

# Towards a Deep-Learning based Inventory of Retrogressive Thaw Slumps across the Arctic



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# Retrogressive Thaw Slumps





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## Significance

- PF disturbance
- Influence on BG Cycles
- Geohazard
- Downstream Ecology
- Indicator of Change





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## Key attributes

- Size & Shape
- Stratigraphy
- Vegetation
- Dynamics
- Definition not clear



# Deep Learning for mapping RTS

## New Data + Techniques available !

### **Global Data:**

- PlanetScope (3m)  
VHR constellations
- Maxar, Skysat, Pleiades
- DEM
- Arctic DEM

### **Methods:**

- Deep Learning
- Machine-learning
- Computer Vision
- Image Augmentation
- ...

### **Computational resources:**

- GPU Processing (Deep-learning)
- High-performance computing
- Cloud Computing
- ...



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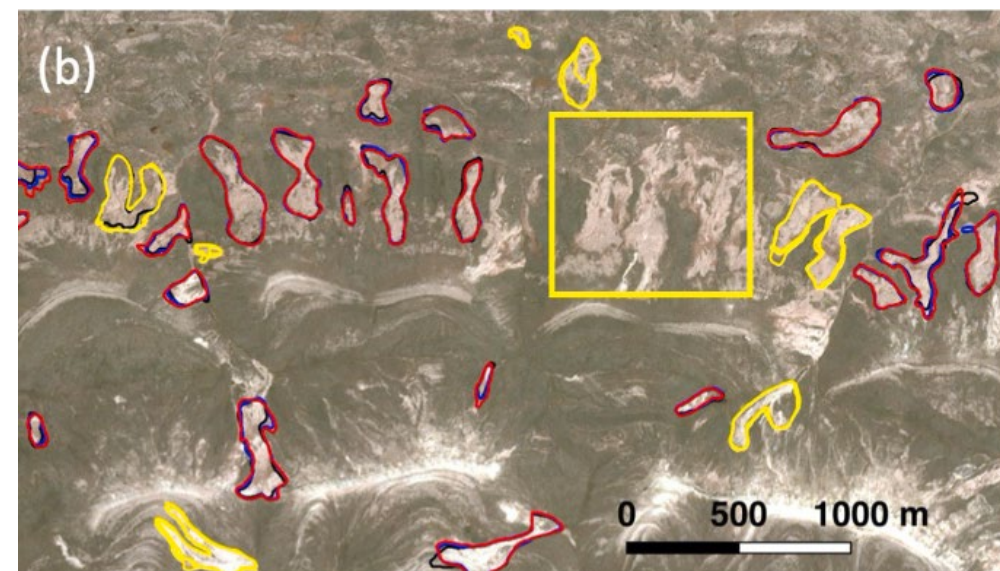
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Only few permafrost DL studies  
-> even fewer for Thaw Slumps



