# Use of Convolution Neural Networks and Object Based Image Analysis for Automated Rock Glacier Mapping

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#### The problem

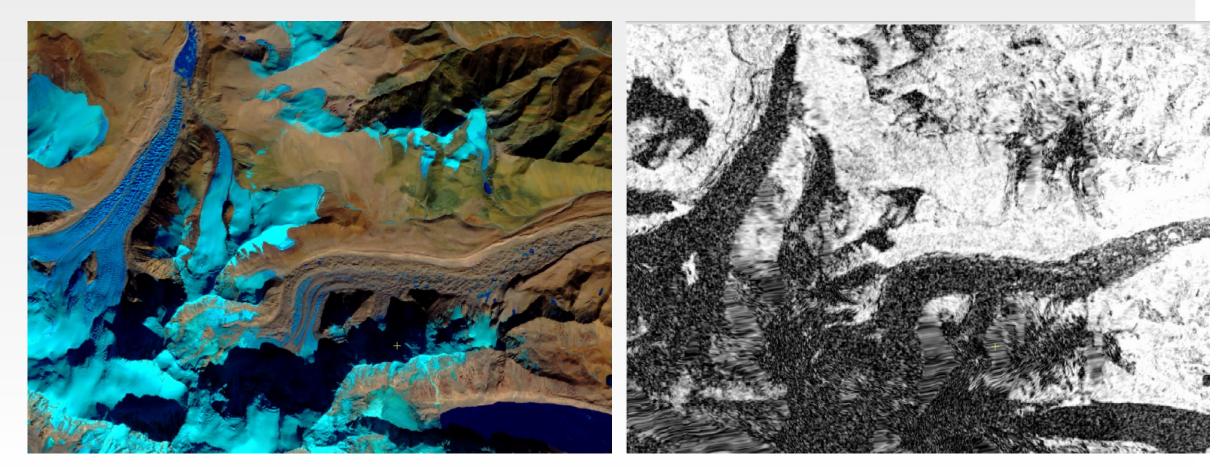
- Rock glaciers can contain significant amounts of ice
- Yet are spectrally inseparable from the surrounding terrain
- As such, most rock glacier inventories are based on time consuming and subjective manual interpretation





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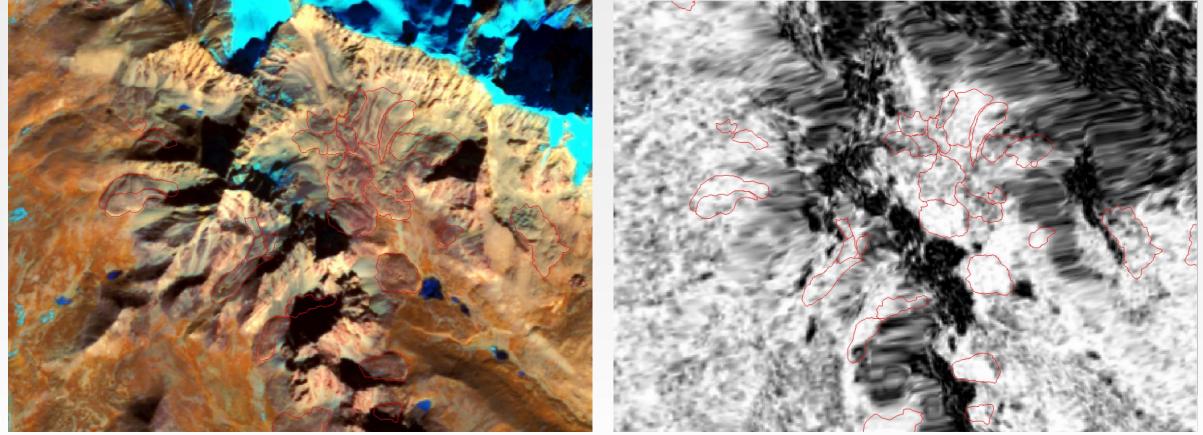
# Clean ice and debris-covered ice can be identified with multispectral and SAR coherence data respectively



