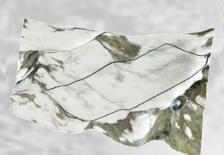
Multistatic Radar Workshop 2025







SnowCAT: an Innovative MIMO SAR Mission for Snow Characterization by SAR Tomography

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Motivation

- Seasonal snow cover affects the global climate system and freshwater availability to billions of people
 - recognized as an Essential Climate Variable by WMO
 - included as one of ESA's Living Planet Challenges
- Accurate assessment of Snow Depth (SD) and Snow Water Equivalent (SWE) is still challenging at operational level, especially in mountain regions with complex topography
 - ☐ high-res optical stereo imagery
 - ➤ SD retrieval by DEM differencing, accuracy about 50/70 cm over few meters, clouds hinder systematic coverage
 - ☐ altimeters are limited by spatial sampling (LiDAR) and coarse resolution (RADAR)
 - ☐ SWE retrieval from SAR radiometry & polarimetry assumes specific snowpack models
 - > accuracy about few cm/1 m in controlled conditions, problematic estimation in heterogeneous areas
 - □ DInSAR-based retrieval allows direct measurement of SWE variations across two dates (for dry snow conditions with highly transparent snowpack)
 - > accuracy reported to be up to few cm, strongly depends on local coherence and compensation of topographic and atmospheric delays
- There is currently no EO mission specifically dedicated to monitor snow mass in mountain regions

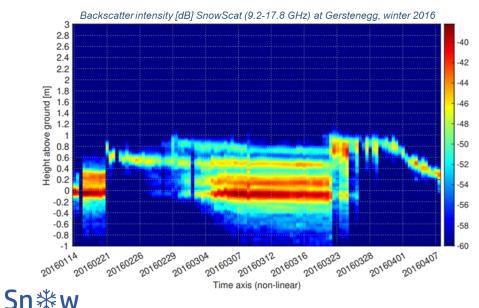


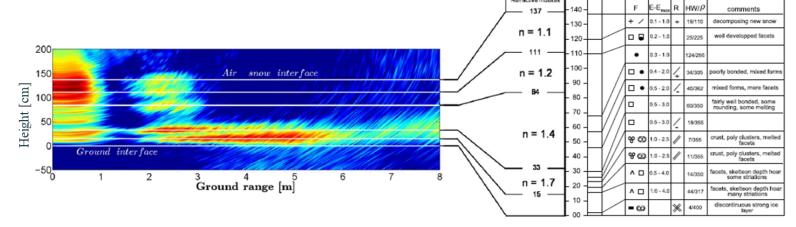
Retrieval of physical snow parameters using TomoSAR

TomoSAR acquires SAR data along *multiple trajectories* and provides a **3D representation** of RADAR backscatter. It is a consolidated technology for remote sensing of forested areas using P- and L-Band SAR data.

TomoSAR at X- and Ku-Band allows for direct measurement of physical parameters of the dry snow pack:

- Total Snow depth
- Refractive index
- SWE
- Internal layering





Data from AlpSAR 2013 (Rennes 1, ESA)

Rekioua et al., Comptes Rendus Physique, 2017

and Estimated

Demonstrated in various ground campaigns:

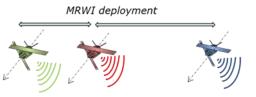
- ESA AlpSAR (2013)
- ESA SnowLab & ESA SnowLab NG (2016-2020)
- NASA SnowEx (2018)
- Altay UAV-SAR Ku-Band experiment (2023)

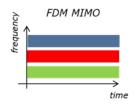
SnowCAT Mission concept

- TomoSAR formation of 3 small satellites, all Tx/Rx (MIMO)
 - reduced costs wrt classic concepts
 - mitigation of temporal decorrelation & atmosphere
- X-Band Radar: large bandwidth (ITU), good compromise between penetration/sensitivity to snow layers
- Dual-Pol
- Horizontal resolution of few meters
- Channel access via Frequency Division Multiplexing (FDM)
 Rainbow system
- Sub-meter vertical resolution by MRWI formation flying
- Incidence angle diversity

 - ➤ transparent snow (snow/terrain or snow/ice interface detectable only) → SWE retrieval by differencing apparent snow depths at two incidence angles

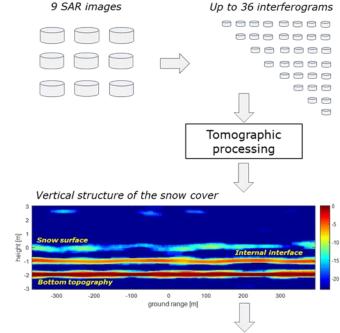








SnowCAT Processing concept





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MIMO Correlation TomoSAR

- Issue: design optimal formation from few real elements (i.e., maximize uniformly spaced virtual elements)
 - optimizing virtual monostatic formation equivalent to MIMO brings limited gain
- Alternative approach: correlation tomography





- tomographic imaging depends on the set of available baselines, i.e. differences in sensor positions
- optimization brings more gain!