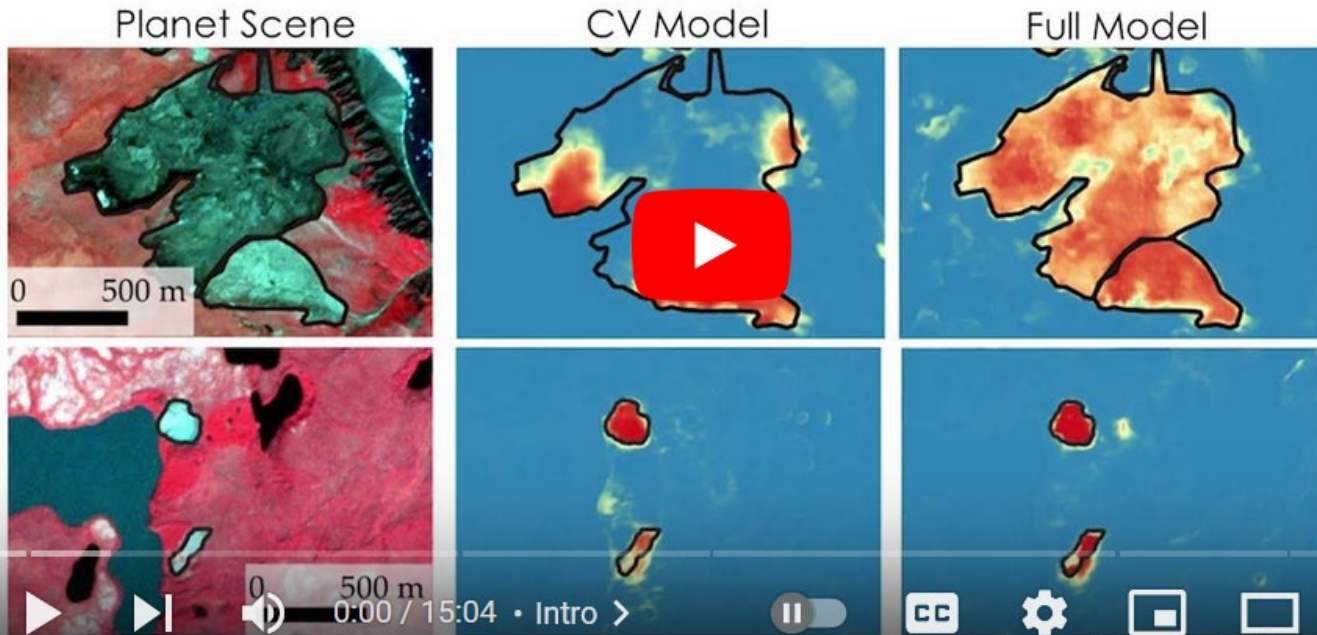
Huang et al., 2021

AGU21 Nitze et al.

