

Brief Communication: Decadal changes in topography, surface water and subsurface structure across an Arctic coastal tundra site

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Abstract. In ice-rich polygonal tundra, spatiotemporal heterogeneity in ground-ice melt reshapes topography, impacting infrastructure, water and carbon cycles. This study evaluates changes in topography and subsurface structure at a coastal Arctic site by comparing data from two surveys conducted a decade apart. Each survey includes electrical resistivity tomography, active layer thickness, photogrammetry, and topographic data. Results reveal subsidence and decrease in permafrost table elevation with varying intensity and spatial distribution across polygons, alongside diverse thermal-hydrological responses, such as thermokarst pool formation in high-centered-polygons and more even subsidence in flat-centered-polygons. The study also underscores the value and limitations of sporadic surveys.

1 Introduction

The Arctic region holds one of the world's largest carbon stocks, is warming three to four times faster than the global average, and is experiencing extreme changes as ground temperatures cross the 0 °C threshold (Rantanen et al., 2022; Tarnocai et al., 2009). Permafrost thaw can lead to major shifts in topography, water and carbon storage and fluxes, energy balance, vegetation, and erosion rates. The presence of complex feedback mechanisms within and between Arctic ecosystems and atmospheric processes introduce considerable uncertainty into climate projections (Koven et al., 2011).

Ice-wedge polygon networks, underlain by abundant and complex ground ice (wedge, vein, segregated, and pore ice), cover large areas of the Arctic tundra, with regional studies reporting up to 65% of terrestrial surface coverage in places such as the Arctic Coastal Plain of Alaska (Lara et al., 2020; Shur et al., 2025). In these environments, ice-wedge aggradation and melt reshape the surface into polygons of various sizes and shapes, typically referred to as low-centered, flat-centered, or high-centered depending on the relative elevation of their centers compared to surrounding rims and troughs (Mackay, 2000). These polygon types correspond to different evolutionary states of ice wedges and are associated with distinct water distribution patterns, vegetation communities, and carbon fluxes (Shur et al., 2025). As ground-ice melts with a warming climate, significant and uneven subsidence due to heterogeneity in ice content and thaw rate within and across polygons leads to changes in surface and subsurface flow, accumulation of surface water (hereafter referred to as pool expansion), snow distribution, and

28 thermal fluxes. These changes, in turn, influence microbial activity, vegetation dynamics, and the partitioning of carbon
29 between carbon dioxide and methane emissions (Lara et al., 2020; Li et al., 2024). Importantly, in low-relief terrain where
30 polygons are abundant, even fine-scale topographic changes can propagate to large-scale shifts in drainage networks and the
31 extent of dry, wet, and inundated areas.

32 Understanding of how surface and subsurface hydrology and carbon fluxes evolve in polygonal tundra require field-
33 based assessments of the co-evolution of microtopography, water distribution and subsurface structure. Such investigations
34 remain rare despite the wide range of environmental conditions that increase system complexity and necessitate broad sampling
35 (Abolt et al., 2024; Jorgenson et al., 2006; Streletskiy et al., 2017). The study by Jorgenson et al. (2006) highlights an abrupt
36 increase in permafrost degradation, primarily in massive wedges of ice, on the central Beaufort Coastal Plain in northern
37 Alaska. The area of degrading ice wedges (negative net ice mass balance) increased from 0.6% in 1982 to 4.4% by 2001. More
38 recently, the work by Abolt et al. (2024) provides insights into thermokarst pool expansion across 27 circumpolar field sites,
39 finding that 44% ($\pm 15\%$) of survey areas experienced pool expansion greater than 3% per year, with an 11% annual increase
40 between 2010 and 2019 at a lowland site near Utqiagvik in the northern Alaska (Abolt et al., 2024).

41 Assessing the co-evolution of permafrost thaw, topography, and water distribution is hindered by the lack of high-
42 resolution and multi-scale observations of terrain and subsurface properties (Dafflon et al., 2016). A particular challenge is the
43 difficulty of quantifying, over space and time, the ice content across the landscape and the associated subsidence driven by ice
44 loss and ground compaction as permafrost thaws (Kokelj & Jorgenson, 2013). Additionally, measurements of ground
45 properties are often referenced to the ground surface – a shifting reference point – introducing uncertainties that are frequently
46 overlooked. For instance, long-term measurements of active layer thickness underestimate the rate of permafrost thaw when
47 subsidence is not accounted for (Rodenhizer et al., 2020; N. I. Shiklomanov et al., 2013; Streletskiy et al., 2017). Rodenhizer
48 et al. (2020) indicate that accounting for subsidence increases estimated permafrost thaw by 19–49% and the estimated amount
49 of newly-thawed carbon by 37–113%. Collectively, these studies underscore the importance of jointly assessing potential
50 changes in topography, water distribution, and subsurface structure.

51 In this brief communication we explore how changes in surface water distribution are associated with changes in
52 topography and permafrost thaw across various polygon types. To this end, we evaluate decadal changes in surface water
53 distribution, thaw layer depth, terrain elevation, and subsurface structure from the meter to hundreds-of-meters scale over a
54 12-year period across a polygonal tundra area near Utqiagvik, Alaska. Besides evaluating the organization of these changes,
55 this study aims to discuss the value and limitations of ground-based methods for sporadic assessment of the evolution of
56 polygonal ground.

