

Development of a new forest snow mapping algorithm using MODIS data, machine learning and time-lapse photography

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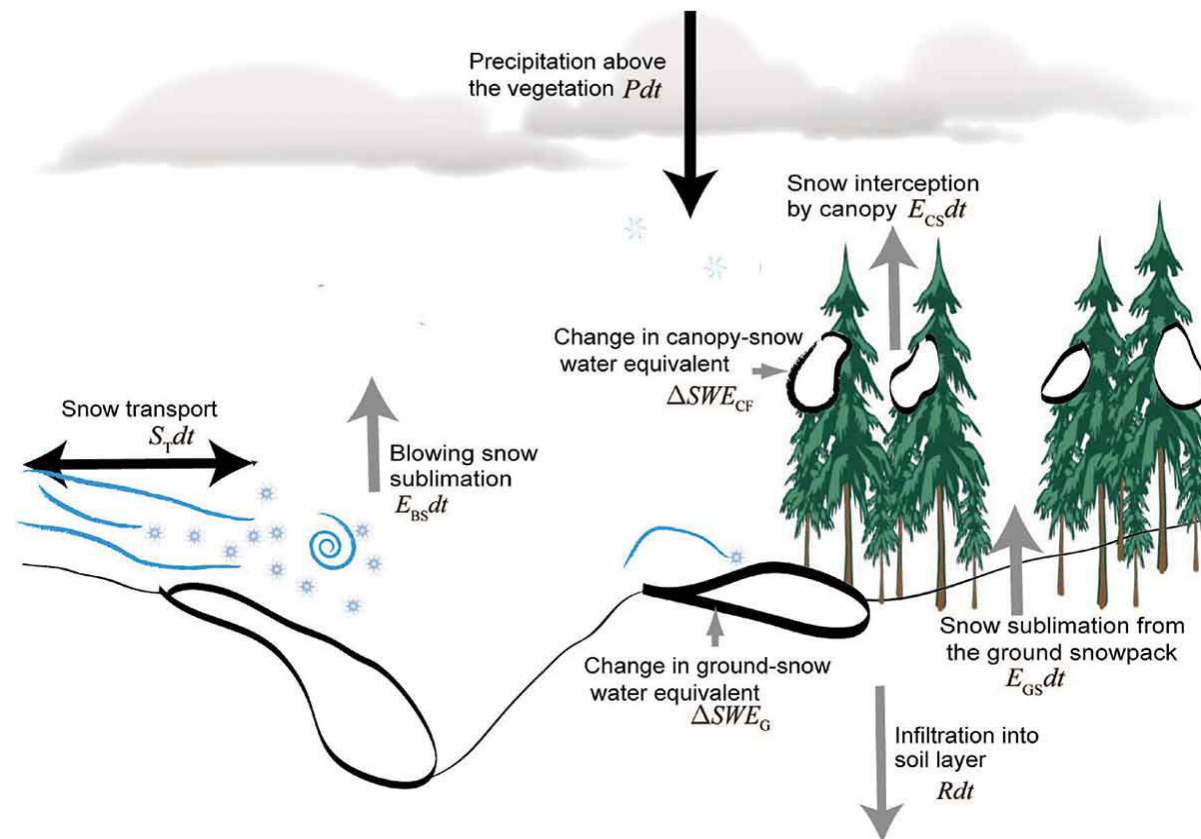
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Motivation