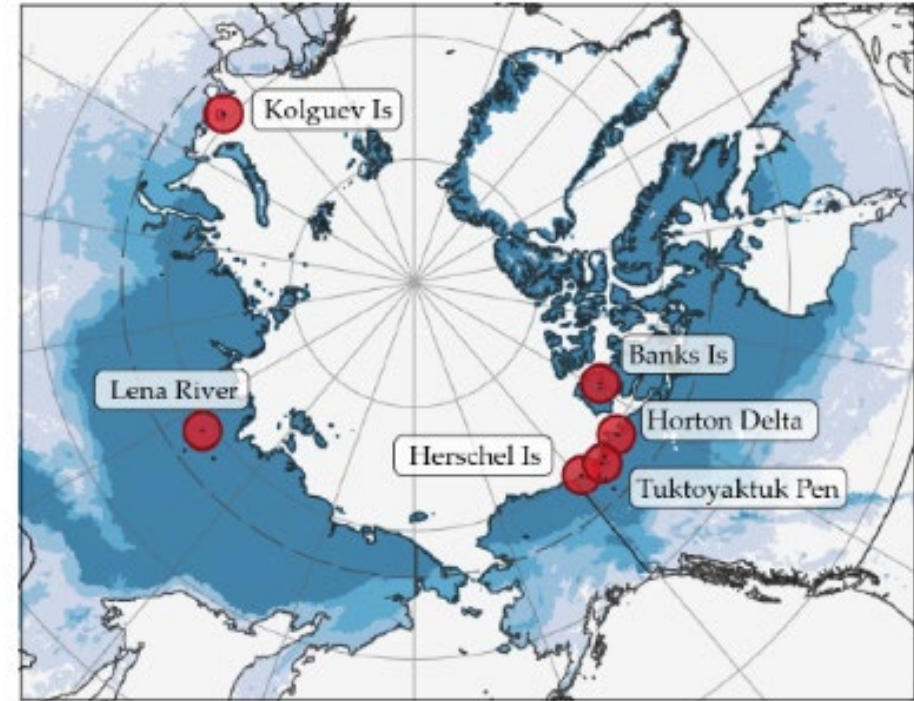
<https://youtu.be/Cuva8OebyrA>



## Detailed Examples



Adapted after Nitze et al. 2021



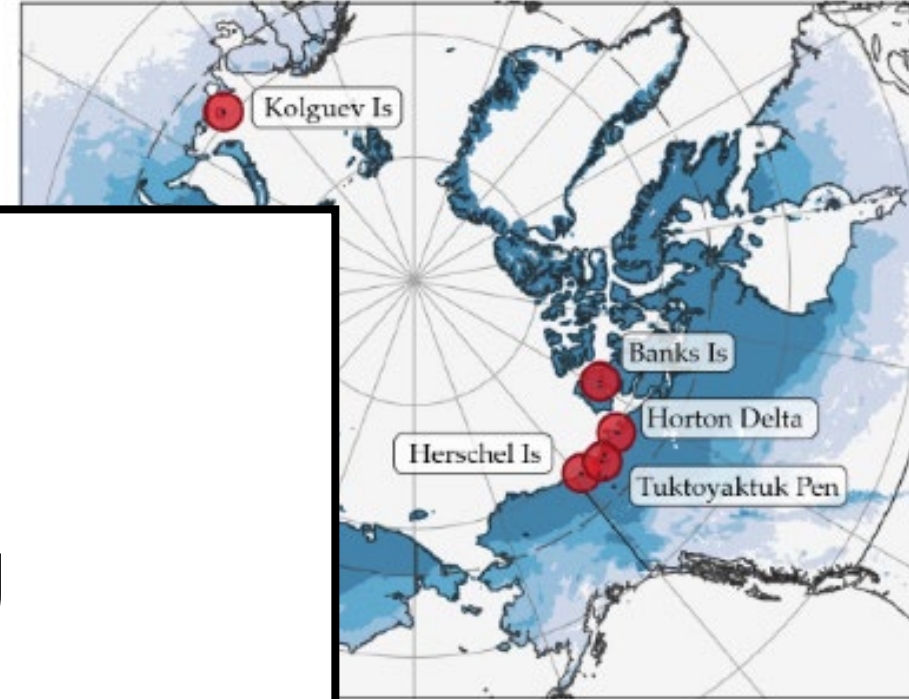
Adapted after Nitze et al., 2021  
(Remote Sensing)

6 Sites  
Setup of Methodology  
Performance tests



AGU21 Nitze et al.

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Nitze et al., 2021

**Next Goals!**

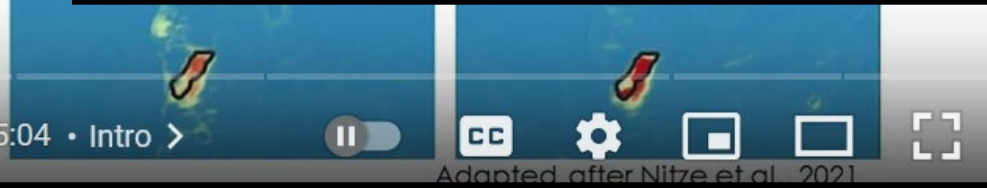
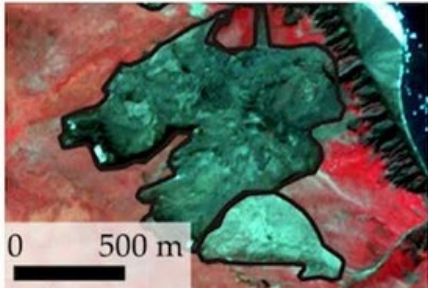
Spatial Scaling

