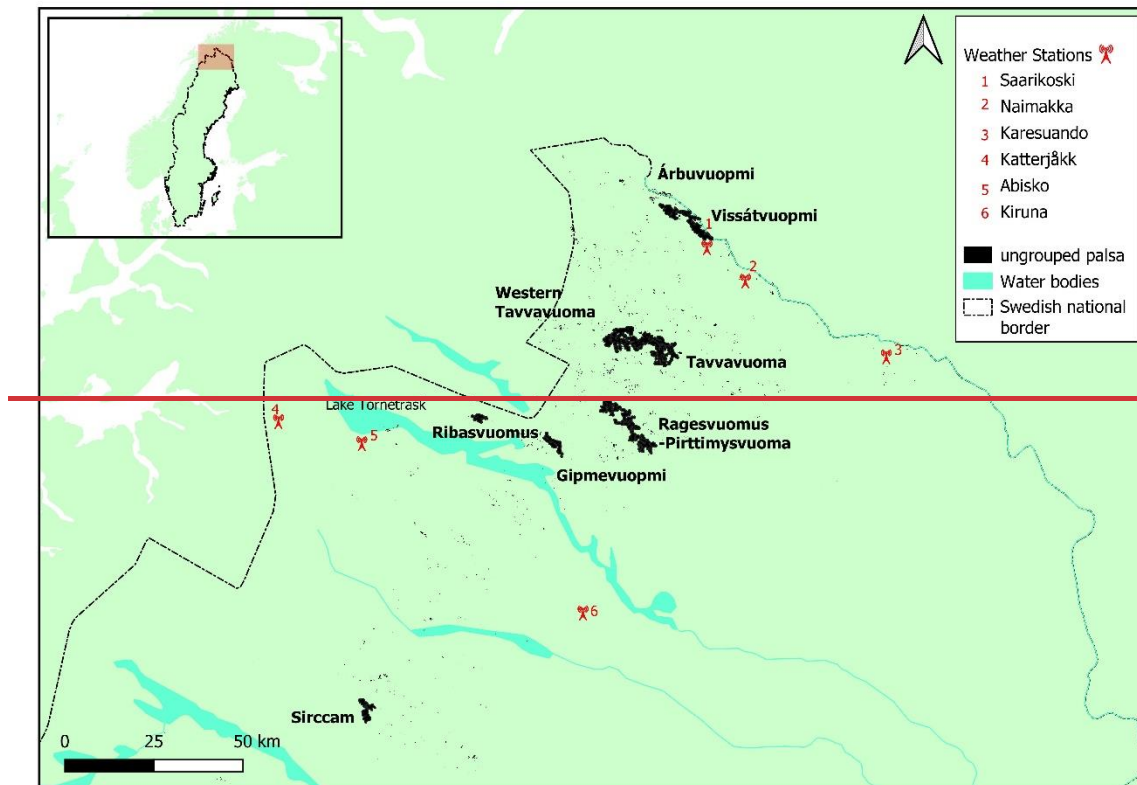
125 2.0 Methodology

126 2.1 Study area

127 This study focused on the northern part of Sweden; a region containing **palsa-peatlands**, located between 68.84-
128 67.64° N and 18.71-21.19° E. The **palsas_peatlands** of the region are confined predominantly to valley bottoms
129 in an elevation range between ca. 350 and 590m asl (Fig. 1). The rest of the study area region is comprised of
130 forests and/or mountain land covers (Siewert, 2018; [Åkerman & Johansson, 2008](#)). Of all the palsas in the
131 region, the eight largest **palsa-peatlands** complexes **of concentrated palsa** -range between 50 and 273ha in area
132 (Table 1). These were located across the region, which covers a ca. 20,000km² area, with the largest palsa sites
133 located **in the north-western parts of the region**. Smaller **palsas_peatlands** occur scattered in distribution right
134 across the region. The climate varies **across the region** from north to south (www.smhi.se). The mean January
135 and July temperatures in Karesuando in the northern part **of the study region** is -16 and 12.8°C, respectively,
136 while in Kiruna, slightly further south, the mean **JanuaryJanuary**, and July temperatures is -11.6 and 13.4C
137 (1991-2020 average). Mean annual precipitation is 443 and 560mm in Karesuando and Kiruna, respectively.

138





139

140 *Figure. 1: Map of the palsa peatland complexes in Sweden which were investigated in the study focusing on the*
 141 *eight named palsa peatlands complexes. The black regions show all the palsa which has been reported to exist*
 142 *(Backe, 2014,) with the larger named areas displaying the 250m buffers around the palsa areas which have*
 143 *created continuous expanses. The black regions show where 250m buffers around the palsa areas have created*
 144 *continuous expanses (Backe, 2014). Meteorological station positions used in the study are also indicated.*

145 A previous national palsa mapping dataset provided raster cells at a spatial resolution of 100m, with the % palsa
 146 cover computed and a 250m buffered output to provide continuous palsa area outputs (Backe, 2014). This afforded
 147 analyses at a spatial resolution suitable for analysis with Sentinel-1 yet provide practical representation of the
 148 condition of the palsa in this region. All these data were analysed in this study, but the eight largest continuous
 149 areas of these palsa (Backe, 2014) were focused on, hereon in referred to as palsa complexes, a term reflecting
 150 their mosaic nature of raised palsa and/or peatland plateau, interspersed with lower lying fen or thermokarst
 151 areas. These eight sites account for the majority of the palsa areas in Sweden, the sites are listed in Table 1 along
 152 with some associated information on their status and total and raised palsa plateau areas.

153

154

155

156 We selected larger palsa areas of the region to focus our analysis. This was in line with focus areas by the mapping
 157 of palsas undertaken as part of a previous national palsa peatland mapping effort (Backe, 2014). The resultant
 158 palsa peatland mapping dataset has a spatial resolution of 100m, with the % palsa cover for each pixel computed,
 159 and these pixels given a 250m buffer to produce continuous area outputs. The eight largest continuous areas of
 160 these palsa peatlands from the national palsa mapping dataset were selected for this study (Backe, 2014), hereon
 161 in referred to as palsa complexes, a term reflecting their mosaic nature of raised palsa plateau, interspersed with
 162 lower lying fen or thermokarst areas. This afforded analyses at a spatial resolution suitable for analysis with
 163 Sentinel 1 yet provide practical representation of the condition of the peatland in the region. These eight sites
 164 account for the majority of the palsa peatland areas in Sweden, the sites are listed in Table 1 along with some
 165 associated information on their status and total and raised palsa plateau areas.

166

167 *Table 1: Information on the major palsa complexes analysed in this paper (Backe, 2014). The protection status*
 168 *means no or limited direct anthropogenic activities that may influence palsa degradation. Total site area is*
 169 *calculated from the total number of 100m × 100m palsa pixels at each site - these pixels have associated*
 170 *percentages for how much of the 100m x 100m area is palsa. The average of these percentages for each site*
 171 *displays the palsa density at each site. These percentages are then used to calculate the “total palsa area” for*
 172 *each site based on the original report estimates.*

Site Name	Protection Classification	Total site area (ha)	Average extent palsa in these areas (%)	Total palsa area (ha)	LiDAR Collection year	Central location (Latitude, Longitude)
Árbuvuopmi	Not protected	327	26.3	86.06	2018, 2016	21.03464, 68.83842
Vissátvuopmi	Not protected	867	31.6	273.75	2015, 2018	21.19497, 68.79412
Tavvavuoma	EU Natura ae 2000 SPA, SAC. Site of National Importance for Nature conservation	1719	15.8	271.25	2018	20.85043, 68.51132
Western Tavvavuoma	EU Natura ae -2000 SPA, SAC. Site of National Importance for Nature conservation	813	13.0	105.74	2018	20.57727, 68.53953
Gipmevuopmi	Pristine mountain forest, Nature reserve, EU Natura ae 2000 SCI	303	23.0	69.62	2013	20.09767, 68.28377
Ragesvuomus-Pirttimysvuoma	Pristine mountain forest, Nature reserve, EU Natura ae 2000 SCI	881	6.55	57.74	2013	20.48660, 68.3741
Sirccam	EU Natura ae 2000 SCI	397	12.8	50.70	2015	18.71528, 67.64537
Ribasvuomus	Pristine mountain forest, Nature reserve, EU Natura ae -2000 SCI	216	23.2	50.13	2014	19.60100, 68.36116