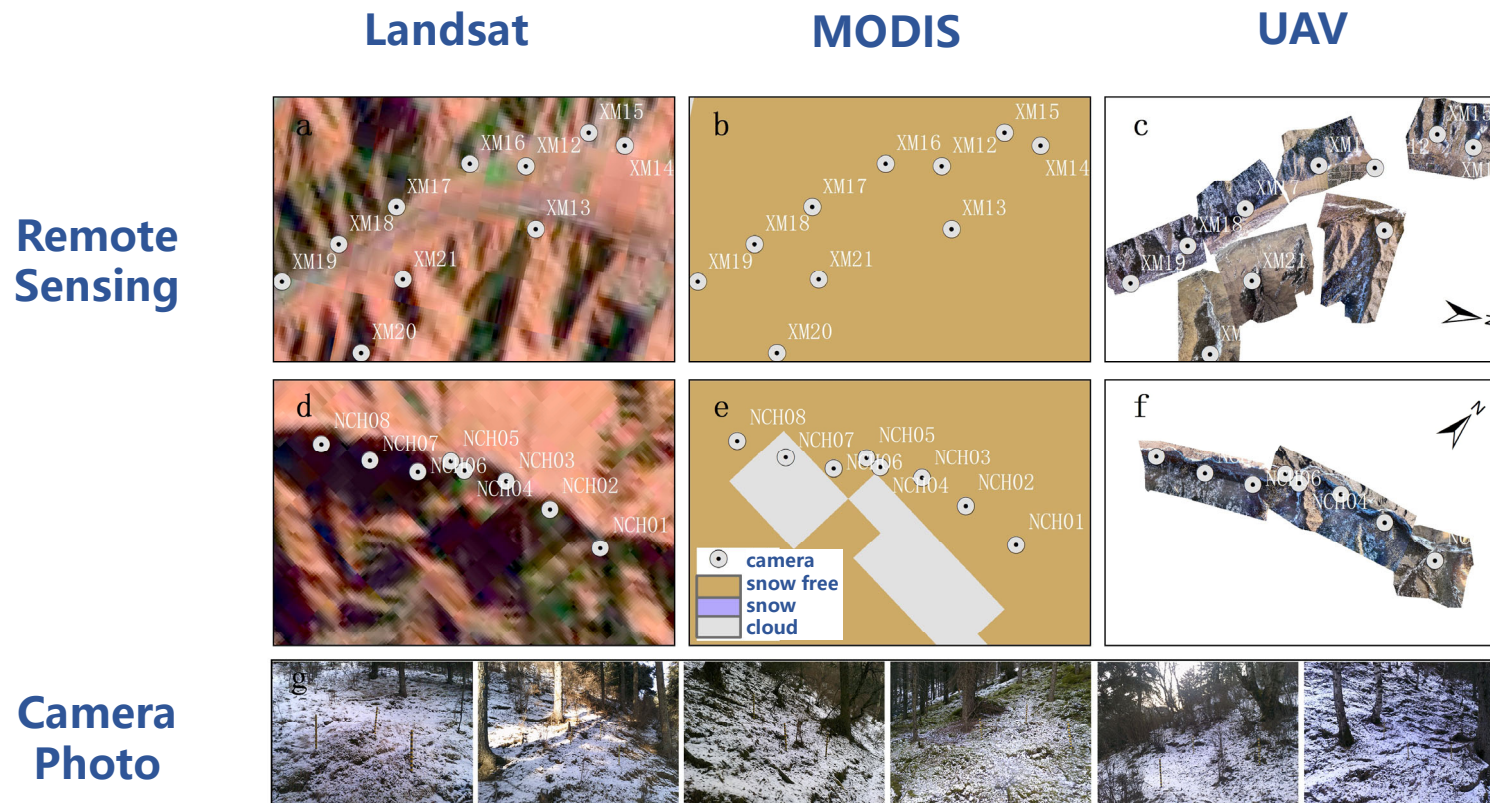
Complex forest snow processes



(Suzuki et al., 2011 HSJ)

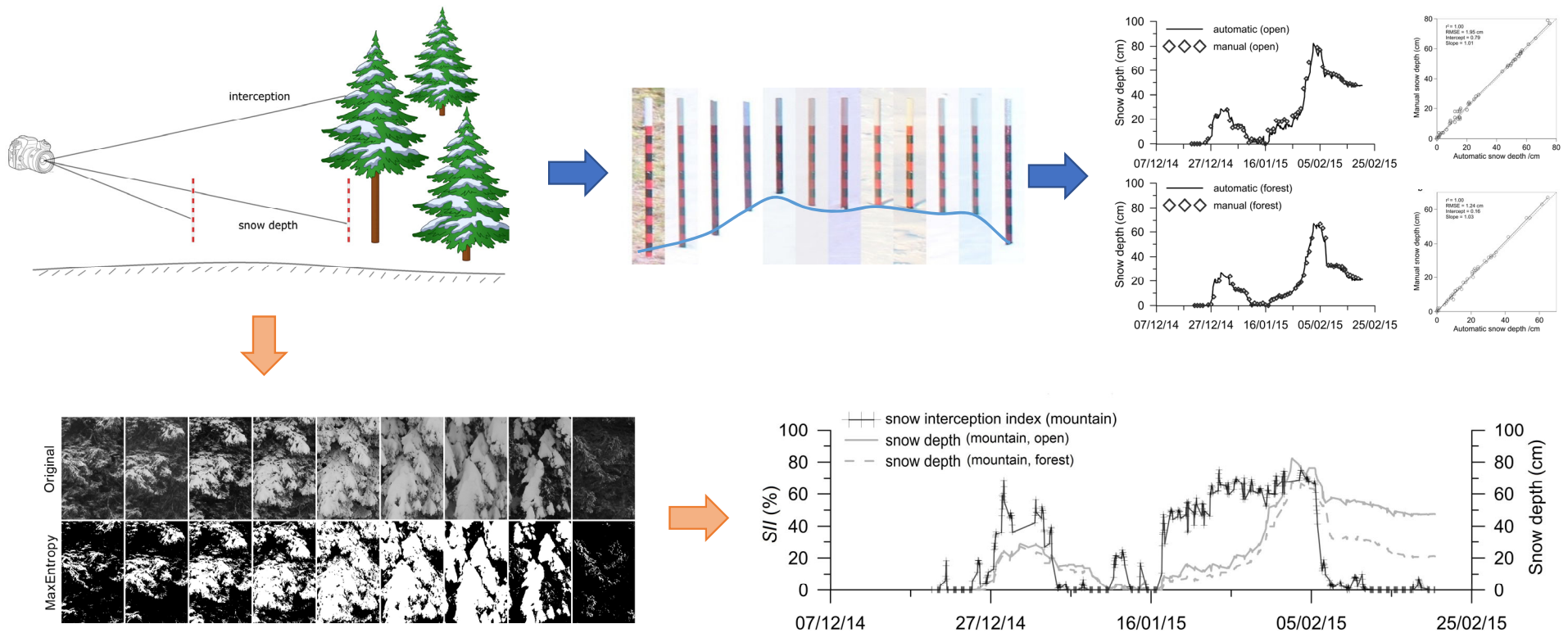
Ungauged Basins
Remote Sensing?