57 **1.1 Study Site**

58 The study site, located within the Barrow Environmental Observatory (BEO) on the Alaskan Arctic Coastal Plain, lies
59 approximately 6 kilometers from the Beaufort Sea near Utqiagvik, Alaska (Fig. S1). The landscape is characterized by low

60 topographic relief, with elevations ranging from 2 to 6 meters, and is dominated by polygonal ground formations (Hinkel et
61 al., 2001). The polygon evolution processes defines the microtopography, while larger topographical patterns are shaped by
62 processes such as ice accumulation, thermokarst lake, drainage network evolution, and landscape uplift (Kanevskiy et al.,
63 2017). Polygons at the site include high-centered polygons having elevated centers associated with drier conditions, low-
64 centered polygons exhibiting wet centers with well-defined rims and troughs, and flat-centered polygons representing an
65 intermediate form, with flat terrain and wet troughs (Mackay, 2000). The permafrost temperature at the study site remains
66 well below 0°C, fluctuating between -12°C and -5°C at a depth of 3 meters, yet the permafrost is not always entirely frozen
67 due to the presence of saline permafrost. Core samples reaching 3 to 6 meters depth around Utqiagvik, AK indicate salinity
68 up to twice that of seawater (Brown, 1969; Dafflon et al., 2016).

69 This study takes place in a 475 m long, 20 m wide corridor crossing high-centered, flat-centered and low-centered
70 polygon areas (Dafflon et al., 2016). This area was intensively studied between 2011 and 2014 to improve the understanding
71 of the co-variability of microtopography with surface and subsurface properties (Dafflon et al., 2016; Wu et al., 2018), the
72 development of zonation approaches to estimating subsurface properties (Wainwright et al., 2015), and the advancement of
73 parameterization and evaluation of terrestrial ecosystem models (Bisht et al., 2018). Dafflon et al. (2016) evaluated terrain
74 elevation using UAV-based photogrammetry, subsurface structure using Electrical Resistivity Tomography (ERT), and soil
75 properties by analyzing soil core and measuring thaw depth. Their study indicates that the saline permafrost is spatially
76 continuous, depth-variable, and that the increase in salinity is correlated with a decrease in ice content. Soil cores reveal
77 volumetric ice content as high as 85% in the top 1 meter of permafrost and as low as 40% at 2.5 meters depth. Further, their
78 analysis shows that the ice content in the top 2 m is greater in high-centered and low-centered polygon areas than in the flat-
79 centered polygon area. These results underline the covariability between topography and permafrost characteristics including
80 ice content and salinity distribution.

81 In the present study, we investigate how terrain elevation, geomorphology, surface water distribution, and subsurface
82 characteristics observed by Dafflon et al. (2016) in 2011-2013 have changed in 2023, this using a similar characterization
83 approach. During that period, meteorological data shows a warming trend at our field site, with decreasing freezing degree
84 days and increasing thawing degree days (NOAA Barrow Atmospheric Baseline Observatory,
85 <https://gml.noaa.gov/data/data.php?category=Meteorology&site=BRW>) (Fig. S2) (Farquharson et al., 2022). Data from the
86 Circumpolar Active Layer Monitoring (CALM) site near Utqiagvik indicate an average thaw depth increase of 0.3 cm/year for
87 the same period, though inter-annual variability is significant (N. Shiklomanov, 2023) (Fig. S2). In addition, a study evaluating
88 subsidence in the same region over the 2003–15 period reports a subsidence trend of 0.4–1.0 cm/year, resulting in a net
89 elevation change of 8–15 cm over that period (Streletskiy et al., 2017). These datasets demonstrate the presence of a warming
90 trend and subsequent geomorphological changes in the region. Besides the overall warming trend (Fig. S2), observations from
91 the nearby CALM site indicate that in the years data were acquired for this study the average thaw depths are consistent with
92 the trend in warming, with means of 39.3, 40.3, and 41.7 cm in 2011, 2013, and 2023, respectively (Fig. S2). Summer weather
93 forcing, which can modulate surface water distribution and evapotranspiration, show moderate variations among these years,

94 with June–July–August precipitation totals of 64.6, 94.6, and 83.6 mm, and mean summer air temperatures of 3.64, 3.66, and
95 4.51 °C in 2011, 2013, and 2023, respectively (Fig. S3).

96 **2 Methods**

97 **2.1 Digital Surface Model, Orthomosaic, and Surface Water Distribution**

98 In July 2013, a kite-based aerial system was used to capture high-resolution imagery of the study area (Dafflon et al., 2016).
99 A Sony NEX-5R digital camera was mounted on a Picavet rig positioned approximately 2 meters below the kite line, enabling
100 stable, downward-looking shots from an altitude of about 45 meters. Images were captured automatically at 3-second intervals,
101 covering roughly 73 by 48 meters per frame with a pixel resolution of 0.015 meters. In September 2023, aerial imagery was
102 acquired using a similar methodology, but with a DJI Mini 2 UAV platform. For both surveys (2013 and 2023), ground control
103 targets, spaced approximately 20 meters apart, were positioned along the corridor and surveyed using a centimeter-grade real-
104 time kinematic (RTK) differential GPS. Agisoft Metashape software was used to extract features, align images, and produce
105 georeferenced outputs, including a digital surface model and RGB orthomosaic for each campaign. Analysis of the data
106 products indicated that the digital surface model can be used as a terrain model proxy because of the low-stature vegetation,
107 and that the products achieved a spatial resolution of approximately 0.05 m, with an estimated positioning accuracy within 6
108 cm across all axes (x, y, z) based on multiple RTK GPS surveys. In addition, inaccuracies in RTK base station setup and
109 correction, which are difficult to assess in evolving landscapes, can introduce a constant shift of a few centimeters between
110 surveys. The digital surface model and RGB orthomosaic were used to evaluate changes in water distribution and topography
111 using Python and Matlab. Visualization Toolkit (VTK) and ParaView were used to generate a 3D animation showcasing
112 topography, vegetation, and water distribution in 2013 and 2023 [https://youtu.be/1_5yZp7yw94].

