



1 **Multitemporal UAV LiDAR detects seasonal heave and**
2 **subsidence on palsas**

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6



7 **Abstract**

8 In the context of the accelerating impacts of climate change on permafrost landscapes, we use uncrewed
9 aerial vehicle (UAV) LiDAR technology to investigate seasonal terrain changes in palsas – mounds of frozen
10 peat – since other remote sensing methods have struggled to capture the full dynamics of these landforms.
11 We investigated two palsas (4–5 m in height) in Sweden's largest palsa mire complex, where we performed
12 five field campaigns between September 2022 and September 2023 to track intra-annual frost heave and
13 thaw subsidence. Our approach allowed us to create digital terrain models (DTMs) from high density point
14 clouds (>1,000 points/m²) and analyze elevation changes over time. We found that both palsas heaved on
15 average 0.15 m (and up to 0.30 m) from September to April and subsided back to their height from the
16 previous year, or slightly below, over the course of the following summer. At one of the palsas, we observed
17 notable lateral degradation over the study period in a 300 m² area, with 0.5–2.0 m height loss, likely initiated
18 during the preceding warm and wet summer months. Part of this degradation occurred between September
19 2022 and April 2023, suggesting that the degradation of these palsas is not limited to the summer months.
20 Our study shows the substantial value of using UAV LiDAR for understanding how permafrost areas are
21 changing. It helps in tracking the ongoing effects of climate change and highlights palsa dynamics that
22 would not be captured by annual measurements alone.

23



24 1 Introduction

25 In the face of accelerating climate change, permafrost – defined as ground that remains at or below 0 °C
26 for at least two consecutive years (Harris et al., 1988) – is warming at a global scale (Biskaborn et al.,
27 2019). The thickness of the active layer, which is the top layer that thaws and freezes each year
28 (Romanovsky & Osterkamp, 1995), is also increasing at many polar sites (Smith et al., 2022). Palsas, which
29 are peat mounds with a core of perennially frozen soil, are indicative of permafrost presence and serve as
30 particularly vulnerable indicators of climatic changes. They are generally found in subarctic wetlands
31 (Seppälä, 1986), in the discontinuous or sporadic permafrost zone. The climatic space for palsas, typically
32 with a mean annual air temperature between -3 °C and -5 °C and mean annual precipitation <450 mm,
33 according to Luoto et al. (2004a), is projected to disappear in Fennoscandia by the end of the 21st century
34 (Fewster et al., 2022). The degradation of palsas leads to significant changes in this landscape and
35 ecosystem. In recent studies, an increasing lateral degradation rate of palsas is reported (Borge et al.,
36 2017; Mamet et al., 2017; Olvmo et al., 2020), which may have far-reaching consequences for the
37 biodiversity of the subarctic (Luoto, et al., 2004b; Swindles et al., 2015). The transition from an elevated,
38 often dry, palsa bog to a lower lying, wet fen, is associated with an increase in CH₄ and CO₂ emissions (e.g.
39 Łakomic et al., 2021; Pirk et al., 2023; Swindles et al., 2015; Voigt et al., 2019). Due to the tendency to
40 enhance atmospheric greenhouse gas emissions, degrading palsas therefore not only indicate but can also
41 contribute to climatic change. This climatic feedback mechanism further highlights the need for continued
42 monitoring of these environments. Therefore, palsa mires are a priority habitat of the EU Species and
43 Habitat Directive (*EUNIS -Factsheet for Palsa Mires*, 2013). However, in Sweden only about half (47%) of
44 the total palsa area is situated within some kind of protected area (Backe, 2014).

45 Permafrost degradation of palsas is indicated by both lateral erosion and vertical subsidence. In addition to
46 this, the elevation of permafrost terrain fluctuates cyclically due to annual freeze-thaw cycles within the
47 active layer (Gruber, 2020; Iwahana et al., 2021), though this has not been explicitly studied on palsas.
48 During fall, freezing of water in the soil and its expansion, in addition to the formation of seasonally
49 segregated ice, can cause heaving of the terrain, while terrain subsidence, caused by the melting of both
50 pore ice and segregated ice, occurs during spring and summer (Fu et al., 2022). The melting of excess



51 ground ice (i.e., ice that is in excess of the total pore volume of the ground in unfrozen conditions; Harris et
52 al., 1988) below the active layer causes longer-term subsidence as result of permafrost thaw. Therefore,
53 thaw-subsidence rates are generally higher in ice-rich than in ice-poor permafrost soils (Gruber, 2020;
54 Zwieback & Meyer, 2021).