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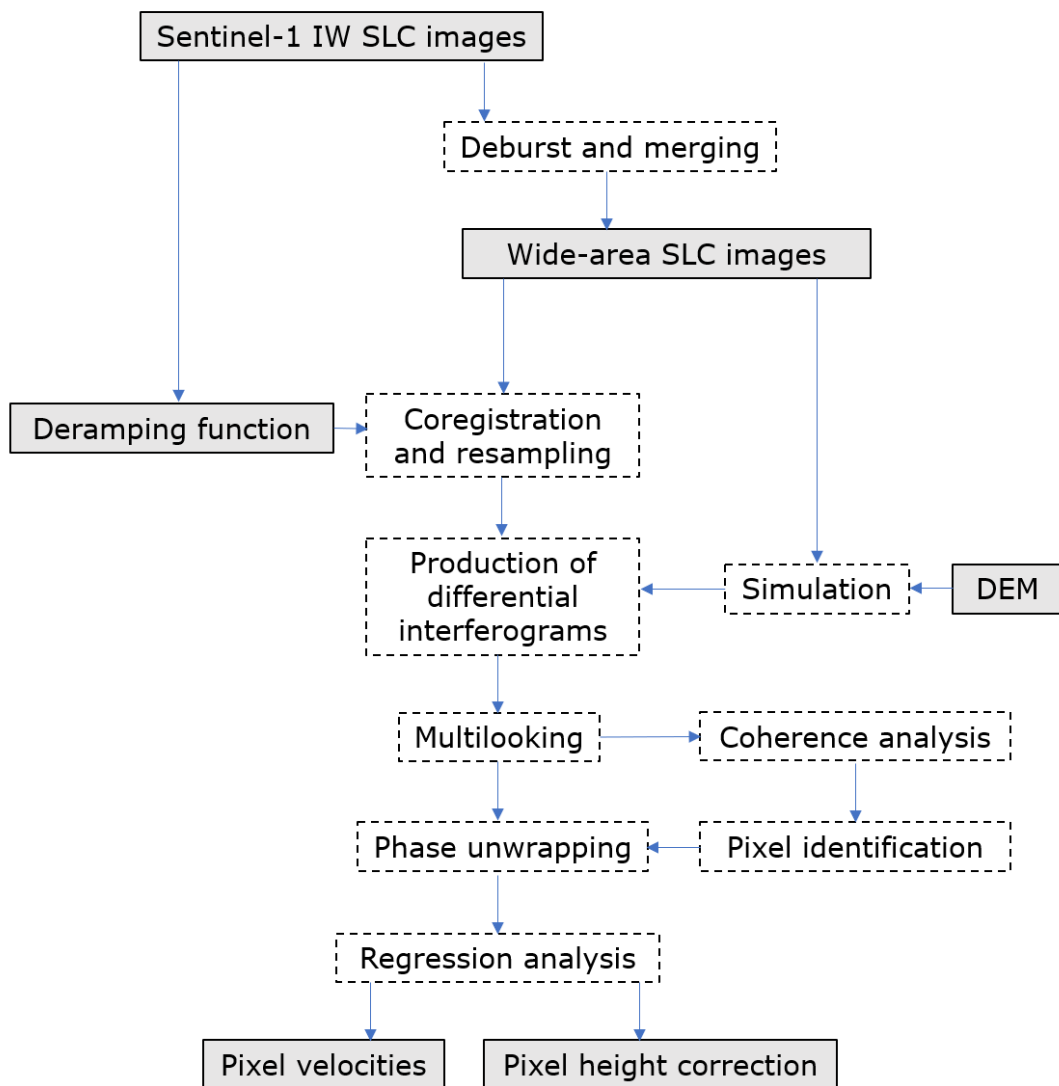
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175 2.2 Datasets

176 The InSAR-derived dataset of surface motion over this northern Sweden region of study was calculated for the
 177 period between 2017 to 2021, from single look complex ([SLC](#)) C-band SAR data, captured in Interferometric
 178 Wide ([IW](#)) Swath mode by the Sentinel-1 constellation (European Union’s Copernicus Programme; Torres et al.,
 179 2012). SAR data input were from the thaw season when there was minimal coverage of snow and ice (i.e., between
 180 April and October in each year). Data from descending tracks 168 and 66 were used to cover the target area. Four
 181 stacks were processed independently with one from track 168 and three from track 66, which was split into a
 182 northern, middle, and southern subsets. The APSIS (formerly ISBAS) method (Sowter et al., 2013; Sowter et al.,
 183 2016) was used to characterize surface motion which relaxes the need for consistent phase stability and therefore
 184 enables near-complete spatial and temporal coverage over vegetated surfaces (Alshammari et al., 2020;
 185 Alshammari et al., 2018; Bradley et al., 2022; Cigna & Sowter, 2017; Gee et al., 2017; Sowter et al., 2016),
 186 including those found across snow-free permafrost regions.

187 InSAR processing [of Sentinel-1 IW SLC imagery involves a series of steps summarised in Figure 2. Firstly,](#)
 188 [deburst and merging involved combining individual sub-swaths into a single wide-area SLC product. Secondly,](#)
 189 [the process of deramping produced a product where the problem of an ambiguous and rapidly changing phase](#)
 190 [with azimuth was solved for: a specific deramping function is available for Sentinel-1 data \(ESA, 2015\). Next](#)
 191 [step](#) involved the co-registration of each Sentinel-1 image to a common slant range coordinate system and multi-
 192 looking of data by factors of [5 m⁷](#) in range and [20 m by²](#) azimuth. This produced a dataset with an approximate
 193 spatial resolution of 20m × 20m. Using a perpendicular baseline of 250m and maximum temporal baseline of 183

194 days ~ 2100 interferograms were generated per stack. The temporal baseline was chosen to balance the need to
 195 reduce the baseline to minimise phase ambiguities and best maintain coherence across the region, whilst also using
 196 a baseline long enough to generate season-to-season pairs over consecutive years. This is required over permafrost
 197 regions to capture more subtle trends of surface motion during the thaw period (de la Barreda-Bautista et al., 2022;
 198 Liu et al., 2010). The interferograms were unwrapped using a modified version of the SNAPHU algorithm (Shen
 199 et al., 2002), which converts circular phase data into a linear measure of deformation. The algorithm was modified
 200 in order to allow the ability to parallelise and to spread the calculation across multiple cores (Chen and Zebker,
 201 2002). ~~by ... because ...~~ The multi-annual average velocity was calculated for pixels which maintained a
 202 coherence greater than 0.45 in a minimum of ~ 650 interferograms, with respect to stable reference points located
 203 in the town Kautekenio (N°69.00, E°23.04) for track 168 and Narvik (N°68.44, E°17.42), Kvikkjokk (N°66.95,
 204 E°17.72), and Rognan (N°67.09) for the subsets of track 66. The line-of-sight measurements were converted to
 205 vertical surface displacement using a cosine correction and finally mosaicked into a single deformation product.
 206 Localised UAV studies at sites in Sweden have verified the ~~accuracy-ability to of using~~ InSAR as a tool to
 207 monitor permafrost degradation (de la Barreda-Bautista et al., 2022). ~~We summarise the steps for InSAR data~~
 208 ~~processing in Figure 2.~~



209
 210 *Figure 2: A flowchart summarising the steps undertaken for InSAR processing using the APSIS method to create*
 211 *a surface motion product from Sentinel-1 IW SLC imagery. Boxes shaded grey represent data sets, boxes with*
 212 *dotted borders represent processing steps. Modified from Sowter et al. (2016).:*

213 In order to interpret the resultant surface motion dataset produced by the APSIS InSAR method, two sets of
 214 additional data were sourced: (i) higher resolution remote sensing data and (ii) meteorological data. The former

215 included orthophotos captured of the eight target areas by occupied airborne surveys commissioned by the
216 Swedish ~~h~~ Mapping, Cadastral and Land Registration Authority Survey (www.lantmateriet.se; © Lantmateriet).
217 The ~~orthophotos~~ orthophotos (Lantmateriet, 2021) have scenes ~~were panchromatic, with each scene~~ covering a
218 5km × 5km area, at a 0.5m spatial resolution, the majority were captured in 20~~14~~¹⁶, although gaps were filled
219 with imagery from 201~~08 for two sites~~^{08 and 2008}. The Swedish National Digital Elevation Model (DEM), was also
220 used in this study. The DEM was derived via occupied airborne LiDAR data capture ~~between 2013 in 2016~~ and
221 ~~2018 (Table 1)~~ processed to compute elevation at 2m spatial resolution across Sweden (www.lantmateriet.se; ©
222 Lantmateriet). The orthophotos and DEM provided elevation and landscape characteristics (geomorphic features)
223 for use in this study. The meteorological data was captured by the Swedish Metrological and Hydrological Institute
224 (www.smhi.se) at meteorological stations across the region. Specifically, the air temperature, precipitation, and
225 snow depth data, were sourced and used from specific stations, i.e., those located closest to the palsa complexes
226 under investigation namely at ~~Katterjåkk~~, Abisko, Kiruna, ~~and~~ Karesuando, Saarikoski, ~~and~~ Naimakka (Fig. 1).

227

228 2.3 Data analyses

229 2.3.1 Surface motion statistics

230 The ASPIS InSAR surface motion dataset was clipped to the 100m x 100m spatial resolution of the whole palsa
231 dataset and separately to the eight palsa complexes~~resampled using the mean value from the original 20m × 20m~~
232 to match the 100m × 100m spatial resolution of the palsa peatland dataset which makes up the eight palsa
233 complexes (Backe, 2014). From this the frequency distributions of ASPIS InSAR surface motion at these eight
234 palsa complexes, and over all individual palsa peatland ~~pixels~~ raster cells in the region, were produced. Using
235 these data, the maximum and minimum rates of surface motion at each site was determined, as well as the sum of
236 the pixels with palsas that showed subsidence. These derived data relating to surface motion were further
237 interpreted using the orthophotos and DEMs, supported by the meteorological data.

238 2.3.2 Roughness thresholds

239 The DEM tiles were joined together and clipped to the eight palsa complexes. Following this, the degree of
240 elevation roughness was calculated, via the native topographic roughness index function (Riley, DeGloria, &
241 Elliot, 1999). This roughness index was thresholded at > 0.5 to provide a visual depiction of palsa landform edges
242 in the otherwise typically even terrain of the valley bottoms where the palsas occur. The roughness data was
243 visually compared to the orthophotos from a subset of areas to assess its potential for delineating palsas and this
244 allowed us to determine a threshold value that connected these continuous terrain variables to the specific features
245 of the palsa complexes, such as the raised mound structure of the palsa – so-called palsa mounds (Franklin, 2020).
246 Hillshade was also calculated via the native QGIS function using the default formula, which uses a lighting effect
247 to visualise the roughness of the terrain from differences in local elevation (QGIS, 2022). The roughness,
248 hillshade, and elevation outputs were overlaid on the mapped palsa tiles to provide higher resolution visual
249 interpretation.