N real MIMO satellites Up to
$$N_v = \frac{N(N+1)}{2}$$
 Up to $N_d = \frac{N_v(N_v-1)}{2}$ baselines

The number of available baselines <u>ideally</u> goes with up to the 4-th power of the number of physical satellites

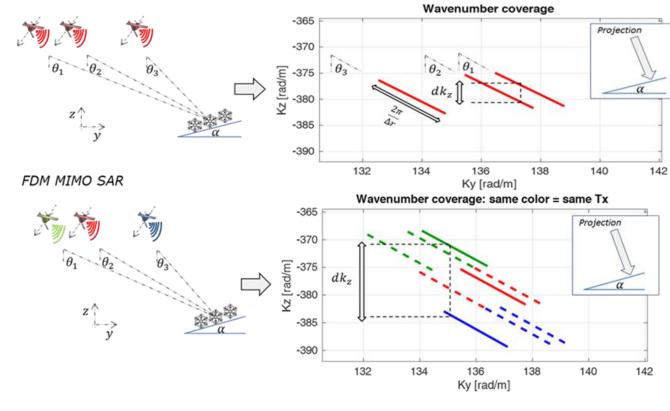
Spaceborne MIMO TomoSAR

Single-frequency monostatic SAR

Problem: MRVA assumes isotropic point scatterers, not accounting for spatial decorrelation as a function of baseline

> not directly applicable to the case of TomoSAR

Solution: design the formation to achieve sub-meter resolution while **accounting for decorrelation due to spatial wavenumber shift**



Using different frequency bands results in InSAR pairs with:

- large common bandwidth (high coherence)
- large InSAR wavenumber (fine vertical resolution)
- impossible to achieve with traditional architectures

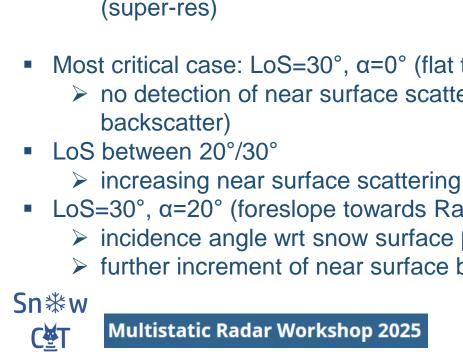
High number of FDM MIMO acquisitions gives also small InSAR wavenumbers:

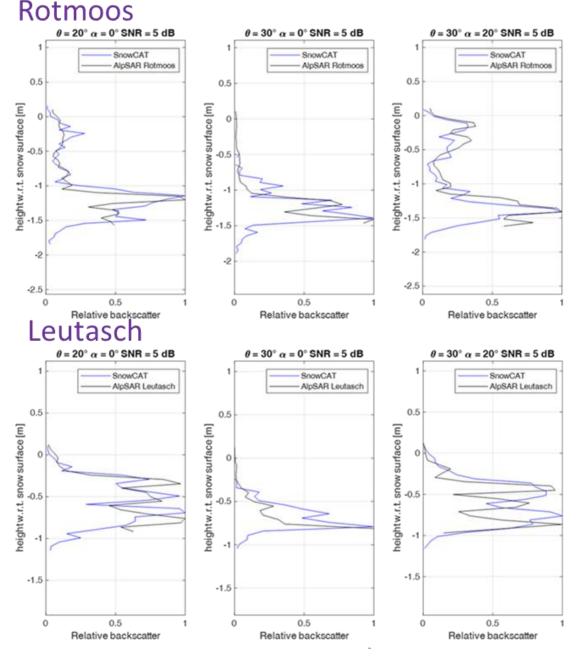
large height of ambiguity (TomoSAR sidelobe mitigation)



