



Rock glaciers identification and characterisation with remote sensing and machine learning

Benjamin Aubrey Robson
benjamin.robson@uib.no

with contributions from Daniel Thomas and Emma Hauglin

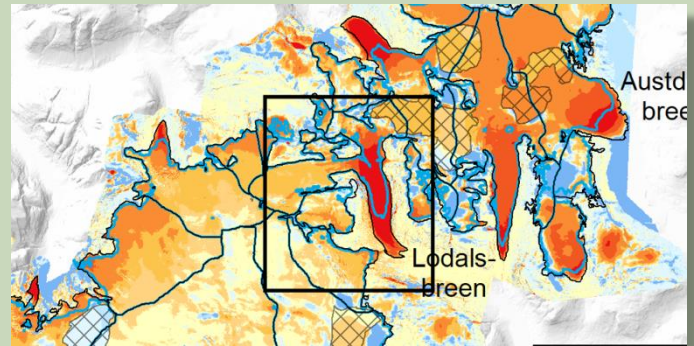
University of Bergen, Norway



Overview of this talk

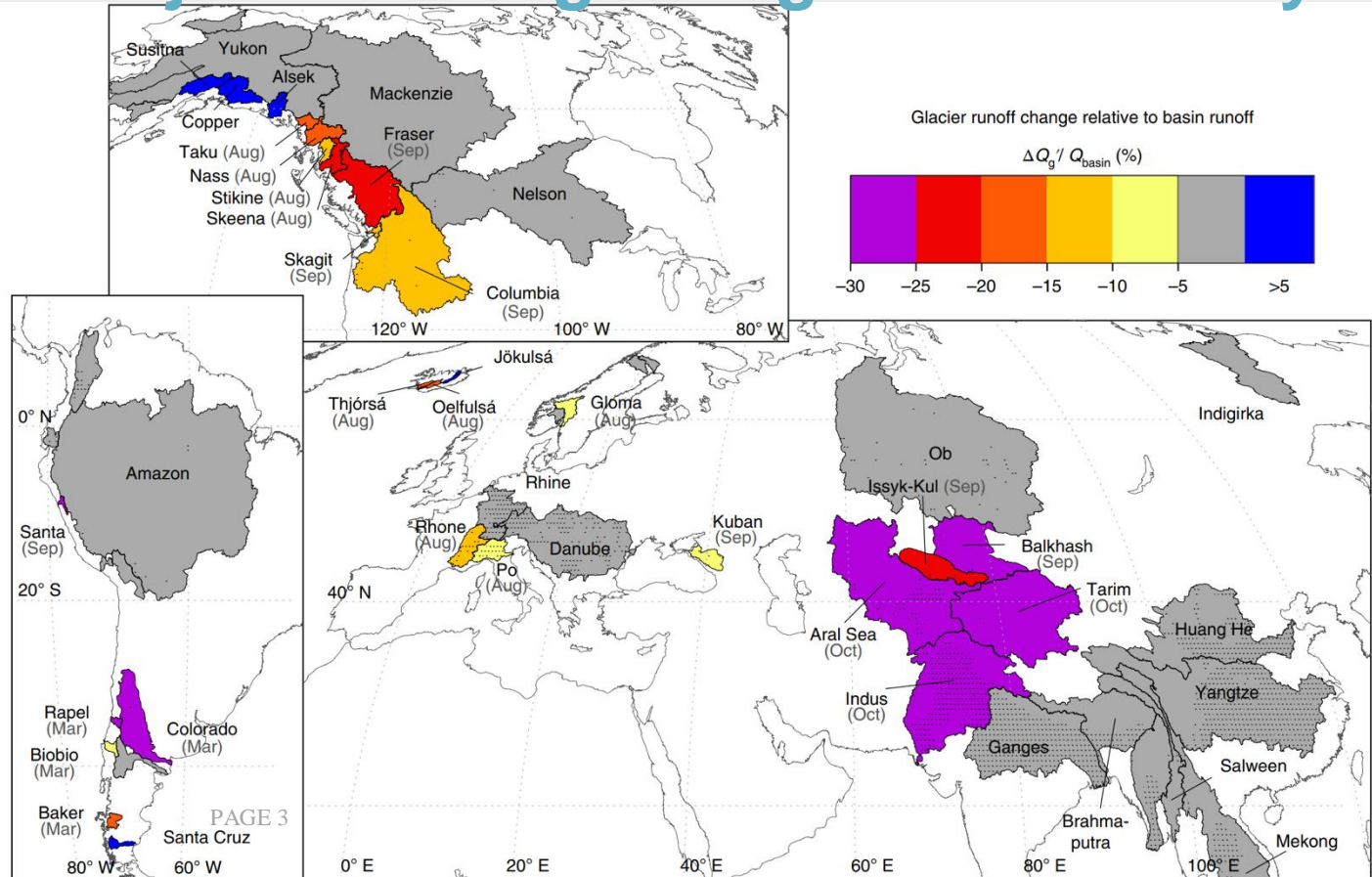


Machine learning based mapping of the cryosphere



Quantifying rock glacier changes using remote sensing

Projected changes in glacial runoff by 2100



PAGE 3

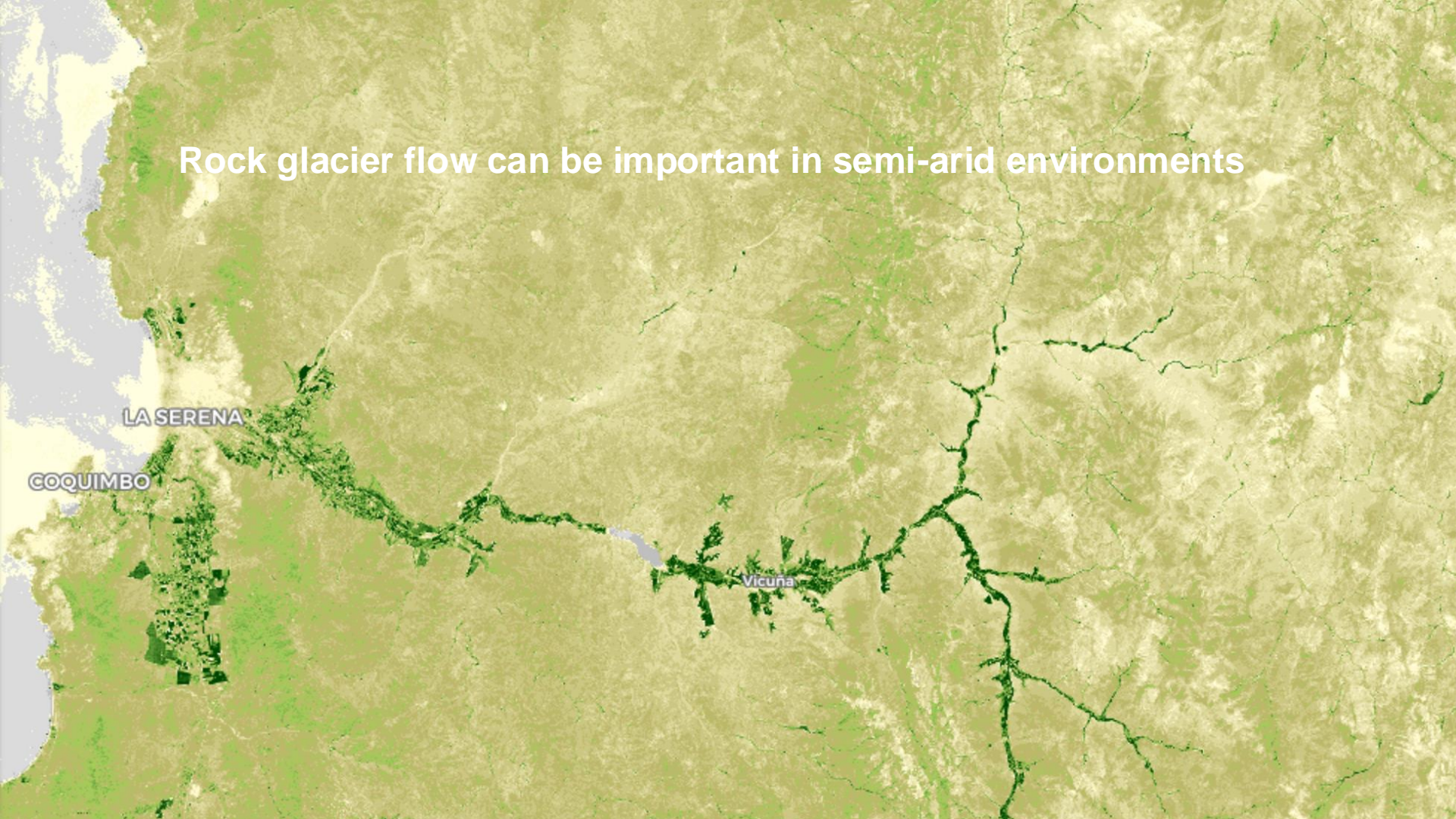


Rock glacier flow can be important in semi-arid environments

LA SERENA

COQUIMBO

Vicuna



An aerial photograph of a rugged mountain range. The terrain is characterized by numerous rock glaciers, which appear as large, fan-shaped, light-colored (tan and grey) deposits. These rock glaciers are interspersed with smaller, more traditional glaciers and patches of snow. The overall appearance is a complex, textured landscape of glacial and periglacial features.

Many of these areas contain more rock glaciers than glaciers

- >80% annual streamflow cryospherically derived
- Up to 13% annual streamflow from rock glaciers*
- Rock glacier inventories are often fragmented

Identifying rock glaciers

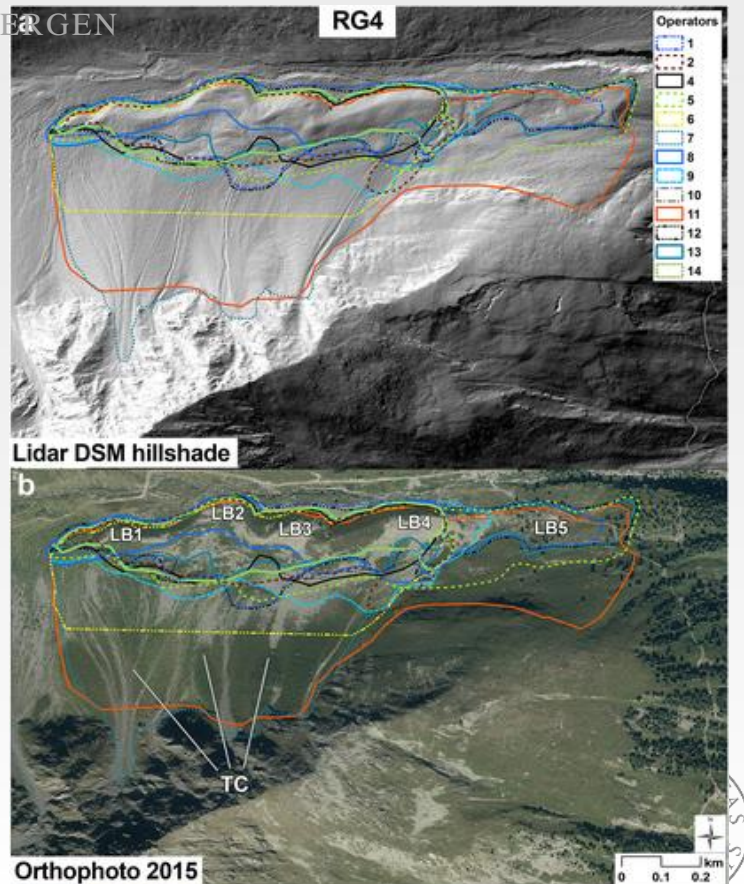
Ridges and furrows,
typical surface
morphology



Rock glacier mapping uncertainty

Large subjective biases

- Up to 70% variation in RG area
- Can machine learning be used to create a more objective inventory?

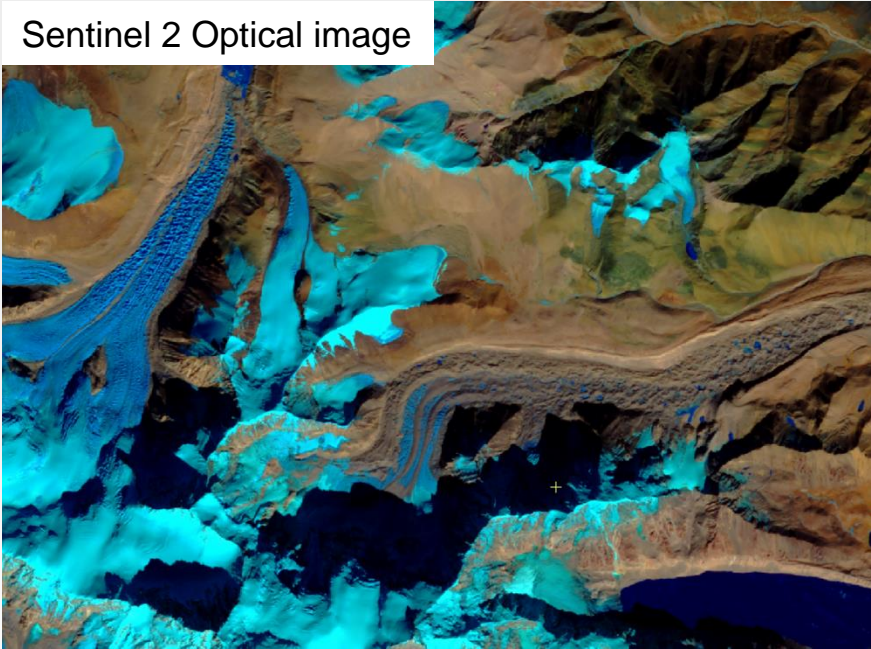


Brardinoni et al. 2019

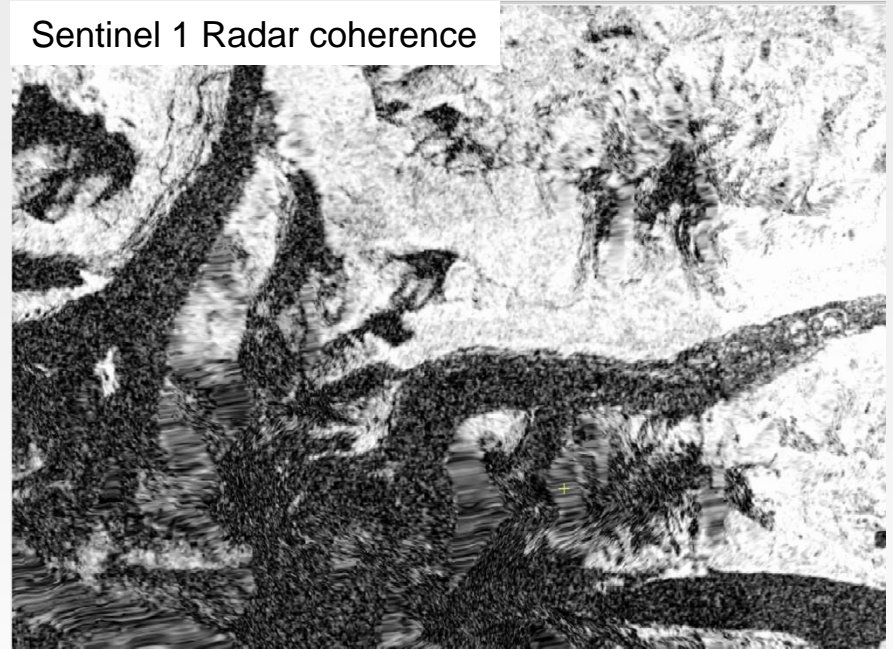
Are methods transferable from other landforms?



Sentinel 2 Optical image



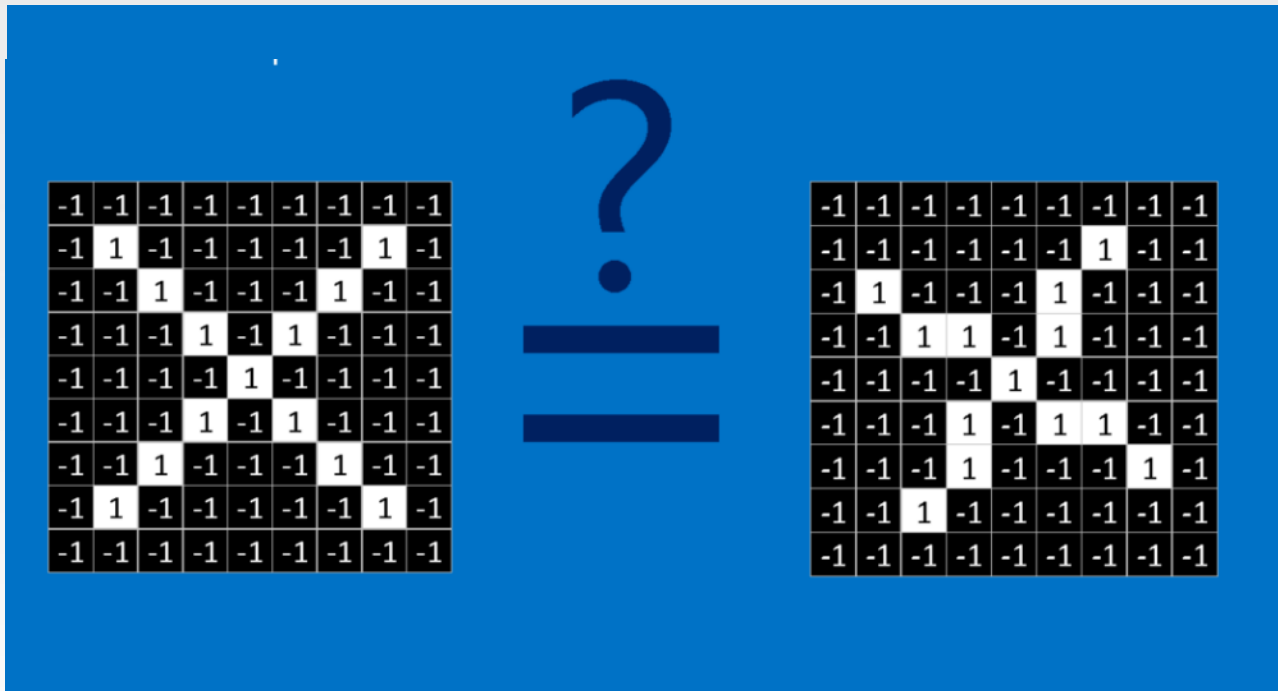
Sentinel 1 Radar coherence



Robson *et al.*, 2020



Deep-learning: recognising rock glaciers by their image texture



0% sure this is an X



100% sure this is an X

Could this be an X?