Issues of remote sensing in forest: **omission error**



Monitoring forest snow processes using time-lapse camera

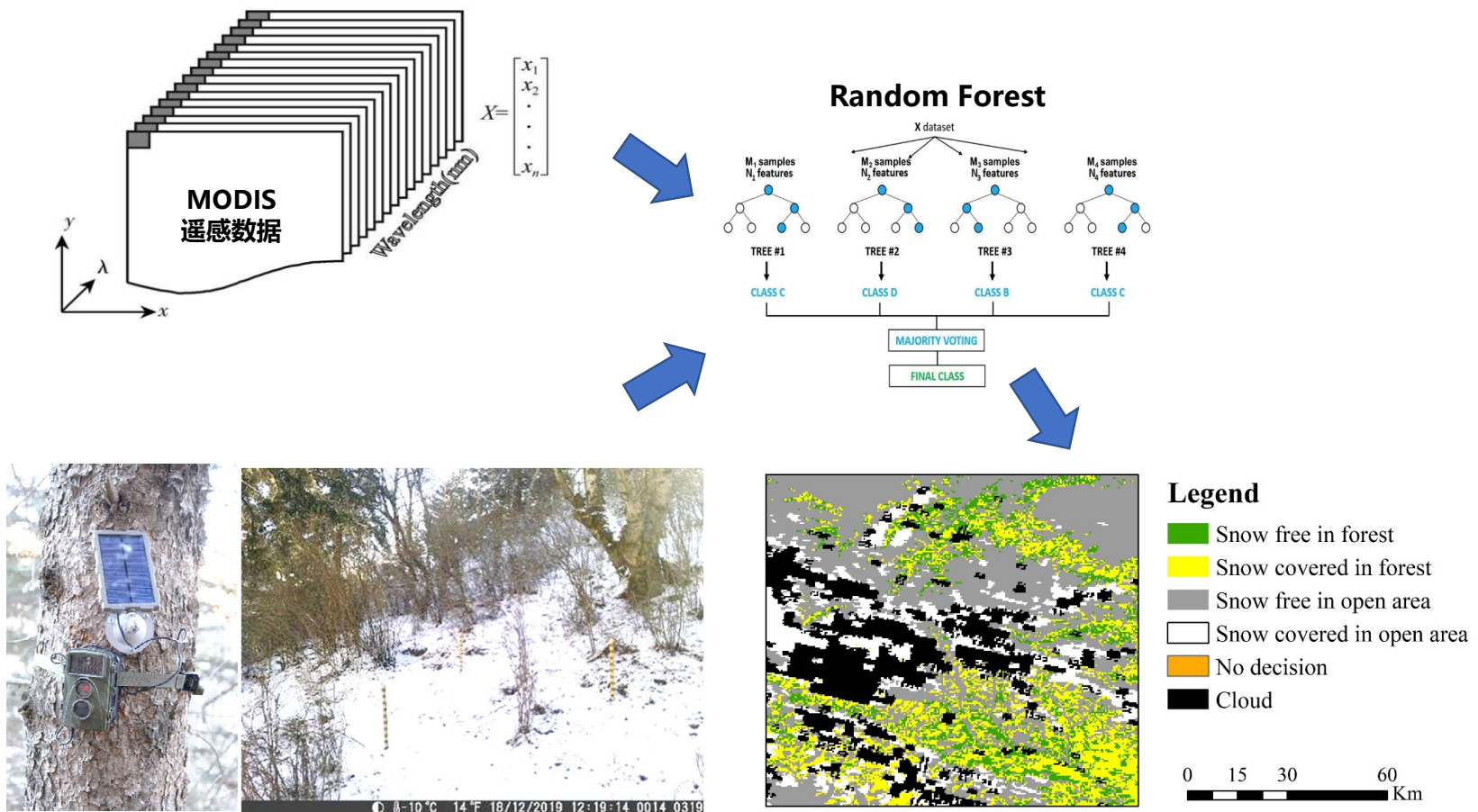
Snow depth in forests and open areas



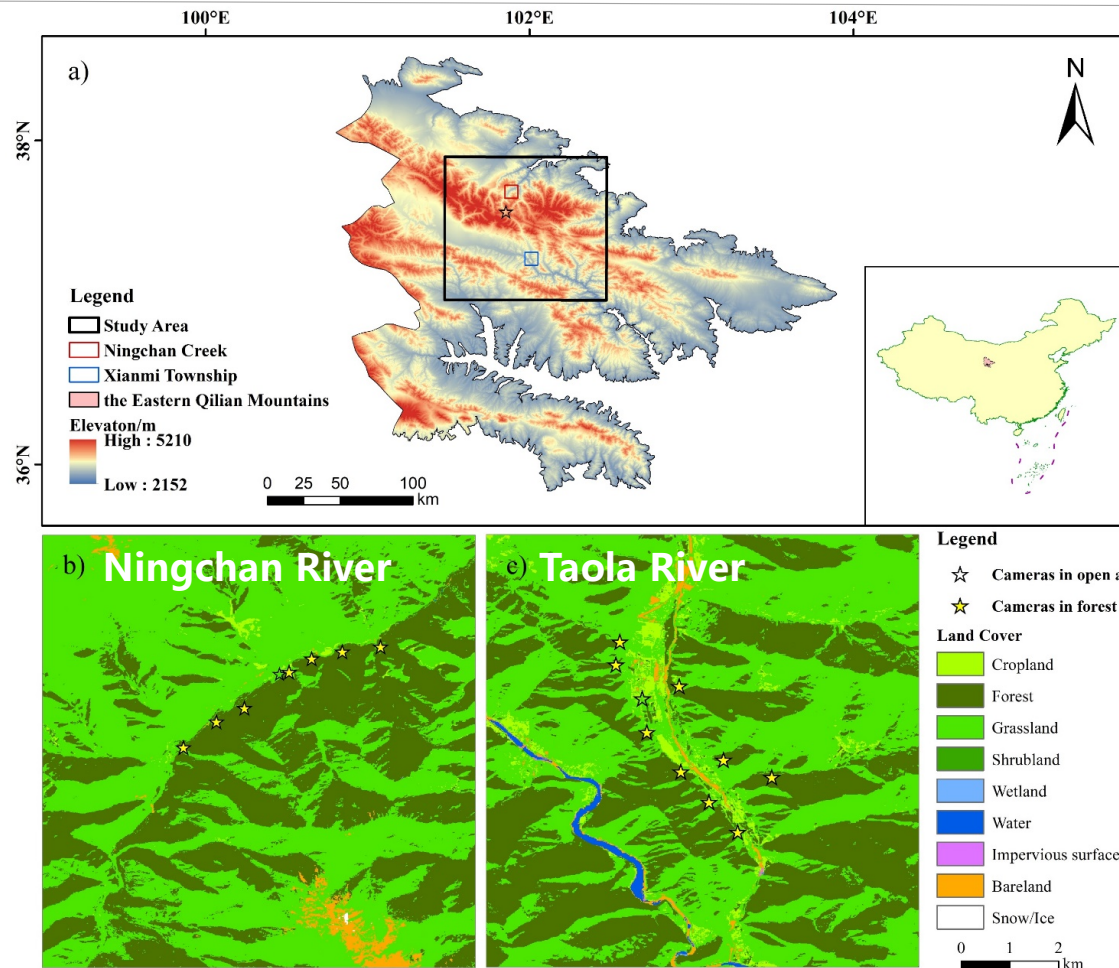
Canopy snow interception

(Dong & Menzel, 2017 HP)