Expected TomoSAR imaging

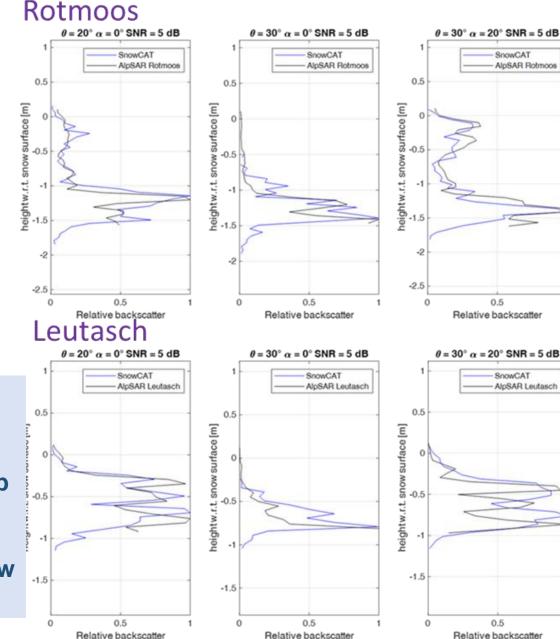
- Data generated with diffractive simulator:
 - AlpSAR TomoSAR giving scatterers distribution
 - Projection to SnowCAT with bistatic delays
 - 5 dB SNR
 - SnowCAT formation optimized for:
 - 1.2 m vertical resolution
 - 8.4 m HoA
 - 20° incidence angle on flat terrain
 - TomoSAR focusing through COMET approach (super-res)
- Most critical case: LoS=30°, α =0° (flat terrain)
 - no detection of near surface scattering (lower backscatter)
- LoS= 30° , $\alpha=20^{\circ}$ (foreslope towards Radar)
 - \triangleright incidence angle wrt snow surface β =10°
 - further increment of near surface backscatter.





Expected TomoSAR imaging

- Data generated with diffractive simulator:
 - AlpSAR TomoSAR giving scatterers distribution
 - Projection to SnowCAT with bistatic delays
 - 5 dB SNR
 - SnowCAT formation optimized for:
 - 1.2 m vertical resolution
 - 8.4 m HoA
 - 20° incidence angle on flat terrain
 - TomoSAR focusing through COMET approach (super-res)
- ➤ SnowCAT expected to meaningfully map dry snow vertical structure when LoS wrt snow surface is strictly less than 30°
- Ascending/descending passes & different LoS help in mountain areas (complex topography)
- ➤ Sea ice (no background terrain slope) provides a favourable target for SnowCAT, observed at shallow incidence angle (<30°)



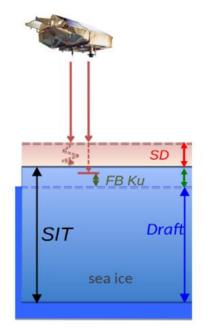


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Objectives

- Characterization of snow cover in mountain regions at fine spatial resolution:
 - > Snow Depth and Snow Water Equivalent (SWE) can be mapped in mountainous regions, and across hydrological basins.
 - > Quantify winter snow accumulation on glaciers and ice caps
 - > Identify snow pack conditions including areas with wet snow
- Evolution of snow stratigraphy by assimilation into land surface models:
 - Understand assimilation of high-resolution snow stratigraphy into weather prediction models
 - ➤ Assess the role of EO snow stratigraphy for avalanche forecasting, to simulate evolution of the soil temperature, assess thermal insulation, ecosystem evolution, predict crust thickness
- Better understanding of the influence of the snow cover on sea ice:
 - ➤ High-resolution (few meters) and precise freeboard height and snow depth determination on smooth and rough ice, further declined as:
 - > Separate rough and smooth ice by estimation of ice topography at fine resolution
 - > Characterization of snow structure by identification of scattering horizons, for example caused by layers of superimposed ice or a layer of snow ice







Mission scenarios & Radar payload

- SnowCAT is intended to provide systematic observations over Alpine and Arctic regions
 - total coverage about a million square kilometers
 - selected AoIs, based on availability of reference validation data & scientific relevance
 - optimization of retrieval performance wrt orbits and range of incidence angles
- The two phases will be characterized by different sets of InSAR baselines (shorter for land snow, larger for snow-covered sea ice)
- Target is to provide weekly or sub-weekly revisit & about 4/5 Stripmap images per orbit