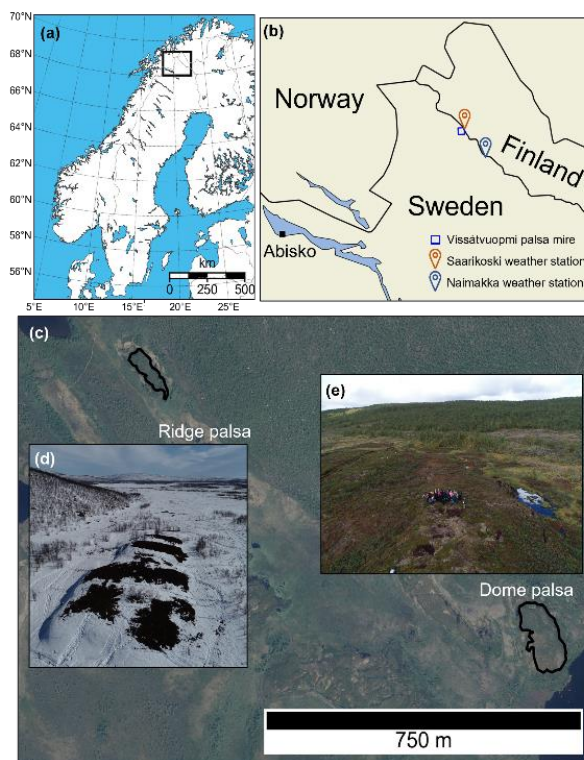
55 While there is growing awareness of the importance of monitoring *palsa* mires, there is a lack of quantitative
56 measurements of their intra-annual heave and subsidence patterns. de la Barreda-Bautista et al. (2022)
57 used InSAR data to analyze thaw-season subsidence on *palsas* in northern Sweden but found only sub-
58 cm scale surface level changes. They suggest that this method likely highly underestimated actual
59 displacement rates as a result of spatial averaging. In other recent studies with attempts to quantify both
60 lateral and vertical changes in *palsas* or peat plateaus, the focus was on multi-year timescales (e.g., Martin
61 et al., 2021; Verdonen et al., 2022) and not on changes occurring within a year. UAV photogrammetry to
62 create digital surface models (DSMs) to study *palsas* and peat plateaus has been applied more frequently
63 in recent years (de la Barreda-Bautista et al., 2022; Krutskikh et al., 2023; Martin et al., 2021; Verdonen et
64 al., 2023). However, as described in Verdonen (2023), change analysis from DSMs created with UAV
65 photogrammetry is sensitive to relatively minor changes in vegetation and light conditions. Advances in
66 uncrewed aerial vehicles (UAV or drone) in combination with light detection and ranging (LiDAR) technology
67 (Ostrowski et al., 2017) have made it possible to collect accurate, high-resolution (cm-scale) digital terrain
68 models (DTMs) and DSMs. LiDAR sensors can prove advantageous over regular photographic Red-Green-
69 Blue (RGB) imagery, in that LiDAR can penetrate through small gaps in the vegetation allowing creation of
70 DTMs of the underlying ground, whereas RGB cameras require clear sight to the ground surface. LiDAR
71 can also be used in low-light conditions as opposed to RGB imagery, which can be beneficial in the Arctic,
72 where daylight is limited during parts of the year. Another advantage that UAV LiDAR data holds is the
73 absence of need of ground control points (GCPs) due to the potentially low bias of UAV LiDAR data
74 positional errors. This advantage drastically saves time and thus costs on repeated visits (Harder et al.,
75 2020). Therefore, the use of repeat UAV LiDAR scanning is a promising tool for accurate change detection
76 (Curcio et al., 2022; Harder et al., 2020; Jacobs et al., 2021; Lin et al., 2019), but has not yet been widely
77 used to monitor permafrost features.



78 In this study, our objective is to detect and quantify the intra-annual vertical heave and subsidence of two
79 palsas using repeat measurements from UAV LiDAR data. Our findings show seasonal dynamics of the
80 palsas and highlight the substantial potential of UAV LiDAR scanning to study permafrost-mediated
81 landscape changes.

82 **2 Study site: Vissátvuopmi palsa complex**

83 Located near the Finnish border, and just southwest of the Könkämäeno river, Vissátvuopmi is the largest
84 coherent palsa mire complex in Sweden (ca 150 ha of palsa area; Backe, 2014) at N 68°47'50", E 21°11'30"
85 (Fig. 1). Surrounded by mountains up to 700 m a.s.l., the valley has several larger lakes, thermokarst
86 features and fens. Two distinct palsas, one ridge-shaped and one dome-shaped (called the "Ridge" and
87 "Dome" palsas from hereon) situated at the foot-slope of a bedrock hill, are the focus of this study and are
88 located between 443 and 452 m a.s.l. (Fig. 1c). In September 2022, the Dome palsa was approximately
89 170 meters in length and 75 meters in width, with its highest point about 4 meters above the surrounding
90 mire. The Ridge palsa measures about 125 meters in length and 40 meters in width, with its highest point
91 being roughly 5 meters above the adjacent mire terrain. The Dome palsa is taller on its northern and eastern
92 sides, while it flattens out on the western and southern sides. An ATV track runs over the northeastern part
93 of the palsa, which most certainly contributes to amplified degradation of the underlying permafrost by
94 collecting snow and water. The track is visible in aerial photos from 1994 but is absent in the 1963 photos.
95 A natural depression in the center part of the palsa has the same effect and further fragments this palsa.
96 The Ridge palsa is smaller in area but slightly taller. This palsa consists of several crests of similar elevation
97 with depressions in between. The southeastern margin of this palsa is 'tail-shaped' and of lower elevation.



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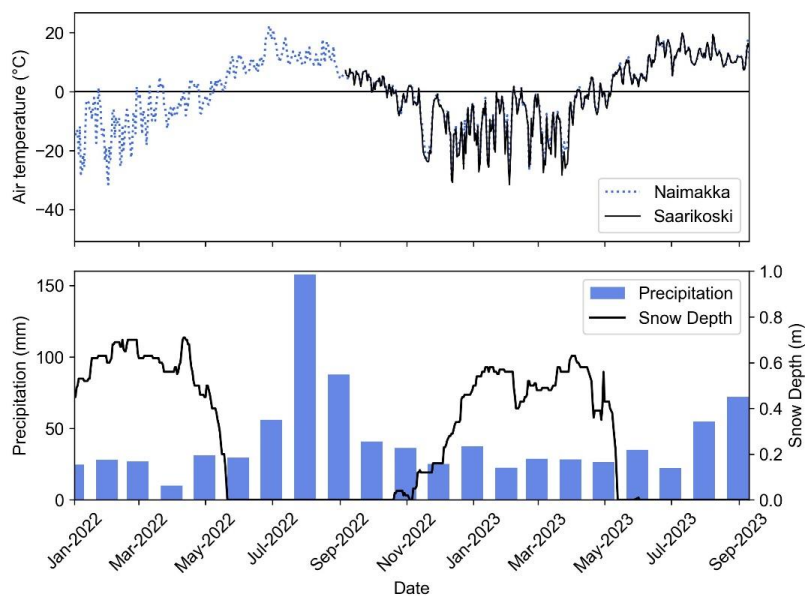
99 **Figure 1. (a,b) Location of the Vissätvuopmi palsa mire complex and used weather stations in Northern**
100 **Sweden. (c) Orthophoto from 2021 (© Lantmäteriet, 2021) showing a segment of the palsa mire and the location**
101 **of the two studied palsas. (d) Oblique UAV image highlighting the partly snow-covered Ridge palsa in April**
102 **2023. (e) Aerial view of the Dome palsa in September 2023.**