250 2.3.3. Causes of surface motion

251 To test for the causes of surface motion palsa ASPIS InSAR surface motion was compared against roughness,
252 elevation and palsa percentage provided with the palsa raster cells (Backe, 2014). The roughness, ~~and~~ elevation
253 and InSAR outputs were ~~also~~ resampled to the resolution of the mapped palsa tiles (100m x 100m) to enable
254 statistical comparison. The zonal statistics tool was used to extract mean average values from the resulting
255 roughness and elevation outputs for the 100m spatial resolution mapped palsa tiles.

256 To analyse the relationships between surface motion, roughness and percent palsa in each 100m by 100m pixel
257 stratified by palsa complex, SciPy statistics (Virtanen et al., 2020) was used to obtain Pearson's correlation
258 statistics with an alpha value of 0.05 used to test for significance. Pandas (McKinney, 2011) and NumPy (Harris
259 et al., 2020) were used for data management. All scripts are available on the project GitHub
260 (https://github.com/SamValman/Permafrost_Sweden).

261 2.3.4 Climatic factors

262 Mean annual, maximum, and minimum daily air temperature, precipitation, and depth of ground snow for the
263 period 2000 to 2022 from the meteorological station nearest to a correspondent palsa complex were extracted and
264 analysed. The Naimakka station did not ~~provide record~~ snow depth and the Saarikoski station did not ~~provide~~
265 ~~record~~ air temperature, however, it was deemed that at the regional scale of this study these sites were sufficiently
266 close together (18km) to be ~~interchangable~~interchangeable. Subsequently, data was averaged to provide an annual
267 measurement of each meteorological variable for each station/palsa complex. Due to incomplete
268 ~~meteorologica~~meteorological datasets, a longer-term record of the meteorological variables was not possible for
269 all sites. However, long-term climate data (>100years) was available from three meteorological stations in the
270 region: namely, Karesuando, Kiruna, and Abisko. This data was used to assess temporal variability in annual,
271 winter (December, ~~January~~January, and February (DJF)) and summer (June, July, and August (JJA)) temperature,
272 precipitation and snowfall since the start of records across the region. Descriptive statistics (mean, minimum,
273 maximum and inter-quartile range) were produced to express the regional differences between these sites. Lastly,
274 to complement the point based meteorological (both weather and climate) data, we used modelled permafrost
275 probabilities based on climatic conditions to explore relationships between climatic conditions (Obu et al., 2018)
276 and subsidence rates (~~Obu et al., 2018~~). In this context, it is worth noting that there may be a mismatch between
277 the modelled permafrost distribution and permafrost in palsa ~~areas~~peatlands as this can, in some areas, be a relic
278 of cooler climatic conditions. ~~The~~We used the mean values from the roughness and InSAR data to resample to
279 100m spatial resolution in line with the permafrost raster cells and spatially joined the permafrost probability
280 layer, taking the mean value where the 100m cell straddled multiple permafrost probability cells. ~~these data on~~
281 ~~permafrost probability were used to resample to a 100m spatial resolution to enable comparison with the other data~~
282 ~~sets.~~

283 ~~To analyse the relationships between surface motion, roughness and percent palsa in each 100m by 100m pixel~~
284 ~~stratified by palsa complex, SciPy statistics (Virtanen et al., 2020) was used to obtain Pearson's correlation~~
285 ~~statistics. Pandas (McKinney, 2011) and NumPy (Harris et al., 2020) were used for data management. All scripts~~
286 ~~are available on the project GitHub (https://github.com/SamValman/Permafrost_Sweden).~~ The relationship
287 between the meteorological variables both over the last two decades at the weather stations closest to the palsa
288 complexes and duration of the climate record at the three weather stations with the longest data series were
289 assessed using linear regression analysis in Genstat (VNS Ltd). Assumptions of normality and homogeneity of
290 variance of the residuals were assessed using residual plots in Genstat. Some of time series were incomplete, in
291 these instances the analysis was conducted using the slightly shorter time series (see fig. 8).

292

293 **3.0 Results**

294 3.1 Surface motion

295 The ASPIS InSAR-derived surface motion outputs for the time-period of interest (2017-2021), ranged between
296 -9.9 and 7.7 mm yr⁻¹ across all of the palsa raster cells measured in northern Sweden, with a mean of 0.05, median
297 of 0.2 and range of 17.7 mm yr⁻¹. Focusing solely on the eight palsa complexes provided greater insight and
298 excluded the most extreme uplift values from scattered individual palsas (Table 2). 69% of results were within
299 MSE of changing ground motion direction.

300

301

302

303

304 *Table 2: InSAR subsidence and uplift measurements of the palsa complexes defined in Figure 1 and Table 1.*
305 *The total palsa area were used to isolate and extract ASPIS InSAR measurements of surface motion at each of*
306 *the eight sites.*

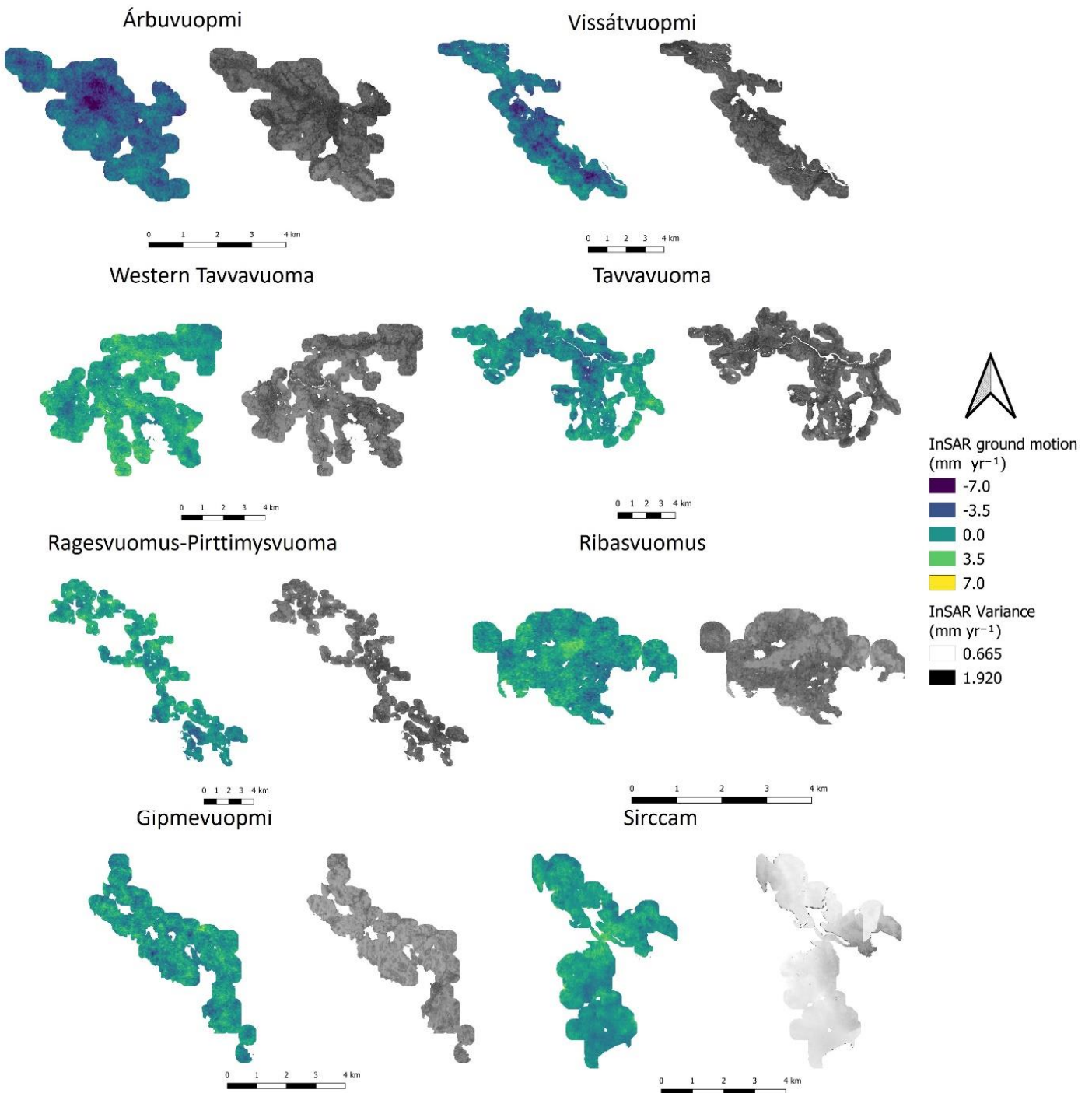
Site	Max subsidence	Max uplift (mm yr ⁻¹)	Subsiding area (ha)	Area subsiding >3.5 mm yr ⁻¹	Mean standard error (mm yr ⁻¹)
------	----------------	-----------------------------------	---------------------	---	--

	(mm yr⁻¹)		(ha)		
Árbuvuopmi	-9.9	1.7	321.3	138.4	<u>1.5</u>
Vissátvuopmi	-8.9	3.5	796.2	204.8	<u>1.5</u>
Tavvavuoma	-6.4	6.6	1009.4	50.9	<u>1.4</u>
Western					
Tavvavuoma	-5.1	6.3	215.0	1.0	<u>1.4</u>
Gipmevuopmi	-6.9	6.3	117.2	1.8	<u>1.2</u>
Ragesvuomus-	-5.9	5.7	358.6	7.4	<u>1.4</u>
Pirttimysvuoma					
Sirccam	-3.1	5.4	135.3	0.0	<u>0.9</u>
Ribasvuomus	-6.5	5.5	93.6	0.7	<u>1.3</u>

307

308

309 The spatial plots of surface motion for each palsa complex displayed in Figure 3, illustrates **patterns of a spatiality**
 310 **in terms of** surface motion (both subsidence and uplift and associated variance) across this northern Sweden
 311 region. This is evident both within the palsa complexes and between the complexes.