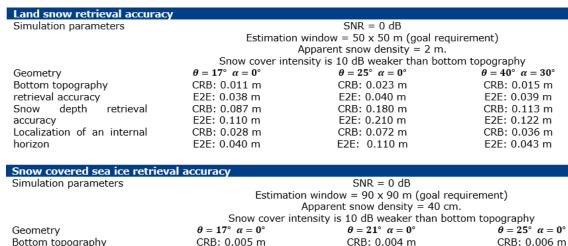
Parameter	Value
Altitude	500 km
Revisit	<1 week
Acquisition mode	Stripmap
Antenna	Offset reflector
Incidence angle	20°/40°
Ground swath	about 10 km
Frequency	X-band
BW	100 MHz
PRF	6KHz
NESZ	< -17 dB (LoS<25°) <-14 dB (LoS> 25°)
Orbital deviation	<100m





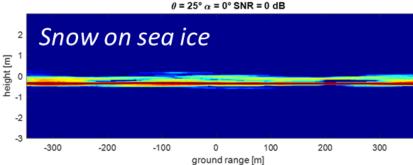
Example Tomo & L2 products

- Layered snow model & E2E simulation
- Layer identification through matching pursuit
- CRB accuracies are compared to L2 retrieval accuracy (standard deviation)
- L2 retrieval accuracy slightly higher than CRB (theoretically consistent)
- Values consistent with scientific targets for relevant applications

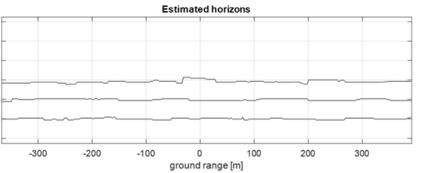


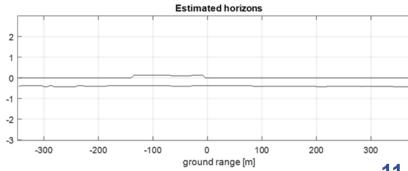
Bottom topography CRB: 0.005 m CRB: 0.004 m E2E: 0.009 m E2E: 0.013 m E2E: 0.019 m retrieval accuracy CRB: 0.042 m CRB: 0.030 m CRB: 0.049 m Snow depth retrieval accuracy F2F: 0.058 m E2E: 0.045 m F2F: 0.042 m

SnowCAT tomography from E2E simulations θ = 40° α = 30° SNR = 0 dB Land snow height [m] -300 -200 -100 0 100 200 300 ground range [m] Estimated horizons



SnowCAT tomography from E2E simulations







Example Tomo & L2 products

- Layered snow model & E2E simulation
- Layer identification through matching pursuit
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- L2 retrieval accuracy slightly higher than CRB (theoretically consistent)
- Values consistent with scientific targets for relevant applications

Land snow retrieval accuracy

Simulation parameters

SNR = 0 dB

Estimation window = $50 \times 50 \text{ m}$ (goal requirement) Apparent snow density = 2 m.

	Snow cover intensity is 10 dB weaker than bottom topography		
Geometry	$\theta = 17^{\circ} \ \alpha = 0^{\circ}$	$\theta = 25^{\circ} \ \alpha = 0^{\circ}$	$\theta = 40^{\circ} \ \alpha = 30^{\circ}$
Bottom topography	CRB: 0.011 m	CRB: 0.023 m	CRB: 0.015 m
retrieval accuracy	E2E: 0.038 m	E2E: 0.040 m	E2E: 0.039 m
Snow depth retrieval	CRB: 0.087 m	CRB: 0.180 m	CRB: 0.113 m
accuracy	E2E: 0.110 m	E2E: 0.210 m	E2E: 0.122 m
Localization of an internal	CRB: 0.028 m	CRB: 0.072 m	CRB: 0.036 m
horizon	E2E: 0.040 m	E2E: 0.110 m	E2E: 0.043 m

Snow covered sea ice retrieval accuracy

Simulation parameters

200

SNR = 0 dB

Estimation window = $90 \times 90 \text{ m}$ (goal requirement) Apparent snow density = 40 cm.

Snow cover intensity is 10 dB weaker than bottom topography

Geometry	$\theta = 17^{\circ} \ \alpha = 0^{\circ}$	$\theta = 21^{\circ} \ \alpha = 0^{\circ}$	$\theta = 25^{\circ} \ \alpha = 0^{\circ}$
Bottom topography	CRB: 0.005 m	CRB: 0.004 m	CRB: 0.006 m
retrieval accuracy	E2E: 0.009 m	E2E: 0.013 m	E2E: 0.019 m
Snow depth retrieval	CRB: 0.042 m	CRB: 0.030 m	CRB: 0.049 m
accuracy	E2E: 0.058 m	E2E: 0.045 m	E2E: 0.042 m

-300

Snow on sea ice

-200

-100

SnowCAT tomography from E2E simulations

θ = 40° α = 30° SNR = 0 dB

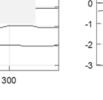
ground range [m]