113 A Random Forest model was applied to the 2013 and 2023 RGB orthomosaics to classify water versus non-water
114 pixels. Given the distinct visual characteristics between the two years –with the 2023 image containing more visible surface
115 water, and the 2013 image showing denser vegetation with underlying water– manual pixel selection for training was required
116 to improve model accuracy. The training dataset consisted of approximately one hundred thousand manually labeled pixels
117 from a subset area of the orthomosaics. Each year’s labeled pixels were split into an 80 % training set and a 20 % held out test
118 set. A random forest classifier was then trained on four features, the RGB channels plus the blue to green channel ratio. On the
119 independent test set, the models achieved an accuracy of ~98% for both years. The models were applied to the entire
120 orthomosaics to predict water presence at the pixel level, and the resulting maps were subtracted to visualize surface-water
121 changes.

2.2 Subsurface Properties and Structure

A 475-meter long electrical resistivity tomography (ERT) transect along the centerline of the studied corridor was acquired in September 2011 and 2023 using a 112-electrode SuperSting R8 system (Advanced Geosciences, Inc.) with a 0.5 m electrode spacing, and a roll-along acquisition technique (Dafflon et al., 2016). Electrode locations were measured using a RTK differential GPS in 2011 and 2023. In 2023, the electrodes were placed along a measuring tape installed between control points every 50 meters that were either wooden poles remaining from the 2011 survey or re-surveyed based on the 2011 coordinates. While the RTK GPS-surveyed x–y coordinates were deemed precise enough for the goals of this study, the RTK-GPS elevation measurements contained some noise and inaccuracies associated with inconsistencies among the short surveys conducted over the three days of ERT acquisition, including in GPS signal quality, base station settings, and GPS pole positioning across organic-rich areas and in the presence of surface water. To increase accuracy, the elevation for each electrode was extracted from the digital surface models, whose ground control targets were all surveyed consistently under ideal conditions

The collected ERT data used a dipole-dipole survey setup with a maximum distance of 18 meters between the nearest injection and potential electrodes and a maximum distance between adjacent injection (or potential) electrodes of 3 meters. Data quality control included the exclusion of low-quality measurements, including those with signal potentials below 3 mV. The acquisition of reciprocal measurements for about 15% of datasets indicates a mean percentage difference of about 6%. ERT datasets were processed and inverted using the open-source *pygimli* library and its *ERTManager* module (Rücker et al., 2017). A parameterized mesh was created for each campaign based on sensor positions, with mesh quality settings and electrode spacing tailored to improve resolution and reduce inversion artifacts. Differences between observed and modeled apparent resistivities were calculated to quantify model accuracy. For the 2011 dataset, a chi-squared value of 1.93 was achieved, along with a mean percentage difference of 12.30% and a log-scale mean percentage difference of 2.41%. The 2023 dataset achieved a chi-squared value of 1.94, a mean percentage difference of 12.03%, and a log-scale mean percentage difference of 2.47%. Although the 2011 and 2023 surveys were both acquired in September, we note that differences in the ERT data need to be carefully interpreted given the inter-annual variability in climate forcing that impacts surface water distribution, thaw layer depth and moisture (Fig. S3), as well as the slight differences in electrode positioning and topography that lead to different mesh grid for the inversion.

In addition, the thaw depth was measured at 3 m intervals along the ERT transect by pushing a tile probe into the ground until reaching frozen soil. Taken at the close of the warm season, these measurements are used as a proxy for active layer thickness, the soil layer overlying permafrost that thaws in summer and refreezes in winter. For data comparison, the nearby electrical resistivity of the active layer was estimated by averaging the resistivity values for mesh cells in the 0.5 m lateral interval around each measurement and from 0 to 0.5 m depth. Resistivity was also extracted from the 0.5 to 1 m and 0.5 to 1.5 m depth intervals using the same approach.

The distribution of active layer thickness, resistivity, and topography was then visualized for each polygon type and surveyed year. The year-to-year differences (2011 vs. 2023) for each polygon type were assessed with the Mann–Whitney U

test, which compares rank distributions of independent samples, with the Wilcoxon signed-rank test, which evaluates paired differences, and with the Hodges–Lehmann estimator of the median difference and its confidence interval. This was applied across all polygon types, without differentiating troughs, rims, and centers due to limited thaw-depth data. This method offers an integrated measure of local changes, expected to be greater in ice-rich areas or those warming faster. Individual datasets (e.g., ERT) can further localize these changes.

