

1 The ABoVE L-band and P-band Airborne SAR Surveys

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41

42 **Abstract**

43 Permafrost-affected ecosystems of the Arctic-boreal zone in northwestern North America are
44 undergoing profound transformation due to rapid climate change. NASA’s Arctic Boreal Vulnerability
45 Experiment (ABoVE) is investigating characteristics that make these ecosystems vulnerable or resilient
46 to this change. ABoVE employs airborne synthetic aperture radar (SAR) as a powerful tool to
47 characterize tundra, taiga, peatlands, and fens. Here, we present an annotated guide to the L-band and P-
48 band airborne SAR data acquired during the 2017, 2018, 2019, and 2022 ABoVE airborne campaigns.
49 We summarize the ~80 SAR flight lines and how they fit into the ABoVE experimental design. We
50 provide hyperlinks to extensive maps, tables, and every flight plan as well as individual flight lines. We
51 illustrate the interdisciplinary nature of airborne SAR data with examples of preliminary results from
52 ABoVE studies including: boreal forest canopy structure from tomoSAR data over Delta Junction, AK
53 and the Boreal Ecosystem Research and Monitoring Sites (BERMS) site in northern Saskatchewan and
54 active layer thickness and soil moisture data product validation. This paper is presented as a guide to
55 enable interested readers to fully explore the ABoVE L- and P-band SAR data.

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57

58 **Short Summary**

59 NASA’s Arctic Boreal Vulnerability Experiment (ABoVE) conducted airborne synthetic aperture radar
60 (SAR) surveys of over 120,000 km² in Alaska and northwestern Canada during 2017, 2018, 2019, and
61 2022. This paper summarizes those results and provides links to details on ~80 individual flight lines.
62 This paper is presented as a guide to enable interested readers to fully explore the ABoVE L- and P-
63 band SAR data.

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66 **Keywords:** Airborne Synthetic Aperture Radar (SAR), Interferometric SAR (InSAR), Polarimetric
67 SAR (PolSAR), Tomographic SAR (tomoSAR), Arctic, tundra, taiga, boreal forest, permafrost, Arctic
68 Boreal Vulnerability Experiment (ABoVE)

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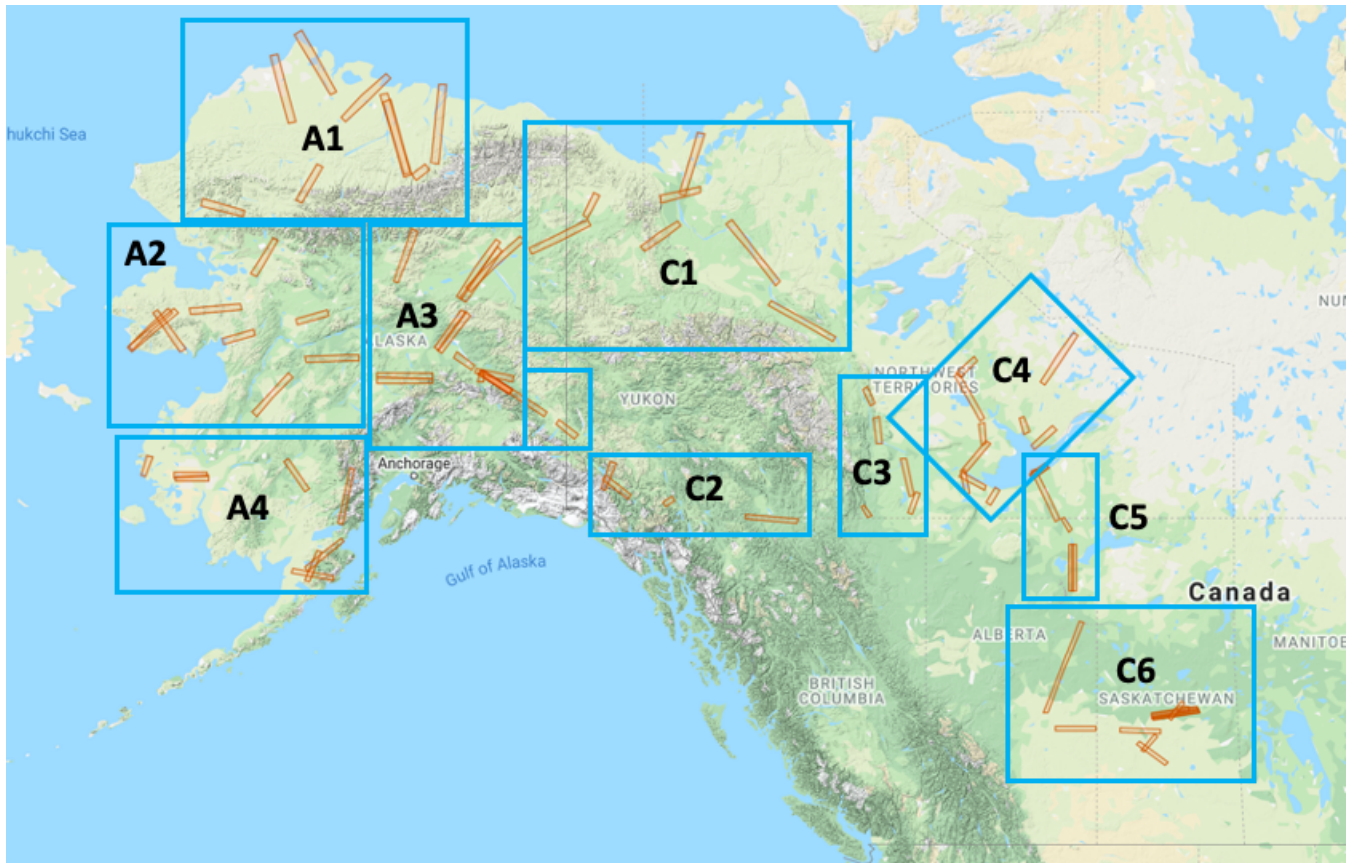
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76 **1 Introduction**

77 The Arctic region contains a remarkable diversity of cold-adapted biota, habitats, and permafrost-
78 affected ecosystems [McGuire et al. 2009; Vincent et al. 2011]. As with other components of the Arctic
79 system, Arctic ecosystems are strongly interdependent and the rapid degradation of the Arctic
80 cryosphere is altering their physical, biogeochemical, and biological linkages in ways that may be
81 irreversible [Vincent et al. 2011; Hinzman et al. 2013]. Understanding characteristics that make Arctic
82 ecosystems vulnerable or resilient to this change is the overarching objective of NASA's Arctic Boreal
83 Vulnerability Experiment (ABoVE, <https://above.nasa.gov/>). Miller et al. [2019] describes how airborne
84 campaigns fit into the broader ABoVE research strategy and how the foundational synthetic aperture
85 radar (SAR) measurements formed the framework around which all other airborne data acquisitions
86 were planned.
87
88



89
90 **Figure 1. Flight lines for the L-band and P-band PollnSAR measurements capture critical bioclimatic, permafrost, and geographic**
91 **gradients as well as key field sites and long-term measurement records across the 4 Mkm² ABoVE domain. The flight lines**
92 **are collected into 10 composites which roughly correspond to the Alaskan (A1-A4) and Canadian (C1-C6) regions sampled on**
93 **individual flight days. © Google Maps**

94 ABoVE SAR flight lines (Figure 1) were planned to leverage legacy L- and P-band SAR transects
95 acquired during the pre-ABoVE period; remotely-sensed permafrost active layer thickness time series
96 derived from satellite interferometric SAR observations (ReSALT) [Schaefer et al. 2015]; SAR data
97 from PALSAR, PALSAR-2, RadarSat, RadarSat-2, and Sentinel-1; historic or planned airborne LiDAR
98 acquisitions; and data from existing field sites [Hoy et al. 2018]. Legacy airborne SAR flight lines
99 include the L-band grid acquired over the Boreal Ecosystem Research and Monitoring Sites (BERMS)
00 area near Prince Albert, SK during SMAP CanEx 2010 [Magagi et al. 2012], the P-band lines over the
01 BERMS area acquired from 2012-2015 during the Airborne Microwave Observatory of Subcanopy and
02 Subsurface (AirMOSS) Earth Ventures Sub-orbital (EV-S1) investigation [Allen et al. 2010;
03 Moghaddam et al. 2016], and a collection of 10 L- and P-band flight lines acquired over the Seward
04 Peninsula, Northwestern Interior, and North Slope of Alaska during 2014 and 2015 [Chen et al. 2019a,
05 2019b]. The BERMS area observations, in particular, link ABoVE to the Boreal Ecosystem–
06 Atmosphere Study (BOREAS) studies of the 1990s [Sellers et al. 1995; 1997].
07