Using time-lapse camera data to develop forest snow remote sensing algorithm



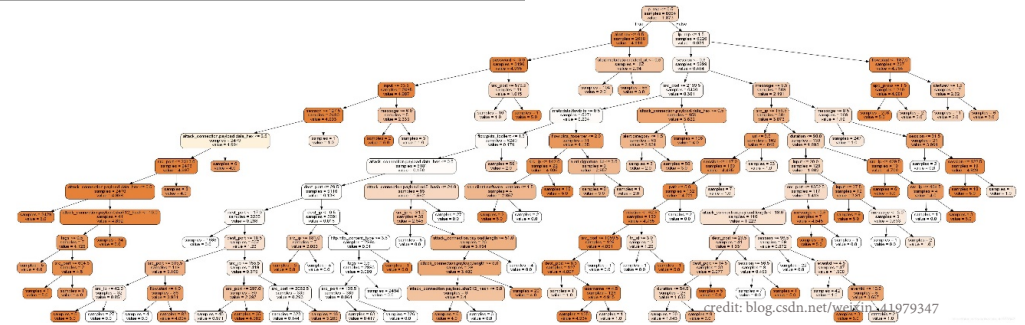
Study area: Qilian Mountain, NW China



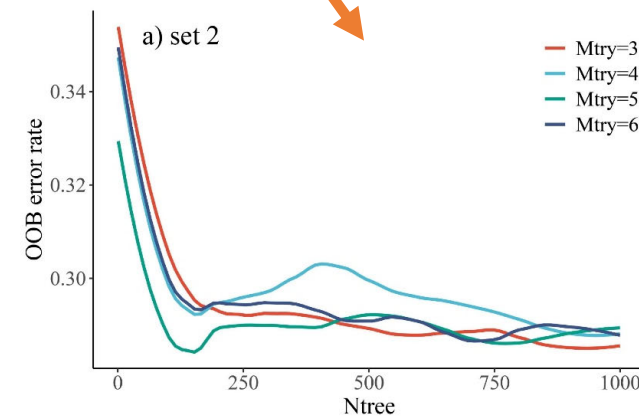
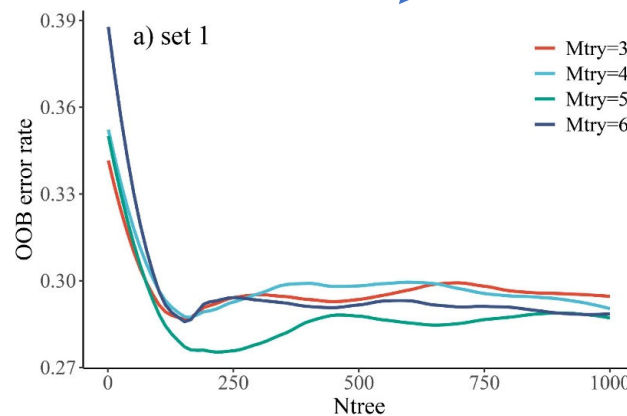
boreal coniferous forest

Method: Random Forest Machine Learning

Two sets of predictors



- Classification: Snow covered/Snow free (Ground truth: time-lapse camera data)
- Predictors: MODIS b1~b7 reflectance, forest cover fraction, slope, aspect