3 Results and Discussion

3.1 Spatio-temporal Variability in Microtopography and Surface Water Distribution

Our results indicate that decadal changes in topography and surface water are highly variable across the landscape, both in magnitude and distribution, with localized hotspots of permafrost degradation and surface water accumulation (Fig. 1 and Fig. 2). Locations with significant increases in trough width and water accumulation are more noticeable in high-centered and flat-centered polygons than in low-centered polygons. In the high-centered polygon area, a limited number of relatively large pools formed in previously dry areas. By contrast, the increase in surface water in the flat-centered polygon area is more evenly distributed, primarily associated with trough widening and small-scale variability in ground subsidence (Fig. 2). These fine-scale changes in the flat-centered polygon area appear to contribute to an overall increase in surface water that is similar to or greater than that observed in the high-centered polygons area.

The observed increase in surface water accumulation along the transect aligns with the findings of Abolt et al. (2024), who estimated an average 11% annual net increase in thermokarst pool extent between 2010 and 2019 at a lowland site near Utqiagvik (Abolt et al., 2024). While their study underlines that some pools shrink while new ones form, our study does not indicate the presence of areas with a significant decrease in surface water. However, our acquisition timing and frequency may limit sensitivity to detecting decreases. Nonetheless, our results highlight varying characteristics of permafrost degradation across areas with different polygonal topography.

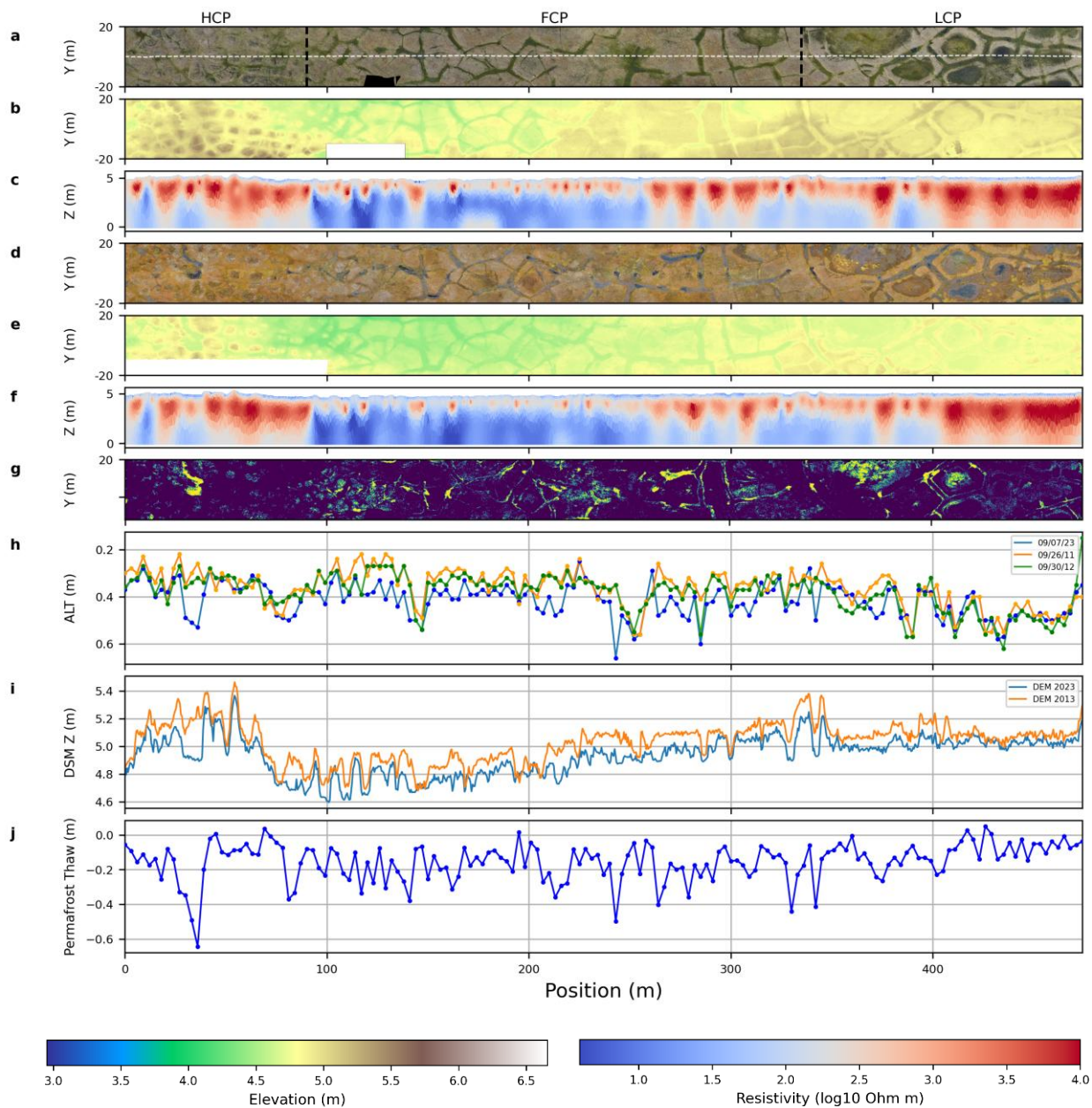
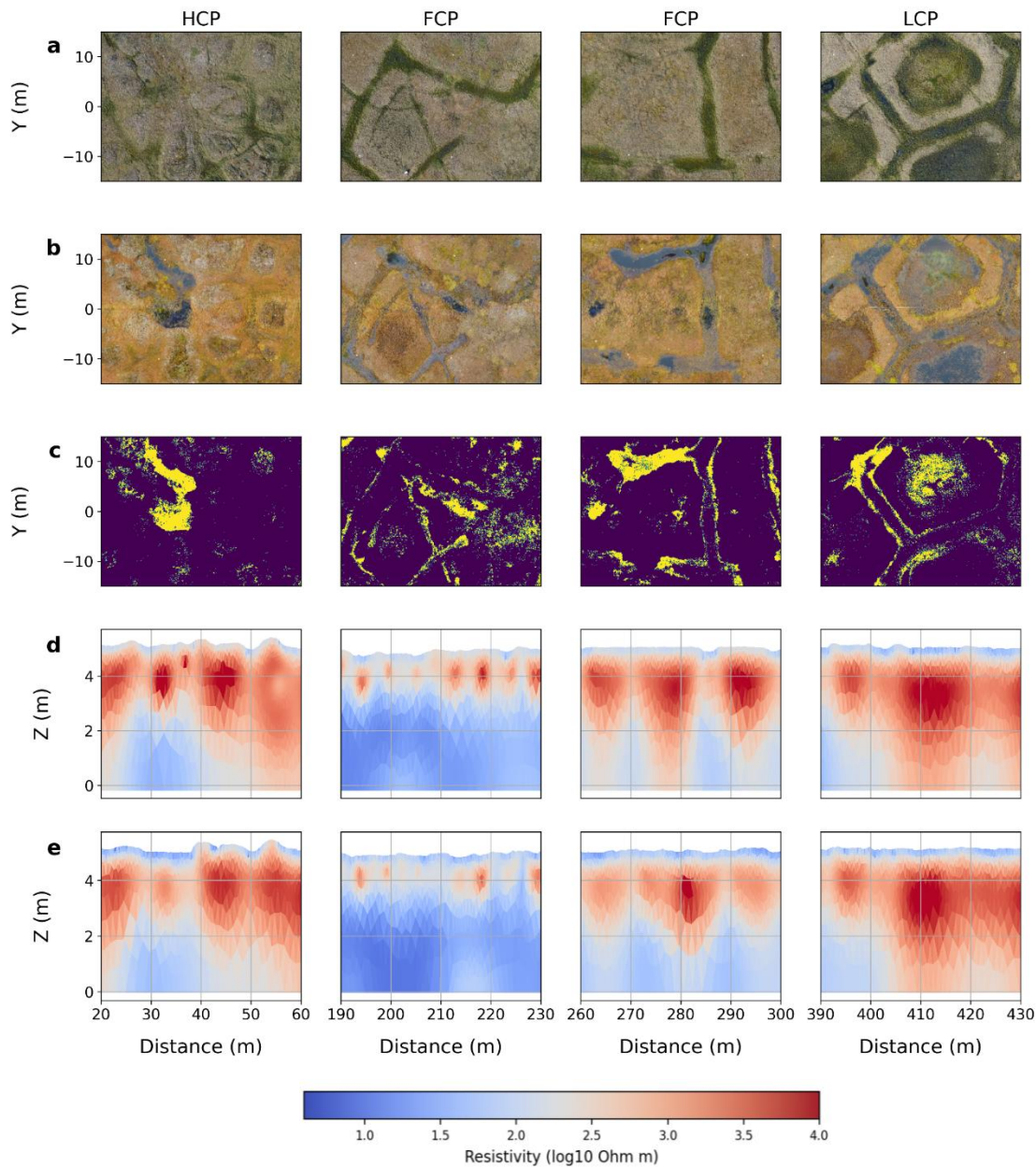