08 Hoy et al. [2018] compiled information on more than 6,700 field sites and previous remote sensing data
09 sets to help plan the SAR flight lines and the ABoVE Airborne Campaigns [Miller et al. 2019]. This
10 compilation is intended to help investigators understand flight line choices and identify ground locations
11 used to anchor individual flight lines. SAR data users may also search for the underlying data available
12 within each flight line. Key anchor points for the SAR flight lines include: Active layer thickness
13 measurements from the Circumpolar Active Layer Monitoring network (CALM); Permafrost
14 temperatures and annual thaw depths from the Global Terrestrial Network for Permafrost (GTN-P)
15 database; Soil moisture and permafrost state data from the Department of Energy’s Next Generation
16 Ecological Experiment-Arctic (NGEE-Arctic) field sites on the Seward Peninsula and near Utqiagvik
17 (formerly Barrow), AK; Extensive in situ terrestrial and aquatic ecosystem data as well as airborne
18 LiDAR and spectral imagery from NSF’s National Ecological Observatory Network (NEON) D18
19 tundra field sites near Utqiagvik (Barrow), AK and Toolik Lake, AK, and from the D19 taiga field sites
20 near Caribou/Poker Creek, AK, Delta Junction, AK, and Healy, AK; Detailed ecological and physical
21 climate time series from NSF’s Long Term Ecological Research (LTER) Arctic (Toolik Lake) and
22 Boreal Forest (Bonanza Creek) sites; Long-term boreal forest inventory data from the Canadian
23 Forestry Service’s (CFS) Climate Impacts on Productivity and Health of Aspen (CIPHA) and High
24 Elevation & Latitude Climate Change Impacts & Adaptation (HELICIA) plots; and Long term
25 permafrost, hydrology and ecology time series records from the Canadian Changing Cold Regions
26 Network (CCRN) sites at Trail Valley Creek, NWT, Havikpak Creek, NWT, Scotty Creek, NWT,
27 Baker Creek, NWT, Wolf Creek Research Basin, YT, and the BERMS site at White Gull Creek, SK.
28

29 Airborne SAR data enable numerous ecosystem and ecosystem change research investigations [NRC
30 2014]. ABoVE researchers are using the airborne L- and P-band data to: Quantify permafrost active
31 layer thickness and soil moisture content [Bakian-Dogahneh et al. 2020]; Complement AirSWOT Ka-
32 band acquisitions to determine water surface elevations in Arctic lakes, wetlands, and rivers [Pitcher et
33 al. 2019a,b]; Investigate boreal forest and tundra fire scars, especially in conjunction with fire
34 disturbance plots [Tank et al. 2018; Walker et al. 2018 a,b; 2019a,b; French et al. 2020; Holloway et al.
35 2020; Loboda et al. 2021]; Map tree density and distribution across the Tundra-Taiga ecotone; Provide

36 control point data for the ArcticDEM [Porter et al. 2018; Meddens et al. 2018]; Investigate lidar-radar
37 fusion remote sensing for boreal forest characterization as a precursor to NISAR/IceSAT-2
38 investigations [Silva et al. 2021]; Quantify expansion and sediment mass flow from massive
39 retrogressive thaw slumps – so-called megaslumps – on the Peel Plateau along the Dempster Hwy west
40 of Fort McPherson [Kokelj et al. 2013; 2015]; Classify Arctic wetlands and habitats [French 2020]; and
41 Support algorithm development for NISAR (L-band) and BIOMASS (P-band) estimates of boreal forest
42 structure and above ground biomass [Quegan et al. 2019; Saatchi et al. 2019]. Goetz et al. [2021]
43 summarizes how the ABoVE airborne SAR data are helping advance Arctic-boreal understanding and
44 the remaining knowledge gaps still to be addressed.

45
46 This paper presents an annotated guide to enable interested readers to fully explore the ABoVE L- and
47 P-band SAR data acquired during the 2017, 2018, 2019, and 2022 ABoVE airborne campaigns. Section
48 2 provides details on the L- and P-band SAR instruments and the flight line catalog. Section 3
49 summarizes the daily sorties from each airborne campaign. Section 4 briefly describes the tomographic
50 SAR (tomoSAR) experiments flown over Delta Junction, AK and the BERMS site near Prince Albert,
51 SK. Section 5 describes some of the ABoVE SAR data products and their validation. Section 6
52 highlights the synergies between the L- and P-band airborne SAR data and other airborne sensors.
53 Section 7 summarizes access to the data products. Section 8 discusses potential future acquisitions and
54 outlooks for exploiting these data. Additionally, we include an Appendix which describes the ~80
55 ABoVE SAR flight lines and how each line fits into the ABoVE experimental design. The Appendix
56 also provides extensive maps and tables for every flight plan and individual flight lines as well as a list
57 of the acronyms and abbreviations. The Supplemental Information includes hyperlinked versions of the
58 tables for direct access to flight lines and flight plans.

59 **2 The L-Band and P-band Airborne SAR Instruments and Data Acquisition**

60 Both the L- and P-band airborne SARs are sensitive to geometrical and material properties of
61 vegetation, soil surface, and subsurface profiles [Saatchi and Moghaddam 2000; Tabatabaenejad et al.
62 2011; 2015]. The joint use of both L- and P-band gives enhanced sensitivity to near-surface (< 5 cm, L-
63 band) and root zone (10-40 cm, P-band) portions of the subsurface profile compared to use of either
64 wavelength alone [Du et al. 2015]. Airborne acquisitions with both SARs provide 6-10 m spatial
65 resolution, ~15 km swaths and transect lengths of 100 – 200 km, making them ideal for surveying
66 above-ground biomass and vegetation canopy structure [Hensley et al. 2014; 2016] as well as the
67 tundra-taiga ecotone [Montesano et al. 2016]. Special tomoSAR data were acquired over the well
68 characterized BERMS site in northern Saskatchewan and the NEON site in Delta Junction, AK to
69 quantify the performance of both SARs in reproducing the structure and biomass of boreal forests.

70

71 **2.1 The L-band SAR Instrument**

72 NASA's airborne L-band SAR (initially named the Uninhabited Aerial Vehicle Synthetic Aperture
73 Radar (UAVSAR) system) is a compact pod-mounted polarimetric instrument for interferometric

74 repeat-track observations that was developed at the NASA Jet Propulsion Laboratory (JPL) and Dryden
75 Flight Research Center (DFRC) in Edwards, CA. It operates at a center frequency of 1.2575 GHz
76 (wavelength = 23.8 cm) with 80 MHz bandwidth. It is deployed on a Gulfstream III aircraft and images
77 the Earth surface from a nominal 12.5 km altitude. The image swath is collected off-nadir in a ~22 km
78 wide with incidence angles ranging from ~22°–67°. The instrument spatial resolution is 0.8m (along
79 flight-line) by 1.7m (slant range, along line-of-sight (LOS) from the antenna to the ground).
80 Topographic information is derived from phase measurements that, in turn, are obtained from two or
81 more passes over a given target region. Its 1.26 GHz frequency results in radar images that are well-
82 correlated from pass to pass. Polarization agility facilitates terrain and land-use classification.

83
84 All L-band SAR data are publicly available at <http://uavsar.jpl.nasa.gov/> as individual InSAR products
85 or as a single look complex (SLC) stack product of coregistered images for individual flight lines.
86 Products are also available from the UAVSAR data portal at the Alaska SAR Facility Distributed
87 Active Archive Center (<https://asf.alaska.edu/data-sets/sar-data-sets/uavsar/>). The L-band SAR provides
88 key pre-launch algorithm development and validation data sets [Saatchi et al. 2019] for the NASA-
89 ISRO SAR (NISAR) mission [Rosen et al. 2017].
90

91 **2.2 The P-band SAR Instrument**

92 The P-band SAR was developed circa 2012 for the Earth Ventures Sub-orbital (EV-S1) Airborne
93 Microwave Observatory of Subcanopy and Subsurface (AirMOSS) investigation [Allen et al. 2010;
94 Moghaddam et al. 2016]. The radar is based on JPL's L-band UAVSAR system. The P-band SAR
95 inherits UAVSAR's existing L-band RF and digital electronics subsystems. New up- and down-
96 converters convert the L-band signals to UHF frequencies (280-440 MHz). The passive antenna is based
97 on the legacy GeoSAR design [Chapin et al. 2012].
98

99 The P-band SAR was flown more than 1200 h from 2012 to 2015, covering regions of 2500 km² spread
00 over nine major biomes in North America during the AirMOSS EV-S1 investigation [Tabatabaenejad
01 et al. 2020]. Legacy acquisitions in Alaska [Chen 2019a, b] and over the BERMS site in northern
02 Saskatchewan [Chapin et al. 2012, 2018] provide an opportunity for extended time series analysis. All
03 P-band SAR data are publicly available at <http://uavsar.jpl.nasa.gov/>. Additionally, ABoVE P-band
04 SAR data will provide valuable insights into the characterization of boreal forest and tundra ecosystems
05 by the upcoming BIOMASS mission [Le Toan et al. 2011; Quegan et al. 2019].
06