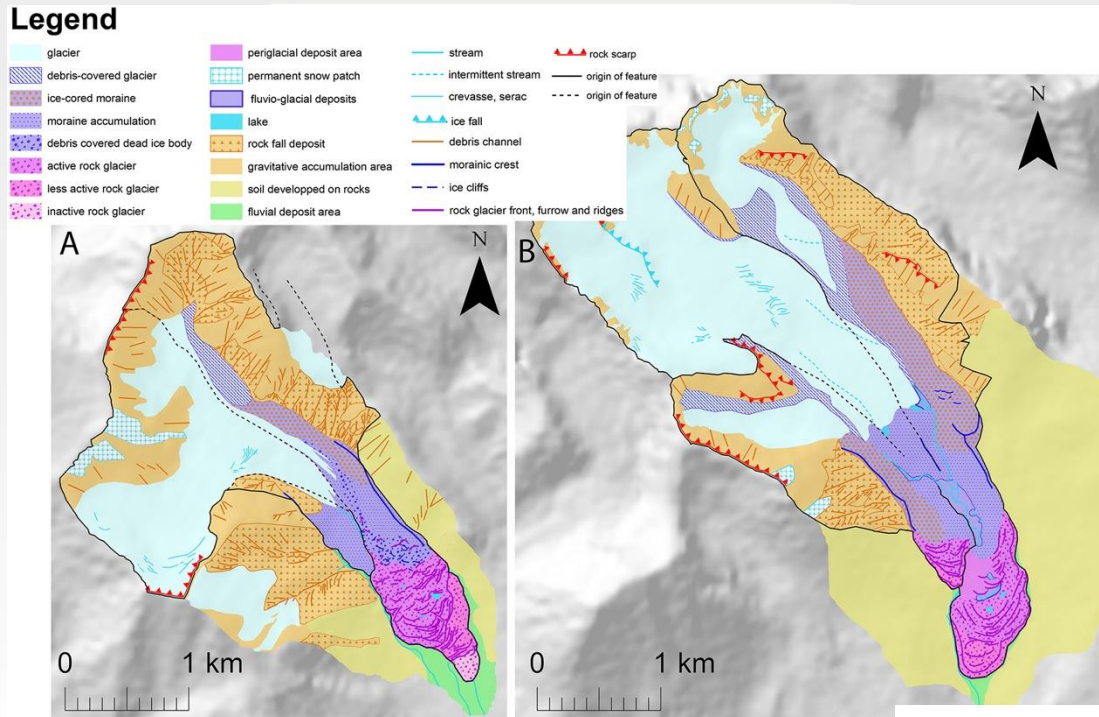
Deep-learning: recognising rock glaciers by their image texture



We can create sample patches and train a model to recognise these patterns

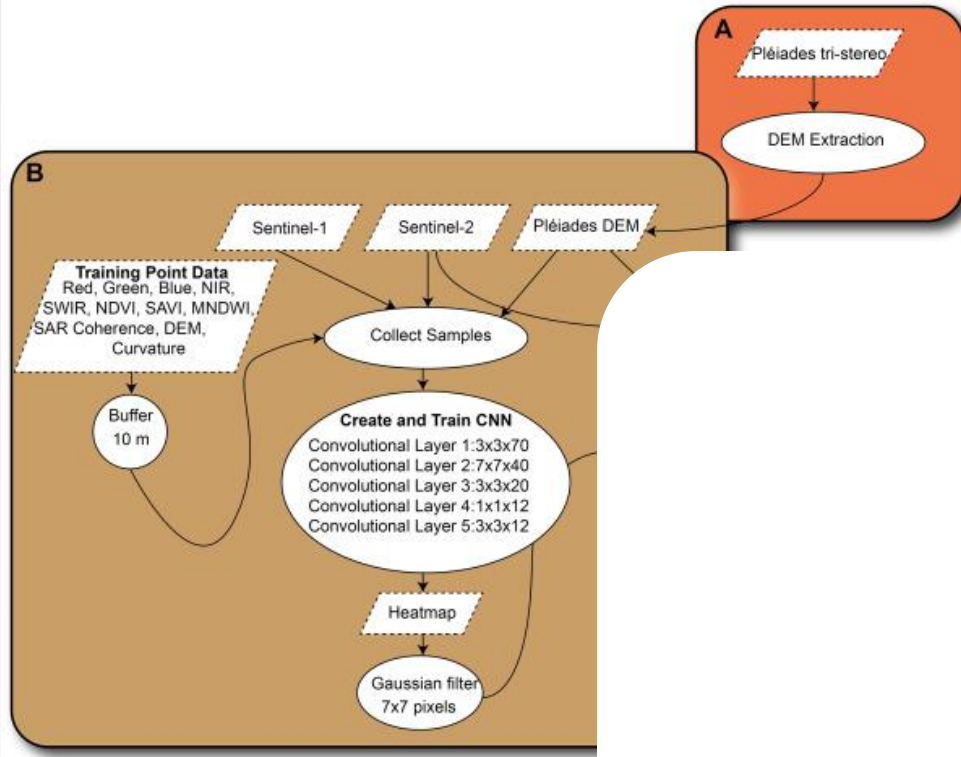


Artificial intelligence needs a lot of data to be trained



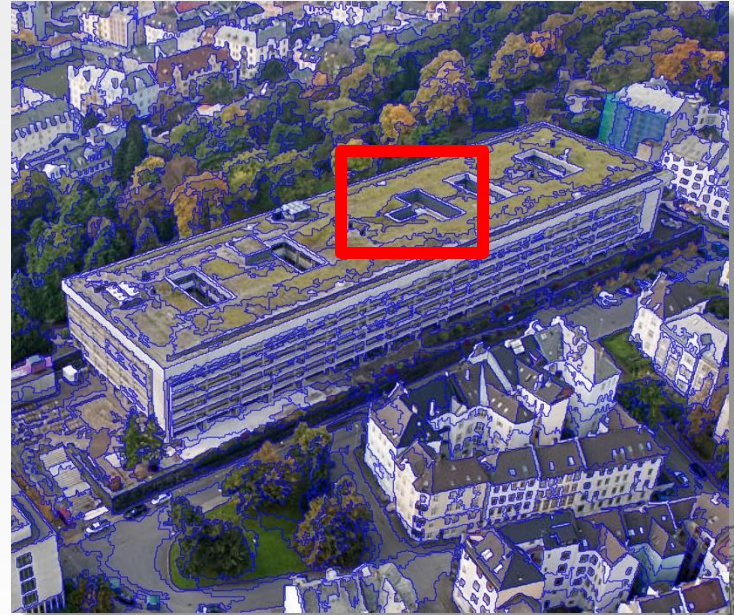
Bolch, T., Rohrbach, N., Kutuzov, S., **Robson, B.A.** and Osmonov, A., 2019. Occurrence, evolution and ice content of ice-debris complexes in the Ak-Shirak, Central Tien Shan revealed by geophysical and remotely-sensed investigations. *Earth Surface Processes and Landforms*, 44(1), pp.129-143.





Most deep learning studies stop here

Image segmentation adds more information to a classifications

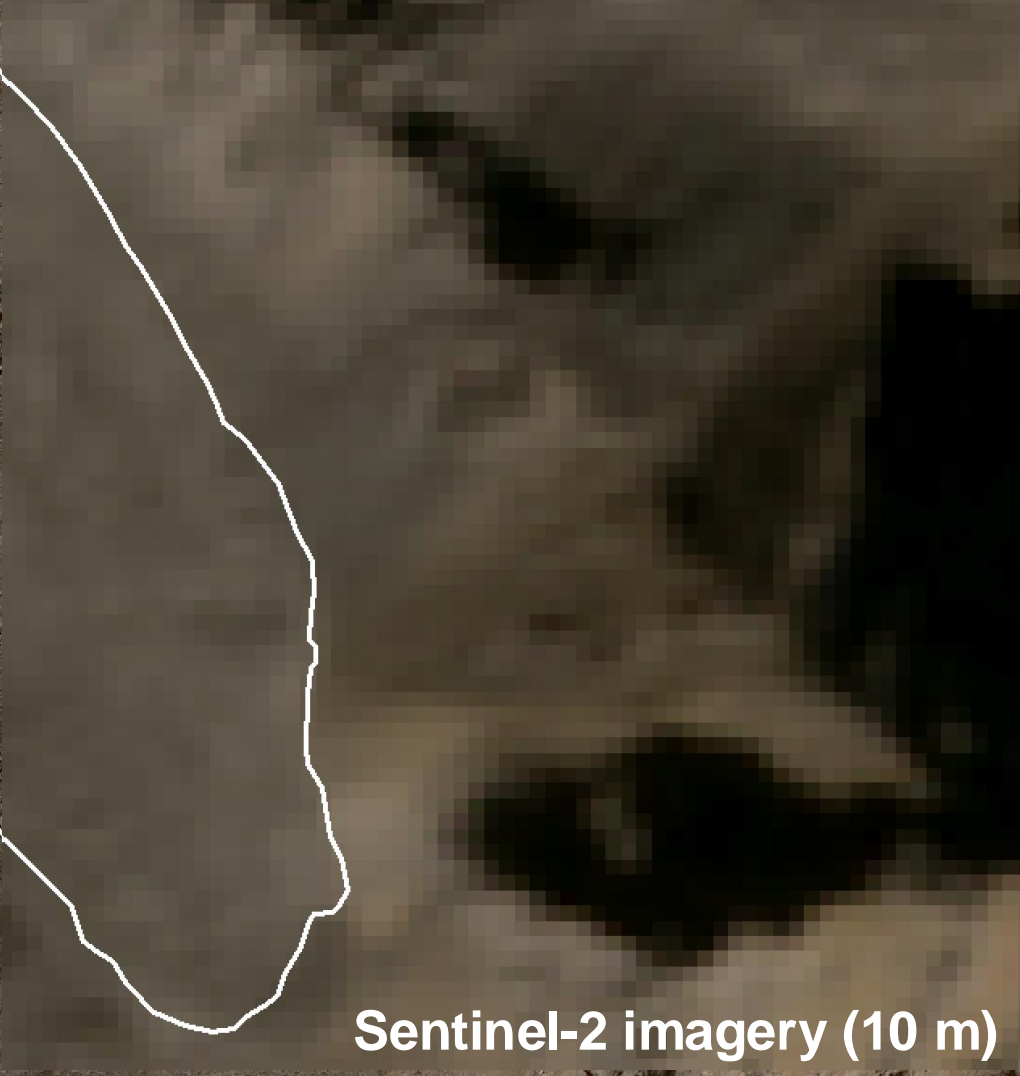


Object Based Image Analysis (OBIA)

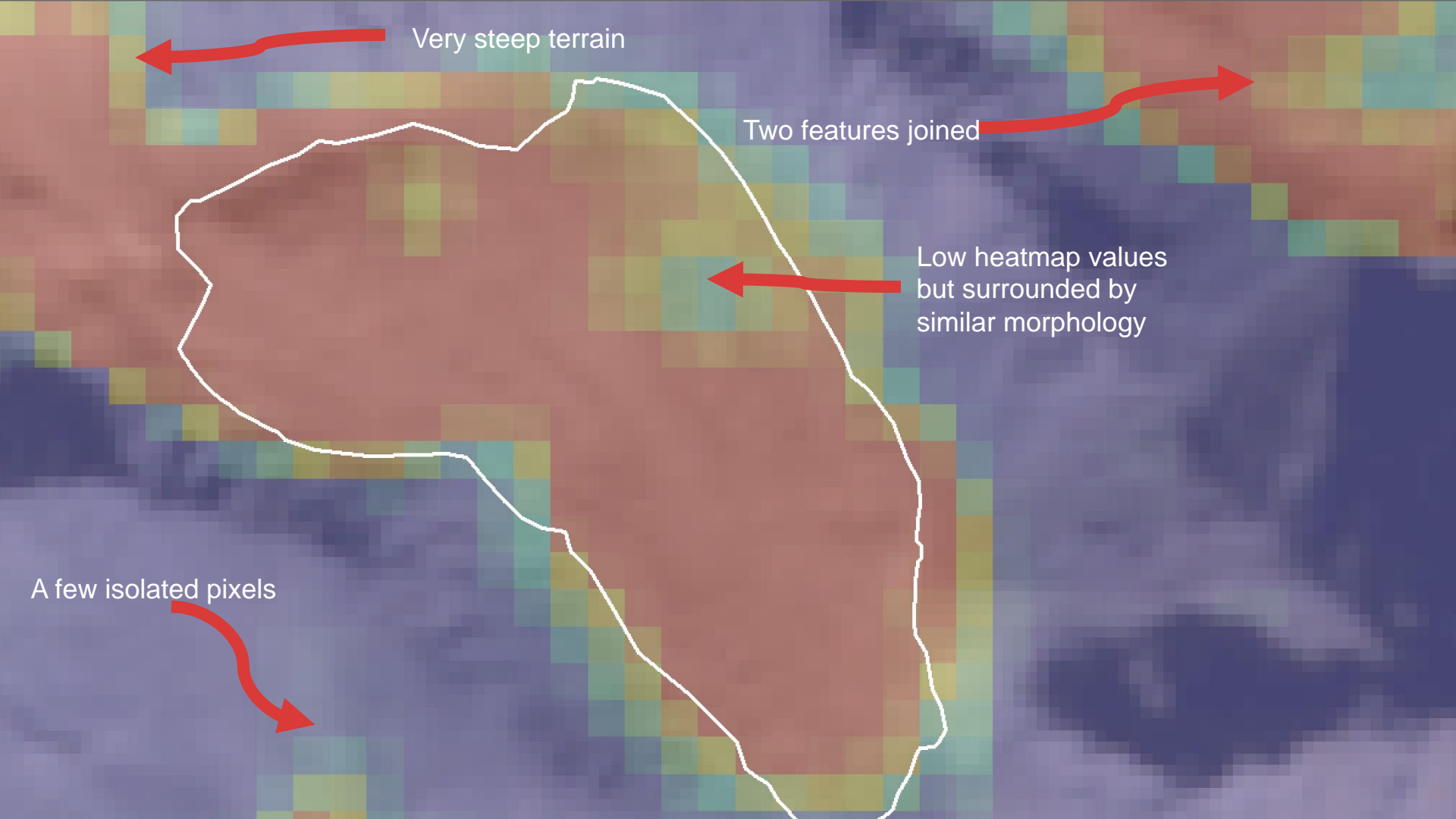




Pleiades imagery (50 cm)



Sentinel-2 imagery (10 m)