Temporal Monitoring

Volumetric Analysis

Detailed Exam

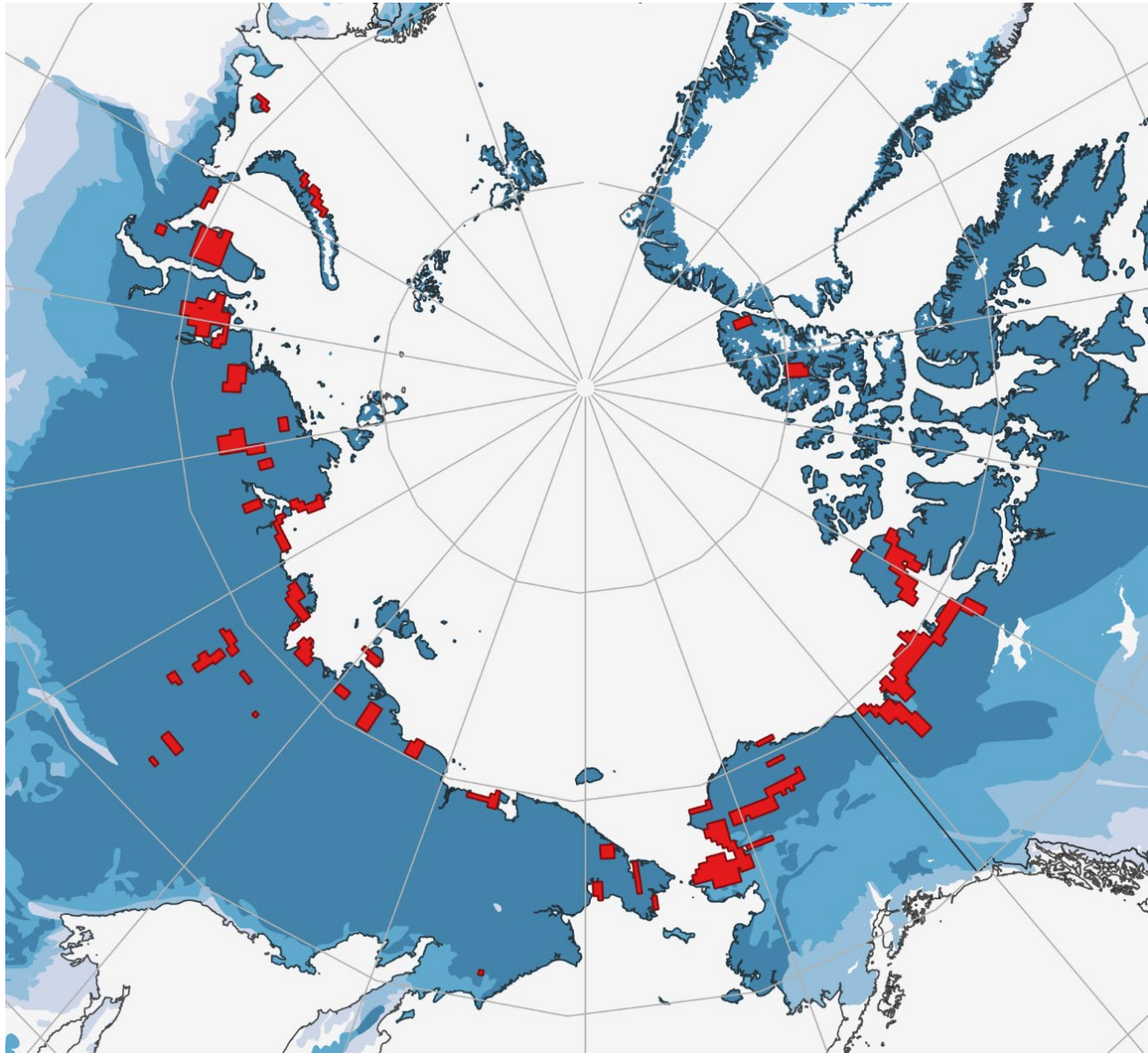
Planet Scene



Adapted after Nitze et al. 2021

Setup of Methodology  
Performance tests

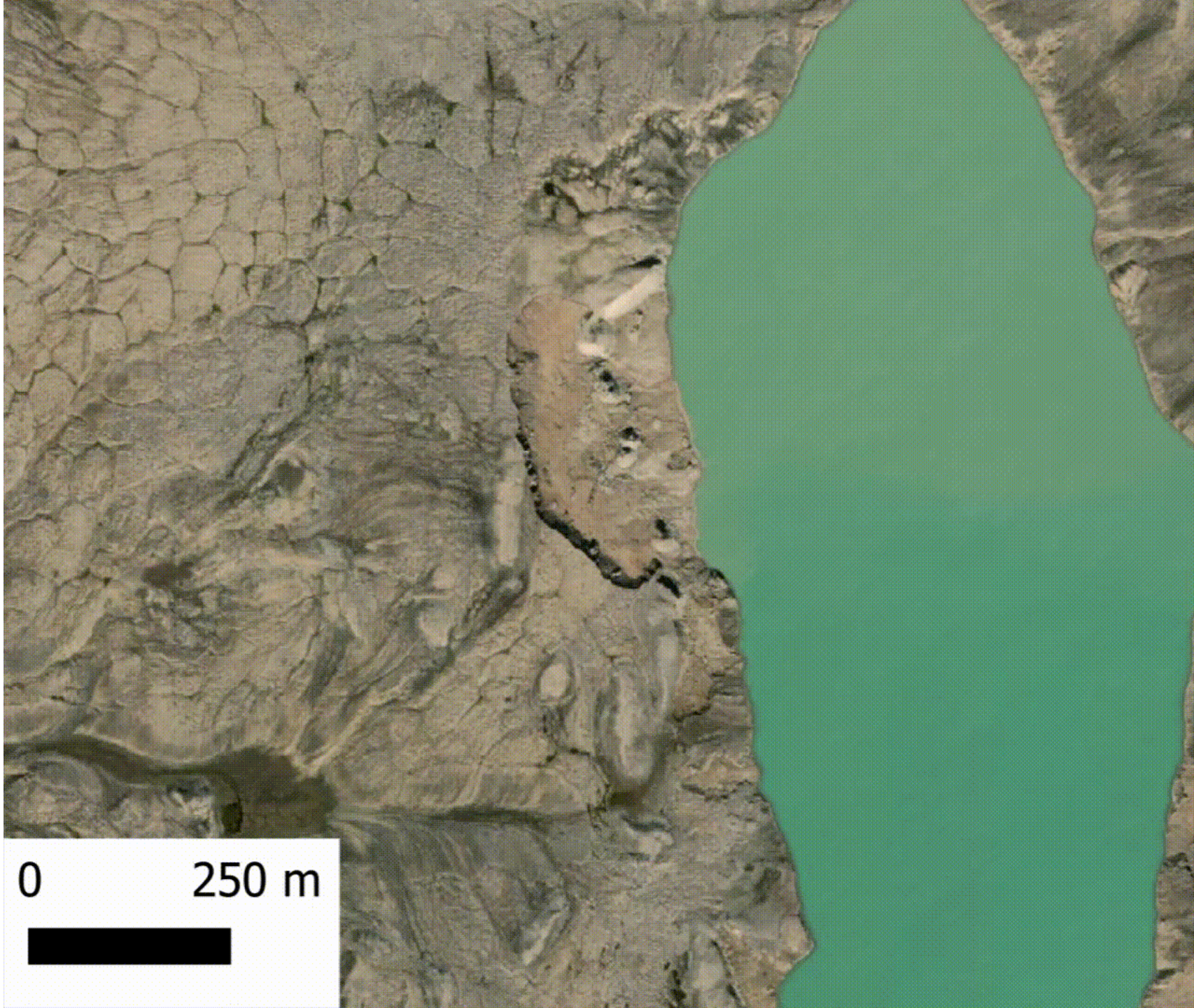




Current data footprint (Dec 2022)

- Total data area 1 Mkm<sup>2</sup> of PlanetScope (limited Quota)
- ~3000 Planet images (Scenes + Orthotiles)
- 849 unique tiles (118 multi-temporal)
  - 2018/19 - 2022
- 4336 training samples (hand digitized)
  - iterative approach
  - digitize - train - analyze - repeat