07 **2.3 The Platform Precision Autopilot (PPA) System**

08 To support cm-precision interferometric land surface characterization, repeat pass measurements
09 acquired by the SARs need to be taken from flight paths that are nearly identical. Both the L- and P-
10 band SARs utilize real-time GPS that interfaces with the platform flight management system (FMS) to
11 confine the repeat flight path to within a 10 m tube over a 200 km course in conditions of calm to light
12 turbulence. The FMS is also referred to as the Platform Precision Autopilot (PPA). Additionally, the

13 radar vector from the aircraft to the ground target area must be similar from pass to pass. This is
14 accomplished with an actively scanned antenna designed to support electronic steering of the antenna
15 beam with a minimum of 1° increments over a range to exceed ±15° in the flight direction.
16

17 ABoVE SAR measurements were typically acquired with platform RMS deviations less than ±3 m. Any
18 platform deviations larger than ±10 m from the programmed flight path resulted in the acquisition being
19 terminated and a real time decision made to reacquire the line from the beginning or to continue with
20 the flight plan and proceed to the next line. This decision balanced the science priority of the flight line,
21 fuel consumption and remaining endurance, the number of flight lines yet to be acquired in the day's
22 flight plan, distance from our base of operations, and whether there would be future opportunities to
23 collect a given line by adding it to an upcoming flight plan. The flight team was extremely efficient in
24 executing these decisions, resulting in 95% flight line acquisition success across the 2017-2019 period.
25

26 **2.4 Airborne SAR Flight Line and Flight Plan Designations**

27 The JPL SAR team devised a convenient and powerful way to identify airborne SAR data acquisitions
28 for the Facility and PI instruments under their charge. Each L-band or P-band SAR flight line receives a
29 unique 5-digit identifier consisting of the three-digit GPS compass heading followed by a two-digit
30 index. A 6-character text string is also associated with each line for ease of identification. The text
31 string proceeds the numerical ID and usually provides abbreviated geographic or infrastructure
32 information that characterizes the line. For example, L-band flight line Teller_04901 identifies the flight
33 line on the Seward Peninsula that overflies the NGEE-Arctic Teller watershed. The flight line identifier
34 is a constant and, once assigned, is used whenever a line is reflown. In some cases, there are
35 overlapping or nearly identical flight lines which differ slightly in their ID number. L-band and P-band
36 flight lines use the same flight line identification system, allowing rapid identification of overlapping L-
37 and P-band data acquisitions.
38

39 Flight Plans are assembled from the composite flight lines for a given sortie. Each flight plan also
40 receives a unique 5-digit identifier based on the year flown (digits 1 and 2) and the flight number for
41 that year (digits 3-5). For example, L-band flight plan 17093 was flown in 2017 and was the 93rd sortie
42 flown that year. Note that there may be more than one sortie flown on a given day, in which case each
43 would have a unique flight plan identifier even though they were flown on the same calendar day and
44 may include some or all of the same flight lines.
45

46 In the Supplemental Information we provide hyperlinks to the JPL UAVSAR data portal
47 (<https://uavsar.jpl.nasa.gov/cgi-bin/data.pl>). This provides links to the individual flight line data, maps,
48 and related flight plans that acquired data over one or more of the individual flight lines. We hope this
49 enables interested readers to explore the ABoVE L- and P-band SAR data more fully. These data and all
50 other airborne data from the ABoVE campaigns may be explored on NASA's EarthData ABoVE Portal
51 (<https://search.earthdata.nasa.gov/portal/above/search>). Ground sites used to design the orientation and
52 locations of the flight lines are archived at the ORNL DAAC [Hoy 2018].

54 **3 The ABoVE Airborne SAR Campaigns**

55 The L-band (Figure 2) and P-band (Figure 3) SARs were considered foundational measurements in the
56 ABoVE airborne campaign strategy [Miller et al. 2019]. The ~80 flight lines described in the Appendix
57 formed the framework for the remainder of the airborne remote sensing acquisitions. The ABoVE SAR
58 strategy was to execute same day acquisitions of both L- and P-band flight lines (Figure 1) for a given
59 sortie during 2017 to optimize dual frequency retrievals; however, technical issues forced us to fly the
60 instruments sequentially. The baseline L-band campaigns were flown in June (DOY 164-173) and
61 September (DOY 251-263) of 2017 to characterize the land surface during periods of minimum and
62 maximum active layer thickness, respectively. Subsequent L-band campaigns in 2018 (DOY 231-241),
63 2019 (DOY 247-260) and 2022 (DOY 226-237) provide a time series synched to maximum annual
64 active layer thickness. P-band campaigns were conducted in May-June (DOY 142-157) and August
65 (DOY 219-227) of 2017. There was a 2-day P-band mini-campaign in October 2017 to extend the
66 legacy time series of early cold season acquisitions over the Seward Peninsula, NW Alaska and North
67 Slope Alaska (DOY 280-283).

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Figure 2. Sahtu students Mandy Bayha (front left) and Joanne Speakman (front center) pose with their mentor Cindy Gilday (front right) and NASA flight crew in Yellowknife, NT after completing a L-band SAR survey flight around the Great Slave Lake Region on 22 August 2018 (Flight Plan 18048). This experience gave these Northerners a new appreciation for how NASA was helping understand, preserve, and protect their lands. Photo Credit: Stephen M. Fochuk, Government of Northwest Territories.

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Figure 3. The P-band SAR team with the NASA JSC G-III (N992NA) on the tarmac in Fairbanks, AK on 18 August 2017 after completing a survey of the Upper Mackenzie Valley (Flight plan 17083). Photo Credit: M. Moghaddam.

80

81 3.1 Alaskan Flight Lines

82 The Alaskan SAR flight lines are broken into four main regional collections: A1) North Slope Alaska,
83 A2) Seward Peninsula and Northwest Alaska, A3) Eastern Interior, and A4) Southwest Alaska and the
84 Yukon-Kuskokwim Delta (**Figure 1**). Individual flight
85 lines were planned based on long-term ground
86 monitoring sites [Hoy et al. 2018], existing or planned
87 field research, recent disturbances, important geographic
88 or ecological gradients, complementary remote sensing
89 data, and consultation with indigenous peoples and
90 governments [Miller et al. 2019]. Legacy L- and P-band
91 flight lines from the AirMOSS EV-S1 investigation
92 [Allen et al. 2010; Moghaddam et al. 2016] in the
93 Seward Peninsula, NW Alaska, and the North Slope
94 were adapted for ABoVE use. Acquisition of P-band
95 flight lines in the central Interior was not possible due to
96 a military radar keep-out zone centered near Clear, AK.
97 The keep-out zone is shown in all P-band flight plan
98 maps (Ex. **Figure 4**).
99

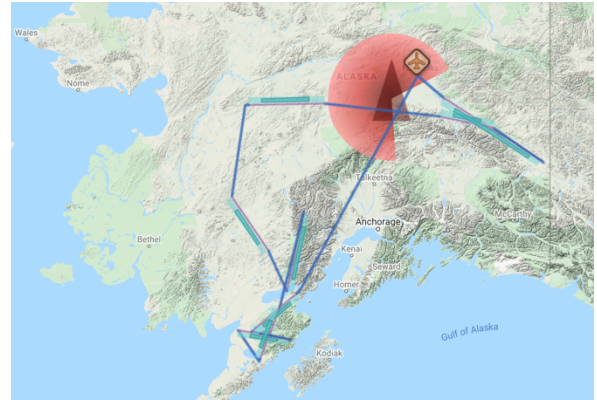


Figure 4. The military radar at Clear, AK creates a large P-band operations keep-out zone in the central Interior (red areas). The aircraft symbol marks our Fairbanks International Airport (PAFA) base of operations. Data acquisitions (blue bars) are from Flight Plan 17054. © Google Maps

00 3.2 Canadian Flight Lines

01 The Canadian SAR flight lines are broken into six regional collections: C1) Lower Mackenzie Valley
02 and Northern Yukon Territory, C2) Southern Yukon Territory, C3) Upper Mackenzie Valley, C4) Great
03 Slave Lake Region, C5) Transboundary Watershed, and C6) Southern Boreal Forest/BERMS.
04 Individual lines were planned based on long-term ground monitoring sites [Hoy et al. 2018], existing or
05 planned field research, recent disturbances, important geographic or ecological gradients,
06 complementary remote sensing data, and consultation with local inhabitants and governments [Miller et
07 al. 2019]. Legacy L- and P-band flight lines in the BERMS area from the CANEX 2010 campaign
08 [Magagi et al. 2012] and the AirMOSS EV-S1 investigation [Chapin et al. 2012, 2018] provide the
09 potential to establish longer time series.
10