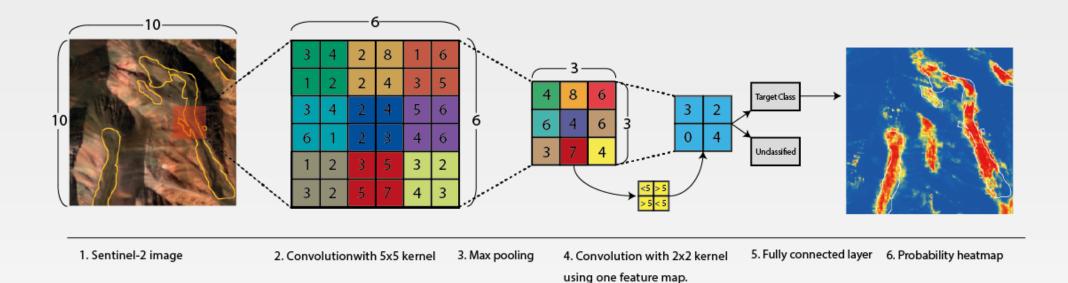
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## The same does not work for rock glaciers





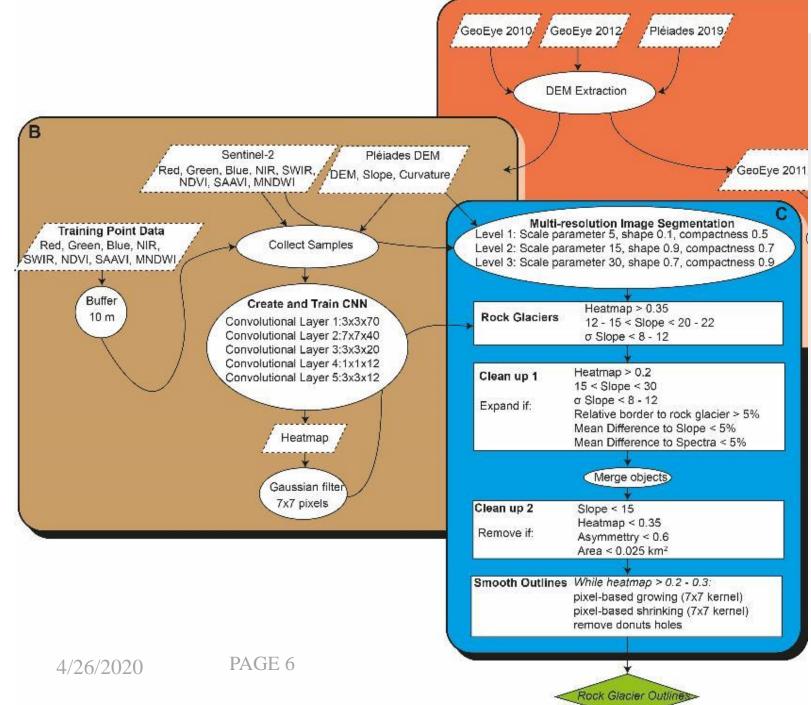


### **Convolutional Neural Networks (CNN)**

A simplified example of a Convolutional Neural Network workflow. A convolution based on a 5x5 moving window is applied to an input image, resulting in the first convolutional layer. Note that the process has been simplified and most CNNs would involve additional convolution and pooling layers, feature maps, and additional input bands.