312
 313
 314

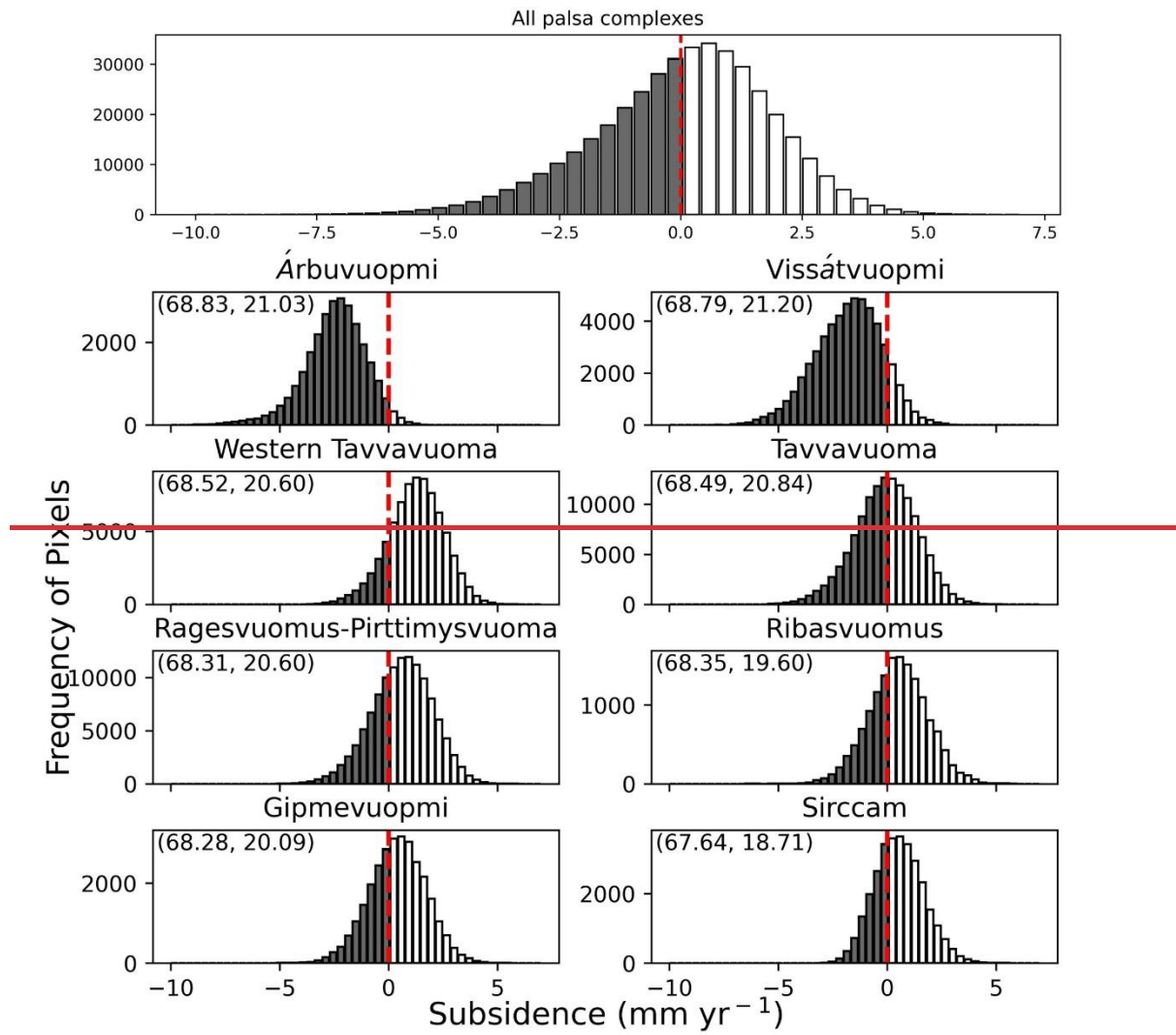
315 *Figure. 3: Palsa ground motion measured using Satellite InSAR **from 2017 to 2021**, showing differing levels of*
 316 *degradation across the eight study sites. Sites are ordered by their latitudinal position. Negative values*
 317 *correspond to subsidence. Note that in order to plot continuous areas the scenes shown are the palsa **peatland***
 318 *area plus a 250m buffer around each 100m × 100m **pixel-raster cell** that cover a minimum of 1% palsa (Backe*

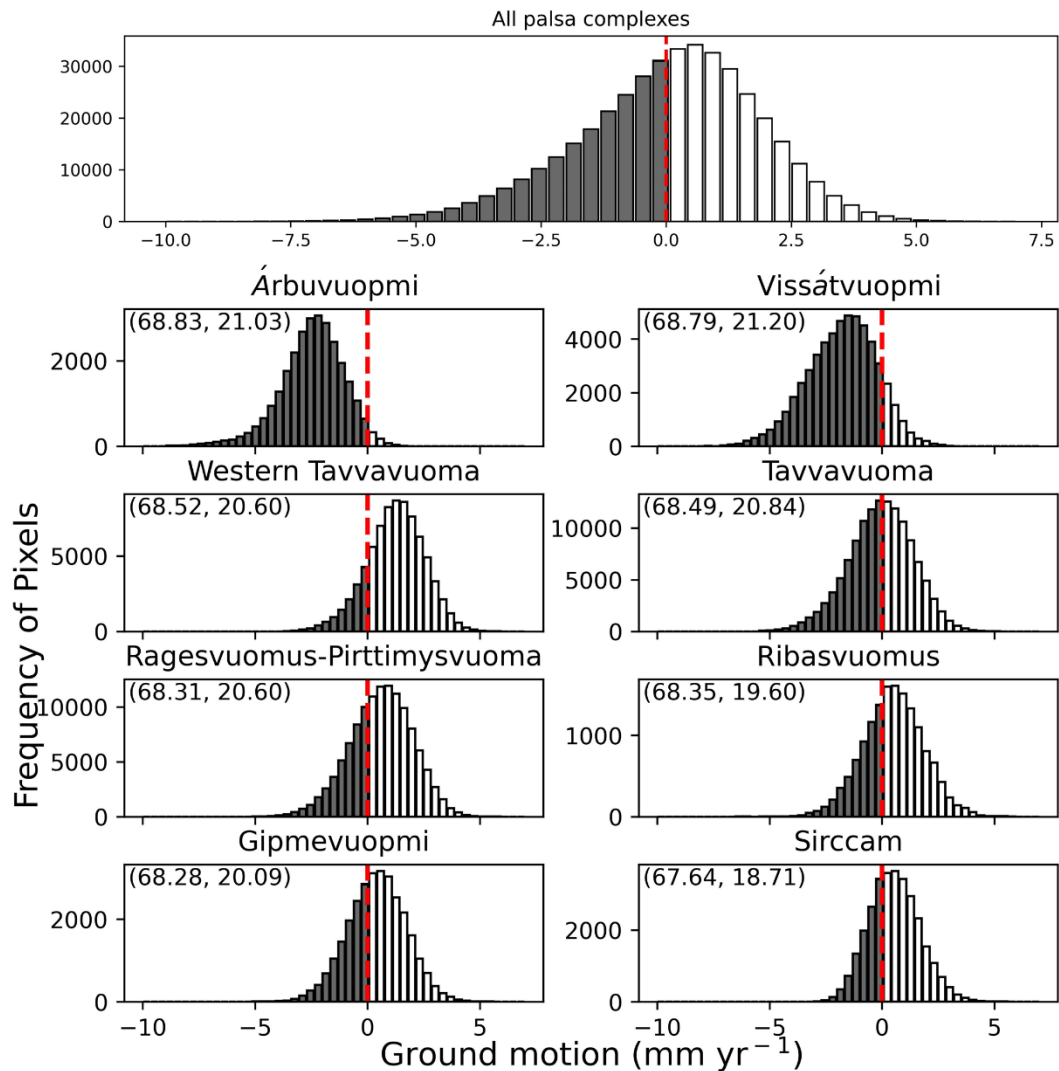
319 2014). This means that areas of non-palsa peatland and some areas with mineral soil are included in the figure.
320 ASPIS InSAR variance were less than 1.5mm yr^{-1} in over 90% of pixels.

321

322 ~~Subsidence was recorded in just under half of the pixels all the eight palsa complexes (Table 2).~~ Across the target
323 sites 3046.6ha (Table 2) out of the total site area of 5523ha (Table 1) were subsiding, which equates to ca 55% of
324 the total palsa complexes' area. Out of the subsiding parts of the palsa complexes, 405ha were subsiding at rates
325 $>3.5\text{mm yr}^{-1}$ at near gaussian distribution. However, it is evident from the frequency distribution plots, that it is in
326 the palsa complexes in the far north of the region that subsidence dominated the surface motion measured (Table
327 2, Fig. 4). At Vissátvuompi and Árbuvuopmi 98 and 92% of the palsa complexes were subsiding with maximum
328 subsidence rates of -9.9 and -8.9mm yr^{-1} , respectively. The measured area affected by high subsidence rates of
329 between ($>3.5\text{mm yr}^{-1}$) were 204.8ha and 138.4ha at Vissátvuompi and Árbuvuopmi, respectively. This means
330 that ca. 30% of the total combined area of these two sites (1194ha) is in the highest range of subsidence. The high
331 degree of palsa subsidence at Vissátvuompi and Árbuvuopmi was confirmed by field observations at these sites
332 (Sofie Sjogersten, pers. Obs.): Both sites showed signs of active lateral erosions, large scale subsidence and
333 thermokarst formation. The more southerly sites also show subsidence, although ground motion rates were much
334 ~~lower~~more stable, with the -1 and 1mm yr^{-1} range being most common (Fig. 4). Areas further to the south and
335 west showed signs of uplift, particularly the western parts of Tavvavuoma and Ribasvuomus with maximum rates
336 of uplift of 6.3mm across some smaller parts of these sites. However, all sites have some degree of subsidence,
337 albeit at a lower rate compared to the heavily subsiding northern sites.