11 Flight planning for the Canadian transects benefited tremendously from consultations with our
12 Canadian colleagues and interested parties in Yellowknife, NT and Whitehorse, YT in 2015 and 2016.
13 Extensive discussions with the Government of the Northwest Territories (GNWT), the Government of
14 the Yukon Territory, First Nations representatives, and scientists from Polar Knowledge Canada
15 (POLAR), the NWT Center for Geomatics, and the Canadian Forestry Service (CFS) Northern Forestry
16 Centre (NoFC) were critical to designing a strategy that captured many of their observing priorities.
17 Subsequent discussions in Yellowknife during 2017 and 2018 enabled us to disseminate preliminary
18 results and coordinate the flights with same-day field data acquisitions.
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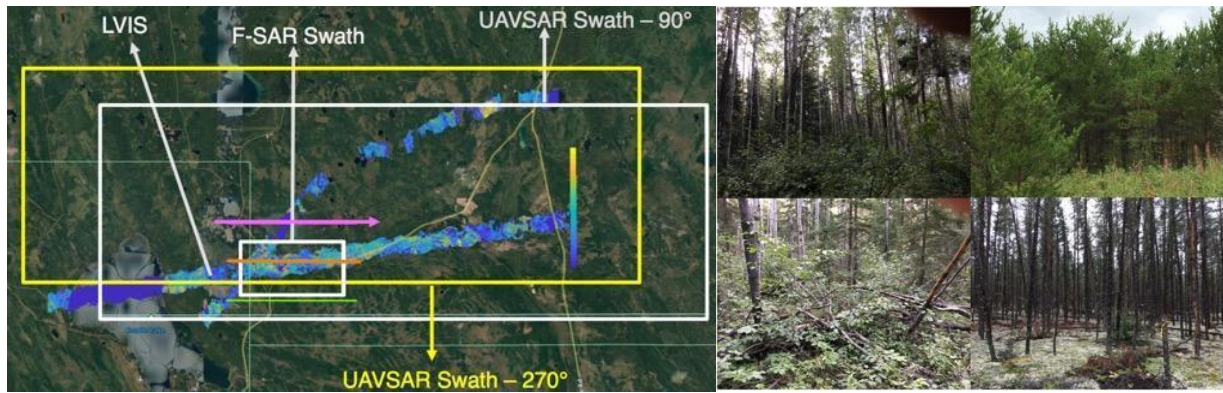
20 4 TomoSAR Measurements of Boreal Forest Structure

21 SAR tomographic methods have proven extremely adept at measuring vegetation vertical structure at a
22 variety of wavelengths including L- and P-bands. The three-dimensional vegetation structure and its
23 changes resulting from either natural or anthropogenic causes are key ecosystem monitoring parameters.
24 ABoVE collected tomographic L- and P-band SAR data over the boreal forest near Delta Junction, AK
25 in September of 2017. UAVSAR (L-band) and the German Space Agency's F-SAR (L- and S-bands)
26 acquired coordinated tomographic SAR data at the BERMS site near Saskatoon, SK in August 2018.
27 Ground truth data sets and LiDAR data from the NASA LVIS system were also acquired at BERMS in
28 2017 [Blair et al. 2018]. We compared L- and P-band tomography at Delta Junction and L-band and S-
29 band tomography from the two systems, to each other, and to the LiDAR data sets at BERMS. Here we
30 provide a preliminary analysis of the data acquired at BERMS.

31
32 BERMS is a southern boreal forest site with gentle topography dominated by Jack Pine and Aspen
33 stands. There is active logging in the area and the site contains clear cut areas and new growth stands in
34 various maturity states. The tomography data acquisition at BERMS was planned jointly in cooperation
35 with the German Space Agency (DLR) who flew the F-SAR radar and acquired data at L-band and S-
36 band. The UAVSAR and F-SAR flight lines were designed to overlap each other and LVIS data
37 acquired at the site in 2017. LVIS reacquired BERMS area data again in 2019 with the LVIS-F and
38 LVIS-C instruments [<https://lvis.gsfc.nasa.gov/Data/Maps/ABOVE2019Map.html>]. Figure 5 (left)
39 shows swaths for the UAVSAR and F-SAR radars along with the LVIS data. UAVSAR acquired L-
40 band tomography data on a racetrack pattern to get multiple incidence angle data for most points in the
41 swath. Because UAVSAR and F-SAR fly at 12500 mAGL and 4200 mAGL, respectively, it is not
42 possible to acquire data with the same incidence angles across the swath. Thus, we configured the flight
43 lines to overlap so that the 40° incidence angle points would coincide. Figure 5 (right) shows photos
44 collected at four of our seventeen ground truth sites during the tomoSAR acquisitions.

45
46 LVIS full waveform LiDAR provides surface elevations and tree height estimates as well as LiDAR
47 echo strength throughout the canopy and thereby information on the canopy internal structure. From
48 LVIS waveforms many products are possible including surface elevation, tree height, moments of the
49 returned waveform distribution and cumulative percentile elevations. We compared these waveforms to
50 radar tomographic profiles for the different radar wavelengths.

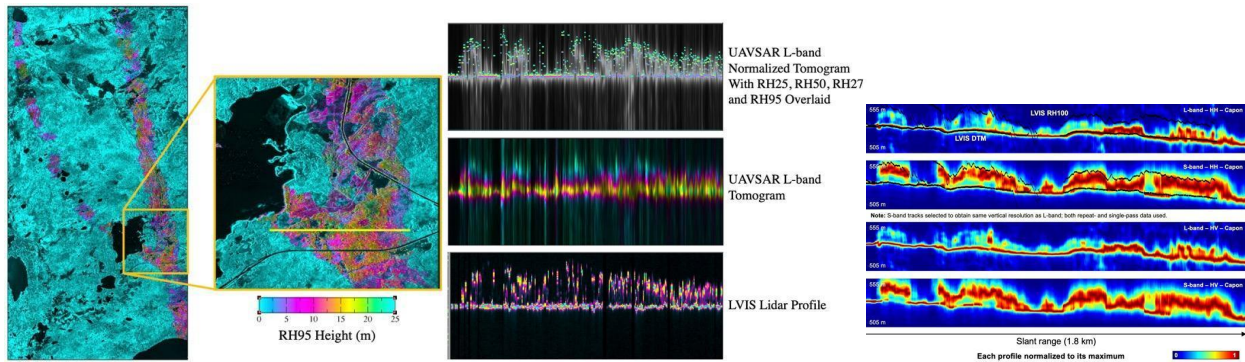
51
52 BERMS field measurements consist of soil moisture measurements at the 17 sites using the average of
53 15 measurements distributed over 60 m × 60 m plots on the day of the UAVSAR radar observations. At
54 BERMS the soil was very dry, roughly 10% volumetric soil moisture or less, during the radar
55 observations. During the summer of 2020 diameter at breast height (DBH) measurements, used to
56 estimate biomass, for a subset of our selected sites was planned but postponed due to COVID-19.
57



58
 59 **Figure 5.** LEFT: Experimental design for the BERMS area TomoSAR flights in August 2018. The large White and yellow
 60 boxes show the ~18 km-wide UAVSAR L-band swaths. The offset is due to the off-nadir viewing angle of the L-band SAR –
 61 it is pointed to the south when flying the 270° swath and pointed north when flying the 90° swath. The small white box near
 62 the center of the image marks the ~3 km-wide F-SAR swath. LVIS LiDAR data are the ~1.5 km-wide colored swaths across
 63 the image; the color scaling reflects the canopy height. RIGHT: Photos from four of the 17 plots used for in situ ground truth
 64 measurement at BERMS. Vegetation at BERMS was mostly Jack Pine and Aspen with many areas having dense understory
 65 vegetation. Most areas have substantial detritus and ground litter left over from previous logging operations. © Google Earth

66 **Figure 6** compares UAVSAR L-band tomography with F-SAR L-band and S-band tomography and
 67 LVIS LiDAR data along a transect shown as a yellow line on the right in the figure. Tree height along
 68 the transect varied from 10-20 m. Middle of Figure 2 is the UAVSAR L-band transect. Top of the figure
 69 shows the LVIS RH25m RH50, RH75 and RH95 profiles overlaid on the UAVSAR tomogram and
 70 below are the radar and LiDAR vertical profiles. On the left of Figure 2 are the F-SAR L-band and S-
 71 band tomographic profiles along with the LVIS RH100 data. The L-band radar profiles exhibit power
 72 concentrated at the base of the canopy whereas the LVIS LiDAR data show more return from the
 73 middle portion of the canopy. S-band obtains greater returns in the upper canopy compared to L-band
 74 and show more uniform scattering within the canopy.