Figure 1. Surface and subsurface properties across various polygon types, including high-centered (HCP; left), flat-centered (FCP; middle) and low-centered (LCP; right). Datasets include a) 2013 RGB orthomosaic, b) 2013 digital surface model (DSM) elevation, c) 2011 ERT, d) 2023 RGB orthomosaic, e) 2023 DSM elevation, f) 2023 ERT, g) pool expansion (in yellow, binary map) based on the RGB orthomosaics and a random forest algorithm, h) active layer thickness (ALT) measurements, i) elevation from the DSMs along the ERT transect, and j) estimated permafrost thaw (defined as change in permafrost table

182 elevation) based on changes in ALT and ground elevation. Note that in (g) field notes and visual observations indicate that the
183 pool expansions in the center of LCPs are classification artifacts, presumably due to the particularly complex, changing mixture
184 of water, solids, moss, and sedge at these locations.

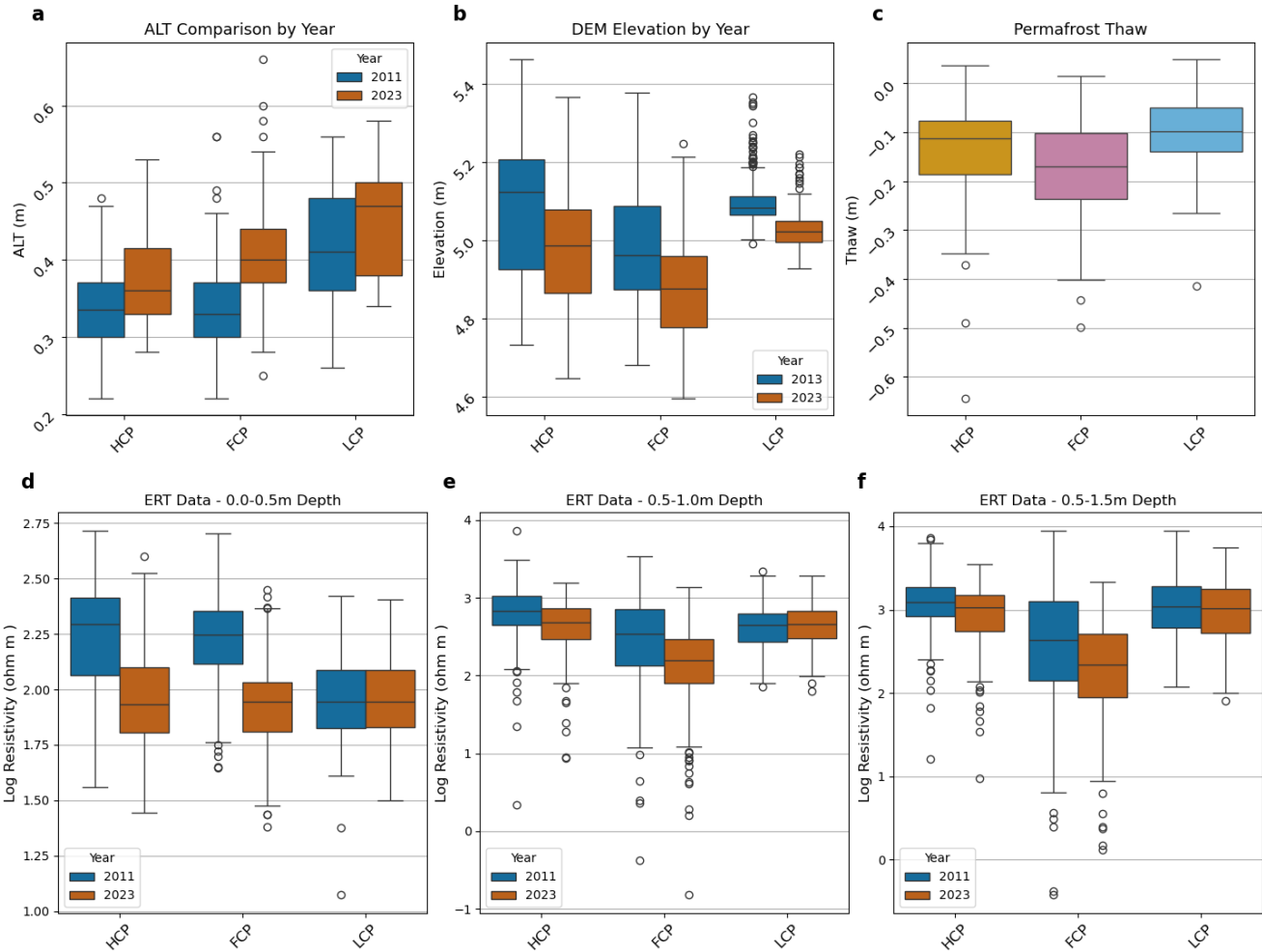


185
186 **Figure 2.** Decadal changes in surface and subsurface properties in four areas. Datasets include a) 2013 RGB orthomosaic, b)
187 2023 RGB orthomosaic, c) pool expansion (in yellow) based on the RGB orthomosaics and a random forest algorithm, d) 2011

ERT, and e) 2023 ERT. Note that in (g) field notes and visual observations indicate that the pool expansions in the center of LCPs are classification artifacts, presumably due to the particularly complex, changing mixture of water, solids, moss, and sedge at these locations.

3.2 Controls on Increased Surface Water, Thawing, and Subsidence

The observed decadal changes in surface water are accompanied by significant variations in active layer thickness and resistivity, which differ by polygon type (Fig. 1). Mann–Whitney U tests indicate significant increases in median active layer thickness over the study period in flat-centered polygons ($p < 0.001$) and high-centered polygons ($p = 0.0491$), but not in low-centered polygons ($p = 0.083$) (Fig. 3; with additional statistical tests results in Table S1). These changes are consistent with the observations of stronger surface subsidence and decrease in permafrost table elevation in high-centered and flat-centered polygons than in low-centered polygons (Fig. 1 and Fig. 3).



199 **Figure 3.** Changes by polygon type (HCP = high-centered, FCP = flat-centered; LCP = low-centered): a) Active layer thickness
200 (ALT), b) elevation, c) permafrost thaw (defined as change in permafrost table elevation), and average resistivity for the d) 0-
201 0.5, e) 0.5-1 and f) 0.5-1.5 m depth interval. p-values from the Mann–Whitney U test comparing 2011 and 2023: a) HCP:
202 0.049, FCP: <0.001, LCP: 0.083 b) HCP: <0.001, FCP: <0.001, LCP: <0.001, d) HCP: <0.001, FCP: <0.001, LCP: 0.749, e)
203 HCP: <0.001, FCP: <0.001, LCP: 0.316, and f) HCP: <0.001, FCP: <0.001, LCP: 0.315. See Table S1 for additional statistical
204 test results.