0

250 m



**Location:**  
Banks Island

**Background layer:**  
ESRI Satellite





**Location:**  
Peel Plateau (NW Canada)

**Background layer:**  
ESRI Satellite



**Location:**  
Novaya Zemlya (NW Russia)

**Background layer:**  
ESRI Satellite

Good performance in well known regions

Temporal evolution detected

Well defined headwall



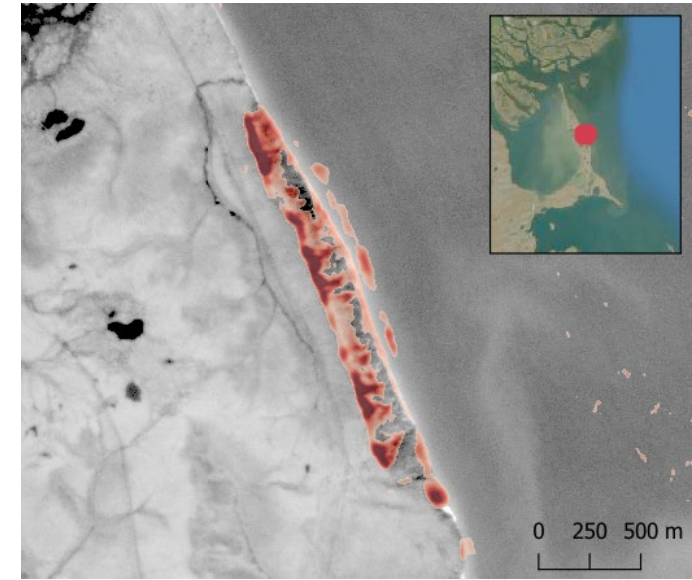
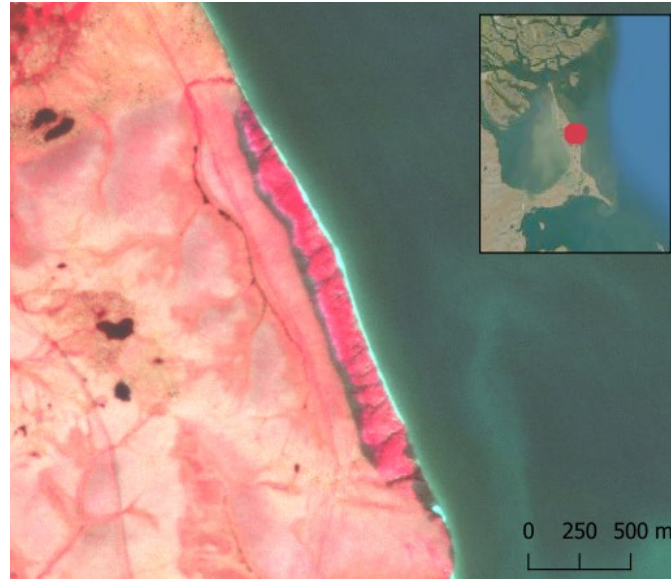
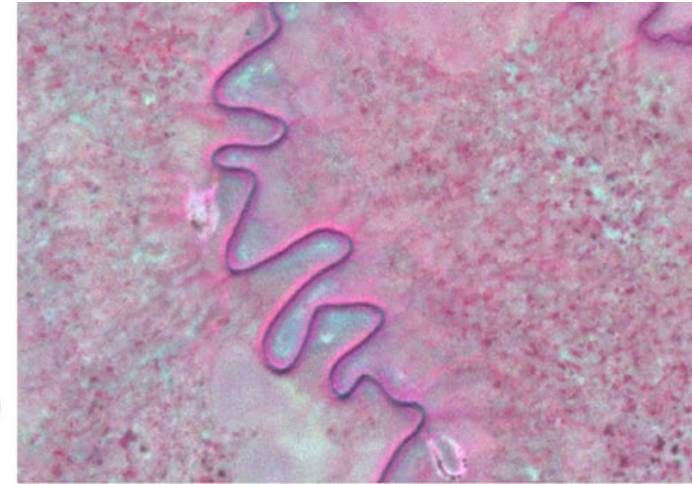
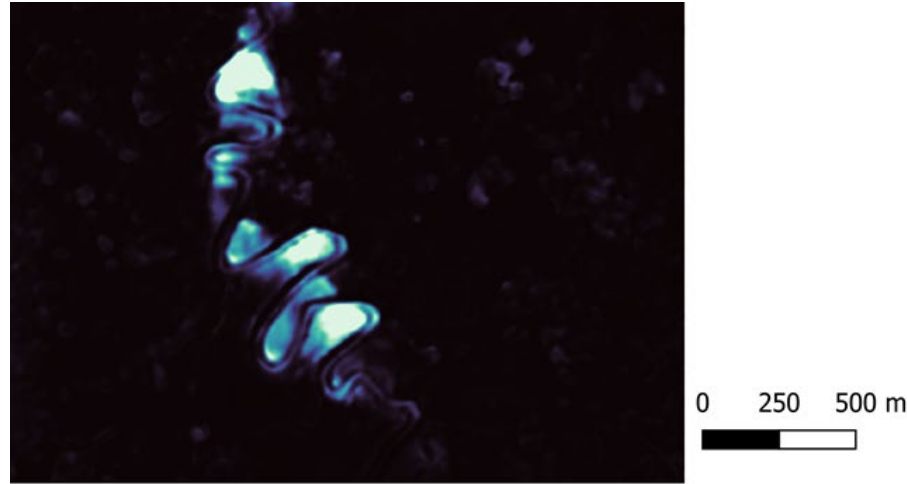
# Scaling to the Arctic?