75



76
 77 **Figure 6.** On the left shows location of transect as yellow line overlaid on UAVSAR imagery in grayscale and LVIS RH95 data is color
 78 where UAVSAR L-band, DLR F-SAR L and S-band and LVIS LiDAR profiles are compared. Center figure shows UAVSAR L-band
 79 tomographic profiles along the transect along with the corresponding LVIS LiDAR profiles. On the right are the corresponding F-SAR L
 80 and S-band tomographic profiles. The L-band radar profiles exhibit power concentrated at the base of the canopy whereas the LVIS LiDAR
 81 data show more return from the middle portion of the canopy. S-band obtains greater returns in the upper canopy compared to L-band and
 82 show more uniform scattering within the canopy.

83 5 ABoVE SAR Data Products

84 Here we highlight some data sets enabling or derived from the ABoVE L- and P-band airborne SAR
85 acquisitions. They represent the current state of the art for the study of permafrost-affected ecosystems
86 using SAR. The ABoVE science team continues to develop additional products and the insights from
87 these studies will be published separately. Links to the repositories for each of these data sets are
88 provided below.

89

90 5.1 Active Layer Thickness (ALT)

91 The Permafrost Dynamics Observatory (PDO) data product estimates seasonal subsidence, active layer
92 thickness (ALT), soil Volumetric Water Content (VWC), and uncertainties at 30-m resolution for 66
93 flight lines across Alaska and Northwest Canada [Michaelides et al. 2021; Chen et al. 2021a,b]. The
94 PDO retrieval uses L-band Synthetic Aperture Radar (SAR) data acquired by the Uninhabited Aerial
95 Vehicle Synthetic Aperture Radar (UAVSAR) instrument and P-band data acquired by the Airborne
96 Microwave Observatory of Subcanopy and Subsurface (AirMOSS) instrument. The PDO results for
97 each flight line appear in separate netcdf files. Each line has a spatial resolution of 30 meters on the
98 ABoVE common grid with a width of 22 km based on the swath width of the AirMOSS instrument. The
99 flight lines as a whole cover many ecosystem types and provide north-south and east-west gradients in
00 ALT and soil moisture across the ABoVE domain ([https://daac.ornl.gov/cgi-
01 bin/dsviewer.pl?ds_id=1796](https://daac.ornl.gov/cgi-bin/dsviewer.pl?ds_id=1796)).

02 **Table 1** defines all variables in PDO data files. The first eight variables represent the four primary
03 outputs of the PDO algorithm and associated uncertainties. Subsidence and ALT represent one-time
04 measurements for the 2017 thaw season [Schaefer et al. 2021]. VWC, defined as the ratio of water
05 volume to total soil volume, represents soil moisture at maximum thaw for 2017. We assume a vertical
06 profile of VWC and estimate Sw0 and wtd, the parameters that define the exact shape of the assumed
07 profile. The product includes a Python script that will create a map of VWC averaged over any user
08 specified depth range. We included maps of VWC averaged over depth ranges of hand-held soil
09 moisture probes commonly used in ABoVE fieldwork.

10

11 **Table 1. Variables in the Permafrost Dynamics Observatory (PDO) data files**

Variable	Full Name	Units	Description
alt	Active Layer Thickness	M	Maximum thaw depth at the end of summer
sub	Subsidence	M	Surface subsidence from start of thaw after snow melt to maximum thaw depth in August or September
Sw0	Surface Saturation Fraction	m ³ /m ³	The ratio of water volume to pore space volume at the surface or zero meters depth
wtd	Water Table Depth	M	The depth from the surface to the level where the soil is 100% saturated

alt_unc	Uncertainty ALT	M	Uncertainty of estimated ALT
sub_unc	Uncertainty Subsidence	M	Uncertainty of estimated seasonal subsidence
Sw0_unc	Uncertainty Surface Saturation Fraction	m3/m3	Uncertainty of estimated surface water saturation fraction
wtd_unc	Uncertainty Water Table Depth	M	Uncertainty of estimated water table depth
mv_6cm	VWC from 0 to 6 cm	m3/m3	The ratio of water volume to soil volume averaged over zero to 6 cm depths
mv_12cm	VWC from 0 to 12 cm	m3/m3	The ratio of water volume to soil volume averaged over zero to 12 cm depths
mv_20cm	VWC from 0 to 20 cm	m3/m3	The ratio of water volume to soil volume averaged over zero to 20 cm depths
mv_alt	VWC from 0 to ALT	m3/m3	The ratio of water volume to soil volume averaged over the entire active layer, from zero to ALT

12 **5.2 Alaska Active Layer and Soil Moisture Properties from Airborne P-band SAR**

13 Chen et al. [2019b] synthesized the P-band polarimetric synthetic aperture radar (PolSAR) data
14 collected in August and October of 2014 and 2015 during the AirMOSS EV-S1 investigation with the
15 ABoVE P-band measurements collected in August and October of 2017 to estimate soil geophysical
16 properties over 12 study sites in Northern Alaska (see Figure S2). Soil properties reported include the
17 ALT, soil dielectric constant, soil moisture profile, surface roughness, and their respective uncertainty
18 estimates at 30-m spatial resolution (<https://doi.org/10.3334/ORNLDAAAC/1657>).

19
20 Most of the study sites are located within the continuous permafrost zone and where the aboveground
21 vegetation consisting mainly of dwarf shrub and tussock/sedge/moss tundra has a minimal impact on P-
22 band radar backscatter. These data were used as inputs to the L-band ReSALT data described in Section
23 5.1.

24 **5.3 In Situ Soil Moisture and Thaw Depth Measurements**

25 In situ measurements of soil moisture, thaw depth, and other quantities are essential to calibrate and
26 validate ABoVE SAR retrievals. The ABoVE team established a set of standardized measurement
27 protocols for field plots to ensure uniform data products and quality in measurements collected by
28 different groups across the ABoVE domain and across multiple years. Numerous teams collected in situ
29 data during the initial 2017 Airborne Campaign, with more targeted field acquisitions conducted in
30 2018 and 2019 [Bourgeau-Chavez et al. 2019a,b, 2021; Bakian-Dogaheh et al. 2020; Loboda et al.
31 2021].

32
33 Bourgeau-Chavez and coworkers [2019a,b; 2021] collected soil moisture at 6, 12, 20, and 50 cm depths,
34 ALT, soil profiles and biophysical measurements of aboveground canopy and ground layers in the
35 Greater Slave Lake Region (C4). These data provide vegetation community characteristics and

36 biophysical data collected in 2018 from areas that were burned by wildfire in 2014 and 2015, and from
37 nine unburned validation sites. Vegetation data include vegetation inventories, ground cover, regrowth,
38 tree diameter and height, and woody seedling/sprouting data at burned sites, and similar vegetation
39 community characterization at unburned validation sites. Additional measurements included soil
40 moisture, collected for validation of the UAVSAR airborne collection, and depth to frozen ground at the
41 nine unburned sites. This 2018 fieldwork completes four years of field sampling at the wildfire areas.
42

43 Bakian-Dogaheh et al. [2020] measurements included soil dielectric properties, temperature, and
44 moisture profiles, active layer thickness (ALT), and measurements of soil organic matter, bulk density,
45 porosity, texture, and coarse root biomass from the surface to permafrost table in soil pits at selected
46 sites along the Dalton Highway in Northern Alaska (A1). Their investigation sites included Franklin
47 Bluffs, Sagwon, Happy Valley, Ice Cut, and Innavait Creek. Measurements collected at Franklin Bluffs
48 were concurrent with an August 2018 ABoVE L-band flight.
49 (<https://doi.org/10.3334/ORNLDAAAC/1759>).
50

51 Loboda et al [2022] collected field measurements from unburned sites and single and repeated burns in
52 the Noatak River valley and the Seward Peninsula regions of the Alaska tundra in July-August in the
53 years 2016-2018. The data include ocular assessment of vegetation cover, soil moisture at 6 and 12 cm,
54 soil temperature at 10 cm, organic soil thickness, thaw depth, and weather measurements.
55 (<https://doi.org/10.3334/ORNLDAAAC/1919>)
56

57 The strong partnership between the ABoVE and NGEE-Arctic projects also resulted in coordinated
58 same-day acquisition of airborne L- and P-band SAR data with in situ soil moisture and thaw depth
59 measurements over the NGEE-Arctic study site at Barrow (Utqiagvik), AK and the Seward Peninsula
60 watersheds near Teller, AK, Council, AK, and Kougarak, AK [Wilson et al. 2018]. These data provide
61 critical calibration for the ABoVE SAR retrievals under continuous (Utqiagvik) and discontinuous
62 (Seward Peninsula) permafrost conditions. Version 2 (V2) of the in situ soil moisture and thaw depth
63 measurements covering years 2017-2019 was released in November 2020.
64 (<https://doi.org/10.5440/1423892>)

65 **6 Synergy with Other Airborne Sensors**

66 Miller et al. [2019] described the overall ABoVE Airborne Campaign design strategy and anticipated
67 airborne sensor synergies. Here, we highlight three SARs and a LiDAR – AirSWOT (NASA), F-SAR
68 (DLR), LS-ASAR (ISRO) and LVIS (NASA) – whose acquisitions in the ABoVE domain were
69 specifically designed to complement and leverage the ABoVE L- and/or P-band SAR acquisitions.
70 Many other airborne sensor synergies are being exploited by the ABoVE science team and are reported
71 separately.