103 The vegetation over the palsa mire complex is characterized by a mosaic of marshes with grasses and
104 sphagnum mosses, wet heaths with willow, and drier areas with subalpine shrub heath. Birch forest occurs
105 primarily on the slopes surrounding the mire, however, birches are also growing within the palsa mire
106 complex. The palsa vegetation consists of dry heath and mesic heath (Andersson et al., 1985) where the
107 field layer is predominantly low-growing *Betula nana* (< 35 cm in mean height), *Empetrum nigrum* ssp
108 *hermaphroditum* and *Rubus chamaemorus*. The bottom layer consists of lichens in higher and drier areas
109 of the palsa, while sphagnum mosses are in lower-lying, wetter areas. At the edges of the palsa, particularly
110 where pools of water have formed, grasses such as *Carex rotundata* and *C. saxatilis* are common, in



111 addition to *Eriophorum russeolum* (Backe, 2014). The ridge palsa has a few birch trees growing on the
112 north and south parts of the palsa.

113 The Köppen climate type is Subarctic (Dfc) and a weather station operated by the Swedish Meteorological
114 Institute (SMHI) in Naimakka, ca 18 km east of the study site observed a mean annual air temperature of
115 of $-1.5\text{ }^{\circ}\text{C}$ and mean annual precipitation was 460 mm in the 1991–2020 standard period. In September
116 2022, a HOBO weather station was set up in the nearby settlement of Saarikoski, ca. 1.5 km from the study
117 site (see Fig. 1b for location of the two weather stations). Both precipitation and snow depth have been
118 monitored since 2015 at Saarikoski. It is notable that in July 2022 the monthly precipitation was 158 mm,
119 which is more than a third of the mean annual precipitation (Fig. 2).



120

121 **Figure 2. Weather data in the study region between January 2022 and September 2023. (a) Air temperature at**
122 **ca. 1.5 km (Saarikoski) and at ca. 18 km (Naimakka) from the studied palsas. (b) Monthly precipitation and snow**
123 **depth observed at Saarikoski.**

124 Based on aerial photo interpretation, Olvmo et al. (2020) reported average areal decay rates of $-0.71\%a^{-1}$
125 and $-1.25\%a^{-1}$ in the Dome and Ridge palsa respectively between 1955 and 2016. Additionally, they



126 concluded that palsas in the Vissátvuopmi complex have most likely been in a phase of degradation since
127 the early 20th century.

128 **3 Data and Methods**

129 **3.1 LiDAR data acquisition**

130 We used repeat UAV-borne LiDAR scanning to obtain point clouds and create raster-based digital terrain
131 models (DTMs) at a high temporal and spatial resolution. Flights were performed on 4 September 2022, 26
132 April 2023, 18 June 2023, 19 July 2023, and 7 September 2023. We conducted the scans on these dates
133 to capture the state of the palsas at the end of summer, then observed them when frozen, and continued
134 tracking them throughout the proceeding thawing season. A DJI Matrice 300 RTK UAV was equipped with
135 a YellowScan Mapper (YSM) LiDAR scanner in September 2022 and a YellowScan Mapper + (YSM+) in
136 all following flights (Fig. 3). These are lightweight Livox LiDAR scanners with an Applanix GNSS/INS
137 system. Table 1 shows the properties of the flights and scanner(s). For both of the palsas, a flight with high
138 overlap was done and supported by a second flight with additional orthogonal trajectories. The same flight
139 plans were executed for all five scanings to ensure equal spatial coverage and resolution.



140
141 **Figure 3. DJI Matrice 300 RTK equipped with the YellowScan Mapper+ in front of the Ridge Palsa.**

142 The April 2023 flights were performed during a period of snowmelt, so that parts of both palsas were snow-
143 free, while other parts remained snow-covered. Consequently, in order to investigate changes in terrain
144 elevation, the snow-free parts needed to be isolated for analysis. For that reason, flights with a second UAV
145 with an RGB camera to create orthomosaics were performed on the same day as the LiDAR scanning.



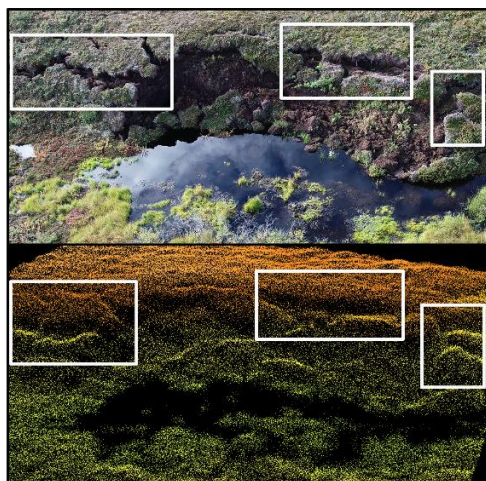
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147 **Table 1 LiDAR scanner and flight parameters of the flying missions. Where values for YellowScan Mapper differ**
148 **from YellowScan Mapper +, they are shown within parentheses.**