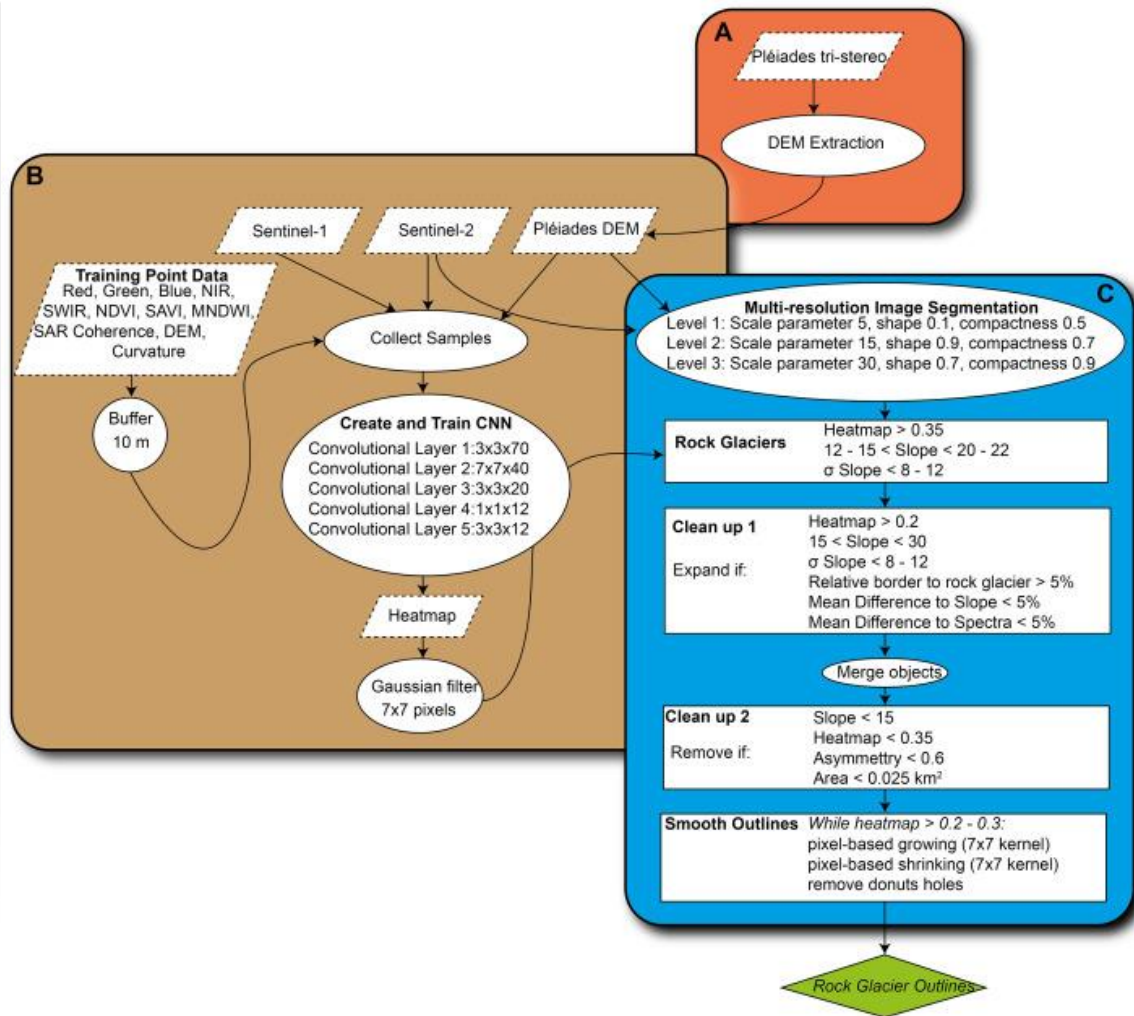


Very steep terrain

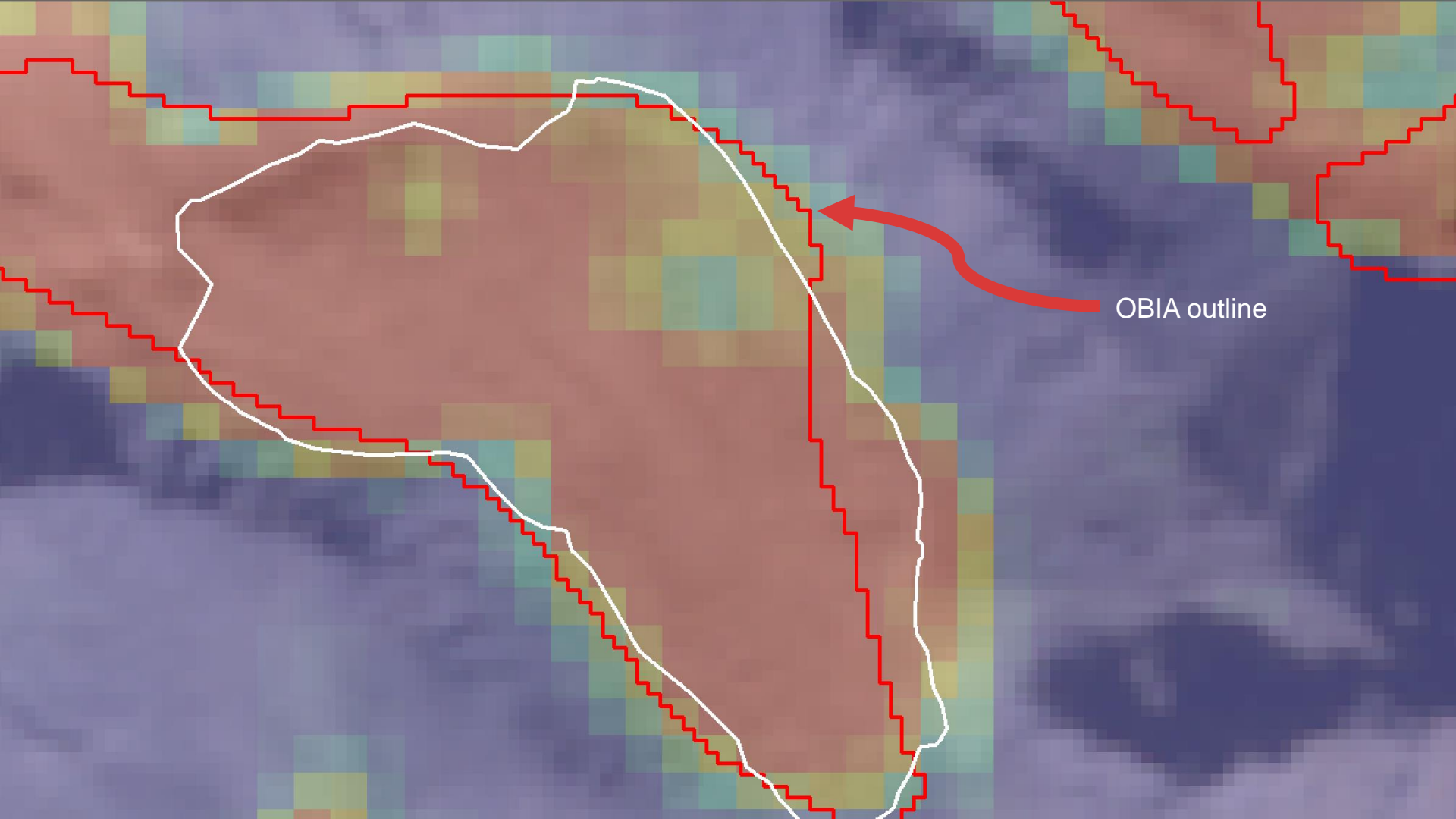
Two features joined

Low heatmap values
but surrounded by
similar morphology

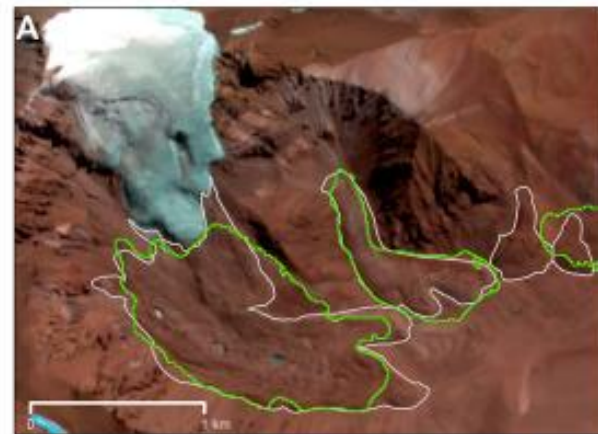
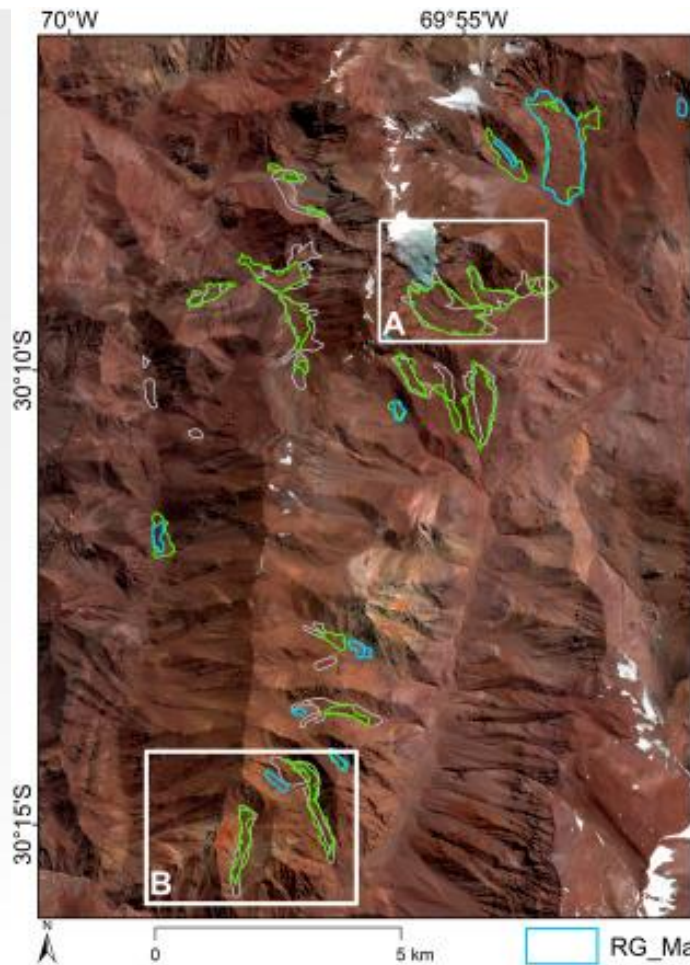
A few isolated pixels



Refinement due to image segmentation

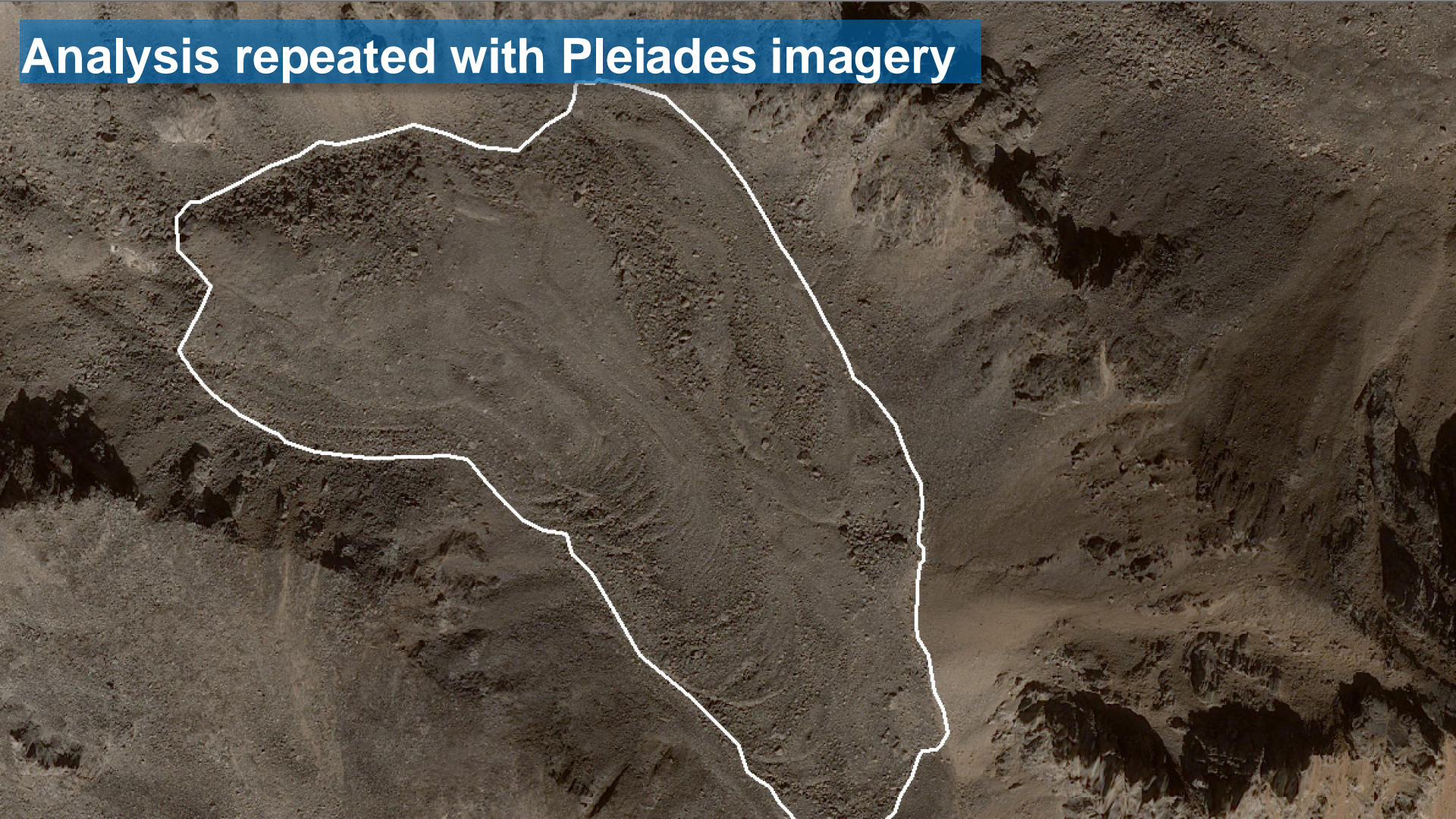


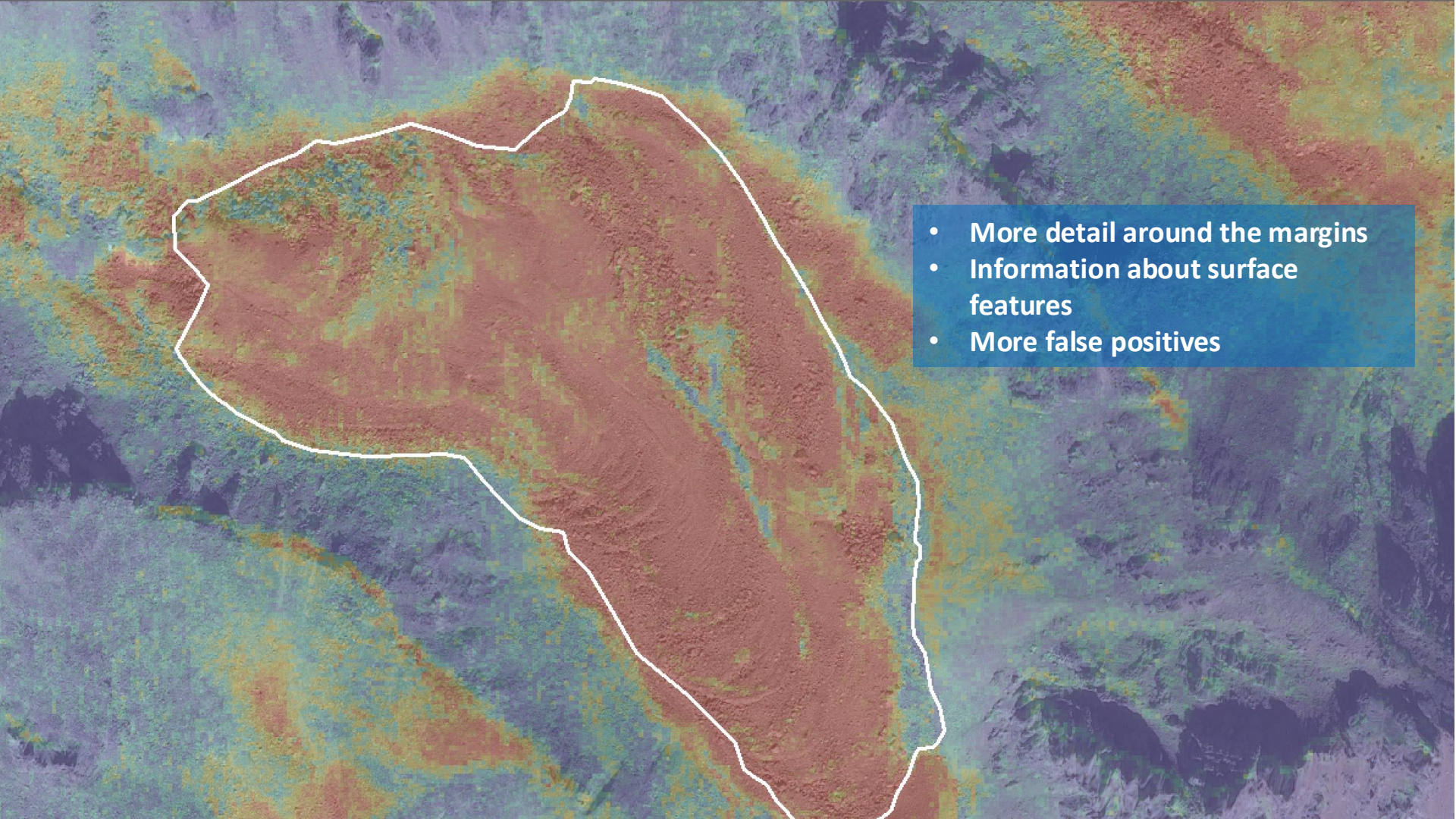
OBIA outline



RG_Man (Training)
 RG_Man (Validation)
 CNN_OBIA

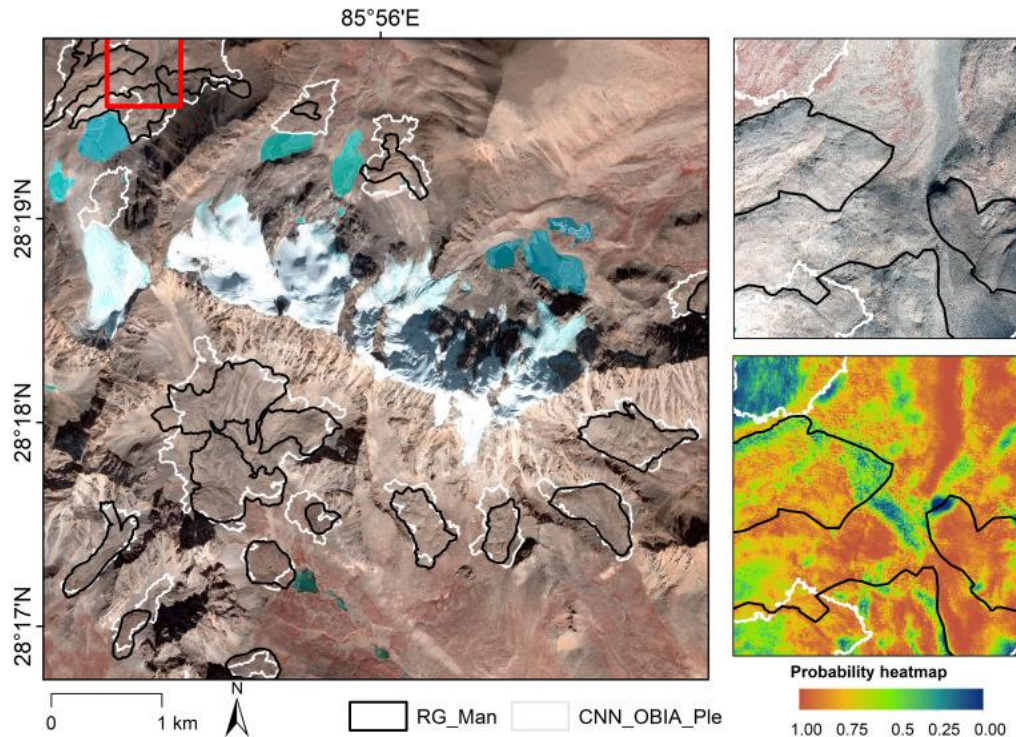
Analysis repeated with Pleiades imagery





- More detail around the margins
- Information about surface features
- More false positives