72 6.1 AirSWOT

73 NASA's AirSWOT airborne instrument suite has been developed to support the Surface Water and
74 Ocean Topography (SWOT) mission. The heart of AirSWOT is the Ka-band SWOT Phenomenology
75 Airborne Radar (KaSPAR). KaSPAR collects two swaths of across-track interferometry data: one swath
76 from nadir to 1 km and a second swath that extends from 1 km to 5 km off-nadir. AirSWOT flight lines
77 for ABoVE were designed to center the AirSWOT swath on the center of the P-band swath for
78 maximum overlap. KaSPAR is complemented by a high-resolution color-infrared (CIR) Digital Camera
79 System [Kyzivat et al. 2019a,b] and a Precision Inertial Measurement Unit (IMU) for accurate attitude
80 and positioning information. In 2015 AirSWOT made pre-ABoVE deployments to the Tanana River
81 Valley [Altenau et al. 2017] and the Yukon Flats [Pitcher et al. 2019a, 2019b] in Region A3.

82

83 In 2017, AirSWOT deployed to acquire
84 early season (May-June) and late season
85 (August) WSEs across the ABoVE
86 domain. **Figure 7** shows the concentration
87 of AirSWOT lines in wetlands complexes
88 in the boreal forest, across the Canadian
89 Shield, along the Mackenzie River Valley,
90 and into the Arctic tundra. All of these
91 regions contain overlapping L- and P-
92 band acquisitions. Of special interest are
93 the lines in the Peace-Athabasca Delta
94 (36000: PADeE and 18035: PADeW)
95 and the Yukon Flats (21508: YFlatW,
96 21609: YflatE, and 04707: FtYuko) and
97 Trail Valley Creek, NT (01703: TukHwy)
98 where extensive on-water measurements
99 were made [Pitcher et al. 2020]. Future
00 joint analyses of the Ka- and L-band data
01 will highlight the advances possible in
02 pan-Arctic hydrology from the upcoming NISAR and SWOT missions.

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Figure 7. AirSWOT flight lines acquired during the 2017 ABoVE airborne campaign sampled wetlands ranging from the Arctic Ocean coast to the southern boreal forest. AirSWOT's Ka-band acquisitions were designed to overlap with the L- and P-band SAR near-field acquisitions (See Fig. 1). © Google Earth

13
14
15

16 6.2 F-SAR

17 The German Space Agency (DLR) developed
18 the F-SAR instrument as an advanced
19 airborne SAR testbed for technology and
20 remote sensing applications [Reigber et al.
21 2013]. F-SAR operates fully polarimetric at
22 X-, C-, S-, L- and P-bands and features
23 single-pass polarimetric interferometric SAR
24 (PolInSAR) capabilities in X- and S-bands
25 [Reigber et al. 2013]. The radar covers an
26 off-nadir angle range of 25 to 60 degrees and
27 provides sub-meter scale spatial resolution
28 from flight altitudes up to 6000 mAGL.
29

30 During August 2018 and April 2019, F-SAR
31 was deployed to northern Canada as part of
32 DLR's permafrost airborne SAR experiment
33 (PerMASAR). It was configured in X-, C-, S-
34 and L-band mode and flew onboard a Dornier
35 Do 228-212 research aircraft. Measurements
36 were acquired from ~4500 mAGL.

37 Coordinated tomoSAR transects were flown
38 over the BERMS site in the southern boreal
39 forest on 18 August (UAVSAR) and 23
40 August (F-SAR). Preliminary results [Hensley
41 et al. 2020] are summarized in Section 6. F-
42 SAR also acquired data over the Scotty Creek
43 watershed, flux towers, and AOIs (Figure 8),
44 the Smith Creek flux tower (Wrigley, NT),
45 Baker Lake, Havipak Creek, Trail Valley Creek,
46 and Herschel Island, providing extensive opportunities to cross-
47 compare F-SAR and the ABoVE SAR
48 acquisitions.
49

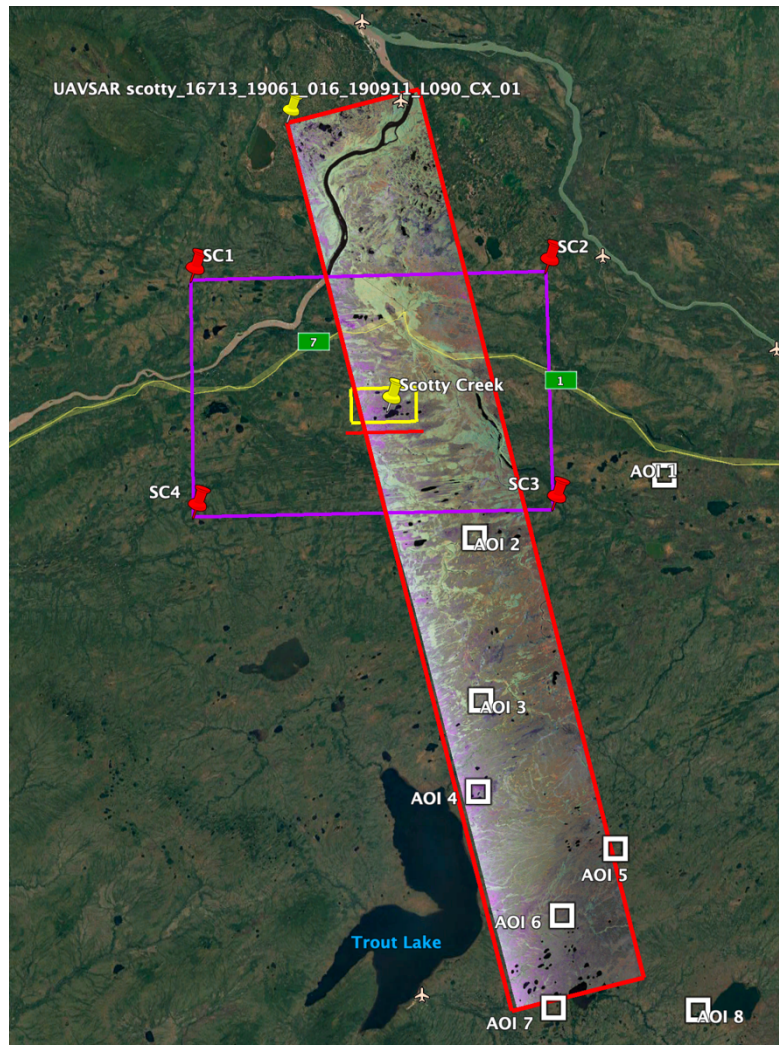


Figure 8. Overlap of the F-SAR acquisition at Scotty Creek, NT (yellow box) and the ABoVE L-band SAR line 16713 (red box & polarized SAR false color). The ABoVE line also captures Scotty Creek AOIs 2 – 6 (see Region C3 details, Sec. 5.3). The SAR data will complement and benefit from the extensive ground-based data acquired in this area [Quinton et al. 2019]. © Google Earth

50 6.3 LS-ASAR (ISRO)

51 The Indian Space Research Organisation (ISRO)
52 and NASA are jointly developing the NASA-
53 ISRO Synthetic Aperture Radar (NISAR), which
54 will map Earth's surface in L-band and S-band
55 every 12 days [Rosen et al. 2017]. As a precursor
56 to the NISAR mission, ISRO has developed a L-
57 and S-Band-Airborne SAR (LS-ASAR) to
58 prepare the community to maximize the scientific
59 and societal benefits of NISAR data [Ramanujam
60 et al. 2016; 2019; Mehra et al. 2019]. LS-ASAR
61 operates in Dual, Quad, and Hybrid Polarization
62 modes in both L- and S-bands. It covers
63 incidence angles from 24°-77° with swaths
64 ranging from 5.5 km to 15 km.