Parameter	Value, YellowScan Mapper + (Mapper)
Vertical Accuracy (RMSE, m)	0.021 (0.028)
Precision (m)	0.024 (0.032)
Number of returns	3 (2)
Altitude (m.a.g.l.)	60
Velocity (m/s)	8
Overlap (%)	60

149

150 Regarding the UAV LiDAR flights, the average flight time was 6 minutes and 49 seconds over the Dome
151 palsa, and slightly shorter over the Ridge palsa at 5 minutes and 13 seconds. Similarly, flights over the
152 Dome palsa yielded a slightly larger coverage area of 53993 m² compared to the Ridge palsa, which
153 covered an area of 42072 m², including part of the surrounding mire and forested mountain. The mean
154 point densities over the Dome palsas were 1327 (YSM) and 1462 (YSM+) points/m², and 1201 (YSM) and
155 1519 (YSM+) points/m² at the Ridge palsa. These pulse densities can be regarded as very high, easily
156 allowing the creation of DTMs with high spatial resolution. Terrain features such as surface cracks and
157 blocks of peat are well-identifiable in the LiDAR point cloud data, which emphasizes the high quality of the
158 collected data (Fig. 4).



159

160 **Figure 4. (a) RGB photo of a degrading edge and formation of a thermokarst lake at the Dome palsa, taken with**
161 **a DJI Phantom 4 UAV in September 2022. (b) Dense UAV LiDAR point cloud taken on the same day with the**
162 **DJI Matrice 300 RTK equipped with the YellowScan Mapper. Morphological features such as cracks in the**
163 **surface and individual blocks of peat can clearly be recognized in the point cloud. This section of the palsa**
164 **reaches a height of ca. 4 meters.**

165 **3.2 Point cloud processing and DTM creation**

166 The processing of a LiDAR point cloud requires several steps, including GPS correction, strip adjustment,
167 classification of ground points, and the creation of a DTM raster. The position data captured by the LiDAR
168 system during the flights are postprocessed in PosPac UAV v. 8.2 (*Applanix*, 2023) using PP-RTX for
169 trajectory correction. PP-RTX for UAV uses the Trimble CenterPoint® RTX™ correction service, which
170 computes corrections to satellite orbit and clock data for trajectory correction and positioning, based on a
171 global network of tracking stations. This cloud-based solution gives centimeter-level positioning accuracy
172 without the requirement to set up a local base, which makes PP-RTX particularly advantageous for UAV
173 surveys in remote regions. YellowScan's processing software, CloudStation (*YellowScan*, 2023), is then
174 used for strip adjustment to reduce the relative adjustment error. The point clouds were compared visually
175 for alignment in areas where changes were least likely. Following this, CloudCompare v. 2.12.4 (*GPL*
176 *software*, 2023) is used to perform the classification of points into ground and non-ground points using the
177 Cloth Simulation Filter (CSF) (Zhang et al., 2016). The CSF method simulates a virtual cloth dropping onto



178 the inverted point cloud. Points that are close to where the cloth settles are classified as ground, while those
179 that are farther away are considered non-ground. The software allows three adjustable parameters for the
180 CSF classification. Several parameter combinations were tested and the best result, based on visual
181 inspection of the resulting point cloud, was achieved with the following parameters: 'Cloth resolution' = 0.1
182 m (matching the resolution of the created DTMs), 'Max iterations' = 500 and 'Classification threshold' = 0.1
183 m (controlling the distance of points to be classified as ground), resulting in a detailed ground surface. Only
184 ground points are then used to create the DTM, and in this project a grid cell resolution of 0.1 m was chosen.
185 Since the total thickness of the point cloud along the ground is between ca. 0.10 and 0.30 m, the minimum
186 elevation per grid cell is used during the rasterization in order to ensure that the raster represents the ground
187 elevation. This process is carried out for all five time points, creating a DTM for each scanning and palsa.
188 Finally, the resulting DTMs are used for change detection by subtracting values of one raster from another.

189 The snow free parts from the April 2023 data were singled out for the computation of heave and subsidence.
190 This was done with georeferenced orthophotos taken on the same day as the LiDAR data, in combination
191 with a hillshade image from the DTM (since the hillshade is smooth where there is snow). For each DTM,
192 the mean elevation of the surrounding mire is computed and subtracted from the DTM, which normalizes
193 the elevation of the palsa into the height above the mire. This simplifies the comparison between the data
194 from the different flights. Since the mire in April was snow-covered, the mire elevation from the closest date
195 (June) was taken for normalizing the April DTM.