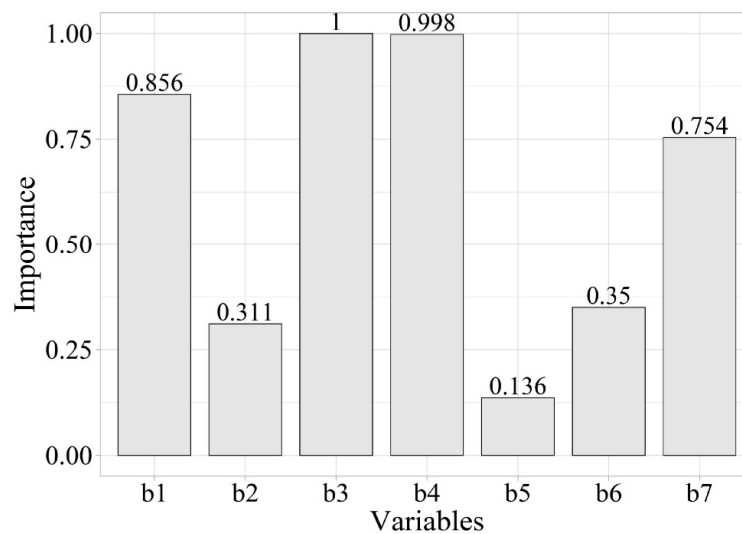


Relative importance of predictors

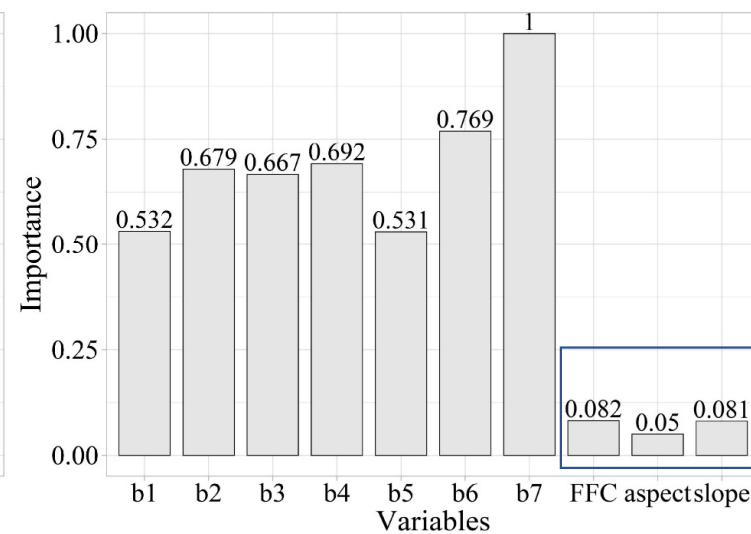
Set1: MODIS b1~b7 reflectance

Set2: MODIS b1~b7 reflectance, forest fraction, slope, aspect

a) PBVI for set 1



b) PBVI for set 2



Forest fraction, slope, and aspect contribute little to the model accuracy

Accuracy Validation

Evaluation Indices

Precision $PC = \frac{TP}{TP + FP}$

Recall $RC = \frac{TP}{TP + FN}$

Accuracy $AC = \frac{TP + TN}{Total}$

F-score $FS = 2 \times \frac{PC \times RC}{PC + RC} = \frac{2TP}{2TP + FP + FN}$

False alarm ratio $FAR = \frac{FP}{FP + TN}$

Kappa coefficient $CK = \frac{AC - Pr(e)}{1 - Pr(e)}$

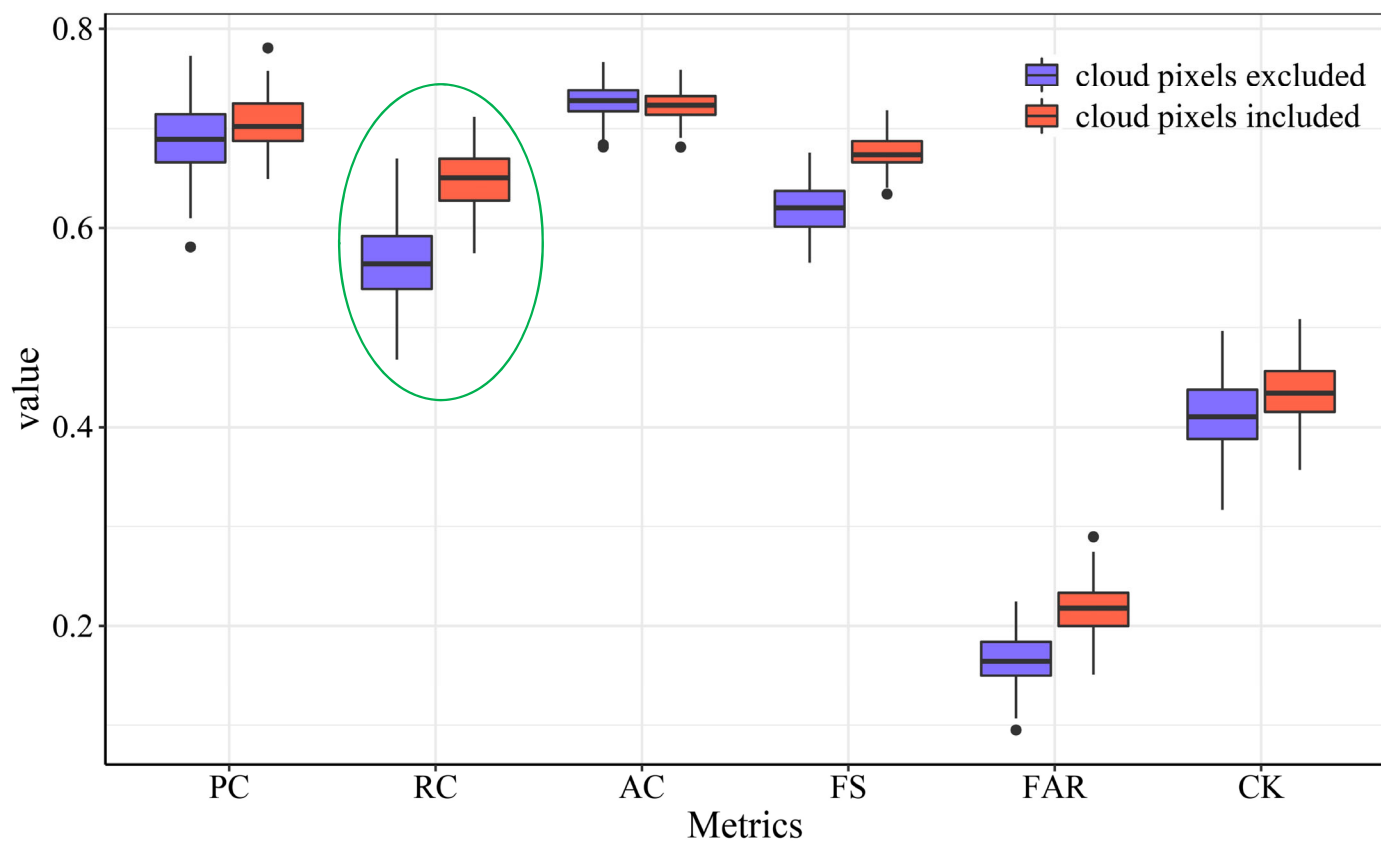
$$Pr(e) = \left(\frac{TP + FP}{Total} \times \frac{TP + FN}{Total} \right) + \left(\frac{TN + FP}{Total} \times \frac{TN + FN}{Total} \right)$$

$$Total = TP + FP + TP + FN$$

Confusion matrix

		Prediction	
		Snow	No snow
Observation	Snow	True positive (TP)	False negative (FN)
	No snow	False positive (FP)	True negative (TN)