Using higher-resolution imagery increases the user accuracy



Classification	User accuracy (%)	Producer accuracy (%)
Poiqu subset (Sentinel-2)	62.9	87.4
Poiqu subset (Pléiades)	72.0	88.4

Automated debris-covered glacier mapping

An integrated deep learning and object-based image analysis approach for mapping debris-covered glaciers

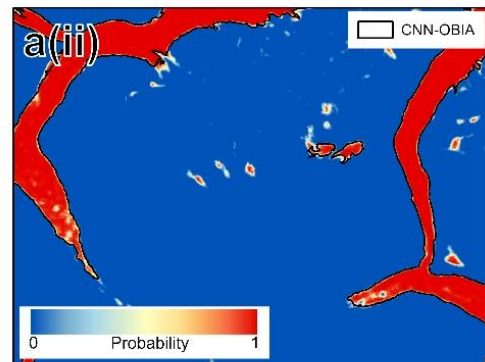
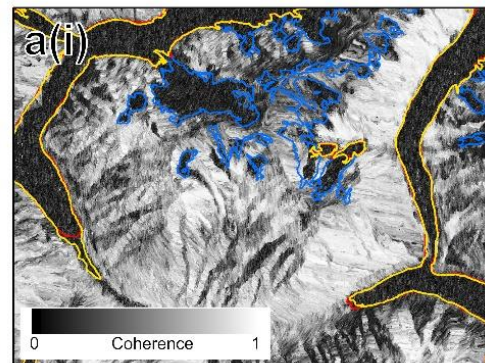
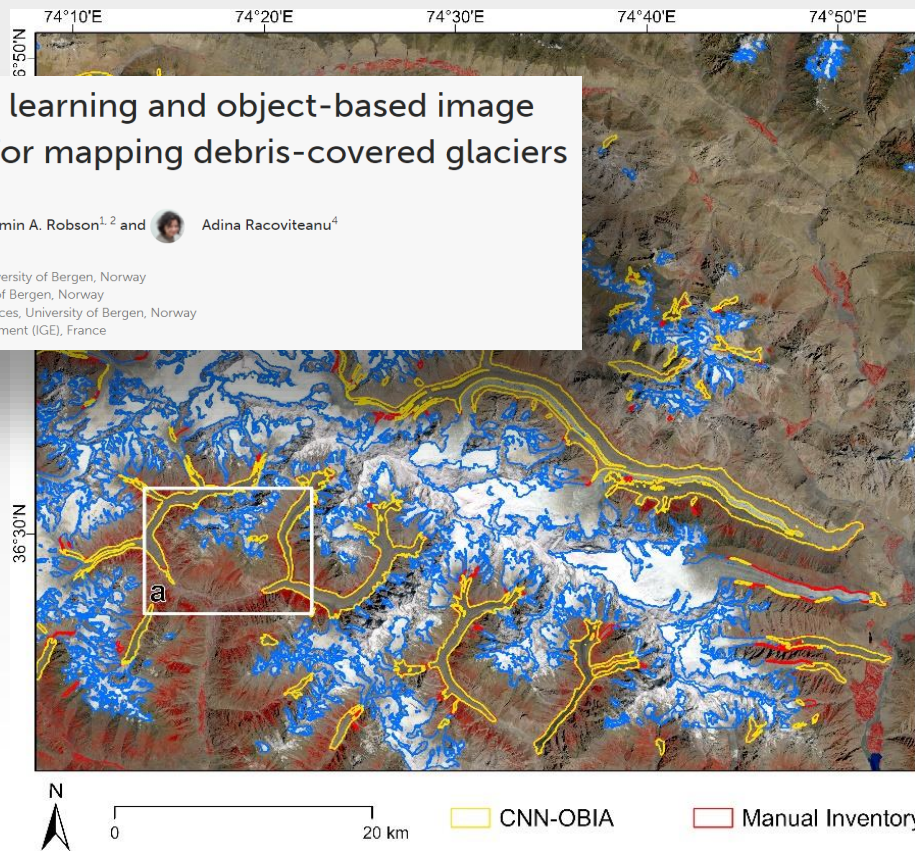
 Daniel J. Thomas^{1, 2, 3*},  Benjamin A. Robson^{1, 2} and  Adina Racoviteanu⁴

¹ Faculty of Mathematics and Natural Sciences, University of Bergen, Norway

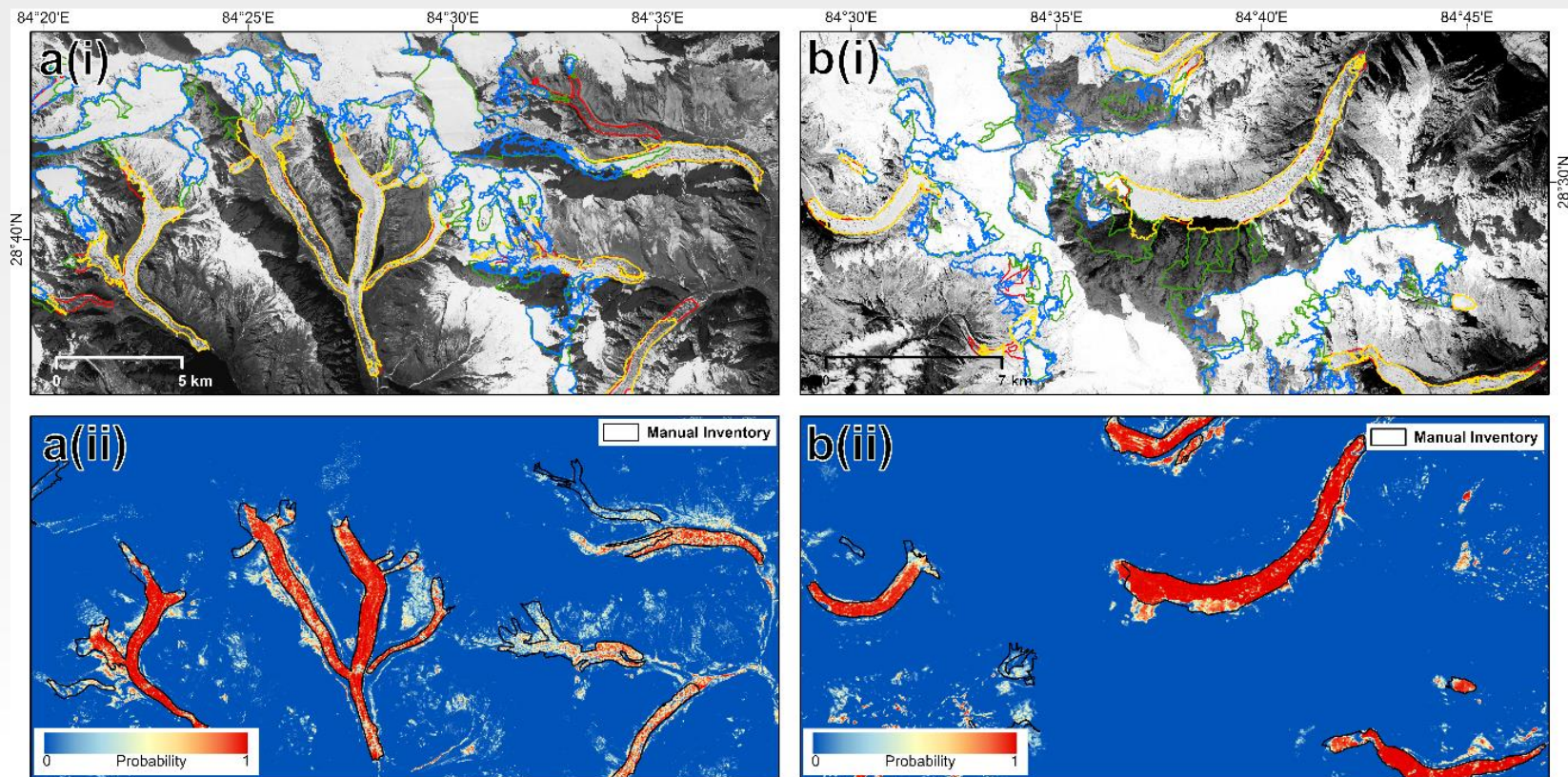
² Bjerknes Centre for Climate Research, University of Bergen, Norway

³ Department of Geography, Faculty of Social Sciences, University of Bergen, Norway

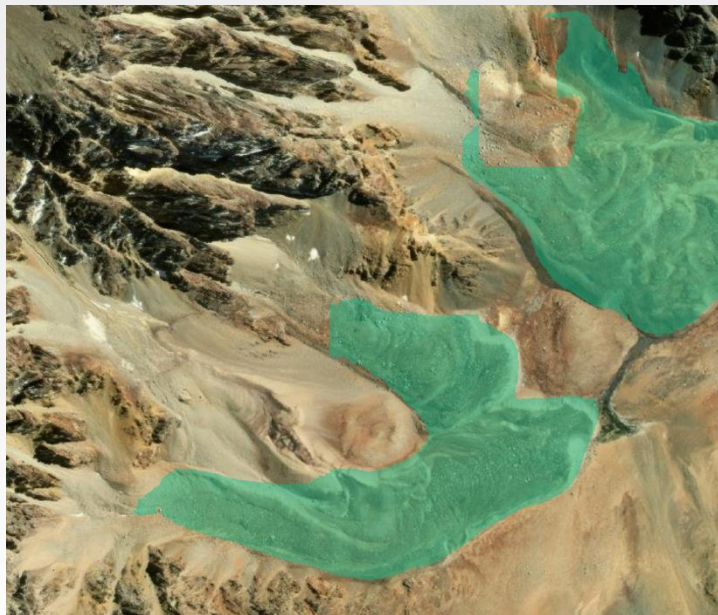
⁴ UMR5001 Institut des Géosciences de l'Environnement (IGE), France



Deep learning for multi-temporal analysis



Scaling up to a regional scale



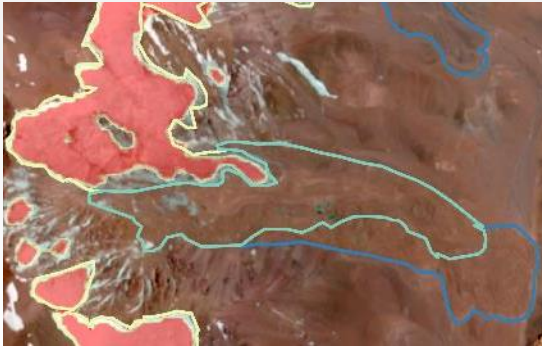
- DeepLabV3 improves image segmentation accuracy through atrous convolution and multi-scale context.
- Used to produce Tibetan Plateau RG inventory (TroGI) (Sun et al, 2024)
- Retrained and being applied to the Andes (PhD thesis Daniel Thomas)



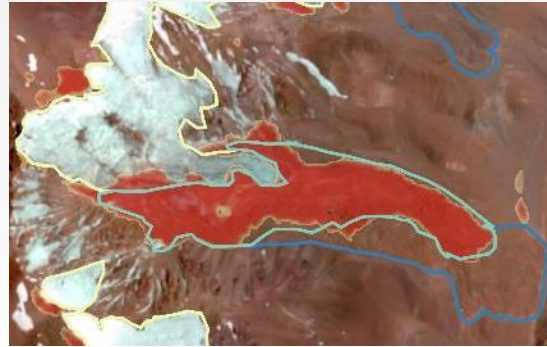
Future directions of AI-based mapping

- Multi-class deep learning for cryospheric mapping in the Andes
Daniel Thomas – PhD student at UiB
- Trustworthiness and ground truthing of AI-based mapping
Sunil Tamang – PhD student at UC/GRI

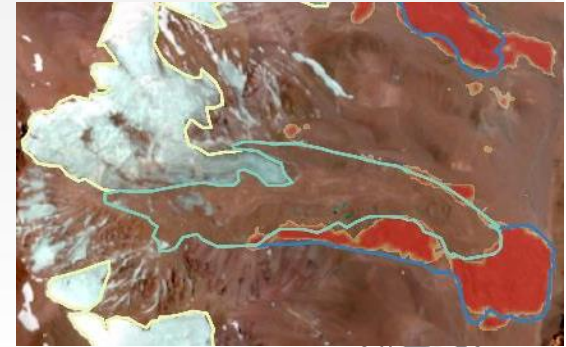
Heatmap – clean ice



Heatmap – debris-covered ice



Heatmap – rock glaciers

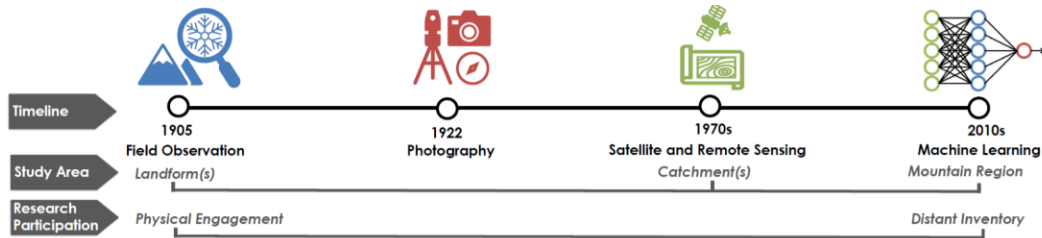


Yellow outline: Clean Ice Manual Inventory

Green outline: Debris-covered Ice Manual Inventory

Blue outline: Rock Glacier Manual Inventory

Trustworthiness and Ground-Truthing of Machine Learning Approach for Regional Scale Rock Glacier Inventories



Practices of Rock Glacier Inventory

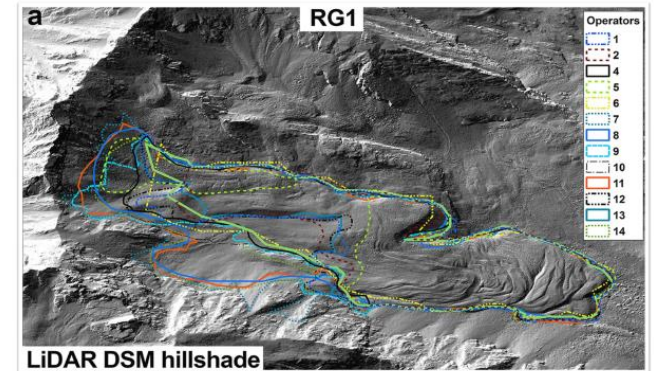


Figure adapted with permission from Brardinoni et al. (2019).