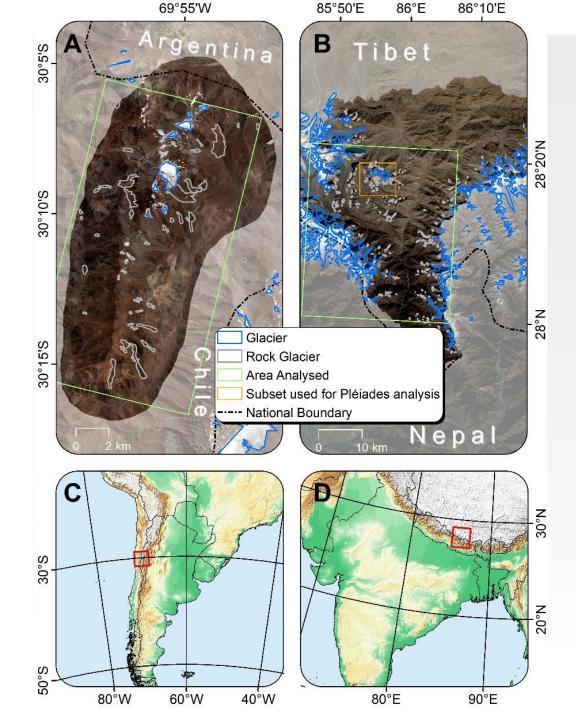


5



# Combining CNNs with Object OBIA

- Split rock glacier inventory into 30% validation and 70% training
- Also used RGI for glacier outlines
- Train CNN to generate
  probability heatmap
- Reshape and refine in OBIA



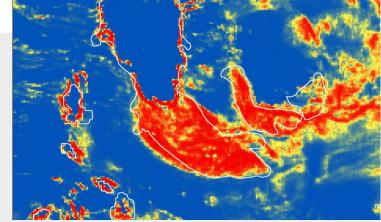
## Trialed in two distinct

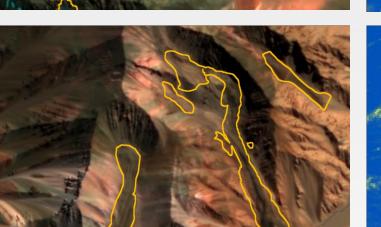
### periglacial catchments

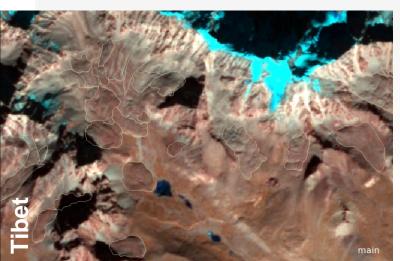
- La Laguna Catchment, Chile
- Poiqu Catchment, Tibet

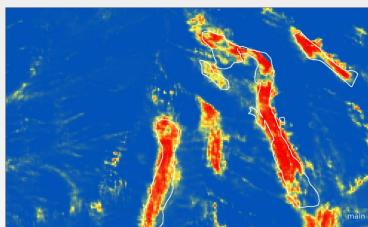
• <u>Data</u>: Sentinel 2, Sentinel 1 coherence, DEM

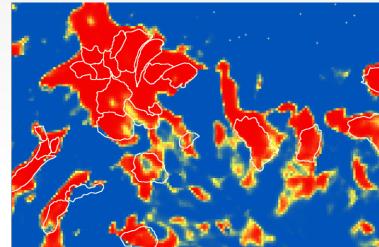










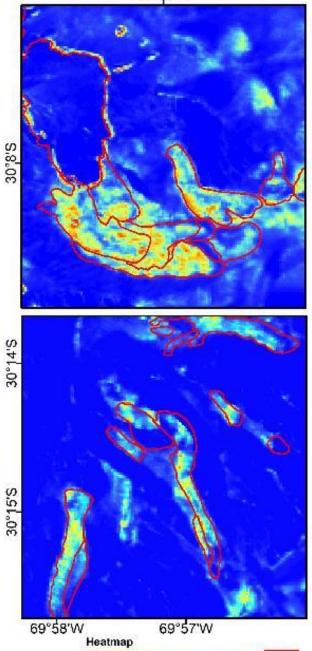


Results: Generation of probability map

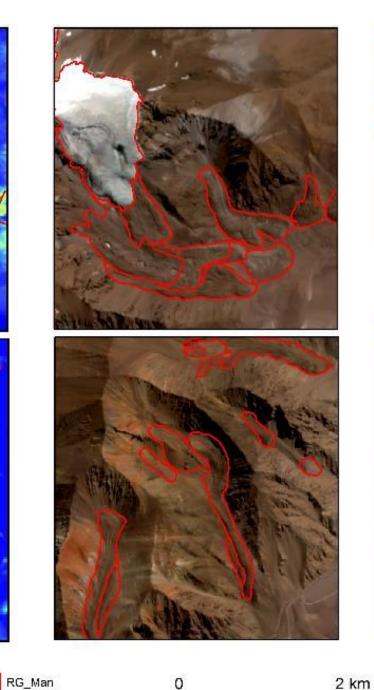
Higher pixel values (shown in red) indicate a higher probability a pixel is a rock glacier