205

206 Electrical resistivity in the 0–0.5 m and 0.5–1 m depth intervals decreased significantly in both high-centered and
207 flat-centered polygon areas ($p < 0.001$), consistent with a deepening of the active layer. This aligns with the expectation that,
208 while the top organic layer may dry slightly during thaw, the overall volume of thawed material increases, with the lower
209 portion of the thaw layer remaining wet due to ice melt and/or other hydrological processes (Dafflon et al., 2017; Rodenhizer
210 et al., 2020; Tran et al., 2018). In contrast, no significant resistivity changes were observed in these depth intervals in the low-
211 centered polygons area ($p > 0.3$), corroborating a slower rate of permafrost degradation. Beyond the changes classified by
212 polygon type (Fig. 3), visual inspection along the ERT transects (Figs. 1 and 2) indicates that while some large changes in
213 resistivity are collocated with wide troughs and ice wedges, many others are associated with more subtle depressions or
214 expressed diffusely across the landscape.

215 Though high- and flat-centered polygon areas both show relatively similar changes in active layer thickness, ground
216 surface elevation, and surface water accumulation, the spatial patterns of these changes are distinct due to different subsurface
217 structures. ERT and soil core data from the site show that the high-centered polygon area is characterized by large troughs
218 relative to polygon diameter, a large volume of ice wedges beneath the troughs, and numerous ice veins throughout the
219 polygons (Fig. 1c). This ice presumably originates from freshwater and is therefore more resistive compared to surrounding
220 saline material (Dafflon et al., 2016). Such zones with large troughs and ice content have been identified as more prone to
221 further degradation (Kanevskiy et al., 2017). Indeed, the newly formed pools in the high-centered polygons area observed in
222 Figure 2 shows surface characteristics similar to those reported at a different site (Figures S12–S16 in Kanevskiy et al., 2017).
223 Although the exact conditions that led to pool formation are unknown, Kanevskiy et al. (2017) reported that these features are
224 typically triggered by an increase in active layer thickness during exceptionally warm and wet summers. Additionally, a snow
225 survey across various polygon types at a nearby site found that high-centered polygons are associated with much higher spatial
226 variability in snow depth than flat- and low-centered polygons (Wainwright et al., 2017), which could participate to the
227 particularly heterogeneous subsidence observed in the high-centered polygons area.

228 In contrast with the high-centered polygon area, the flat-centered polygon area exhibits smaller ice wedges beneath
229 the troughs, less ice veins, ice-rich permafrost limited to shallow depths, and a strong salinity gradient with depth (Fig. 1c)
230 (Dafflon et al., 2016). The smaller amount of ice in flat-centered polygon compared to high- and low-centered polygons areas
231 may contribute to faster degradation as less latent heat must be absorbed. The relatively low amount and heterogeneity of ice-
232 content in this area are also expected to constrain the magnitude and variability of subsidence. Our observations of subsidence,

233 along with the fact that the topography is relatively flat, align closely with this conceptual model. The flat-centered polygon
234 area shows relatively even subsidence across the area, with small changes occurring at many locations, including in trough,
235 rim and center.

236 Finally, the low-centered polygon area shows less significant increases in active layer thickness, degradation, or
237 subsidence. Low-centered polygons tend to have a large amount of surface water, and geophysical and soil core data tend to
238 show that the low-centered polygon area has a greater ice-content than the flat-centered polygon. While their ice-content is
239 more similar to high-centered polygon, the permafrost structure is more ataxitic with a lower presence of ice veins and ice
240 wedges (Fig. 1c and Fig. 2) (Dafflon et al., 2016). The low-centered polygon landscape and subsurface structure appear to
241 contribute to more stable conditions during the period of this study.

242 Overall, while this study does not attempt to fully disentangle the effects of surface water (dominant in low-centered
243 polygons), microtopography-driven snow distribution, and the role of ice-rich material on latent heat and subsidence, the
244 results demonstrate that subsidence occurs with varied patterns across the landscape, including varying intensity and
245 unevenness. Spatially variable subsurface characteristics and thaw rates are key drivers of the amount and distribution of
246 subsidence. In addition, the results of this study reinforce the conclusion of Kanevskiy et al. (2017) that understanding the
247 contrast between degradation and stability across areas, and the associated evolution of the landscape, remains complex. This
248 complexity remains a major challenge to understanding trajectories of ground stability and distribution of surface and
249 subsurface water, processes that are essential for infrastructure management, land transition assessment, and the
250 parameterization of Earth system models.

251 **3.3 Challenges in Capturing Variability in Thermal and Hydrological Trajectories**

252 The results of this study highlight the value of collecting independent datasets that capture above-ground, ground-surface, and
253 subsurface properties to gain a comprehensive understanding of hydrological changes in permafrost systems. Multiple datasets
254 are needed due to the complexity of geomorphological and hydrological changes occurring during permafrost degradation, as
255 well as the uncertainties inherent in individual measurements, which may lead to misinterpretation when considered
256 independently. For example, integrating thaw depth measurements with ground elevation changes for estimating permafrost
257 degradation is crucial to avoid the underestimation of thaw rates, ice volume loss, and soil mass loss (Rodenhizer et al., 2020).
258 In the present study, where decadal changes in thaw depth are expected to be limited compared to lower latitudes where both
259 permafrost and air temperatures are warmer, measuring variation in topography has still proven to be important for capturing
260 small changes that may accelerate as ice melts. Subsidence is consistent with variations in thaw depth when averaged across
261 the various polygon types, but they are not strongly correlated across each individual point due to the heterogeneity of ice
262 content and thaw rates. Overall, the convolution of spatially varying thaw rates and ice content creates large heterogeneity in
263 hydrological and topographic changes.