Subjectivity in Rock Glacier Inventory

Sunil TAMANG
PhD Candidate

University of Canterbury, New Zealand
sunil.tamang@pg.canterbury.ac.nz





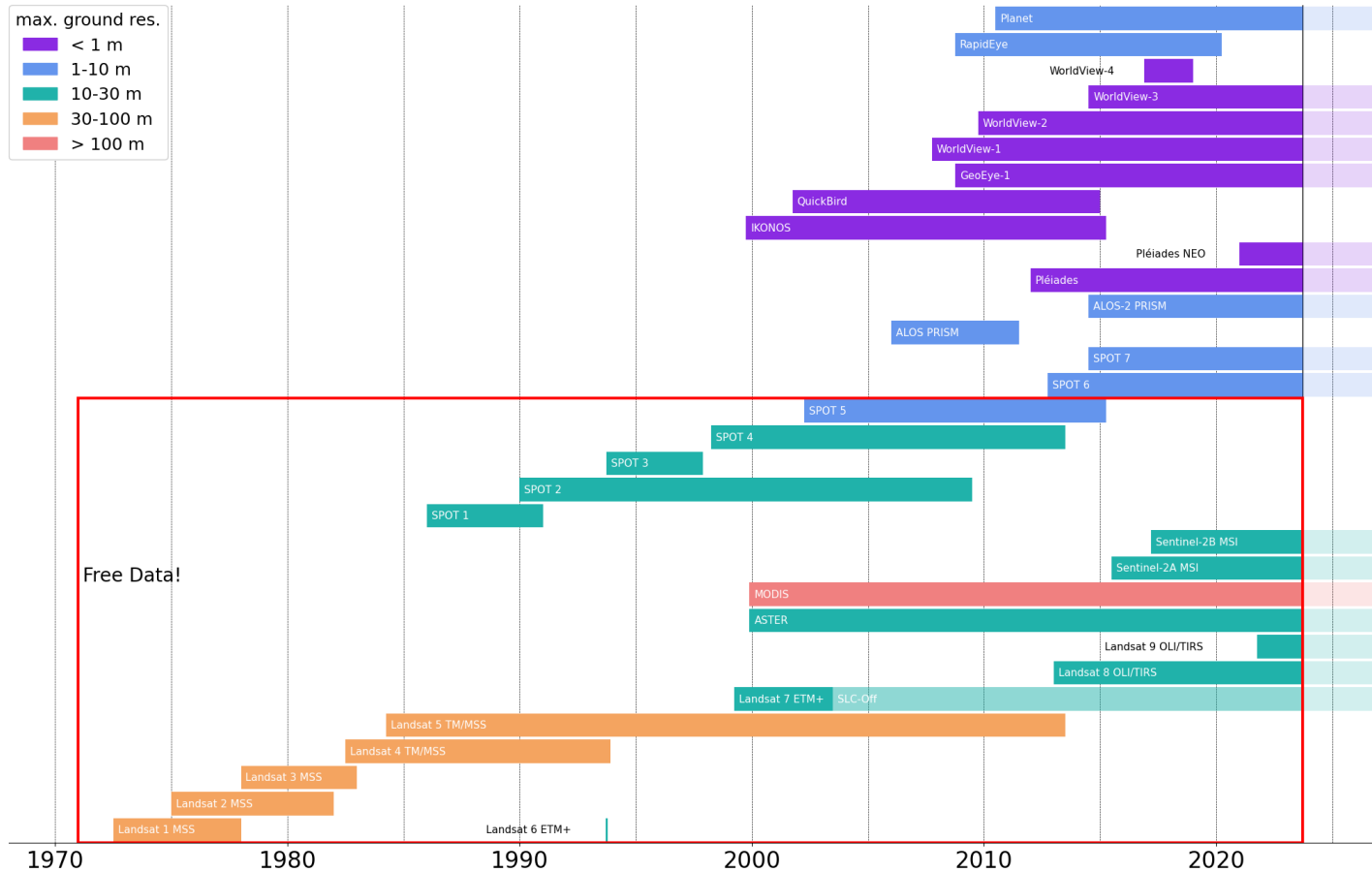
Quantifying changes using remote sensing data



UNIVERSITY OF BERGEN

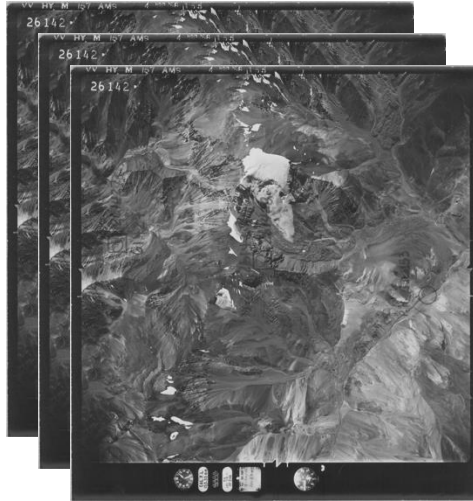
max. ground res.

- < 1 m
- 1-10 m
- 10-30 m
- 30-100 m
- > 100 m



Data available for studying rock glacier change

UNIVERSITY OF BERGEN



Historical Aerial Photography



Stereo Satellite Imagery



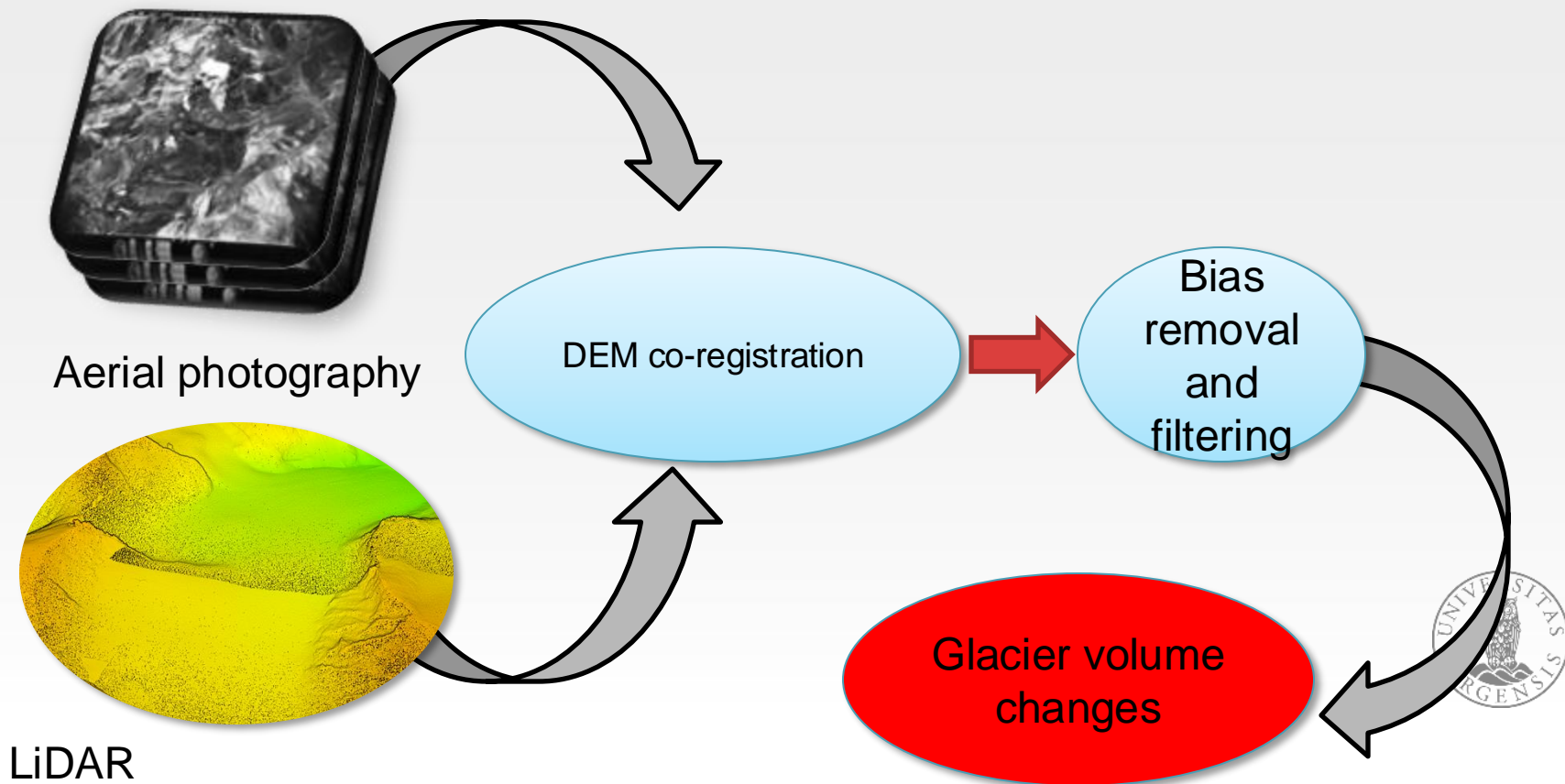
Terrestrial Laser scanning



UAVs

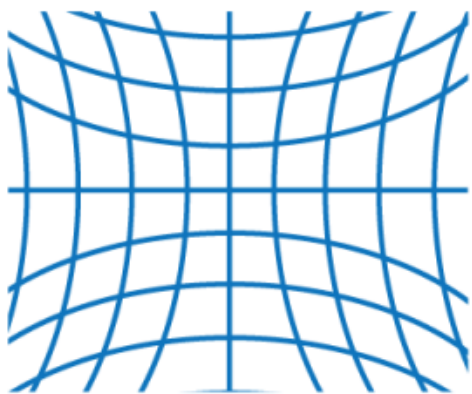


Methods

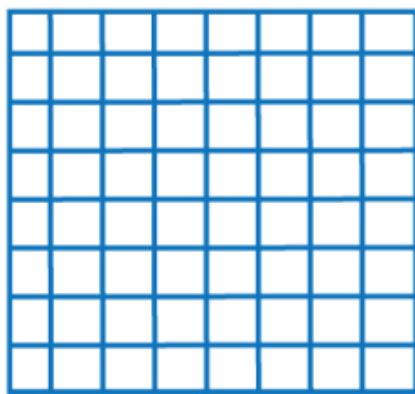


Problems with the data

- Aerial photographs had no camera calibration report
 - Focal length written on the photographs
 - Fiducial marks can be estimated (ish)
 - ...radial lens distortion?



negative radial distortion
"pincushion"

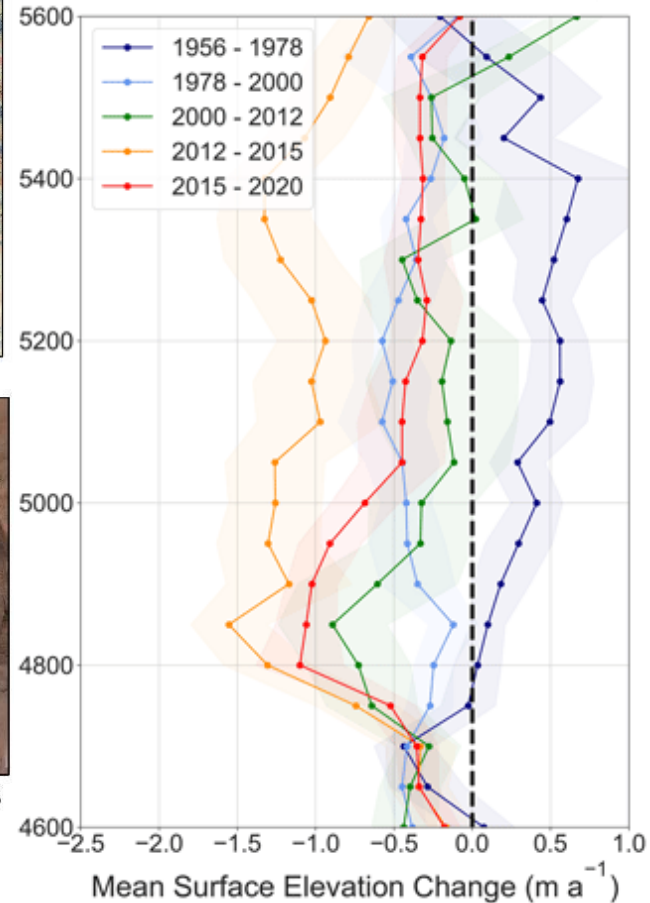
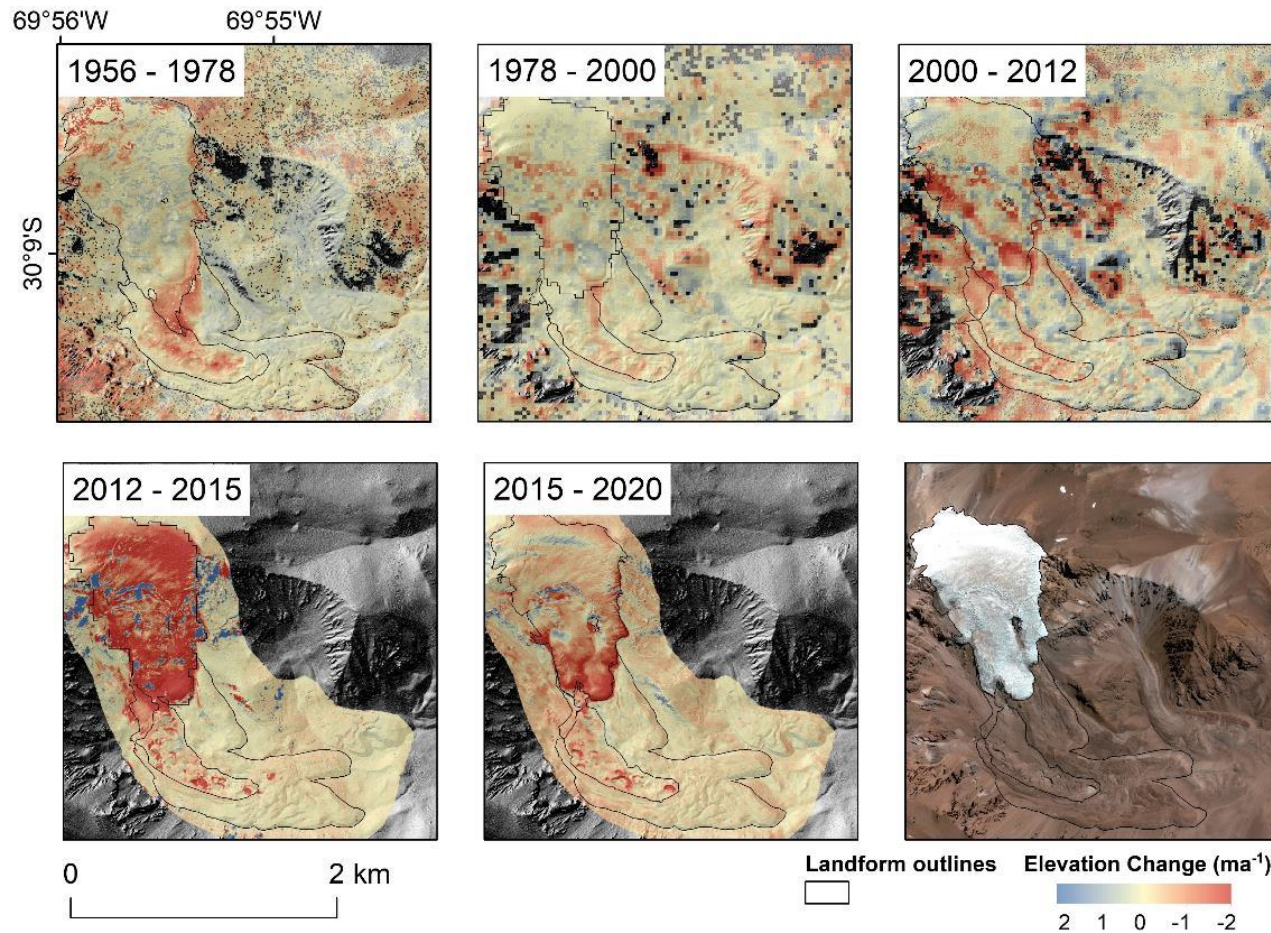


no distortion



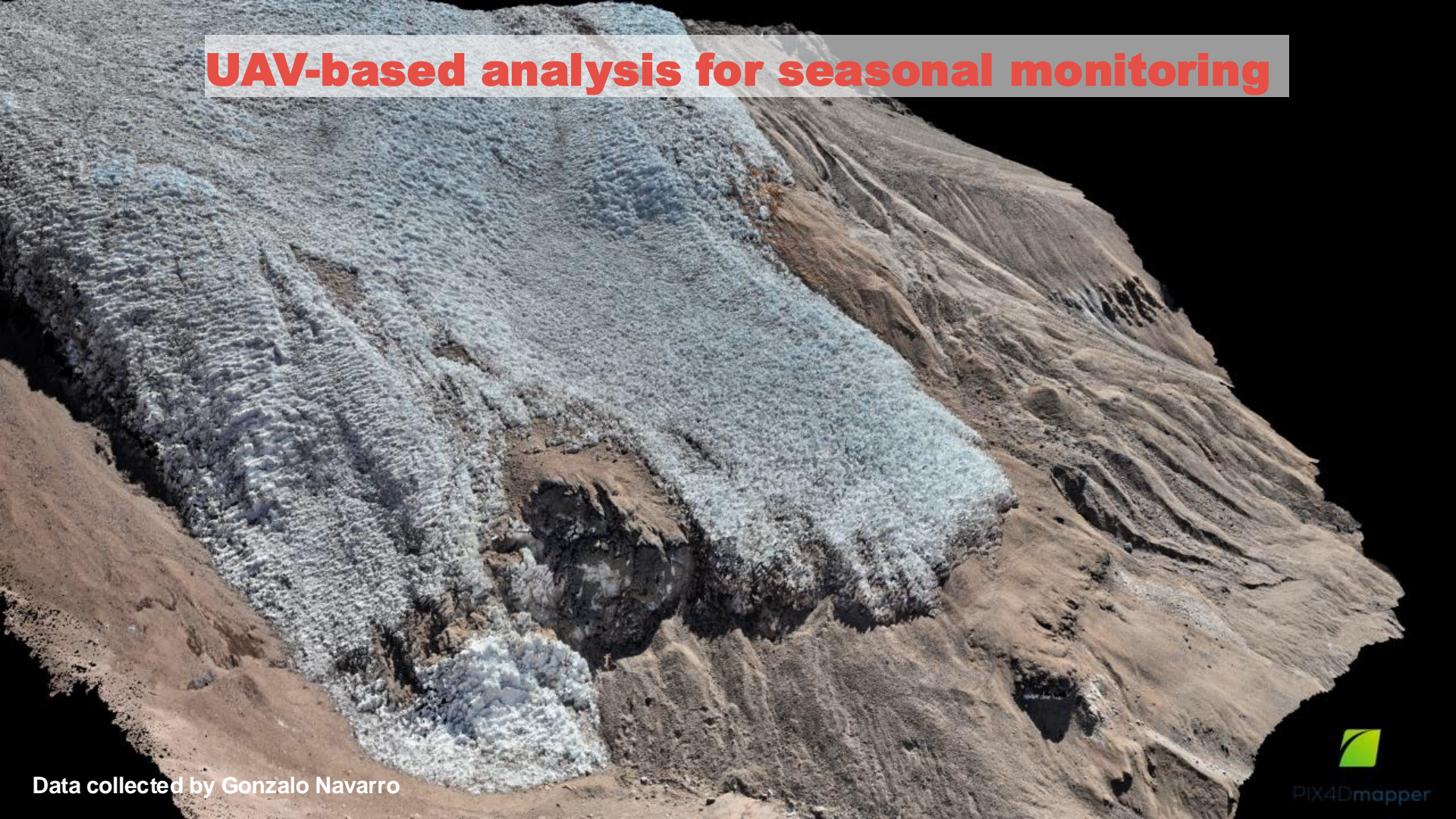
positive radial distortion
"barrel"





Robson, B.A., MacDonell, S., Ayala, Á., Bolch, T., Nielsen, P.R. and Vivero, S., 2022. Glacier and rock glacier changes since the 1950s in the La Laguna catchment, Chile. *The Cryosphere*, 16(2), pp.647-665.