## Positives

- Good performance in trained areas
- Fast and established processing pipeline

## Challenges:

- Many False positives
  - Spatial diversity
- Semantics: What is an RTS?
- Data access (commercial data)

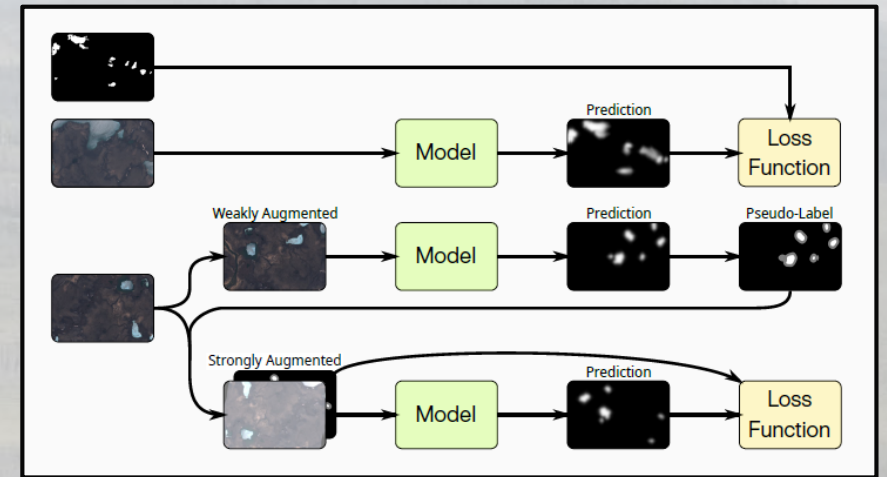




# Outlook

## Self supervised learning

- Less training data required





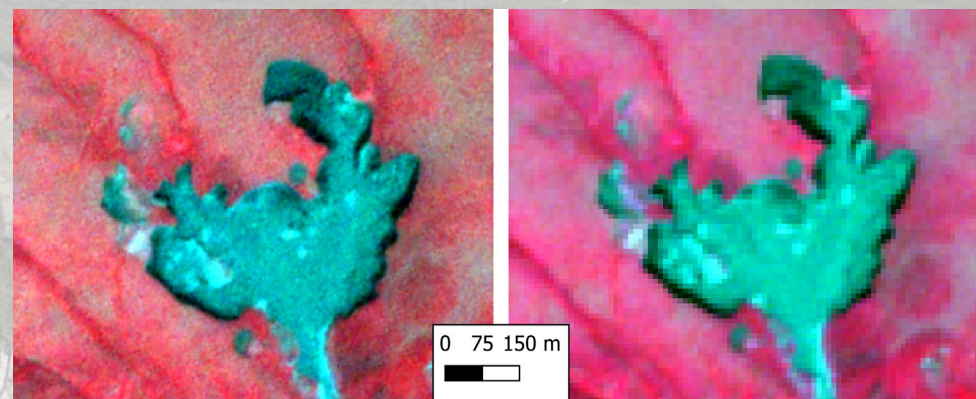
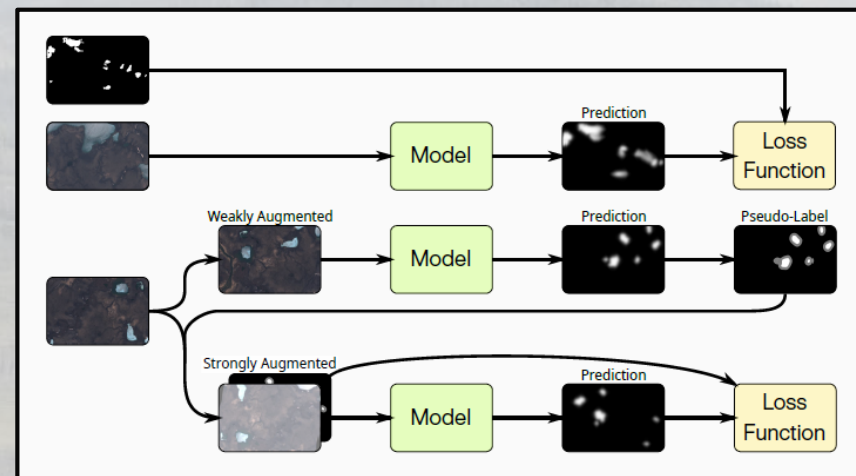
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## Additional Data Sources