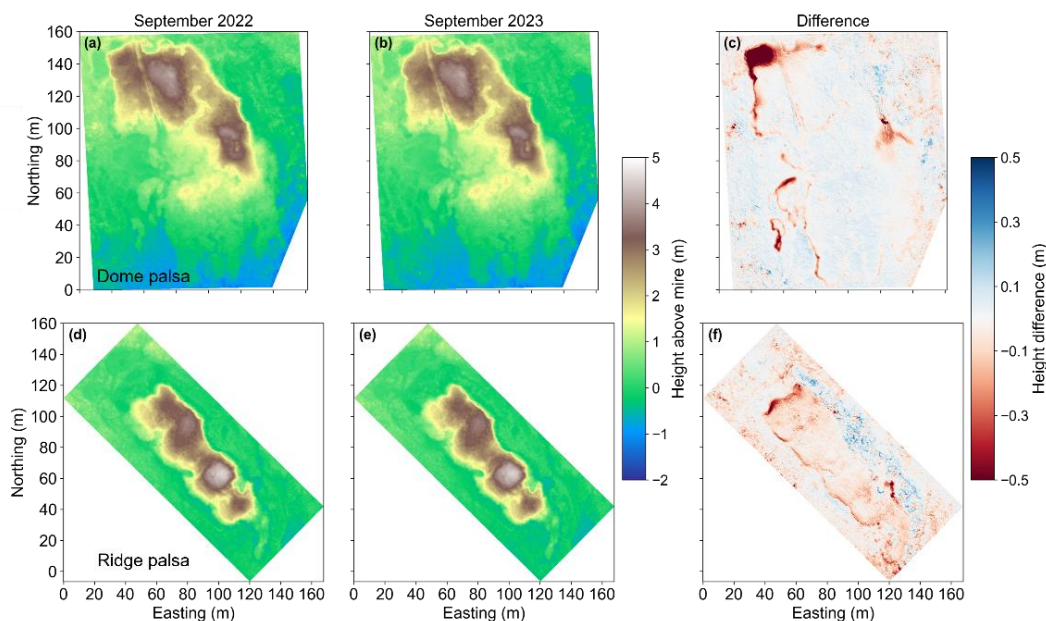
196 **4 Results**

197 **4.1 Annual terrain changes from UAV LiDAR**

198 Between September 2022 and September 2023, the maximum height difference was 0.03 m, (from 4.21 m
199 to 4.18 m) on the Dome palsa and 0.05 m (4.67 m to 4.62 m) on Ridge palsa (Fig. 5). The largest change
200 in height was detected along the northwest edge of the Dome palsa where an area of ca. 300 m² subsided
201 between from 0.5 and 2.0 m. Degradation also occurred along the margins of the Dome palsa, as well as
202 within the ATV track that borders the eastern side of the new 300 m² degraded area. The height of the



203 Ridge palsa decreased slightly over the entire landform, with most loss along the margins in the form of
204 lateral degradation.



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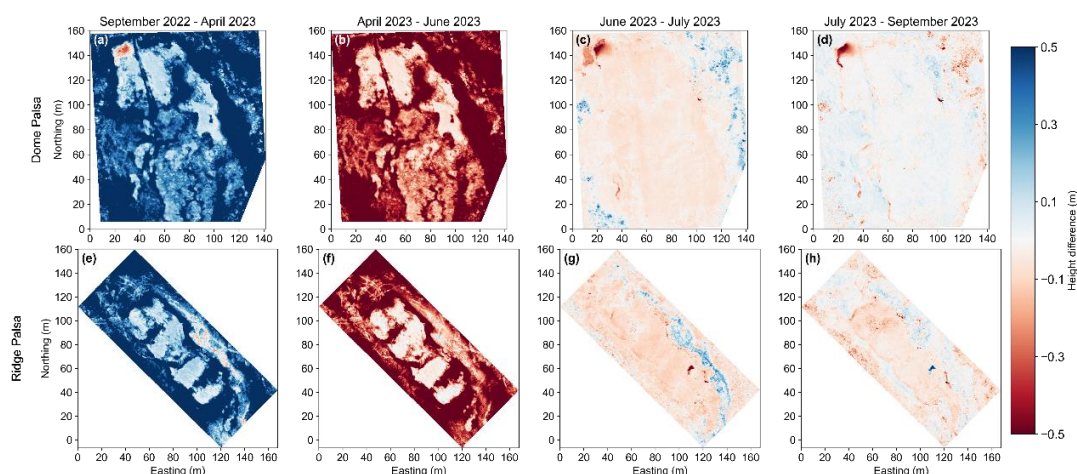
206 **Figure 5. DTMs of the Dome (a,b,c) and Ridge (d,e,f) palsas palsa on UAV LiDAR scans in September 2022 and**
207 **September 2023. The images highlight the topographical changes over one year.**

208 **4.2 Seasonal terrain changes from UAV LiDAR**

209 By comparing each DTM with the following, we observed intra-annual terrain variations, i.e., frost heave
210 and thaw subsidence (Fig. 6). The heterogeneous snow cover that affects this pattern is more clearly visible
211 in Fig. 7c and d.. The snow thickness is up to ca. 2.0 m at the eastern margin of both palsas and 1.0 to 1.5
212 m in the depressions, while the crests remain snow-free (also see Fig. 1d).. Most interestingly, on these
213 snow-free crests (i.e. Fig. 7f), an elevation increase up to 0.30 m and on average of 0.15 m from September
214 2022 to April 2023 is observed. Comparing these two closeup parts of the Ridge palsa profile also shows
215 that the subsidence the palsa underwent throughout the year in the depressions was larger than on the
216 crests. Between June and July, both palsas clearly subside all over (Fig. 6c and 6g), on average 0.05 m on