338



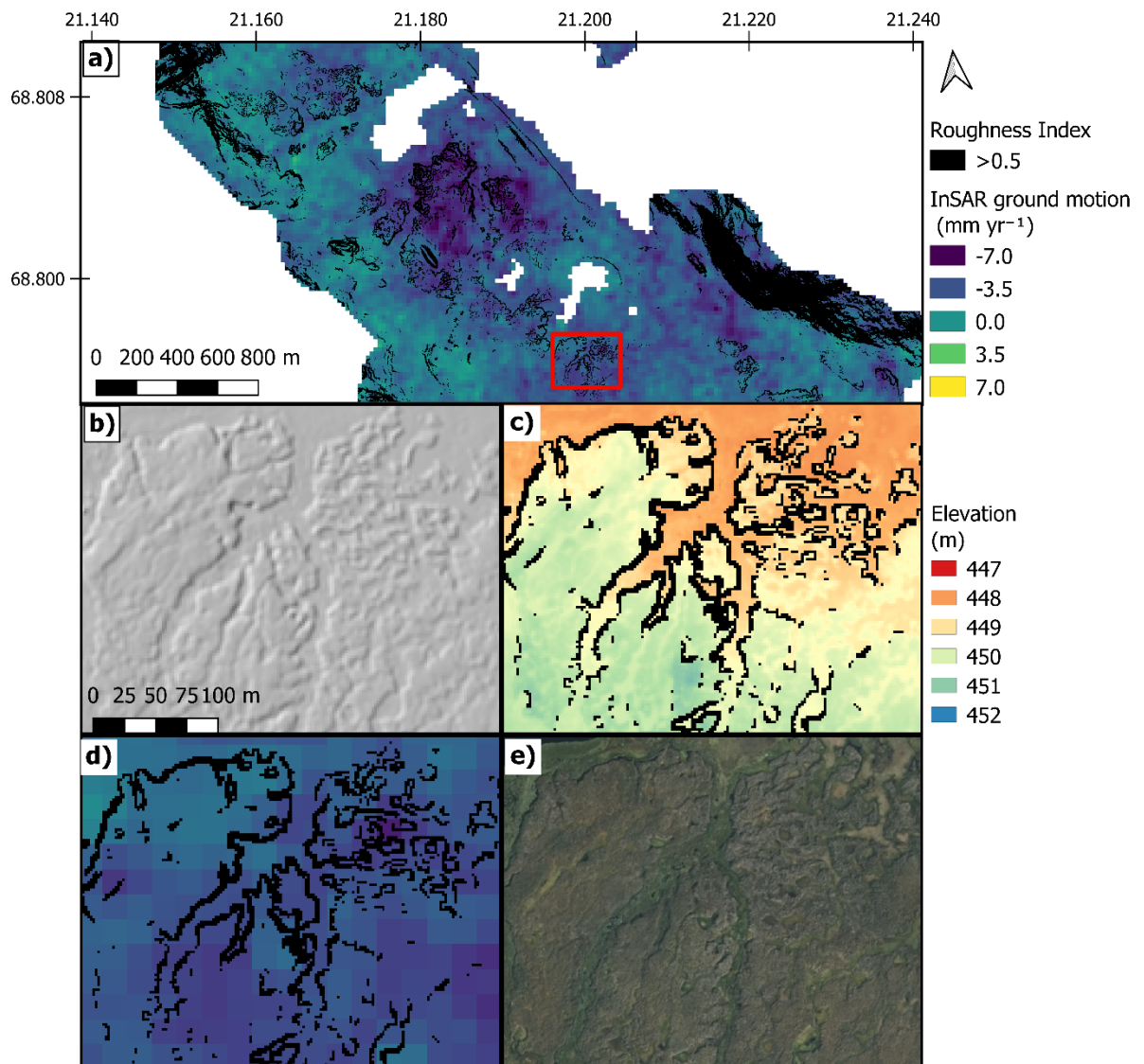


340

341 *Figure 4: Distribution of 20m × 20m ASPIS InSAR pixels within each of the palsa complexes in this study and*
 342 *the overall trend of the dataset according to the distribution of pixel moving in a particular direction and a given*
 343 *rate. Shaded areas with negative values correspond to subsidence. The dashed central lines indicate pixels in*
 344 *stable areas with no motion. Central point latitude and longitude is provided for each site in brackets for each*
 345 *site.*

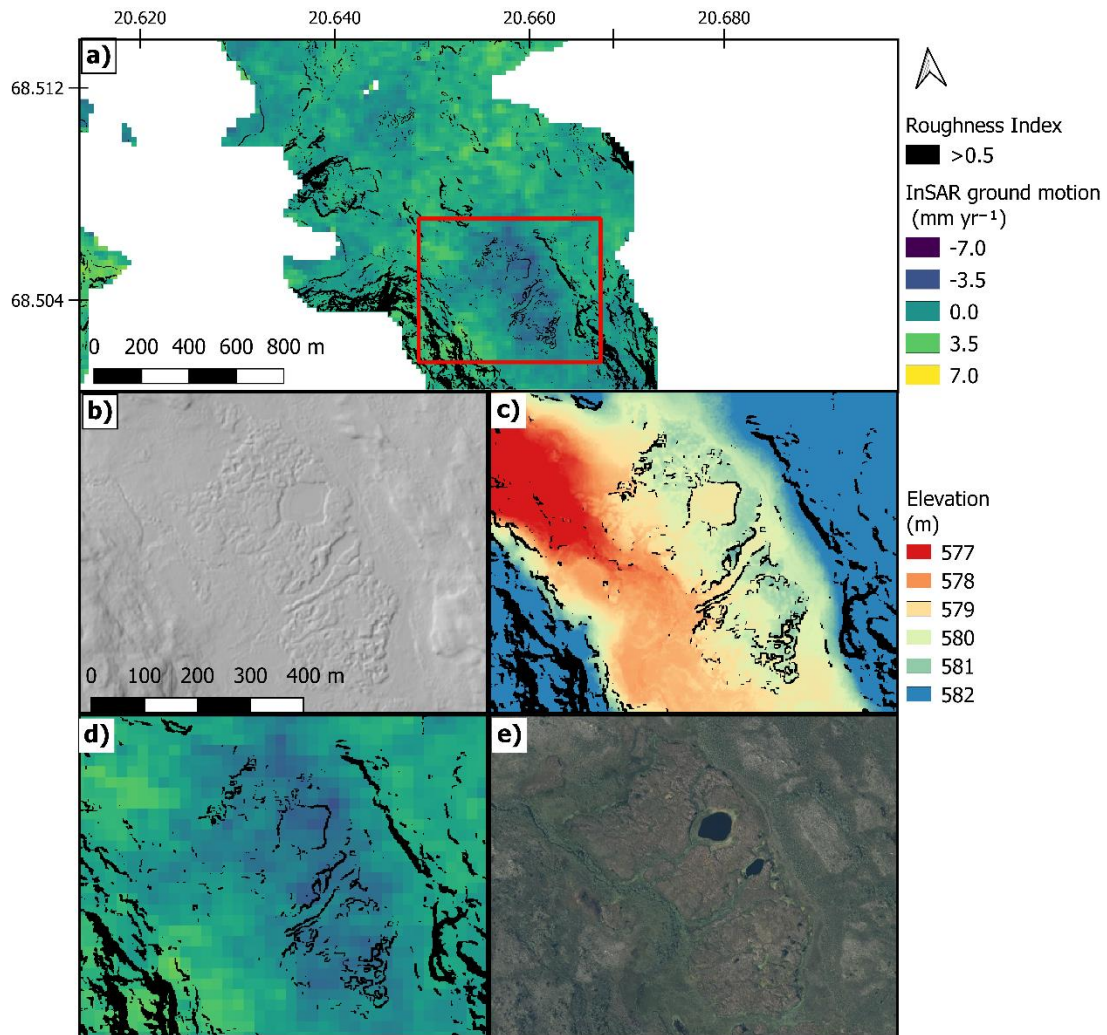
346 3.2 Topographic drivers and indicators

347 Calculating the roughness index from the DEMs at each palsa complex enabled differentiation of palsa from
 348 surrounding lower lying and flat fen areas. Representative example complexes are shown in Figures 5 and 6 -
 349 Vissátvuopmi and Western Tavvavuoma. Overall, the palsa complexes to the north (e.g., Fig. 5b, c) display a
 350 more pronounced topography across the focus areas than the more south-westerly ones (e.g., Fig. 6b, c). There
 351 was clear correspondence between density of palsa and subsidence, i.e., areas with more palsa showed more
 352 subsidence (Fig. 5a, d). Furthermore, the palsa complexes showed greater elevation variation compared to
 353 surrounding fen areas and were more densely clustered to the north than in the more south westerly sites. These
 354 features spatially coincided with higher subsidence. Substantial within site variability in subsidence was evident,
 355 where the pixels with the highest subsidence rates being clustered together and following landscapes features,
 356 e.g., palsa plateau edges. It was evident that many separate palsa complexes in an area resulted in a high degree
 357 of elevation change, causing a high roughness index. In turn, areas with high roughness have the greatest
 358 subsidence (Fig. 5,6). Visual comparison between orthophotos and roughness showed that areas of high roughness
 359 corresponded well with areas of severe permafrost degradation (as indicated by lateral erosion and thermokarst
 360 formation).