65
66 In December 2019 LS-ASAR flew a series of
67 Arctic sea ice sorties from Fairbanks, AK. During
68 this deployment, LS-ASAR also acquired data
69 over a number of the ABoVE flight lines in
70 Regions A1 (North Slope) and A3 (Eastern
71 Interior) as well as over a number of glacier sites
72 in the Alaska Range. The acquisitions are
73 available via the NASA & ISRO ASAR
74 Campaign page ([https://uavsar.jpl.nasa.gov/cgi-
75 bin/deployment.pl?id=L20191101](https://uavsar.jpl.nasa.gov/cgi-bin/deployment.pl?id=L20191101)) and are summarized in Figure 9. These data provide snow-on
76 coverage that was a known deficiency of previous ABoVE airborne campaigns. Additionally, the LS-
77 ASAR data extend coverage of these regions to S-band.

79 6.4 LVIS

80 The Land, Vegetation, and Ice Sensor (LVIS) is an airborne, full waveform scanning laser altimeter
81 which produces topographic maps with decimeter accuracy as well as vegetation vertical height and
82 structure measurements [Blair et al. 1999a,b]. Flight lines for LVIS (~1.4 km swath) were slaved to the
83 centerline of the P-band swath during ABoVE, except where deviations were required to capture critical
84 ground sites. LVIS-C (classic configuration) was deployed in 2017 aboard a B-200 and achieved limited
85 coverage (Figure 10, left panel). During 2019, the new LVIS Facility instrument (LVIS-F) as well as
86 LVIS-C were deployed on the NASA Gulfstream-V and achieved coverage of all legacy SAR lines
87 (Figure 10, right panel).

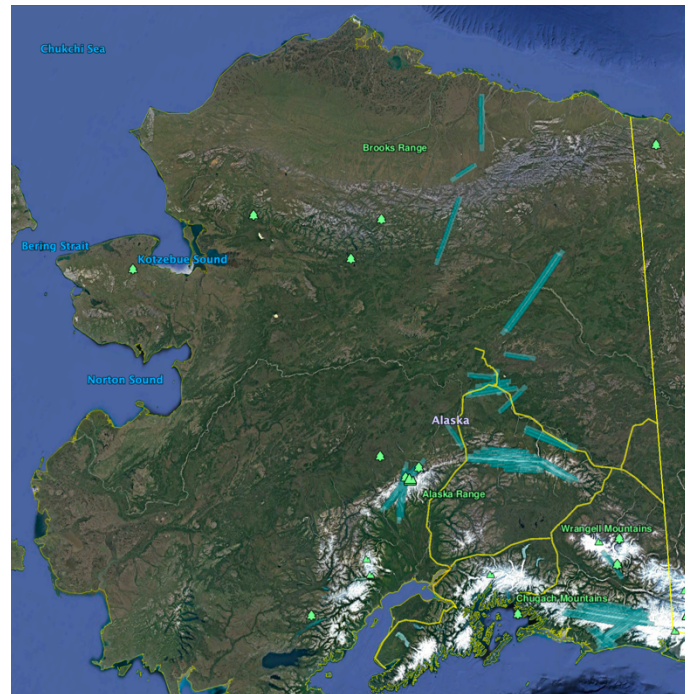


Figure 9. L- and S-band SAR lines acquired over Alaska during the December 2019 ASAR Campaign. ASAR flight lines on the North Slope, in the Yukon Flats, and in the Western Interior exactly overlap ABoVE flight lines. These early winter acquisitions provide a preliminary look at cold season SAR data that will be explored in greater detail in the planned ABoVE-SnowEx campaign. © Google Earth

89 LVIS' unique capability for measuring the sub-meter topography beneath boreal forest canopies
90 complemented the tomoSAR acquisitions over Delta Junction, AK and the BERMS site in northern
91 Saskatchewan [Hensley et al. 2020; Section 6]. LVIS altimetry will also prove valuable in analyses of
92 such variables as permafrost degradation, active layer thickness, and water surface elevation; however,
93 LVIS' significantly narrower swath limits the spatial extent over which these analyses may be
94 performed.

95

96 In June-July 2017, the NASA LVIS Facility was deployed to sites in northern Canada and Alaska as
97 part of NASA's Arctic-Boreal Vulnerability Experiment (ABoVE) 2017 airborne campaign. During the
98 4-week deployment of LVIS-F, a total of 15 flights were flown over diverse science targets based out of
99 multiple airports in Canada and Alaska. Data are available in both Level1B and Level2 formats (**Table**
00 **2**). The Level1b data files contain the geolocated laser waveform data for each laser footprint. The
01 Level2 data files contain canopy top and ground elevations and relative heights derived from the
02 Level1b data. ABoVE LVIS L1B Geolocated Return Energy Waveforms, Version 1 [Blair and Hofton,
03 2018a] and L2 Geolocated Surface Elevation Product, Version 1 [Blair and Hofton, 2018b] may be
04 obtained from the National Snow and Ice Data Center via <https://doi.org/10.5067/UMRAWS57QAFU>
05 and <https://doi.org/10.5067/IA5WAX7K3YGY>, respectively.

06

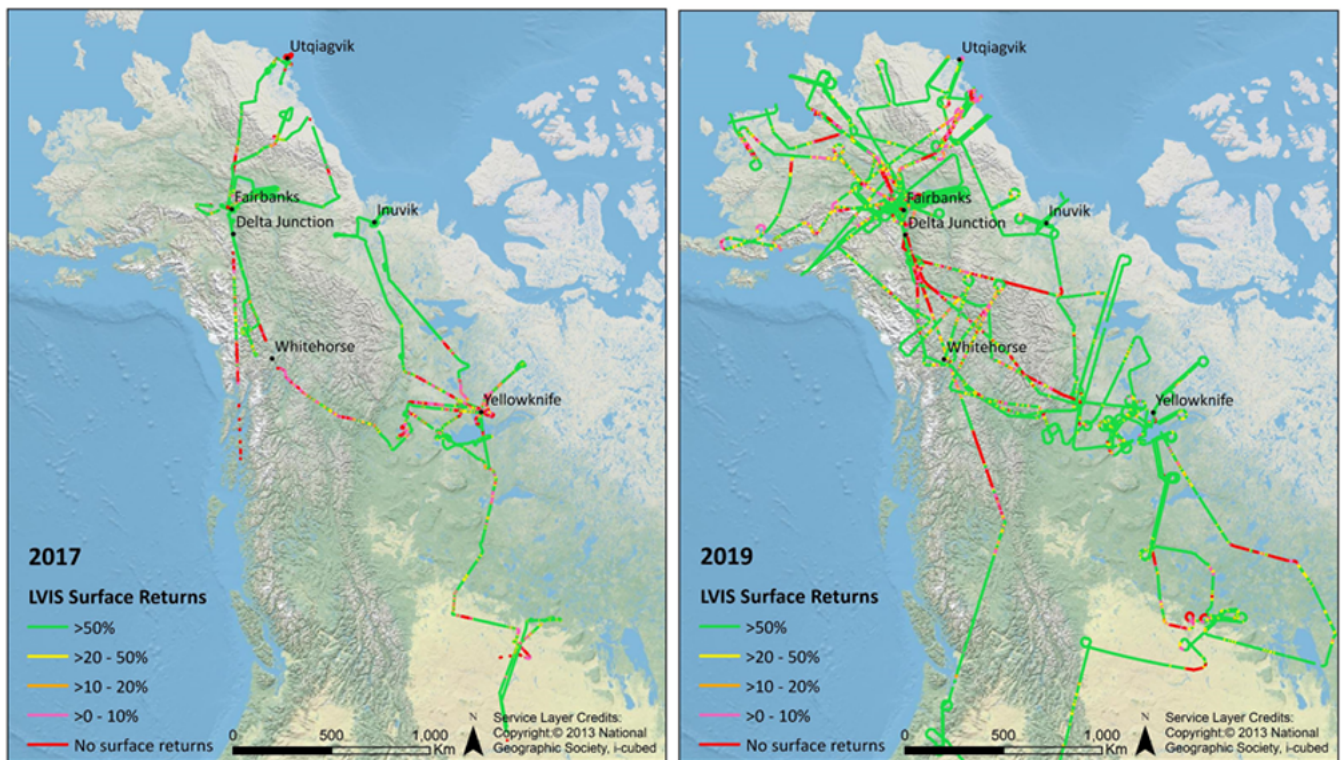


Figure 10. Flight lines for the LVIS 2017 flights (left) and 2019 flights (right) were designed to overlap with the near-filed portions of the L-band and P-band SAR swaths to maximize opportunities for synergistic science. Aircraft and weather limited coverage during the 2017 campaign, but complete coverage of the SAR flight lines was achieved in 2019. These data will enable studies of SAR/LiDAR fusion over the Arctic-boreal regions as a precursor to NISAR/IceSat-2 studies. © National Geographic Society

12

13 **Table 2. 2017 LVIS Data Products**

LVIS Data Products	Format
Flight Trajectories	KMZ
Camera Trajectories	KMZ
LVIS L1A Camera Imagery	JPG*
LVIS L1B Geolocated Waveforms	HDF, LDS 2.0.2
LVIS L2 Elevation and Height Products	ASCII TXT, LDS 2.0.2a

14

15 In July-August 2019, the NASA LVIS Facility and LVIS Classic were deployed to sites in northern
 16 Canada and Alaska as part of NASA's ABoVE 2019 airborne campaign. The increased range and
 17 endurance of the Gulfstream-V platform enabled extensive sampling, including: all L-band SAR lines,
 18 multiple IceSAT-2 underflights, and numerous ABoVE field sites. The available data products are given
 19 in **Table 3**.