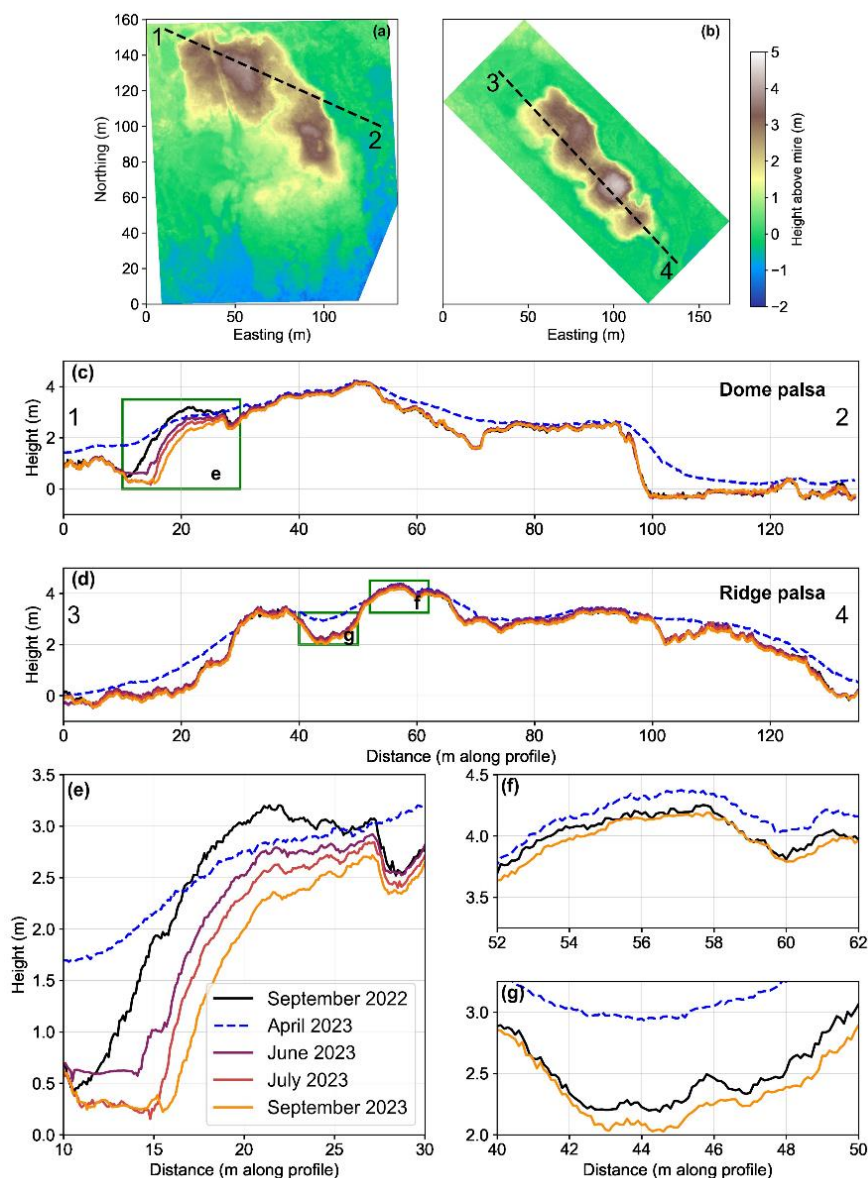


217 the Dome palsa and 0.08 m on the Ridge palsa. Subsidence from July to September is only clear on the
218 Ridge palsa, with 0.05 m on average over the entire palsa.



219
220 **Figure 6. Sequential height difference maps of the Dome (a,b,c,d) and Ridge (e,f,g,h) palsas between each of**
221 **the five UAV LiDAR DTMs from September 2022 to September 2023. Each panel showcases the topographical**
222 **changes over successive intervals. Blue colors indicate elevation gains and red colors indicate elevation**
223 **losses. The snow cover in April is mostly responsible for changes observed to and from those panels.**

224 Even despite the snow cover in April, the northwestern corner of the Dome Palsa shows signs of
225 degradation up to 0.4 m (Fig. 6a), indicating it occurred sometime between September 2022 and April 2023.
226 This degradation remains evident and increases throughout the time series and is visible in Fig. 7c and e,
227 where this area progressively degraded vertically up to 2.0 m between September 2022 and September
228 2023. Additionally, the ATV track that crosses the Dome palsa shows a deepening of 0.2–0.3 m just over
229 the time period in this study (at ca. 27–29 m in Fig. 7e).

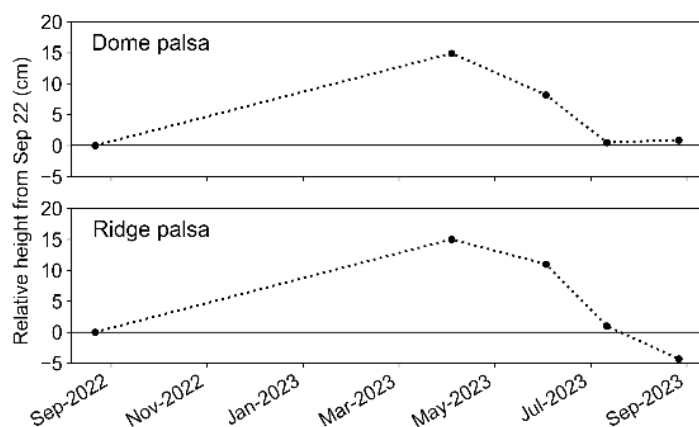


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231 **Figure 7. Elevation profiles of the Dome and Ridge palsa based on UAV LiDAR DTMs from September 2022 to**
232 **September 2023. (a,b) The dashed lines denote the location of the elevation profiles 1-2 and 3-4. (c) and (d)**
233 **provide a general elevation overview of the transect, while (e), (f), and (g) offer zoomed-in views of specific**
234 **areas of interest, indicated by the green boxes. The profile from April 2023 is mostly showing the snowpack,**
235 **although some peaks are snow-free (see also Fig. 1d).**



236 The mean height of both palsas increased 0.15 m between September 2022 and April 2023 (Fig. 8). The
237 subsequent flights throughout the summer season show a successive lowering of the palsa height i.e.
238 subsidence. As shown in Fig. 6, the rate of subsidence is highest between the June and July flights. The
239 dome palsa returned to a similar mean height in September 2023 as compared to September 2022, while
240 the ridge palsa subsided to a 0.04 m lower mean height in September 2023 than in September 2022, which
241 is also shown in Fig. 5. Overall, the heave and subsidence pattern is very similar across the two studies
242 palsas.