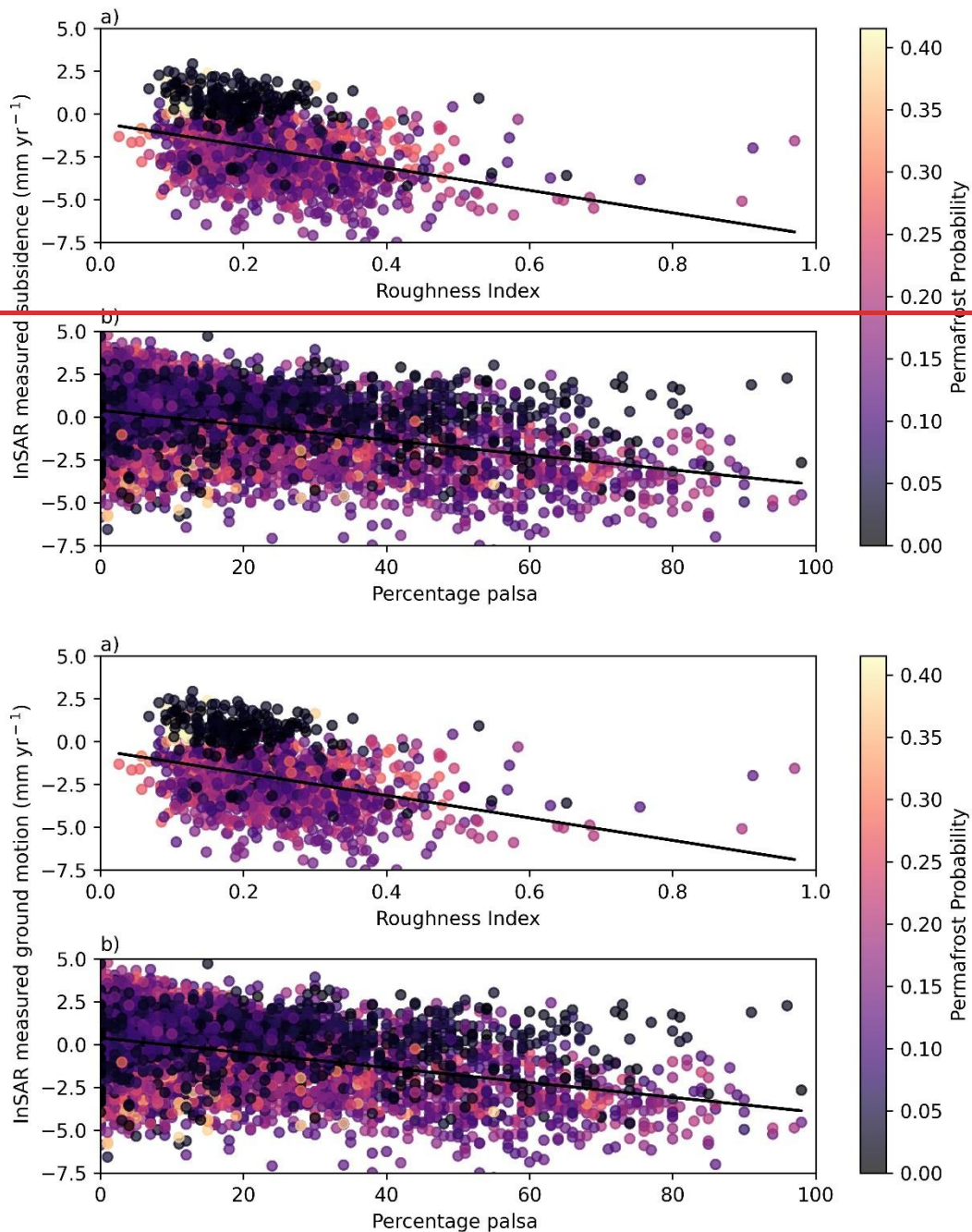
361

362 *Figure. 5: Visual analysis of Vissátvuopmi one of the sites where the most subsidence was found to be*
 363 *occurring. Evaluation of correspondance of hillshade DEM (b), DEM (c) and a close look at InSAR subsidence*
 364 *(d) with Palsa-palsa complexes suggested by roughness overlays and aerial imagery (e). The positioning of b,c,*
 365 *and d within the larger site (a) show bands of subsidence ~~is~~ proximal to roughness patches suggesting*
 366 *Palsapalsa.*
 367



368
 369 *Figure. 6: Visual analysis of Tavvavuoma which was found to have much lower levels of subsidence in*
 370 *comparison to more northern sites. Evaluation of correspondence of hillshade DEM (b), DEM (c) and InSAR*
 371 *subsidence (d) with Palsa complexes suggested by roughness overlays and aerial imagery (e). The positioning*
 372 *of b,c, and d within the larger site (a) show many less “bands” (linear arrangements of palsa across the image)*
 373 *of subsidence and potential palsa than Figure 4.*
 374

375 Regression analysis showed a relationship between roughness and subsidence as sites with greater subsidence
 376 were also found to have greater roughness (Fig. 7a). Higher percentage palsa in a location was linearly related to
 377 subsidence with the greatest subsidence found in areas with the highest percentage palsa cover (Fig. 7b). It was
 378 also clear that the modelled permafrost probability did not correspond to the percentage of palsa, i.e. pixels with
 379 100% palsa are in some instances predicted to have no permafrost (Fig. 7b).



380

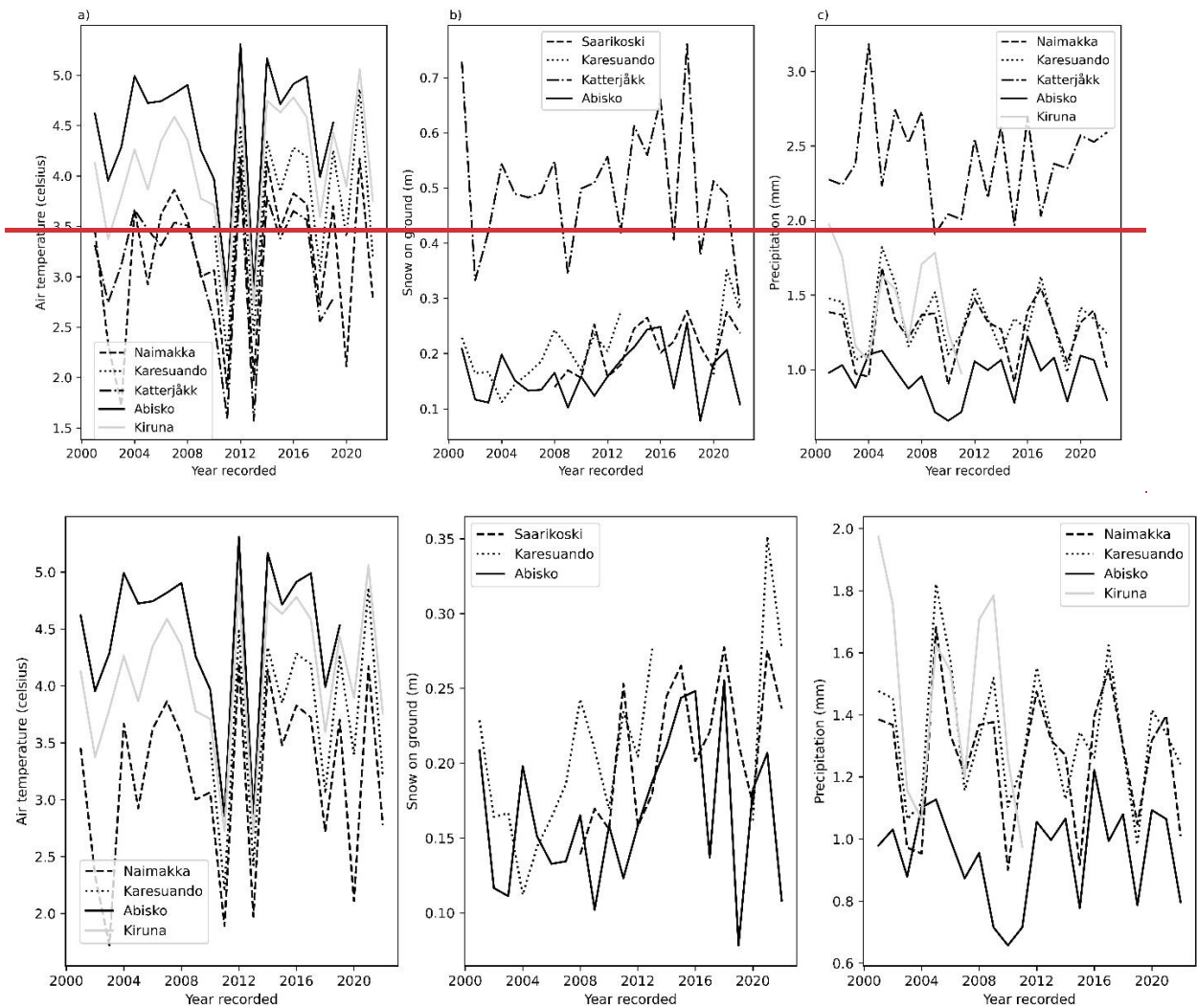
381

382 *Figure. 7: Relationship between a) the roughness index; $p < 0.001$, $R^2 = 0.35$ and b) percentage palsa in a*
 383 *pixel; $p < 0.001$, $R^2 = 0.41$ and subsidence. The colours indicated for each data point are the analysed*
 384 *probability (on a scale from 0 to 1) that an area would include permafrost, (Obu et al., 2018). Note that there is*
 385 *less data for the analysis of roughness as the roughness was characterized only for the eight study sites and not*
 386 *all palsa areas raster cells from (Backe, 2014). Roughness values from valley sides (which at time were included*
 387 *in the buffer areas) are not used in the figure.*