UAV-based analysis for seasonal monitoring

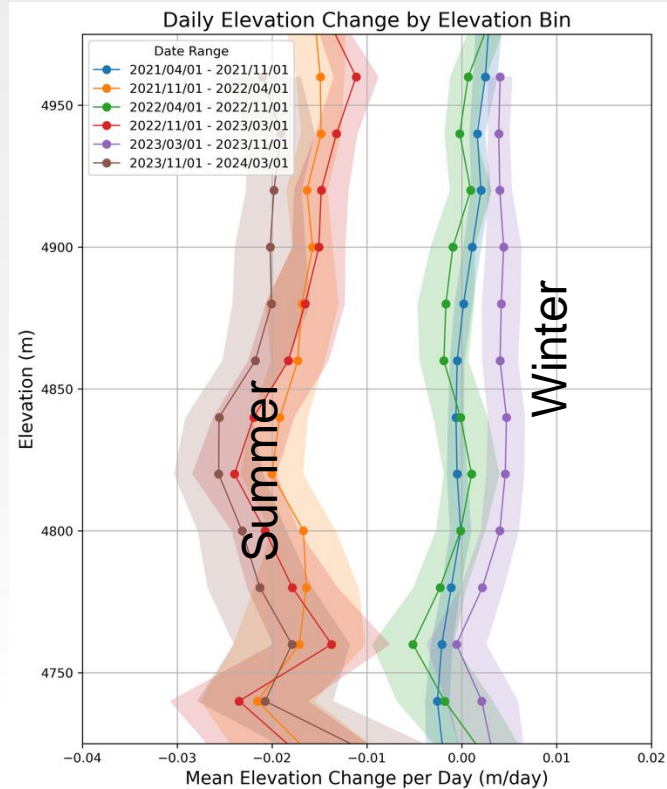
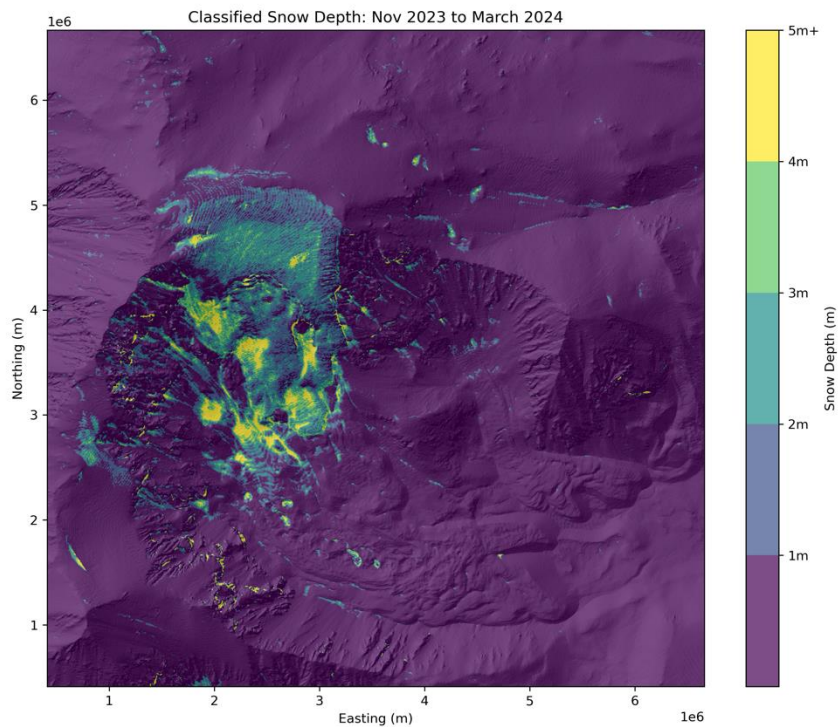


Data collected by Gonzalo Navarro

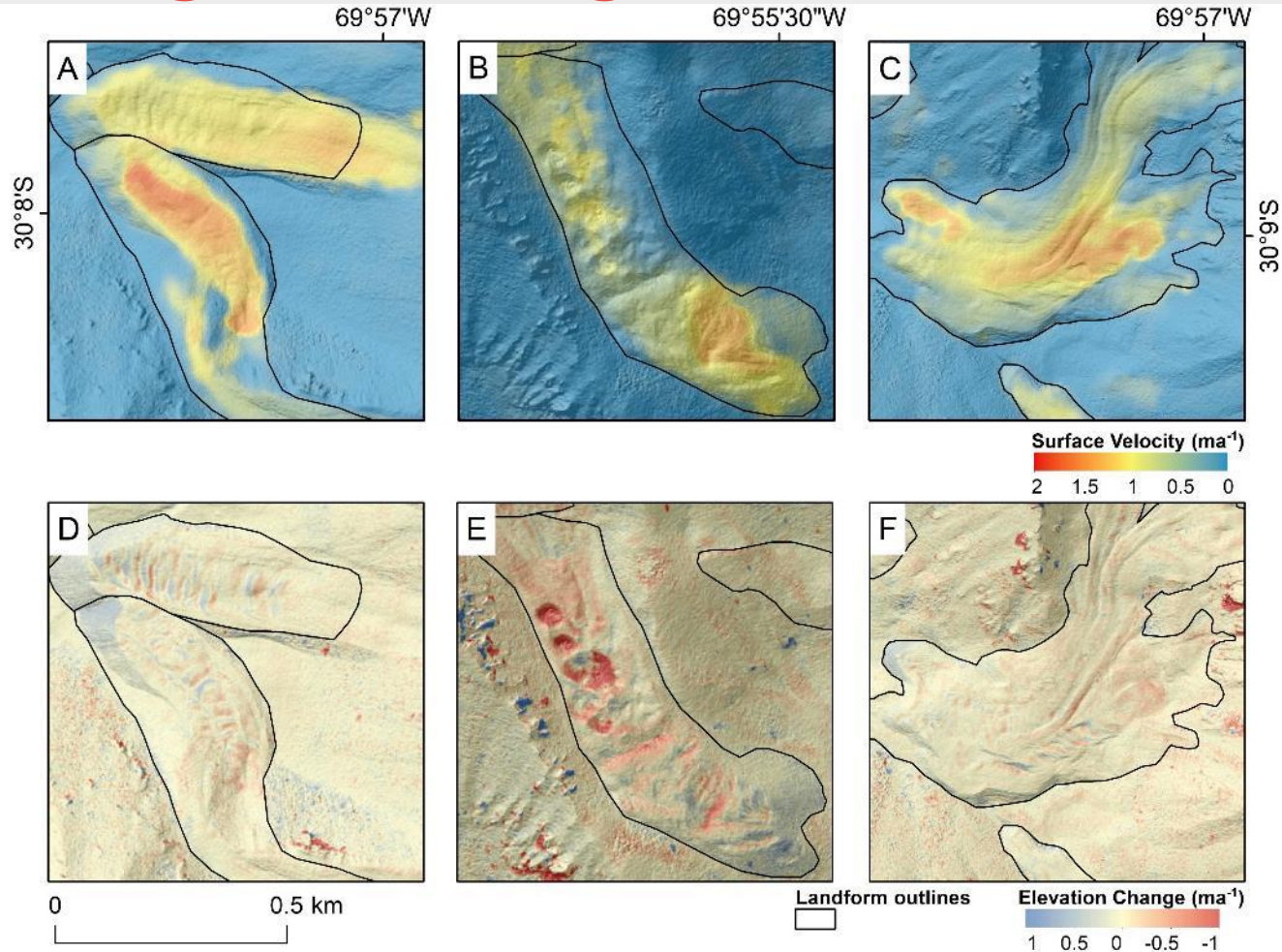


PIX4Dmapper

High-resolution seasonal changes

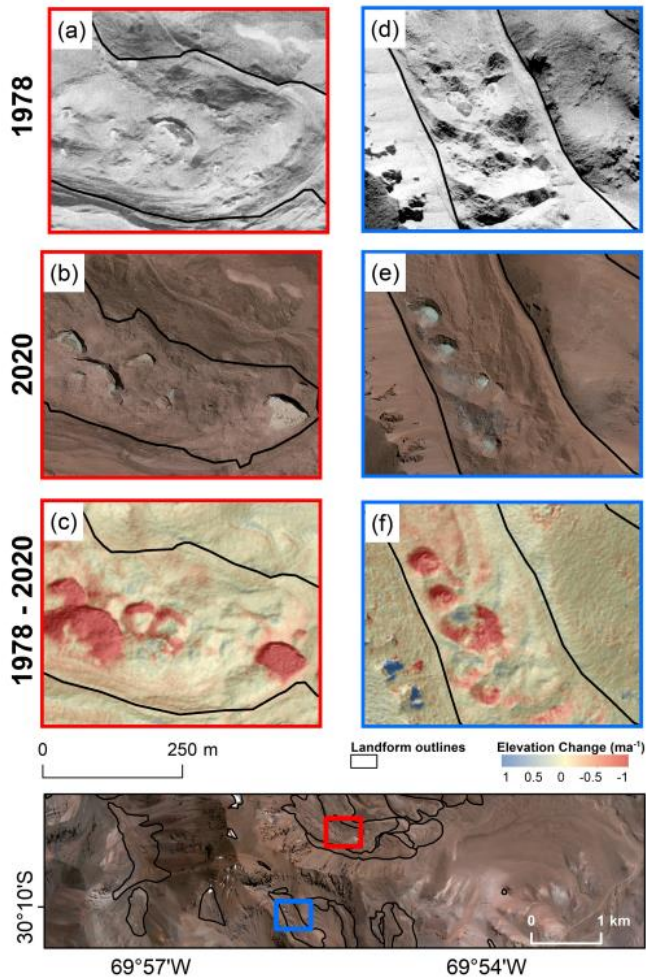


Rock glacier changes in the catchment

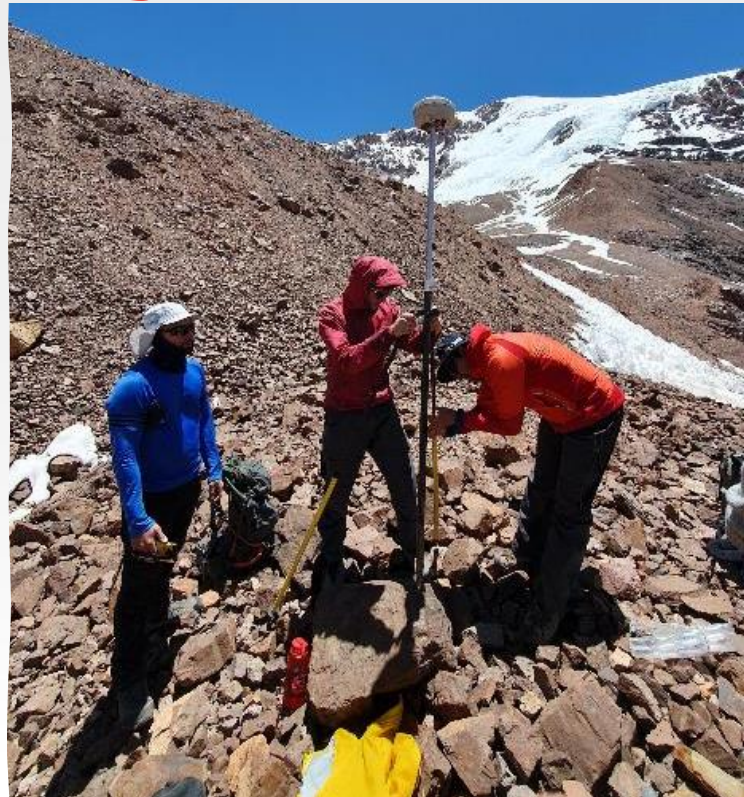
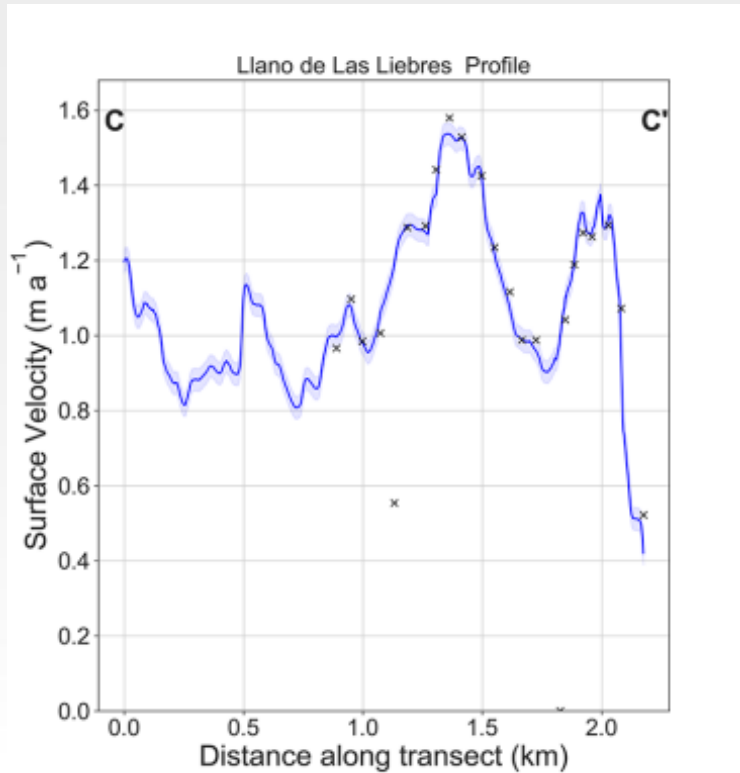


Evidence of glacier-rock glacier transition

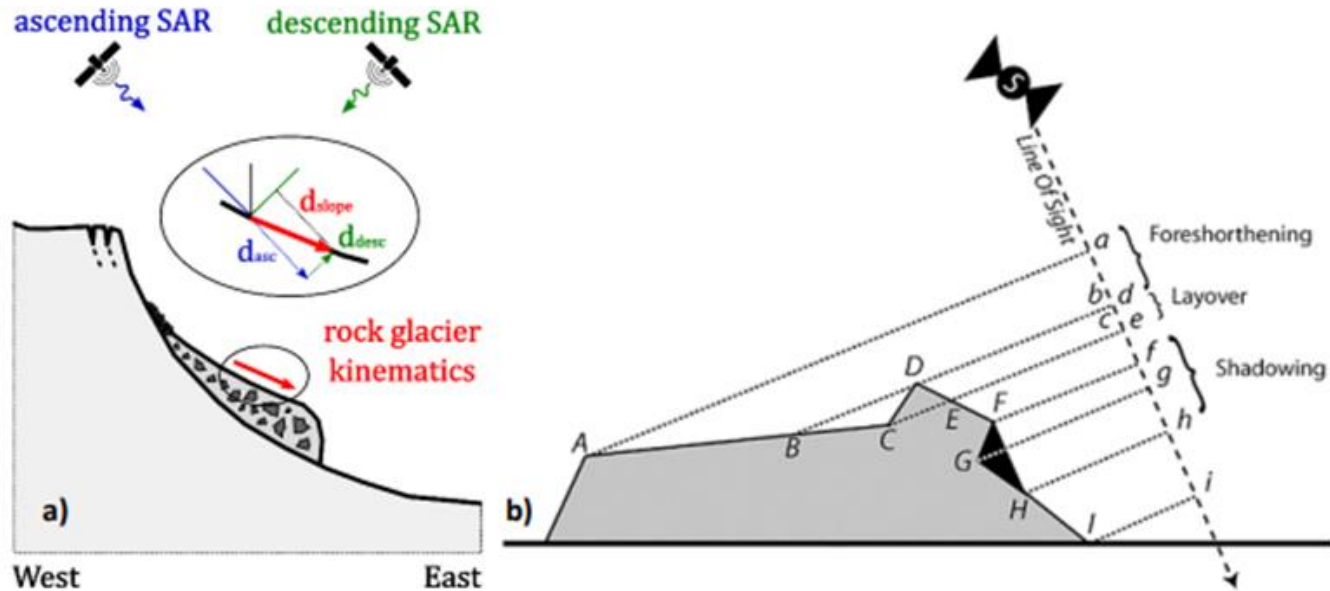
UNIVERSITY OF BERGEN



Validation of velocity data

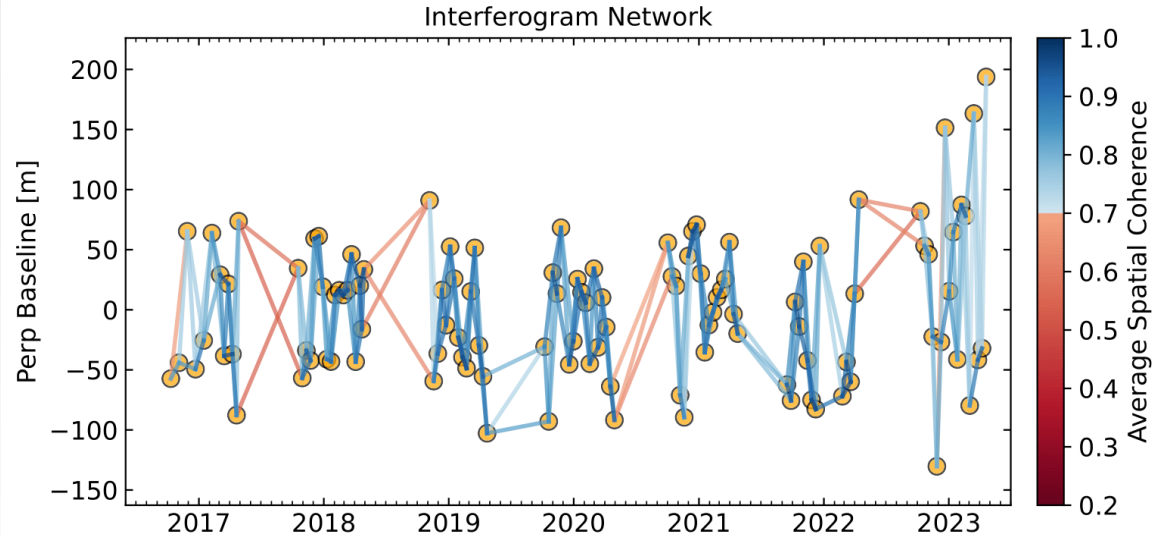


InSAR Analysis of rock glaciers



InSAR Analysis of rock glaciers

- Study areas:
Manaslu (Himalaya)
and Semi Arid Andes
(Chile)
- Sentinel-1 InSAR
processing in ISCE
- SBAS stacking in
Mintpy