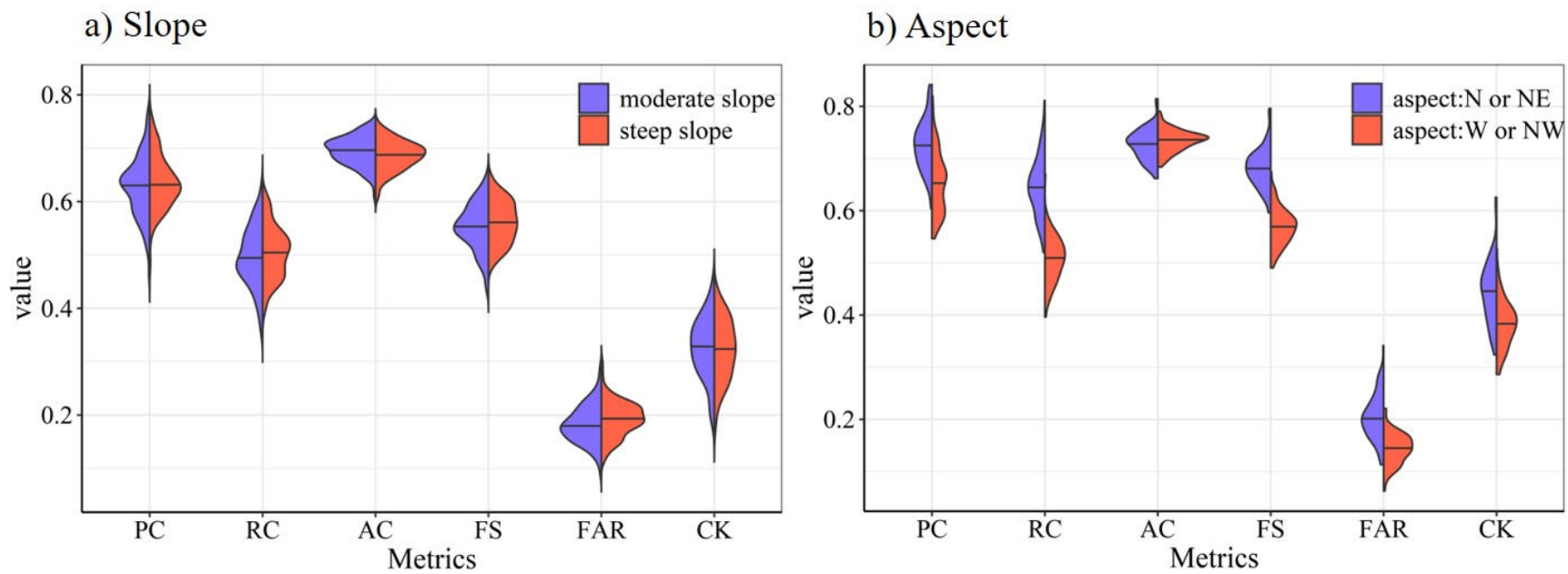
Effect of cloud pixels



Include cloud pixels: Recall 

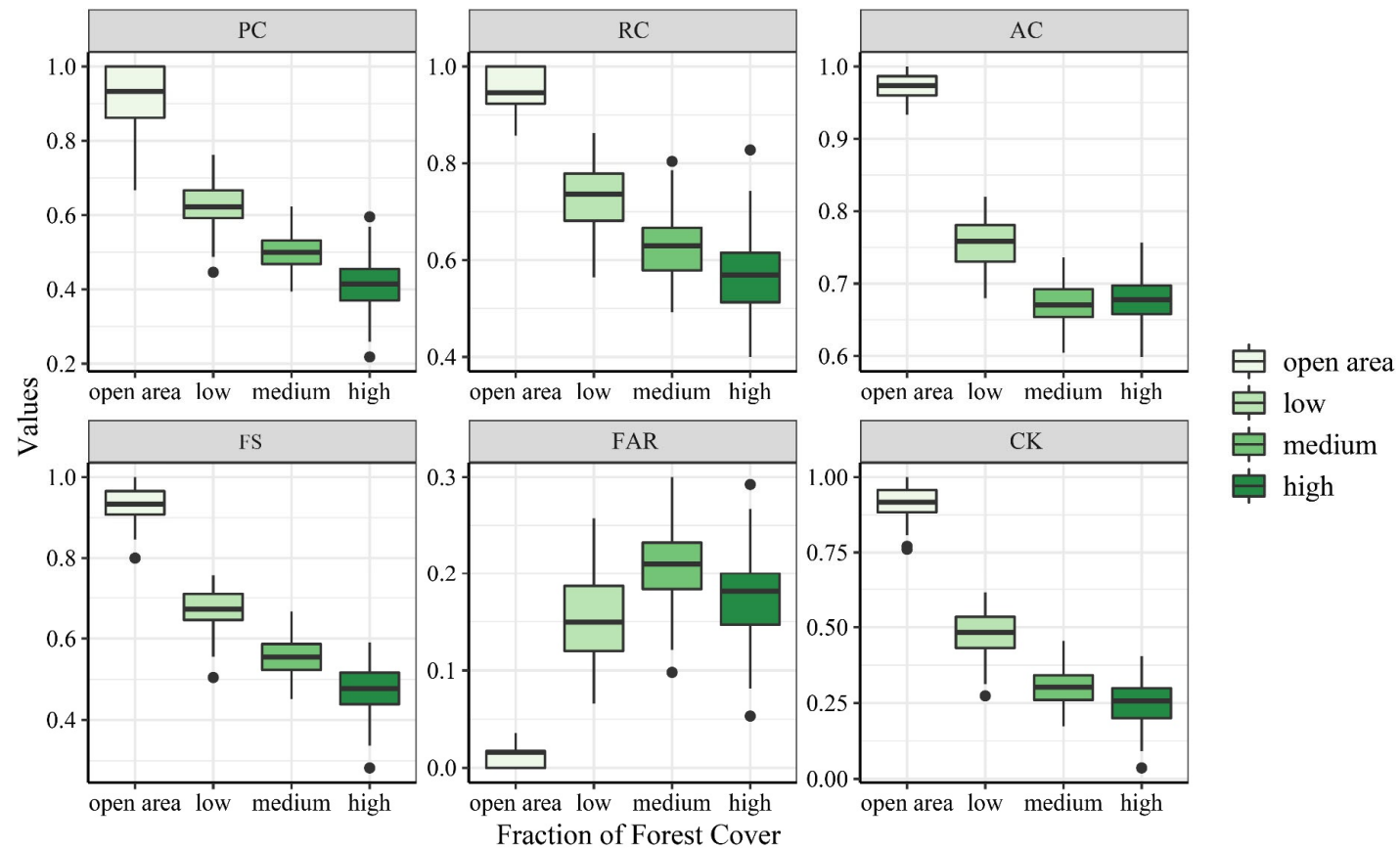
Cloud/Snow confusion in
MODIS cloud mask?