20

21 **Table 3. 2019 LVIS Data Products**

LVIS Data Products	Format
Flight Trajectories	KMZ
Coverage Maps	KMZ
LVIS Classic L1B Geolocated Waveforms	HDF, LDS 2.0.3
LVIS Classic L2 Elevation and Height Products	ASCII TXT, LDS 2.0.3
LVIS Facility L1B Geolocated Waveforms	HDF, LDS 2.0.3
LVIS Facility L2 Elevation and Height Products	ASCII TXT, LDS 2.0.3

22 **6.5 G-LiHT**

23 Zhao et al. [2022] used ABoVE airborne L- and P-band SAR to map boreal forest species and canopy
 24 height in the Tanana Valley State Forest (TVSF) near Delta Junction, AK. They employed machine
 25 learning (random forests) to train separate regression models for canopy height mapping and a
 26 classification model for forest species mapping. Data derived from NASA's Goddard LiDAR,
 27 Hyperspectral, and Thermal Imager (G-LiHT) system [Cook et al. 2013] were treated as ground truth
 28 for the canopy height model (CHM). Forest species prediction were referenced against (TVSF) Timber
 29 Inventory and Forest Inventory and Analysis (FIA) data. The experimental results show the proposed
 30 method yields a root-mean-square error of 1.90 m for forest height estimation and overall accuracy of
 31 79.5% for forest species classification. A significant finding was that PolSAR decomposition
 32 parameters, such as volume scattering and entropy, strongly influenced the canopy height estimates.
 33 Interestingly, topography played a crucial role in the species classification.

34

35

36

37 7 Data Availability

38

39 Links to the ABoVE L- and P-band SAR products, supporting data, derived products, and ancillary
40 measurements are provided in the Appendix. Formal citations to all DOIs are provided in the
41 References. The L- and P-band SAR data may be found at the JPL UAVSAR data portal,
42 <https://uavsar.jpl.nasa.gov/cgi-bin/data.pl>. L-band data may also be accessed via the UAVSAR portal at
43 the Alaska SAR Facility (ASF) DAAC (<https://asf.alaska.edu/data-sets/sar-data-sets/uavsar/>) while
44 AirMOSS P-band data may be accessed via the ORNL DAAC, https://daac.ornl.gov/get_data/#projects,
45 select “AirMOSS”.

46

47 Miller et al. [2023; <https://doi.org/10.3334/ORNLDAAC/2150>] provides a detailed description of all 80
48 SAR flight lines and how each fits into the ABoVE experimental design. Extensive maps, tables, and
49 hyperlinks give direct access to every flight plan as well as individual flight lines. It is a guide to enable
50 interested readers to fully explore the ABoVE L- and P-band SAR data.

51

52

53 8 Summary

54 The ABoVE project conducted airborne L-band PolInSAR surveys in 2017, 2018, 2019 and 2022 across
55 Alaska and northwestern Canada. These were complemented by a P-band PolInSAR survey in 2017
56 along the same transects. This time series provides a powerful data set with which to evaluate the state
57 of permafrost, active layer thickness, soil moisture, boreal forest structure, above ground biomass, and
58 water surface elevation. Additional studies leverage the PolInSAR data to address fire disturbance and
59 recovery, thermokarst feature development, and retrogressive permafrost thaw megaslumps. Many of
60 these analyses are in progress and will be published separately.

61

62 Miller et al. [2023] provides extensive, fully hyperlinked notes on the airborne SAR data. Researchers
63 may discover these data via daily sorties and/or individual flight lines. Alternatively, they may be
64 explored via the interactive map at the JPL UAVSAR data portal, ([https://uavsar.jpl.nasa.gov/cgi-
65 bin/data.pl](https://uavsar.jpl.nasa.gov/cgi-bin/data.pl)) which provides links to the individual flight line data, maps, and related flight plans that
66 acquired data over one or more of the individual flight lines. We have also identified the ground-based
67 anchor points for each flight line to facilitate comparisons with those data. Calibration and validation
68 data sets as well as many derived products produced by the ABoVE Science Team may be found at the
69 Arctic Boreal Vulnerability Experiment (ABoVE) landing page at the ORNL DAAC
70 (https://daac.ornl.gov/cgi-bin/dataset_lister.pl?p=34).

71

72 The example studies (Sections 4 and 5) and multi-instrument synergies (Section 6) described here are
73 only a small portion of the studies currently being undertaken by the ABoVE Science Team and the
74 SAR Working Group. We anticipate many new and innovative uses of the L-band and P-band SAR data
75 as the ABoVE team expands its range of synthesis activities in Phase 3.

76

77 The ABoVE L-band SAR flights planned for 2020 and 2021 were postponed due to the global COVID-
78 19 pandemic and safety considerations; however, flights were resumed in 2022 and we anticipate at
79 least one more thaw season campaign in 2024. Finally, we note that the data and analyses discussed
80 here set the stage for the upcoming NISAR mission (expected launch in 2023). NISAR will deliver
81 global L- and S-band imagery with a 12-day revisit. Its emphasis on snow- and ice-covered surfaces has
82 obvious applications in the ABoVE domain, and its global coverage will allow researchers to test the
83 methods developed for the ABoVE domain across the pan-Arctic. Beyond NISAR, NASA is studying
84 architectures for the Surface Deformation and Change (SDC) Earth System Observatory Mission and
85 ESA are developing the Rose-L Copernicus expansion mission. SDC and Rose-L are also envisioned as
86 an L-band sensors. NISAR, SDC, and Rose-L will all benefit from the ABoVE SAR studies,
87 algorithmic advances, and lessons learned.

88 **9 Supplemental Information**

89 The Supplemental Information (SI) contains detailed descriptions of all L-band flight lines plus tables
90 with hyperlinks to all L-band lines and sorties. Additionally, the SI includes tables and links to the P-
91 band flight lines acquired during the ABoVE campaigns and to all Legacy L-band and P-band flight
92 lines. The SI is identical to the file <Summary ABoVE L- & P-Band SAR Surveys - hyperlinked.pdf>
93 that may be found in the /data folder of the uncompressed data download from Miller et al. [2023;
94 <https://doi.org/10.3334/ORNLDAAC/2150>].
95

96 **10 Author Contributions**

97 SJG, CEM, PCG and the ABoVE Science Definition Team developed the preliminary ABoVE
98 Implementation Plan; this was updated by SJG, CEM, PCG and the ABoVE Science Team. CEM, PCG,
99 and EH developed the initial flight lines based on the ABoVE Implementation Plan and consultations
00 with the ABoVE Science and partners. NSP translated the notional flight lines into the UAVSAR
01 planning system. NSP, YL, MM, PCG, , ELH, and CEM served as Scientist on Board during the data
02 acquisition flights. YL, SH, BC, NP and the JPL Sub-orbital Radar Science and Engineering Team
03 (334F) processed the L- and P-band SAR data. CW, KS, MT, JB, LBC, RHC, MM, SW, MM, RJM,
04 TL, LJ, PS AT, and RD collected cal/val field data. CW and SW coordinated ground cal/val data
05 acquisitions at the NGEE-Arctic sites. SH, PS, NSP performed the tomoSAR analyses and SH
06 contributed the text and images for Section 6. KS, MM and the ABoVE SAR Working Group calibrated
07 the L-band SAR data. MM, AT, RHC processed the P-band data. NSP and YL coordinated the LS-
08 ASAR flights and the joint BERMS area tomoSAR flights with F-SAR. DS coordinated all data product
09 submissions to the ORNL DAAC. CEM wrote the initial manuscript and all co-authors contributed to
10 the final version.

11 Competing Interests

12 The authors declare that they have no conflict of interest

13 12 Acknowledgments

14 The L- and P-band SAR data acquisitions would not have been possible without the indefatigable
15 support of our NASA pilots and flight crews. We thank John McGrath and the NASA AFRC C-20
16 (N30502) team as well as Derek Rutovic and the NASA JSC G-III (N995NA) team. We also thank the
17 instrument scientists, operators, and data processing team from the JPL Suborbital Radar Science and
18 Engineering Team (334F) who were essential to the successful execution of these experiments and rapid
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