SAR Interferometry

- Produces a stack of co-registered and unwrapped interferograms

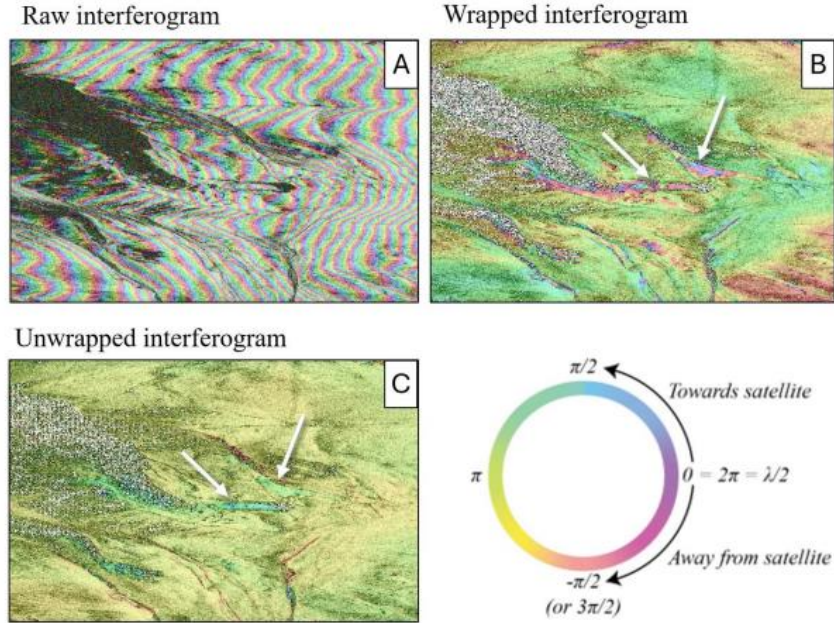
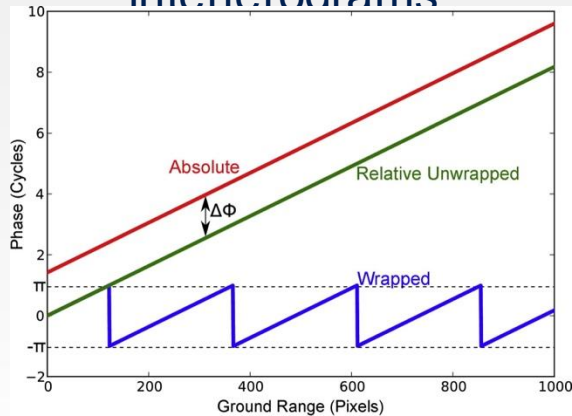
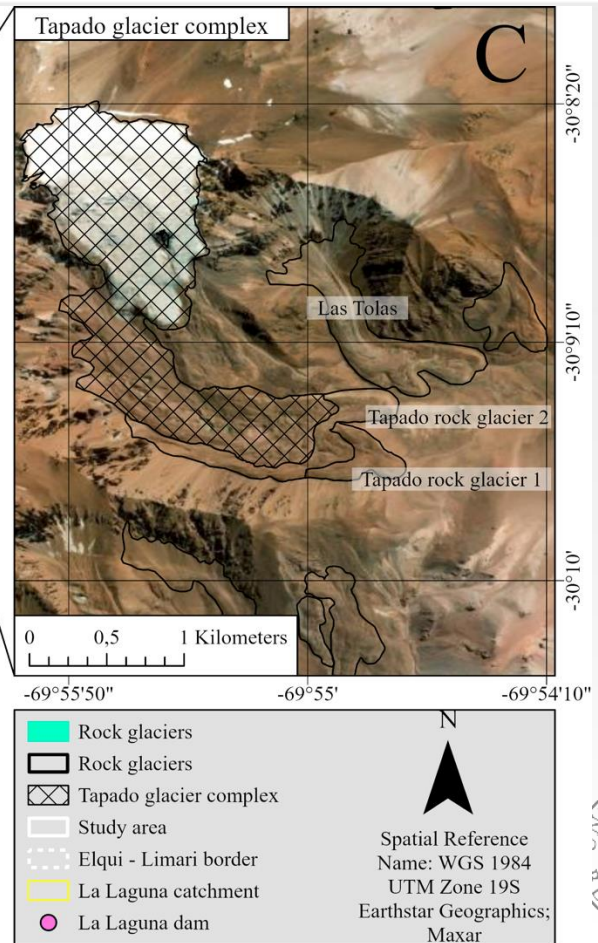
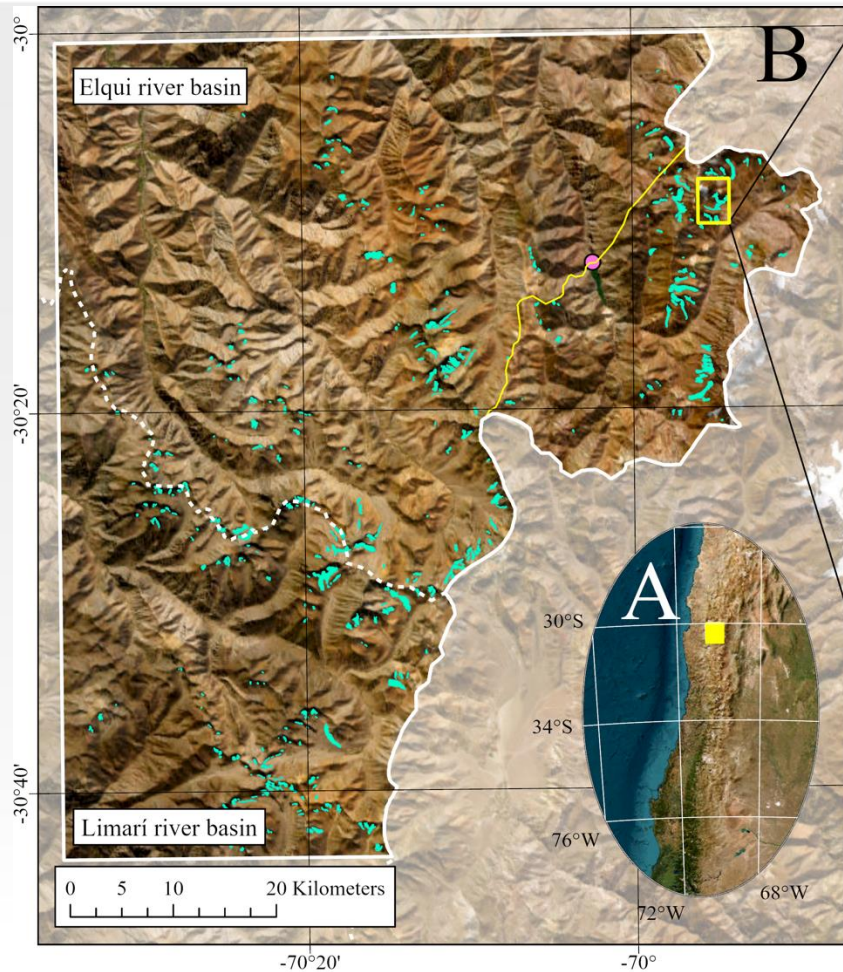


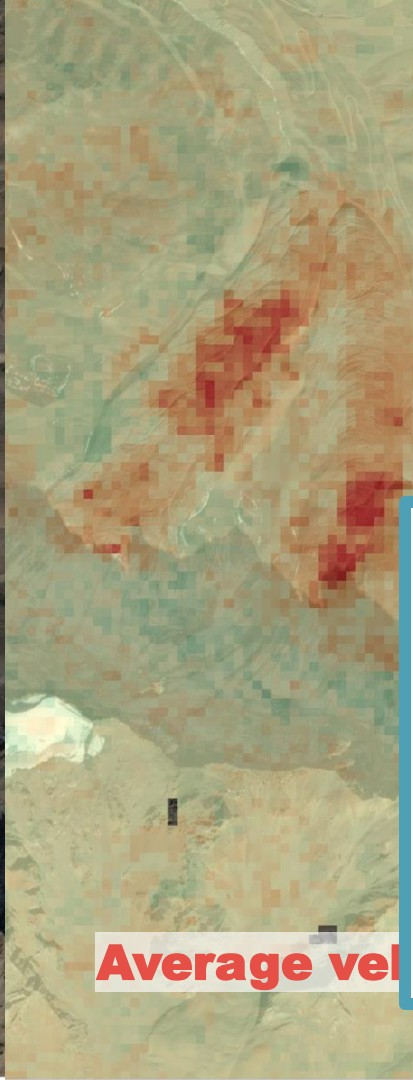
Figure 2-13: Wrapped phase in blue, relative unwrapped phase in green and absolute phase change in red. Figure from (Osmanoğlu et al., 2016).



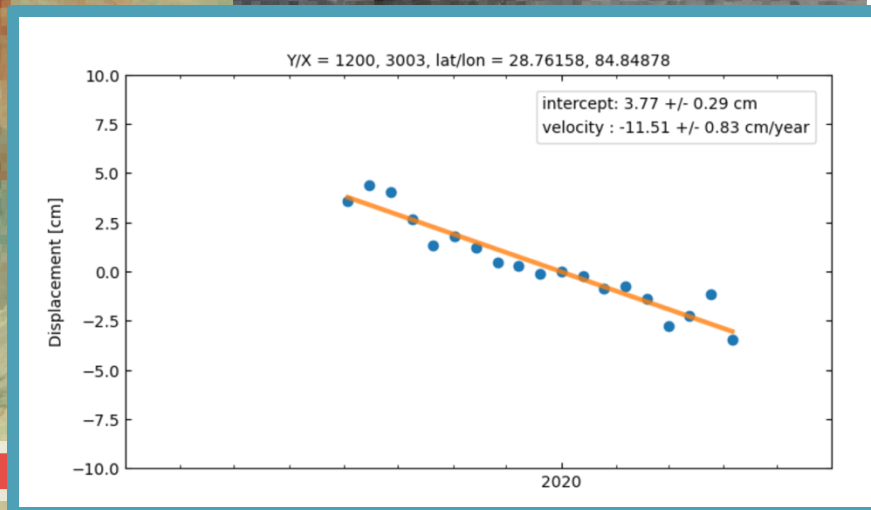
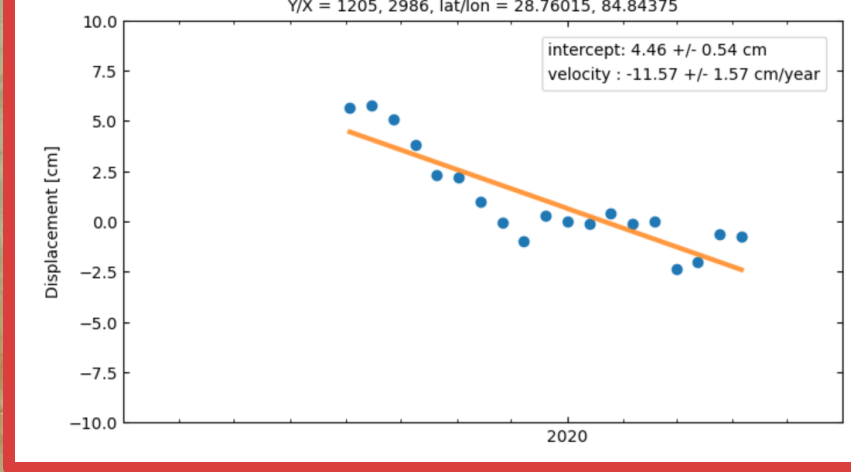




Google Earth



Average vel



Cumulative displacement - Tapado 2

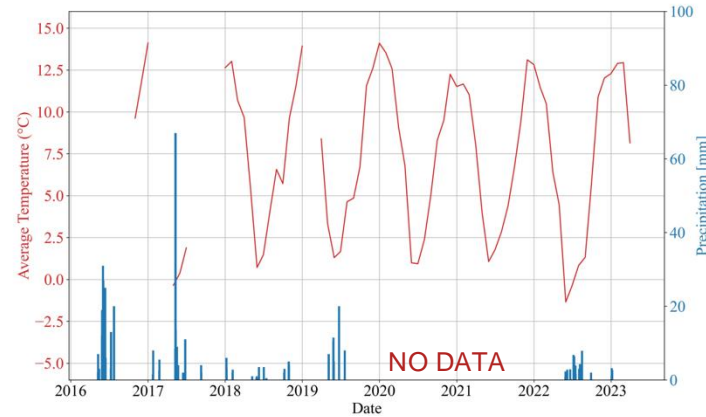
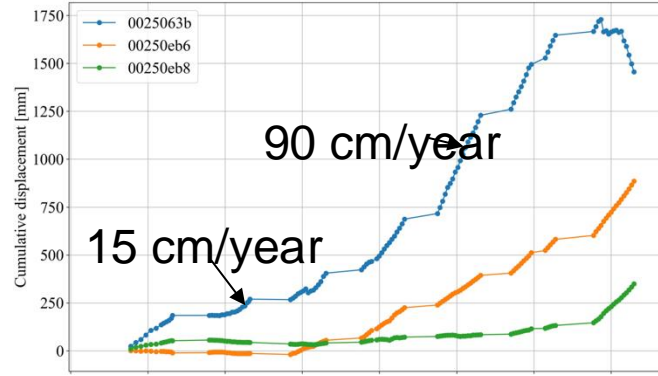
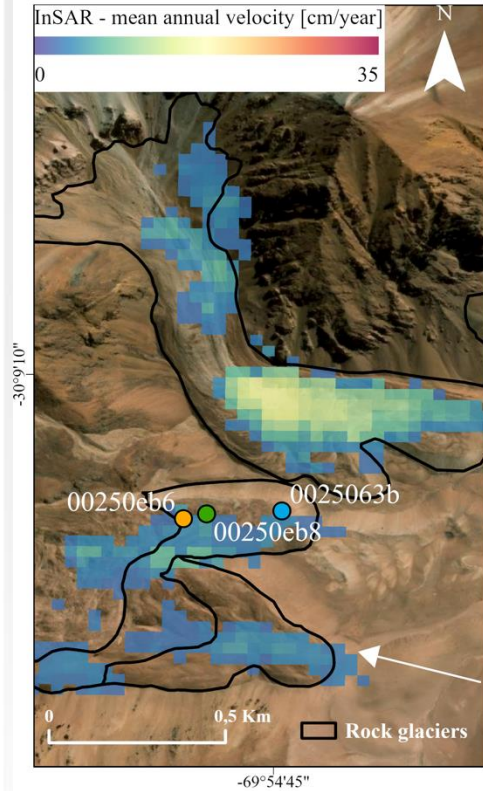


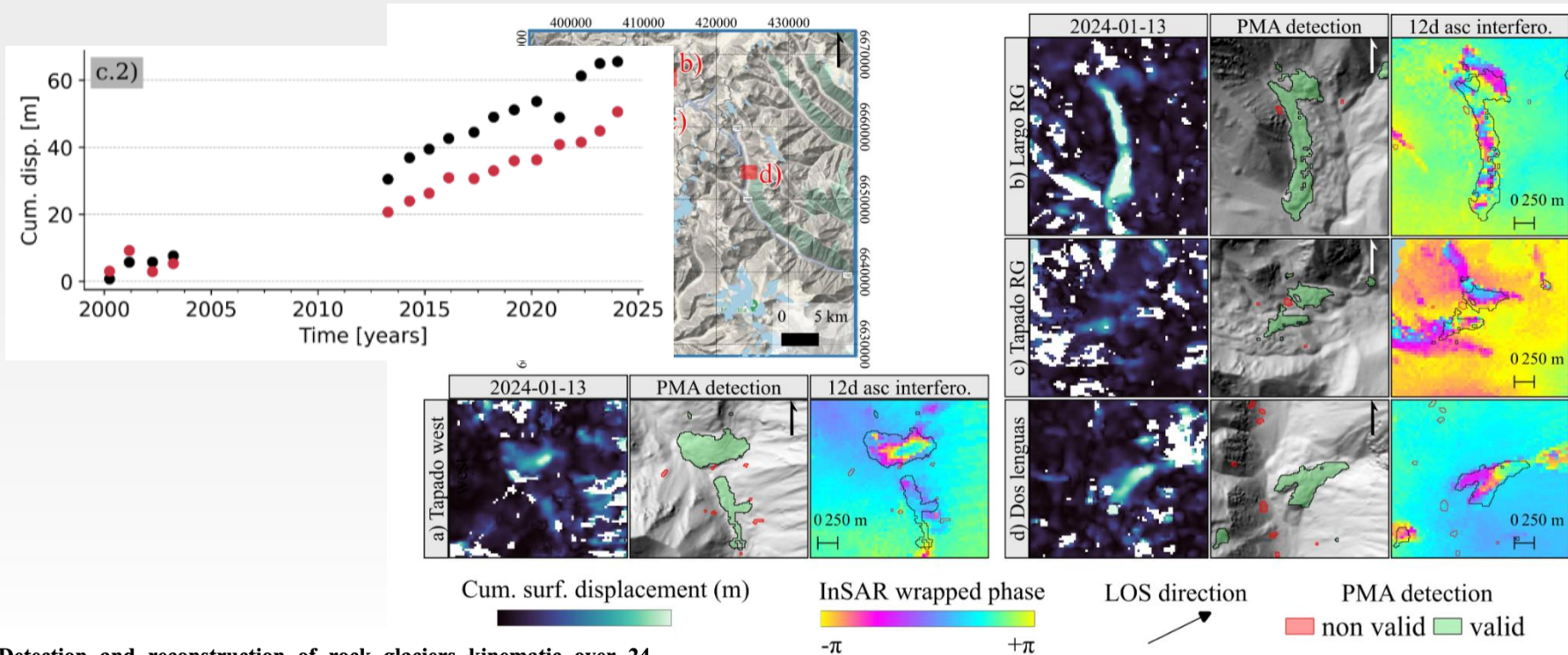
Figure 6-4: Overview map of the three selected rock glaciers in the A) La Laguna catchment. The three rock glaciers are: B) Tapado 2 rock glacier, C) Llano Des Las Liebres and D) Cl104300102. Location of the weather station is located at the La Laguna dam.