243

244 **Figure 8. Heave and subsidence on the two studied palsas during the 2022-2023 year. Mean height changes**
245 **relative to September 2022 are plotted (only for the areas that were completely snow-free in April 2023).**

246

247



248 **5 Discussion**

249 **5.1 Intra-annual terrain dynamics on palsas**

250 The objective of this study was to elucidate and quantify the intra-annual dynamics of palsas, and to achieve
251 this, the use of UAV LiDAR proved to be an extremely suitable tool. Analysis of the DTMs indicated that
252 both palsas had frost heave of ca. 0.15 m on average between September and April and subsequent thaw
253 subsidence. This is the first study that covers these dynamics on palsas in such high spatial resolution,
254 although seasonal terrain variability has been investigated in other permafrost environments (Gruber, 2020;
255 Hu et al., 2022; Iwahana et al., 2021; Lin et al., 2019). For example, the heave and subsidence detected in
256 this study are similar to values (0.10 – 0.14 m) observed in the GNSS positioning survey by Iwahana et al.
257 (2021) on an intact tundra site in Alaska after the warm summer of 2019.

258 The largest rate of subsidence in our study occurred between June and July, which was 0.07 m on the
259 Dome palsa and 0.10 m on the Ridge palsa. Freeze-thaw cycles within the active layer cause frost heave
260 and thaw subsidence, partially due to the density difference between water and ice. However, the average
261 observed heave of 0.15 m cannot be explained alone by this process and requires the formation of seasonal
262 segregation ice (i.e. ice lenses) within the active layer (Fu et al., 2022; Iwahana et al., 2021). Alternatively,
263 processes within the core of the palsa, for example, the infiltration and refreezing along meltwater pathways,
264 might result in seasonal heave and subsidence. To make more conclusive statements about the exact
265 mechanisms, it would be necessary to obtain complementary observations from the interior of these palsas,
266 in the form of soil and ice cores. Taking this step in the future would allow the observed changes in the
267 terrain morphology such as those done here to be better understood in relation to internal palsa dynamics.

268 The 0.15 m heave is computed on the areas that were snow-free in April and are thus biased towards the
269 crests of the palsas that have a thinner active layer as they have a thinner winter snow cover, which limits
270 the insulation of the ground below. Since the magnitude of the heave and subsidence depends on the
271 thickness of the active layer (Iwahana et al., 2021), the areas with a deeper active layer (i.e., those not
272 included in the computation due to remaining snow in April) are therefore expected to have undergone an
273 even larger increase in height between September 2022 and April 2023. A process other than freeze-thaw



274 dynamics that could have affected the LiDAR measurements is the seasonal oscillation of the peatland
275 surface height due to water table fluctuations, called 'mire breathing' (Kellner & Halldin, 2002; Roulet, 1991).
276 By accounting for the elevational changes in the mire surrounding the palsas, we have corrected for this
277 effect.

278 We present seasonal patterns at a high spatial resolution over the full spatial extent of two large palsas.
279 While other studies have applied multitemporal LiDAR for detecting permafrost dynamics, they have been
280 terrestrial LiDAR scanning (Anders et al., 2020) rather than UAV-borne LiDAR, or else inter-annual airborne
281 LiDAR (e.g. Douglas et al., 2021; Jones et al., 2015). Anders et al., (2020) performed three TLS
282 measurements over 14 months (June 2015, Aug 2015 and Aug 2016) to observe thaw subsidence of a
283 permafrost area in Northwest Territories, Canada, finding a total change of 1.4 cm; they determined that
284 the TLS measurements were more accurate than those from field-based surveying. Another approach to
285 quantify the effect of seasonal freeze/thaw dynamics on topography is through the use of InSAR remote
286 sensing (de la Barreda-Bautista et al., 2022; Kou et al., 2021; Yanagiya et al., 2023) which can be an
287 effective method to detect the signal of subsidence. Due to its coarser spatial resolution, however, it is likely
288 to underestimate actual heave and subsidence values of smaller isolated features such as palsas.

289 The time series from this study showed not only the seasonal heave and subsidence patterns, but also
290 revealed a large area of degradation on one of the palsas that happened between September 2022 and
291 April 2023. By comparing the elevation profiles at these two timestamps in the northwest part of the Dome
292 palsa (Fig. 7e), we can see that a drop of up to 0.4 m happened within the autumn and/or winter season,
293 despite the presence of a snow cover in April. It is important to consider that positive air temperatures
294 persisted until the end of October 2022 (Fig. 2), which suggests that the subsidence likely continued into
295 the late autumn period, influenced by prolonged thawing conditions. Whether this entire drop in palsa height
296 occurred solely between the September 2022 campaign and the freeze-up of the entire active layer, or if
297 gravitational processes also contributed after the complete freeze-up, remains uncertain. Regarding the
298 initiation of the degradation at this location, it is likely that it was ongoing earlier in the summer of 2022.
299 Both the air temperatures (>20 °C) and precipitation (158 mm) (both measured at Saarikoski weather
300 station) peaked in July of that year, which could have resulted in a deeply thawed, saturated upper layer of



301 the palsa and initiated a progressive lateral degradation event. The characteristic of this lateral degradation
302 is similar to that described in Martin et al. (2021) as “constant edge degradation”, which they found as the
303 second phase of lateral degradation on peat plateaus following “initial slope adjustment” and preceding
304 “plateau collapse” phases. While both the spatial and temporal scale differ between their and our study,
305 this could indicate that a more widespread collapse or subsidence can follow on the Dome palsa. In order
306 to make more conclusive statements about the reason for this rapid degradation, the monitoring of ground
307 temperatures and knowledge of the internal structure of this palsa is required.

308 While this described degradation could be just an event, the elevation change of the Dome palsa as a whole
309 from September 2022 to September 2023 was minor. The Ridge palsa, however, did subside on average
310 0.07 m within one year. The faster degradation on this specific palsa is in line with observations in Olvmo
311 et al. (2020).