388 3.4 Meteorological trends

389 The analysis of the ~~metreological~~ meteorological data showed variability in both weather and climate across the
 390 study region in part reflecting the patterns in the subsidence data. The warmest minimum and maximum
 391 temperatures, -29.2 and 32.8°C respectively, were recorded for the palsa complexes north of Lake Tornetrask, i.e.
 392 Gipmevuomi and Ribasvuomus (Abisko weather station) (Fig. 1). The temperature in the area of Árbuvuopmi,
 393 Vissátvuopmi, and Tavvavuoma palsa complexes (Saarikoski/Naimaka and Karesuando weather stations) ranged
 394 between -39.4 and 30.5°C (Table 3, Fig. 8a). ~~The Katterjåkk weather station located in the mountains close to the~~

395 Norwegian border recorded the greatest annual snow depth measure of 229cm and a mean of 50cm. Note that in
 396 this far western part of the study area peatland were not present anymore. In contrast, the three other sites
 397 had comparable annual snow depth with a mean of 20-30cm (Table 3, Fig. 8b).



398

399
 400 Figure 8: a) Mean annual daily maximum temperature, b) snow depth on the ground, and c) daily
 401 precipitation at the meteorological stations in the study region (SMHI 2022).

402

403

404 Table 3: Temperature and snowfall descriptive statistics. The snow depth data are estimated from days with
 405 snow on the ground. Mean annual temperature and precipitation are averaged from 2000 to 2021. Maximum,
 406 minimum and the inter-quartile range are of daily maximum temperature and daily precipitation since 2000 are
 407 also shown. Some weather stations lack certain years but were considered to have adequate coverage for this
 408 task while two sites did not have sufficient data collection during the time period to be reliable and were shaded
 409 out.

Weather Station	Temperature (°C)				Snow depth (m)			Precipitation (mm)		
	Mean annual	Max daily	Min daily	IQR daily	Mean annual	Max daily	IQR daily	Mean annual	Max daily	IQR daily
Naimakka	-1.40	29.5	-38.2	15.7				456	50.8	1.0
Saarikoski					76.9	0.85	0.43	422	43.6	0.9
Karesuando	-0.70	30.5	-39.4	16.9	75.1	1.00	0.40	490	53.2	1.1
Katterjäkk	-0.32	29.5	-27.6	13	183.0	2.29	0.97	875	104.3	3

Abisko	0.53	32.8	-29.2	13.5	60.0	1.27	0.42	348	61.9	0.6
Kiruna	0.06	30.3	-30.6	15.6	5.3	1.13	0.45	545	53.1	0.9

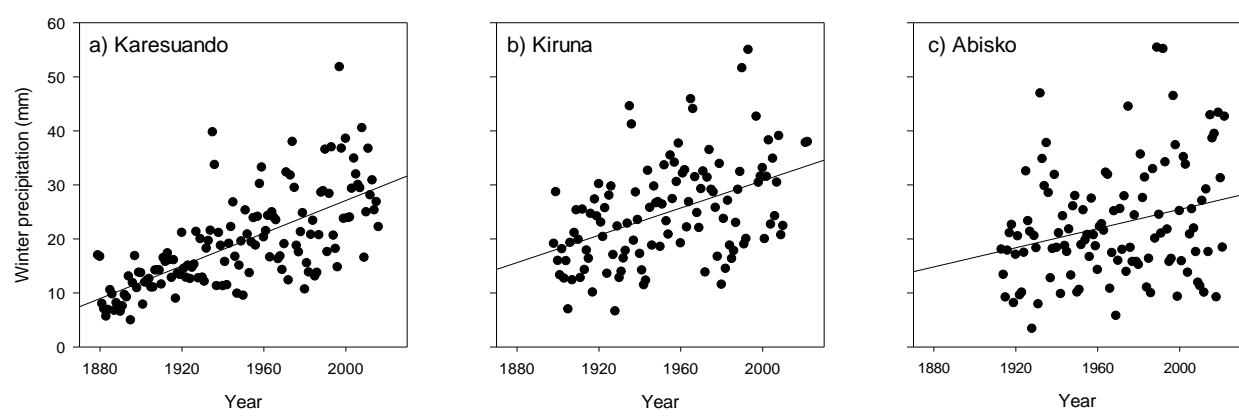
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411

412

413 There was no detectable difference in climatic trends among the meteorological weather stations since 2001 for
414 any of these sites ($p > 0.05$). In contrast, the longer-term climate records show a strong increase in winter
415 precipitation over the last 140 years at Karesuando, the northern most weather station of the three with long term
416 records available ($F_{1,136}=122.33$, $p < 0.001$; $\sigma_{\square}^2=47.0\%$; Fig. 9a). This long-term trend was also evident, albeit
417 less strong, in Kiruna ($F_{1,110} = 28.17$, $p < 0.001$; $\sigma_{\square}^2=19.7\%$; Fig. 9b). In Abisko, the pattern of increasing in
418 winter (DJF) precipitation was less clear ($F_{1,108}=8.29$, $p < 0.01$; $\sigma_{\square}^2=6.3\%$; Fig. 9c). Mean annual Snow depth,
419 temperature, and summer precipitation (JJA) did not show clear temporal trends (~~data not shown~~See
420 supplementary materials).

421



422

423 *Figure 9. Mean winter (DJF) precipitation over time at a) Karesuando, b) Kiruna, and c) Abisko, significant*
424 *trendlines are shown.*

425

426 **4. Discussion**

427 By way of satellite ASPIS InSAR-derived surface motion and associated spatial and statistical analyses, we
428 have demonstrated on-going, subsidence in the ~~palsas~~ peatlands of northern Sweden driven by a warming
429 climate. Based on the compelling agreement of subsidence with ~~palsa~~ landforms and their roughness, we
430 interpret this as permafrost degradation, i.e., thaw of the permafrost core within ~~palsas~~ and disintegration of
431 these landforms. This is in line with a wide range of literature (see introduction) and concurs with the local-scale
432 studies in the area undertaken using both satellite- and field-based methods (de la Barreda-Bautista et al., 2022;
433 Olvmo et al., 2020; Sannel, 2020; Sannel et al., 2016; Sannel & Kuhry, 2011). The findings also agree with, as
434 well as with, what is expected from the severe climate warming impacts on temperatures and precipitation noted
435 in the region (Hänsel, 2020; Irannezhad et al., 2017; Vikhamar-Schuler et al., 2016) and the modelled
436 predictions of total loss of permafrost across the region within decades (Fewster et al., 2022). We suggest that
437 the surface subsidence of the sample ~~palsa~~ complexes measured in this study, together with complementary
438 work in Norway (Borge et al., 2017), can be taken as evidence of ~~significant~~ substantial permafrost degradation
439 in ~~many all~~ palsa ~~peatland~~ areas across northern Sweden and therefore, likely to be also occurring across
440 northern Fennoscandia-

441 The processes driving the degradation of the permafrost, as measured by the ASPIS InSAR-derived subsidence
442 data, are complex. Although permafrost degradation was observed in all the ~~palsa~~ complexes, rates varied both
443 within and among ~~palsa~~ complexes (Table 2, Fig. 3 and 4). Overall, the InSAR subsidence data demonstrates a
444 north-south to south-north gradient in increasing degradation. This indicates that local factors, such as local
445 climate warming responses or permafrost temperature, determine the sensitivity of particular areas and that

446 regional climatic gradients play a role in the long-term trajectory of these ecosystems (Johansson et al., 2011;
447 Olvmo et al., 2020). In particular, winter precipitation is generally considered a strong predictor of permafrost
448 degradation due to the highly insulating properties of snow, preventing heat dissipation during winter (Olvmo et
449 al., 2020; Seppälä, 2011). This points to increased winter precipitation in the part of the northern most part of
450 study areas as a driver of the higher subsidence rates at the northern most palsa complexes (Table 2 and Fig. 8a).
451 Interestingly, climate data from the last two decades did not reveal strong differences in climatic conditions over
452 the area. This suggests that long-term trends combined with a buffered system reaction to change are driving
453 regional patterns in permafrost degradation.

454 It could also be the case that the observed north to south gradient of subsidence rates reflect different phases of
455 progression in an ongoing trend of permafrost degradation across the study region of northern Sweden. It is
456 plausible that the degradation process has progressed further at the more southern sites, reflecting higher
457 permafrost temperatures, and that as a result, subsidence rates have now slowed. All the while at the northern
458 sites, which still have a high cover of palsa: 26.3 and 31.6 % at Árbuvuopmi and Vissátvuopmi respectively,
459 show high subsidence rates. This is supported by research showing rapid permafrost degradation in the
460 southernmost palsa complexes in Sweden (Zuidhoff, 2002; Zuidhoff & Kolstrup, 2000) and in the area around
461 and to the south of Tornetrask, since the 1960's (Åkerman & Johansson, 2008; de la Barreda-Bautista et al.,
462 2022; Varner et al., 2022). However, permafrost degradation in palsa ~~s~~ ~~peatlands~~ ~~have~~ ~~s~~ progressed over longer-
463 time periods even in ~~the far northern Fennoscandia of Scandinavia~~. Here palsas' have decreased in areal extent
464 by 33– 71% over ca. 60 years, with more rapid contraction in recent years in Finmarkvidda, Norway and 54% in
465 Vissátvuopmi, northern most Sweden (Borge et al., 2017; Olvmo et al., 2020) and total loss of palsa complexes
466 has been recorded in the far north eastern parts of Norway (Vorren 2017).