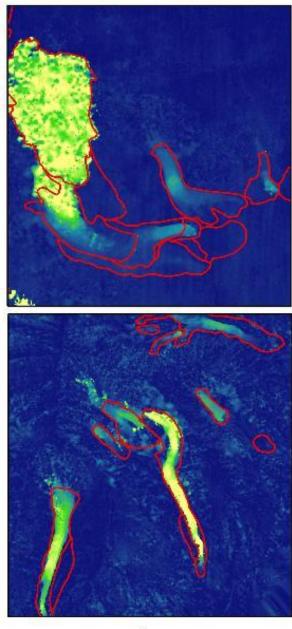


69°55'W



1 0.8 0.6 0.4 0.2 0





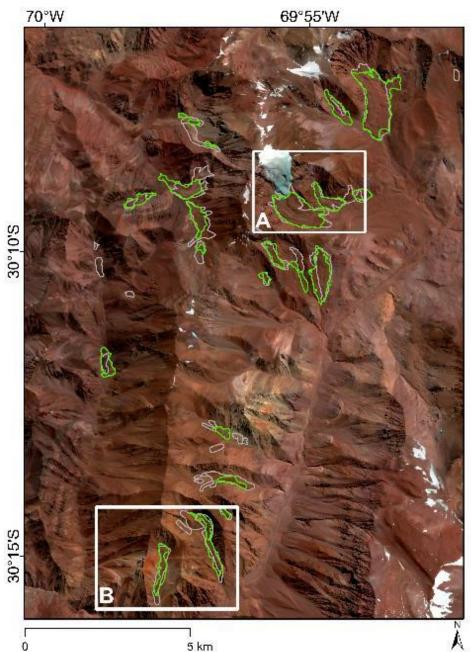


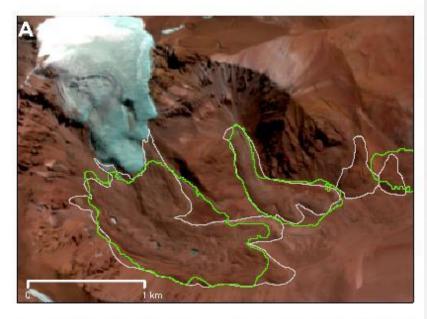
outlines with optical image based surface velocities

of

Velocity (m a 1)							
1	0.8	0.6	0.4	0.2	0		

#### **Comparison with manual rock glacier inventory – La Laguna**





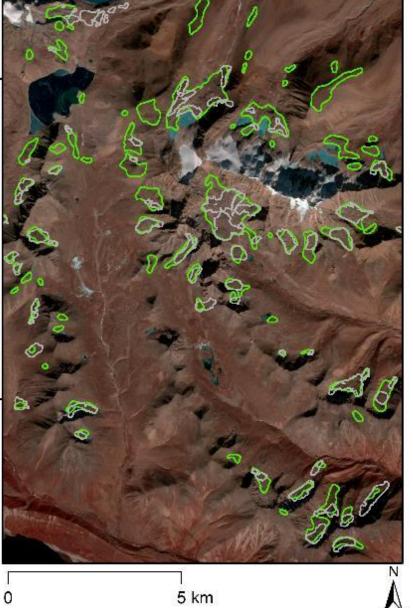


RG\_Man

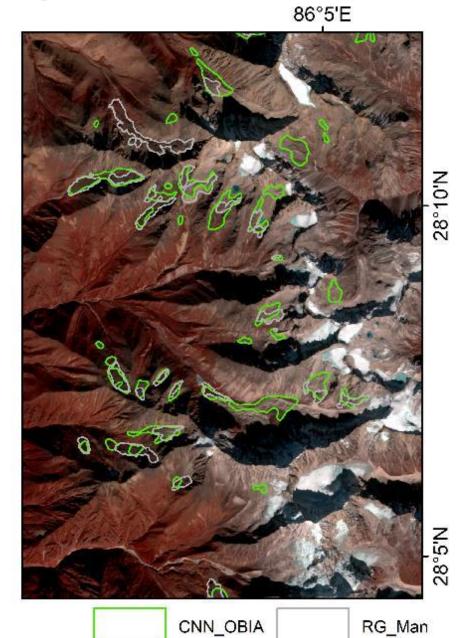
#### **Comparison with manual rock glacier inventory - Poiqu**