312 **5.2 Using UAV LiDAR to monitor permafrost landscapes**

313 The use of UAVs in assessing permafrost landscapes has increased in recent years (de la Barreda-Bautista
314 et al., 2022; Krutskikh et al., 2023; Martin et al., 2021; Siewert & Olofsson, 2020; Sjögersten et al., 2023;
315 Verdonen et al., 2023), although primarily with the use of photogrammetry. Changes in the exterior of
316 permafrost peatlands can be subtle and therefore require the use of highly accurate methods, particularly
317 for vertical subsidence, and when studied over relatively short timescales.

318 A challenge in permafrost environments, especially palsa mires, is the lack of stable elevation points to be
319 used for ground control points (GCPs). In this study, GCPs were not used, however, , all point clouds were
320 obtained using GNSS, post-processed as described in 3.2, and visually checked for consistency. The
321 elevation profiles (Fig. 7) from the five campaigns are well-aligned and thereby confirm that the positional
322 data are highly accurate. As mentioned by Harder (2020), the use of GCPs is not strictly necessary for UAV
323 LiDAR applications, which significantly contributes to the efficient field visits in harsh Arctic conditions.

324 In this study, we employed the YellowScan Mapper (YSM) system in September 2022 and the YellowScan
325 Mapper+ (YSM+) in the four campaigns in 2023. The use of an upgraded LiDAR system after the first
326 scanning introduces a potential source of uncertainty in our measurements. However, both systems



327 achieved high point densities ($>1,000$ points/m²), ensuring detailed surface representation in both datasets
328 despite the difference in the number of recorded returns per pulse. Furthermore, the alignment and
329 positioning of the data were visually verified, showing that the impact of using different systems on our
330 results is minimal. The vertical accuracy of the YSM and YSM+ systems used here have an RMSE of 2.8
331 and 2.1 cm, respectively. This is a potential error source in the data and changes less than these are within
332 the margin of error. Both systems have a minimum distance of 1 m between registered returns at a flying
333 height of 60 m. This means that all objects on the surface shorter than 1 m in height will be measured by a
334 single return. In this case, to measure the ground elevation underneath vegetation cover requires high point
335 density, which was acquired in this study. By using the minimum elevation per grid cell, the resulting DTM
336 should exclude vegetation heights. In our study, the diffuse cover of *Betula nana* ranged from 0-100% per
337 m² over both palsas, yet the ground could almost always be seen between the sparse branches. Careful
338 consideration of acquisition parameters and algorithms is needed when UAV LiDAR is to be used to create
339 DTMs in areas of extremely dense vegetation cover, or when small grid cell sizes are needed to allow
340 determination of the ground elevation (Kucharczyk et al., 2018).

341 The ability of LiDAR instruments to measure ground elevation below the vegetation canopy is one of the
342 major advantages that LiDAR has over photogrammetry. Fig. 3 shows, for example, the Ridge palsa with
343 several large birch trees, where terrain analysis can still be done below the canopy. Photogrammetry
344 creates surface models that consist of vegetation height, which is likely to increase over the vegetation
345 season, adding bias to the models. The observed frost heave and subsidence in our study would not have
346 been soundly established with just the use of orthophotos or aerial imagery, due to the potential confusion
347 between vegetation and ground height. A disadvantage when comparing LiDAR and photogrammetric
348 methods is the higher costs of UAV LiDAR scanners at the present time. For that reason alone, when one's
349 objective is to merely compare the perimeter of palsas (e.g, lateral degradation) or other landforms with
350 low-growing vegetation, photogrammetry might still be preferred. While some UAV LiDAR systems can also
351 be integrated with an RGB camera, due to the additional costs, this was not done in this study. The main
352 advantage that this would give is an improved visualization through use of colorized point clouds and could
353 thereby improve the delineation of, for example, snow-covered (or -free) areas.



354 When compared to 'classic' airborne LiDAR surveys, the potential for higher spatial resolution DTMs, ease
355 of planning and lower associated costs do largely favor the use of a UAV when studying permafrost
356 landscapes at least over smaller spatial extents.

357 **6 Conclusions**

358 This study has provided insights into the intra-annual dynamics of palsas through the use of repeat UAV
359 LiDAR measurements, which we highlight as an effective tool for detailed change detection in permafrost
360 landscapes. We present a unique time series of five UAV LiDAR campaigns during a one-year time span
361 on two large palsas in Sweden's largest coherent palsa mire complex. The study revealed seasonal
362 variations in the palsa's topography, with an average frost heave and thaw subsidence of 0.15 m (and up
363 to 0.30 m), with the highest rate of subsidence on the palsas between June and July. The time series also
364 shows a considerable lateral degradation in a 300 m² section at one of the palsas of 0.5–2.0 m over the
365 one-year study period. This substantial permafrost degradation occurred between September 2022 and
366 April 2023 which suggests that the degradation palsas is not limited to the summer months. To conclude,
367 the use of repeat UAV LiDAR scanning has proven to be a highly effective tool for capturing detailed
368 measurements of permafrost dynamics, which would not have been observed if only annual measurements
369 had been taken.

370



371 **Data availability**

372 The data presented in this article are stored at [10.5281/zenodo.10497093](https://zenodo.org/record/10497093) (Renette., 2024).

373 **Author contribution**

374 CR and HR designed the study and performed the UAV surveys. HR acquired funding. CR handled data
375 processing, analysis, and figure creation. ST and BH assisted in the setup of the weather station. MO and
376 BH contributed expertise and advice. The manuscript was initially prepared by CR, assisted by HR. All
377 authors reviewed and edited the manuscript draft. All authors approved the final version for submission.

378 **Competing interests**

379 The authors declare that they have no conflict of interest.

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384

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