Effect of terrains



Aspect shows high

Effect of forest cover fraction



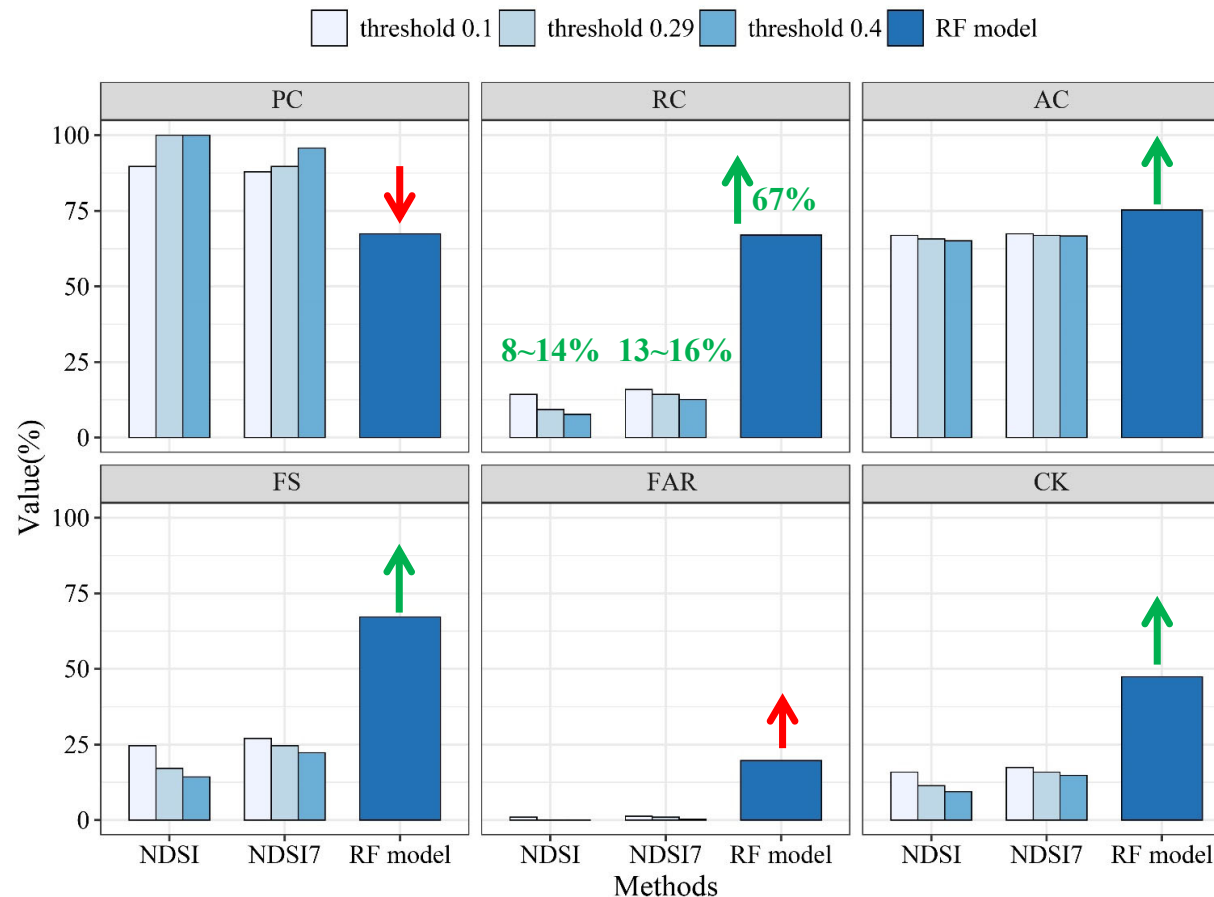
Forest cover fraction



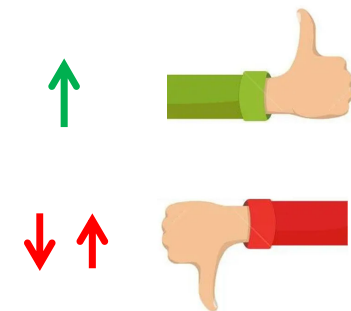
Performance



New algorithm vs NDSI approach



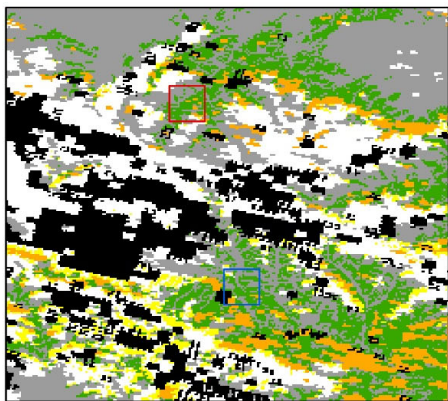
$$NDSI = \frac{b4 - b6}{b4 + b6}$$



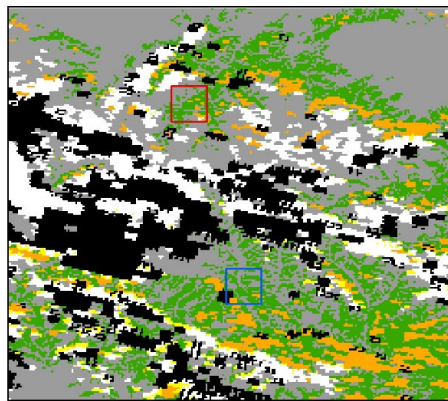
Large increase in forest snow detection of the new algorithm