28°15'N

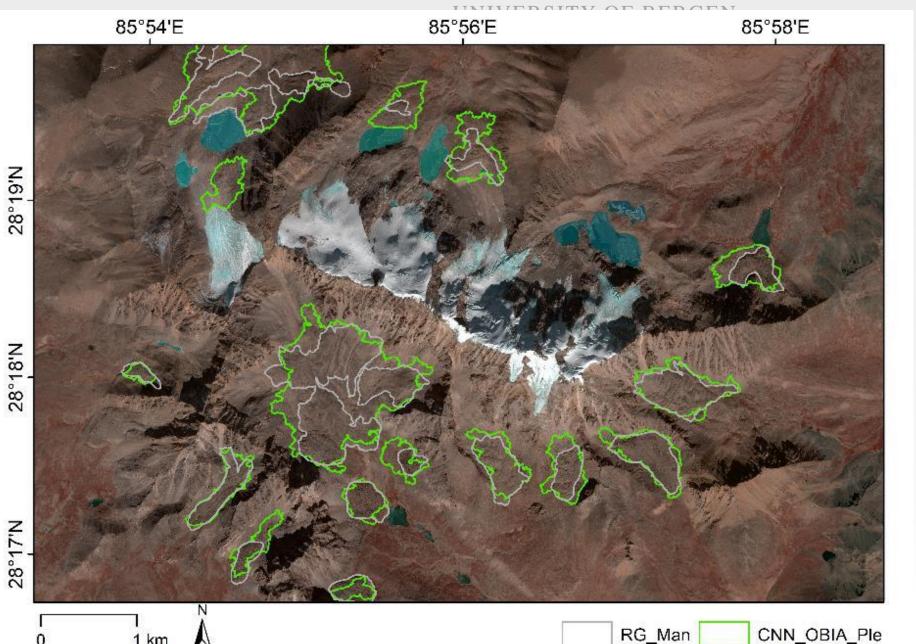
28°20'N



85°55'E



#### **Comparison with manual rock glacier inventory – Poiqu**



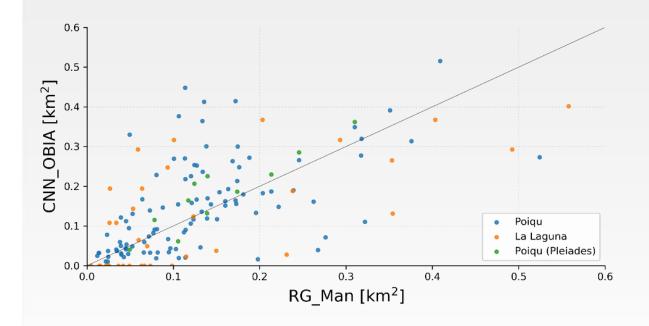
1 km

**Analysis** repeated on **Pleiades** imagery

Same accuracy, but smaller rock glacier mapped



# **Results - accuracy**



Classification	User	Producer	Total	Карра	
	Accuracy	Accuracy	Accuracy	coefficient	
	(%)	(%)	(%)		
La Laguna	63.9	75.4	97.1	0.67	
Poiqu	68.8	75.0	56.5	0.72	
Total	65.9	71.4	72.0	0.68	
(Sentinel)					
Poiqu	72.0	88.4	76.8	0.76	
(Pléiades)					



# Conclusion

- A combination of CNNs and OBIA shows promise for creating rock glacier inventories from remote sensing data
- Future developments needed before the method can be used automatically for rock glacier inventories...
- ...but our method reduces the amount of manual work needed