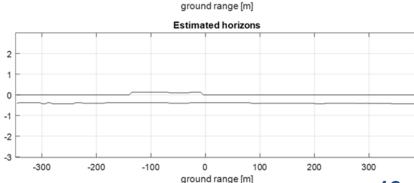
SnowCAT tomography from E2E simulations $\theta = 25^{\circ} \alpha = 0^{\circ} \text{SNR} = 0 \text{ dB}$

Dissemination strategy

- SnowCAT is intended to fit innovative scientific programs, possibly embracing Open Science paradigm (e.g., like Earth Explorer, Scout)
 - ➤ L1/L2/L3 products freely available to users
 - dedicated platform for processing/catalogue
 - specific attention to relevant user communities (e.g., ECMWF, Méteo France, NPI, AWI, Centre for Polar Observation and Modelling)

-300





100

200

300



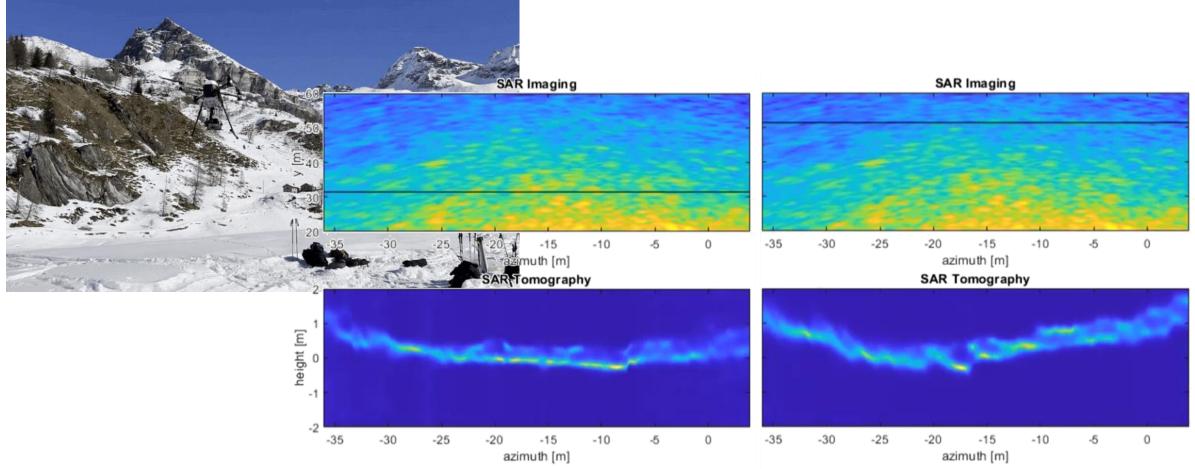
Possible synergies with upcoming EO missions

- Feasibility of InSAR with small satellites is recently demonstrated by Hongtu-1
- SnowCAT to be the first space mission for high resolution 3D snow imaging
- SnowCAT would open to synergistic use with other ESA/Copernicus programs
 - it complements with CRISTAL in solving interpretation problems in complex ice topography
 - retrieval of snow density by integrating the apparent snow depth observed by SnowCAT & CRISTAL
 - SnowCAT snow depth & SWE can serve as a reference for ROSE-L time series, sensitive to snow changing conditions, imperfect phase calibration and unwrapping
 - ROSE-L & SnowCAT can be jointly analysed to derive snow density at local scales
 - SnowCAT stratigraphy can enhance multifrequency capabilities of CIMR (& MetOp-SG) to capture snow structure and ultimately atmospheric information
 - SnowCAT information on snow layering provides insights into radar signal penetration into polar ice, thereby supporting the development of methods to correct *Harmony* penetration bias



A lot of work in progress

X-band snow TomoSAR at Cheneil (Aosta, 2100 m a.s.l.), 40 flights at about 50 m altitude



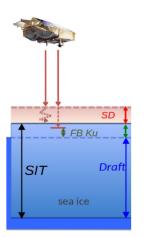


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Thanks for your attention!

Altimeters

- Nadir-looking geometry determines sensitivity to specular scattering
- Horizons are detected by ranging
- Horizontal resolution on the order of hundreds of meters across-track
- Coverage along transects



TomoSAR

- Side-looking geometry determines sensitivity to back-scattering
- o Horizons are detected by SAR tomography
- Horizontal resolution on final products the order of few meters in both directions
- Continuous coverage over the imaged swath