467 Although there are differences in subsidence rates among sites the region wide permafrost degradation reflects
468 ongoing climatic trends (Fig. 3 and 7). Since 1901 ~~Fennoscandia~~~~Scandinavia~~'s climate has become wetter as
469 well as warmer with a greater proportion of the precipitation falling as rain relatively to snow (Hänsel, 2020;
470 Irannezhad et al., 2018; Irannezhad et al., 2017; Vikhamar-Schuler et al., 2016). These trends are reflected in the
471 far north where higher air temperatures, greater precipitation and snow depths has already shifted climatic
472 conditions, in parts of the region, away from those that support permafrost in peatlands e.g. since the 1940's
473 (Åkerman & Johansson, 2008; Borge et al., 2017; Olvmo et al., 2020). Further, deep permafrost boreholes show
474 decadal signals of increasing temperatures in the Scandes mountains suggesting that warmer temperatures have
475 been impacting permafrost since the 1920's (Isaksen et al., 2007). Hence, it seems that climate warming has
476 been impacting permafrost in ~~Scandinavia~~~~Fennoscandia~~ for at least 100 years.

477 As a result of the ongoing trend of increasing permafrost temperatures in palsa ~~s~~ ~~peatlands~~ in
478 ~~Scandinavia~~~~Fennoscandia~~, their permafrost temperatures are now close to 0°C, making them ~~very especially~~
479 vulnerable to decay in response to further increases in temperatures (Christiansen et al., 2010; Farbrot et al.,
480 2013). Palsa formation is closely linked to the mean annual temperature, with temperatures ~~between~~~~below~~ -1 to
481 -2°C ~~over and limited insulating snow cover over~~ consecutive years needed as a threshold for palsas to form
482 (Vorren, 2017). In this context it is important to note that the MAT in the area was between 0.53 and -1.4°C
483 since 2000 suggesting that at least in parts of the study area the climatic conditions do not support formation of
484 palsa anymore while conditions are marginal for palsa preservation in the entire region.

485 Although subsidence dominated in the northern sites, uplift was also noted in the study region. Mechanisms that
486 may explain patterns of uplift are formation of new palsa as well as short-lived frost mounds that can form
487 temporarily in the palsa system (Zuidhoff, 2002). Further mechanisms that may result in uplift are changes in
488 the water level of the flooded parts of the peatlands as well as accumulation of plant residues from the
489 productive fen vegetation parts of the study sites on the peatland surface, reflecting adaptation of the local
490 ecosystem to degraded palsa mounds reflected by changes in remotely sensed terrain surface.

491 In addition to demonstrating regional permafrost degradation in northern Fennoscandia, this work also provides
492 proof of concept for circumpolar assessments of permafrost degradation using ASPIS InSAR. It enables
493 detection of the areas with rapidly degrading permafrost and deepening active layers but also peat consolidation
494 in areas that ~~have~~ ~~s~~ already lost its permafrost (de la Barreda-Bautista et al., 2022). The fact that InSAR data is
495 integrated over 20m × 20m pixels means that the signal of local level degradation may be somewhat dampened
496 (de la Barreda-Bautista et al., 2022). However, the high precision of the change in vertical position means that
497 InSAR is an important tool to employ to detect the initial stages of large-scale permafrost degradation. ~~In~~
498 ~~concurrence with the literature~~ (Alshammari et al., 2020; Alshammari et al., 2018; Bartsch et al., 2016; de la

499 [Barreda-Bautista et al., 2022](#); [Short et al., 2014](#); [van Huissteden et al., 2021](#)), we found the majority (69%) of
500 [our results were within the MSE of direction of ground motion change, providing confidence to locate where](#)
501 [permafrost is degrading](#). Currently, the study of long-term trends and drivers using InSAR is somewhat limited
502 by the short collection period of Sentinel 1, but as more data are continued to be collected, methods such as non-
503 linear time series creation will become viable to compare subsidence directly to longer climatic drivers.
504 However, ~~the our large-scale baseline~~ assessment of permafrost subsidence, [provides a baseline to direct, and](#)
505 [compare, against future fieldwork monitoring in northern Sweden developed here, provides an initial assessment](#)
506 [of ongoing subsidence, would be advantageous should field monitoring be arranged in the future](#). As a
507 complement to the ASPIS-InSAR data, the novel roughness thresholding method used here together with
508 contextual data proved a powerful tool to map and monitor changes (Franklin, 2020; König et al., 2019; Otto et
509 al., 2012). This approach could be developed using machine learning methods to model palsa dynamics to better
510 automate the extraction of palsa landform positions (König et al., 2019; Luoto & Seppälä, 2002). If
511 accomplished, the operating extent of this tool could be vastly increased using the Arctic 2m DEM dataset over
512 area where its quality is high enough to allow high resolution mapping of the degrading edges of raised palsa
513 ~~plateaux~~ (Morin, 2016); [Karlson et al., 2021](#)). [In turn this could be used to remove the stable centre of palsa](#)
514 [plateaux and exclusively compare palsa edges to the roughness index, where we would expect to find a stronger](#)
515 [correlation than there exists with the current palsa raster cells \(Fig. 7a\). It has been suggested that small,](#)
516 [fragmented, and irregularly shaped palsa are more susceptible to erosion \(Borge et al., 2017; Mamet et al., 2017,](#)
517 [Beer et al. in review\). We have not gone as far as to estimate this here but the possible palsa edges inferred from](#)
518 [the roughness index, could be built upon for this understanding. Casual analysis of figures 5, 6 and the](#)
519 [orthophotos provided in the supplementary materials would support these expectations](#). Together the ASPIS-
520 InSAR and the DEM derived roughness index metrics offer novel ways of large scale monitoring of permafrost
521 degradation. This will help to quantify the rate of palsa ecosystem collapse and transition to a non-permafrost
522 state.

523 We conclude that permafrost degradation of palsa ~~peatlands~~ is occurring across northern Sweden, with the
524 greatest rates of degradation and largest areas impacted being Sweden's two largest permafrost ~~peatland~~
525 complexes in the far north. This raises serious concerns that these systems will lose their permafrost entirely in
526 the coming decades especially as climatic conditions are approaching the limits of sustaining palsa ~~peatlands~~
527 (Fewster et al., 2022). The implications of this rapid loss of permafrost is ecosystem collapse and loss, as the
528 permafrost core is fundamental to the existence of palsa ~~peatlands~~. Future research should focus on the
529 implications of this collapse on increased CH₄ emissions (Glagolev *et al.*, 2011; Turetsky *et al.*, 2020; Varner *et*
530 *al.*, 2022), carbon loss (Hugelius *et al.*, 2020), and thus the potential for strong climate feedbacks (IPCC, 2021)
531 as well as using longer-time InSAR data as this becomes available to investigate regional variations in climatic
532 drivers of permafrost degradation. Further, our study demonstrates that InSAR together with terrain data can be
533 applied over continuous natural surfaces at a regional-scale to monitor permafrost degradation in palsa
534 ~~peatlands~~, offering a tool for circumpolar monitoring of climate warming impact on these systems.

535

536 5. Acknowledgement

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538 InterAccess programme and the Swedish research council (VR-2021-05767 to M. Siewert). Associated
539 fieldwork was supported by the Climate Impacts Research Centre (CIRC) at Umeå University. Samuel Valman
540 was supported by the EPSRC funded Geospatial Centre for Doctoral Training (EP/S023577/1).

541

542 **6. Author contributions**

543 SV: Carried out the majority of the data analysis and made a significant contribution to data interpretation, writ-
544 ing and finalising the manuscript text. Both SV and MS can be considered to have contributed equally to this
545 work.

546 MS: Contributed to the conception of the study, contributed DEM and orthophoto data, carried out fieldwork to
547 assess permafrost degradation, contributed and advised on data analysis and interpretation, contributed to struc-
548 turing, writing, and refining the text. Both MS and SV can be considered to have contributed equally to this
549 work.

550 DB: Contributed to the conception of the study, advised on the data analysis, and made a significant contribution
551 to finalising the text.

552 ML: Provided data analysis, support on the InSAR processing, data interpretation, and writing of the text.

553 DG: Carried out the initial InSAR data processing

554 BBB: Contributed to the conception of the study and refining the text.

555 AS: Contributed to the conception of the study and advised on the InSAR data processing

556 SS: Conceived and directed the study, contributed to data analysis, carried out fieldwork to assess permafrost
557 degradation and made a significant contribution to formulating and finalising the text.

558 SS, DB, AS and MS secured the funding for the project.

559 **Code Availability**

560

561 All the python scripts used to carry out these analyses are available at the github repository:

562 https://github.com/SamValman/Permafrost_Sweden.

563

564 **Data Availability statement**

565

566 The Sentinel-1 datasets are freely available and can be obtained by searching and downloading the Interferomet-
567 ric Wide (IW) swath mode products for orbit track numbers [168-??](#) and [??-66](#) through the Copernicus Open Ac-
568 cess Hub (<https://scihub.copernicus.eu/dhus/#/home>). The processed interferometric data and deformation maps
569 are commercially sensitive and may be made available on reasonable request by email addressed to the corre-
570 sponding author. All other datasets produced during this project will be uploaded on zenodo and the DOI pro-
571 vided once the article has been accepted.

572

573

574

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