Studying seasonal changes

- Acceleration
 - 2017/2018 –
 - 2020/2021
- Seems to be linked to changes in precipitation



Assessing RG velocity from Landsat imagery



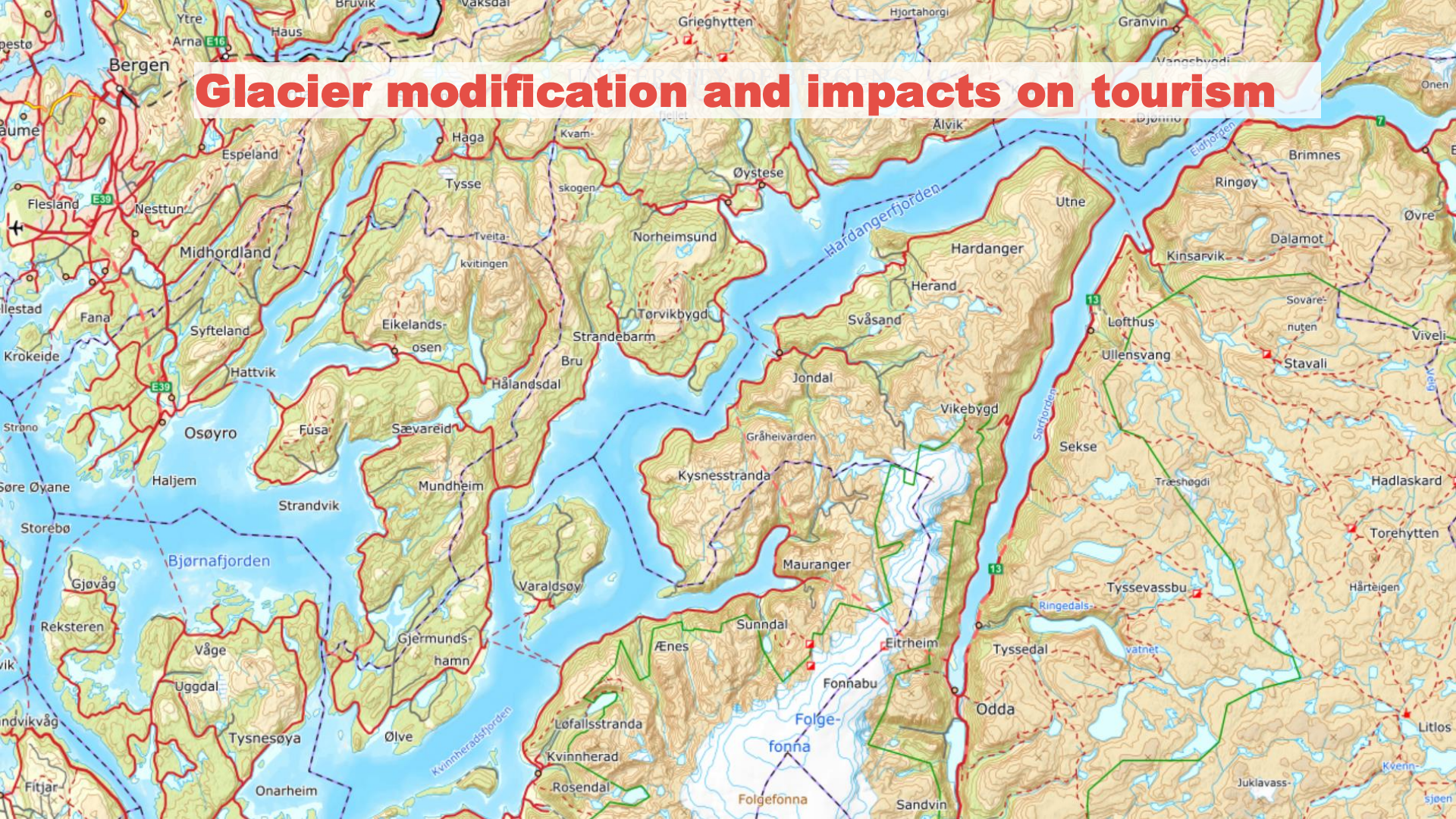
Detection and reconstruction of rock glaciers kinematic over 24 years (2000-2024) from Landsat imagery

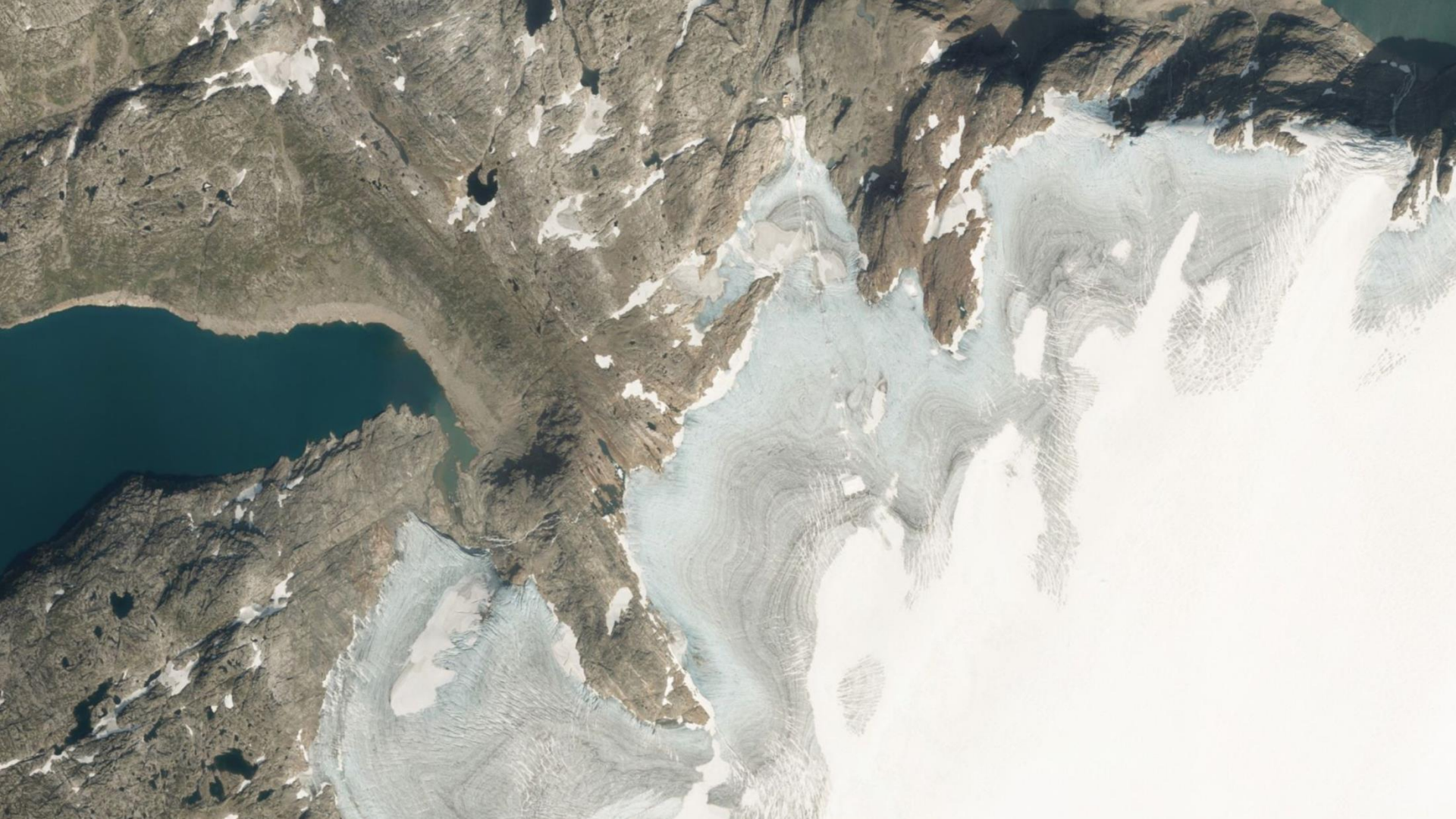


Glaciers a bit closer to home



Glacier modification and impacts on tourism



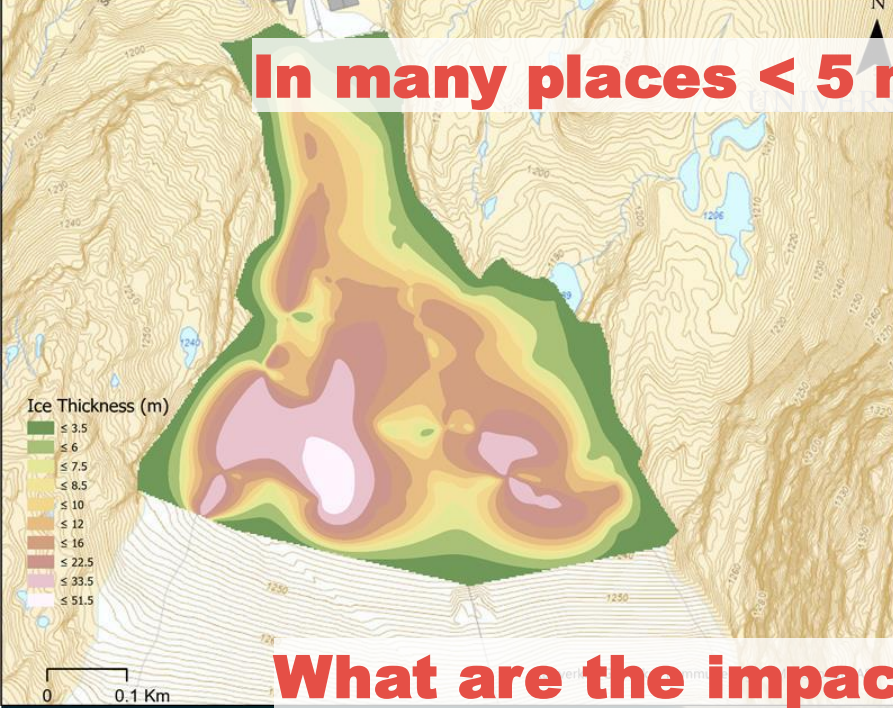




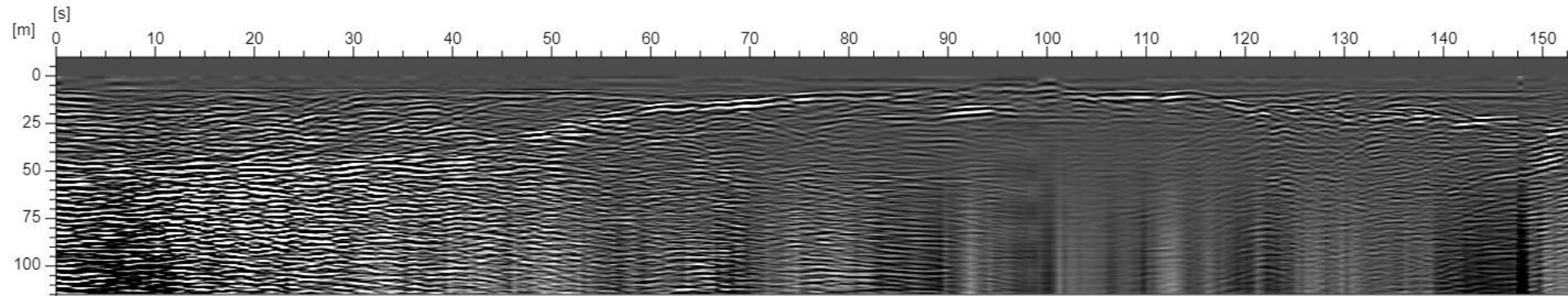




In many places < 5 m of ice on the ski area



What are the impacts of glacier modification?



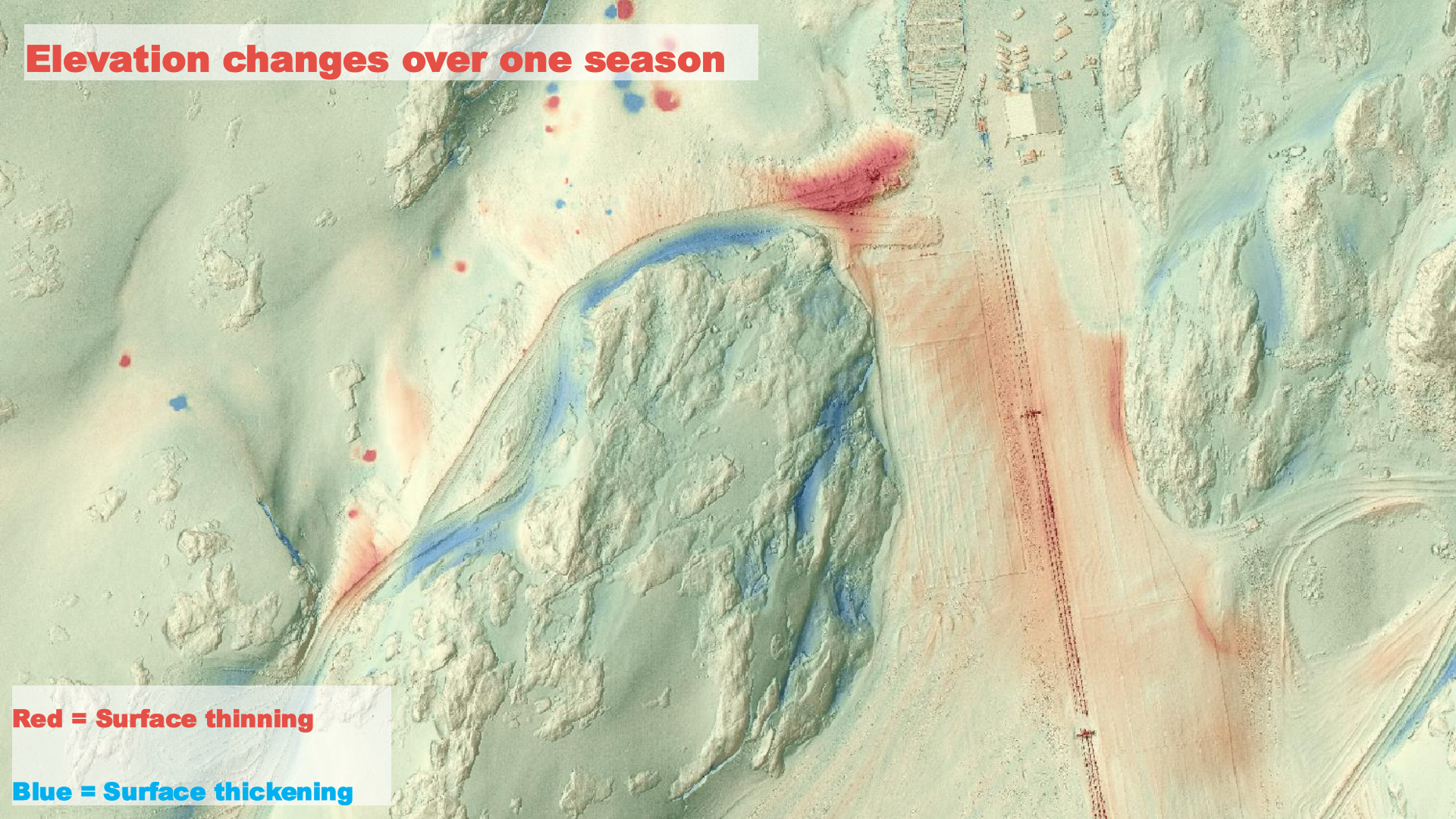


Repeat seasonal measurements with Wingtra drone





Elevation changes over one season



Red = Surface thinning

Blue = Surface thickening

Thank you for your attention!