Capturing fine-scale elevation and its temporal changes to estimate subsidence and changes in permafrost table elevation in Arctic permafrost environments poses significant challenges, particularly concerning uncertainty and accuracy. In-field assessment of ground surface elevation is complicated by the presence of a soft organic layer with highly variable water content and a poorly defined transition between the peat layer and vegetation (e.g., dwarf shrubs, graminoids, and mosses). The presence of surface water poses an additional challenge, and these factors together give rise to considerable uncertainty in locating ground-based measurements, which typically use the ground surface as a reference point. An additional source of uncertainty in ground elevation between repeated surveys arises from potential inaccuracies in the RTK base station coordinates, which are difficult to detect in evolving landscapes and can introduce a constant shift across the dataset. This uncertainty makes relative changes easier to interpret than absolute differences. Caution in acquiring and interpreting elevation measurements is especially critical in contexts like this study, where small spatial and temporal variations in ground surface elevation have a major impact on surface and subsurface hydrology. Overall, ground elevation and its associated uncertainty affect many data products, including ground subsidence, permafrost table elevation, topographic metrics (e.g., surface flow lines), and the comparability of time-lapse ERT datasets. In high-resolution ERT surveys, accounting for topographic variation complicates data comparison over time due to differences in surface conditions, inversion meshes, and the associated regularization and convergence processes.

The above-mentioned complexity underscores the importance of developing strategies to manage uncertainties as effectively as possible, and bridging temporal and spatial scales in coverage and resolution. While sporadic above- and below-ground surveys such as those conducted here are valuable for model evaluation and satellite data product development, their interpretation requires careful consideration of seasonal and interannual variability, as well as the limited spatial coverage that constrains their representativeness across the Arctic. Our study draws on datasets from years that align relatively closely with the long-term fitted trend in active layer thickness at the nearby Utqiagvik Circumpolar Active Layer Monitoring (CALM) site (Fig. S2 and S3). Results may differ, and the detection of broader trends can be limited if one of the acquisitions is conducted during an exceptionally warm or cold year. Improving the temporal resolution and associated understanding in subsurface variables, such as using CALM or other monitoring networks data, is highly valuable to weigh in inter-annual variability. Merging sporadic surveys with temporally continuous measurements is also becoming increasingly feasible due to the growing availability and affordability of autonomous instrumentation (Dafflon et al., 2017; Wielandt et al., 2023), as well as community efforts in discussing best practices in data acquisition, including for ERT (Herring et al., 2023). Further, due to increase in resolution, satellite imagery may become well-suited for integration with ground-based data to facilitate scaling while possibly better capturing changes in shallow surface water distribution with more surveys conducted under similar conditions of water color, turbidity, and vegetation cover (Abolt et al., 2024). The results of this study advocate for further integration of multiple datasets across a range of scales to improve the understanding of controls on pool formation and drying, assess changes in permafrost table elevation and ice content, and disentangle the impact of thawing mechanisms on water distribution from changes in lateral inputs caused by rerouted surface flow.

297 **4 Conclusions**

298 This study documents changes occurring within polygonal tundra landscapes over a decadal timescale in Utqiagvik, Alaska.
299 Our findings demonstrate that changes in surface water distribution occur and are closely linked to subsidence, permafrost
300 thaw, and decreased permafrost resistivity. Additionally, more intense thaw and ice loss is observed in high- and flat-centered
301 polygons than in low-centered polygons. The observed expansion of thermokarst pools and trough widening illustrates how
302 permafrost thaw can trigger local ground subsidence, a primary control on infrastructure hazards, landscape change, and
303 methane emissions. Yet, studies from sites with a wide range of geological history and environmental conditions are needed
304 to improve the predictability of such changes and their implications. The results support the need for future studies that
305 characterize the small- to large-scale feedback mechanisms associated with permafrost thaw. Also, this study underscores the
306 importance of combining multiple data products to capture the intensity and spatial heterogeneity of subsidence and its
307 controls. Such insights are essential for refining models that simulate Arctic ecosystem responses to a warming climate,
308 ultimately improving our understanding of how permafrost thaw contributes to changes in global carbon and water cycling.

309 **Data availability**

310 The 2023 survey, 2011 ERT, and 2013 UAV datasets presented in this study have been submitted to the ESS-DIVE repository
311 at <https://data.ess-dive.lbl.gov/datasets/doi:10.15485/2564382>, <https://data.ess-dive.lbl.gov/datasets/doi:10.5440/1233229>,
312 and <https://data.ess-dive.lbl.gov/datasets/doi:10.5440/1177858>, respectively.

313 **Author contributions**

314 BD, CU and JL planned this study and collected the various datasets in 2011, 2013, and 2023. JB and BD analyzed the datasets
315 and prepared the manuscript. JL and NT reviewed the manuscript. All authors contributed to the study and approved the final
316 version of the paper.

317 **Competing interests**

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

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