- Sentinel-2 → free data
- Panarctic extent
- Monitoring





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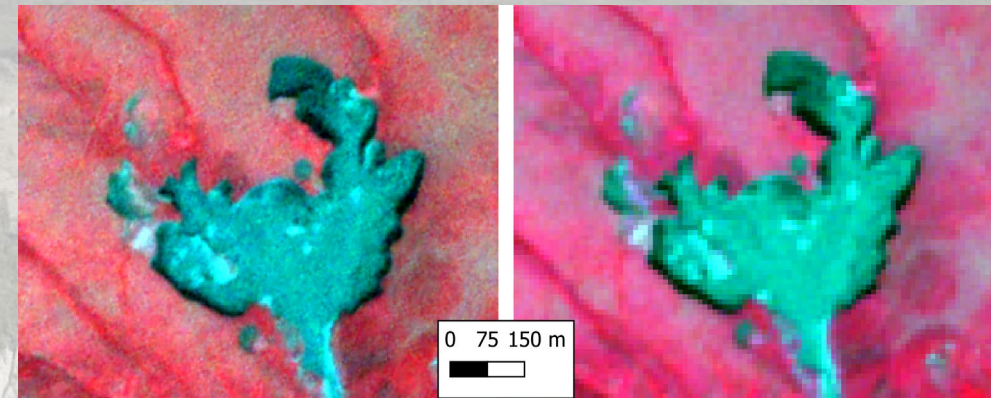
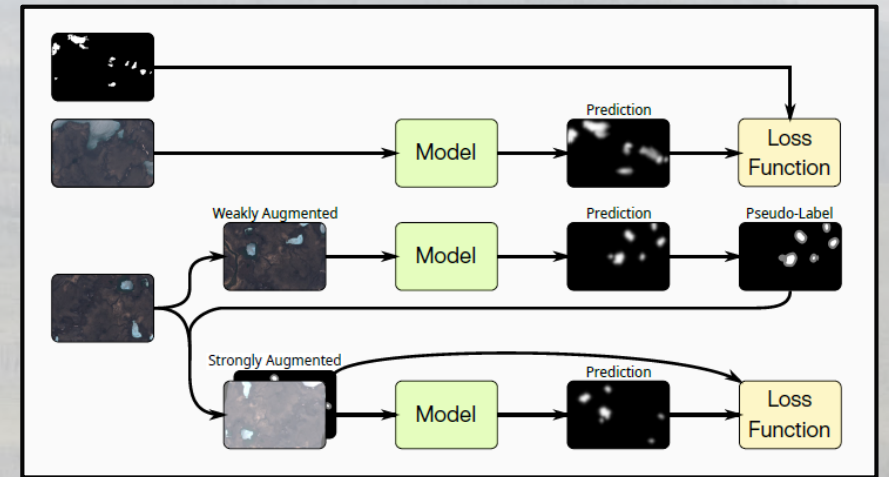
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## Additional Data Sources

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## Volumetric Changes

- RTS are 3D features !!!





# Thank You

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**Twitter:** @i\_nitze **Mastodon:** @i\_nitze@mapstodon.space **Github:** <https://github.com/initze/>

**Code:** <https://github.com/initze/thaw-slump-segmentation>

**Training Labels:** [https://github.com/initze/ML\\_training\\_labels](https://github.com/initze/ML_training_labels)

## References:

Nitze, I., Heidler, K., Barth, S., & Grosse, G. (2021). Developing and Testing a Deep Learning Approach for Mapping Retrogressive Thaw Slumps. Remote Sensing, 13(21). <https://doi.org/10.3390/rs13214294>

Huang, L., Liu, L., Luo, J., Lin, Z., & Niu, F. (2021). Automatically quantifying evolution of retrogressive thaw slumps in Beiluhe (Tibetan Plateau) from multi-temporal CubeSat images. International Journal of Applied Earth Observation and Geoinformation, 102, 102399. <https://doi.org/10.1016/j.jag.2021.102399>



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**ETH zürich**