Snow map example: NDSI approach

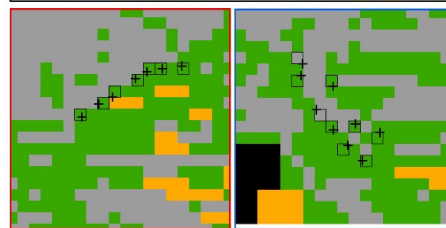
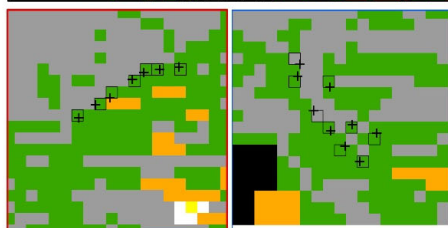
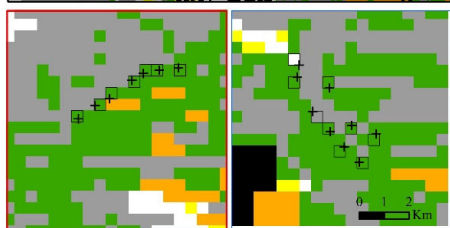
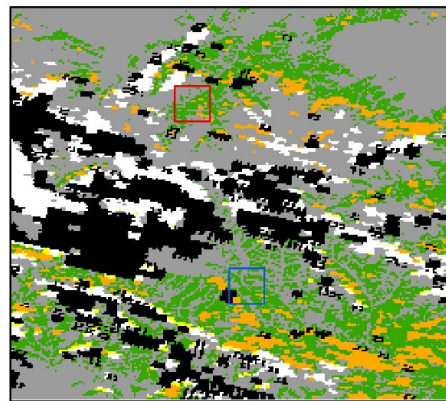
a) MOD10A1 NDSI threshold is 0.1



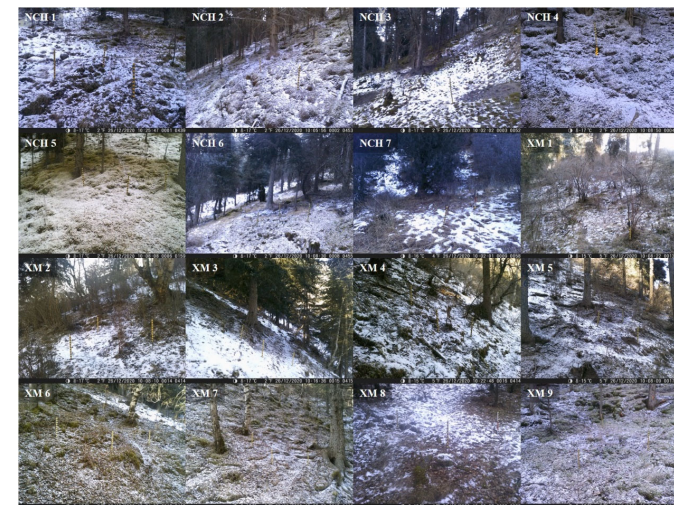
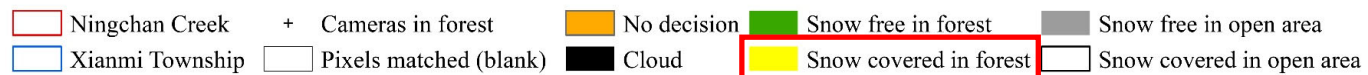
b) MOD10A1 NDSI threshold is 0.29



c) MOD10A1 NDSI threshold is 0.4



Legend

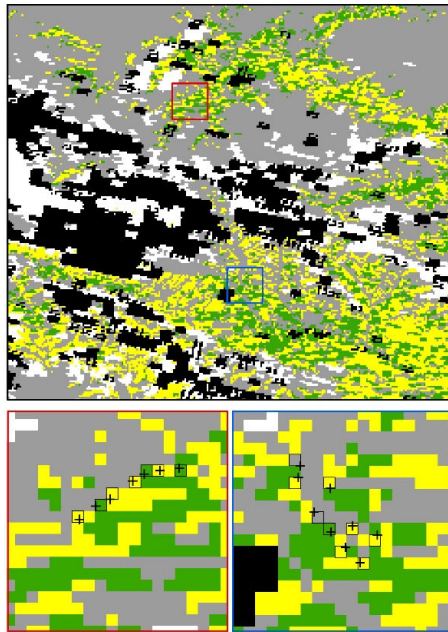


Ground truth: all snow covered

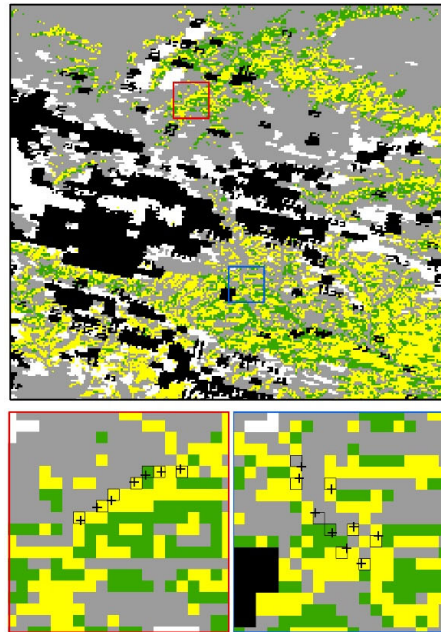
Dec. 26, 2020 in Qilian Mount.

Snow map example: machine learning approach

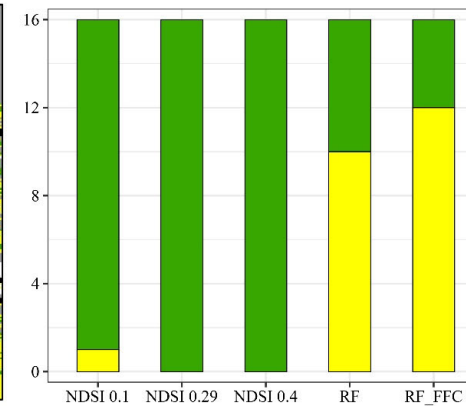
d) RF models



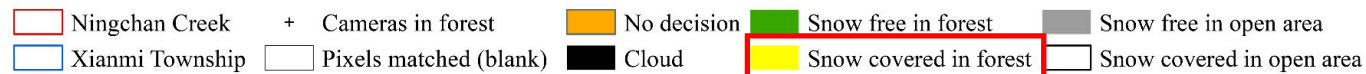
e) RF model with different FFC



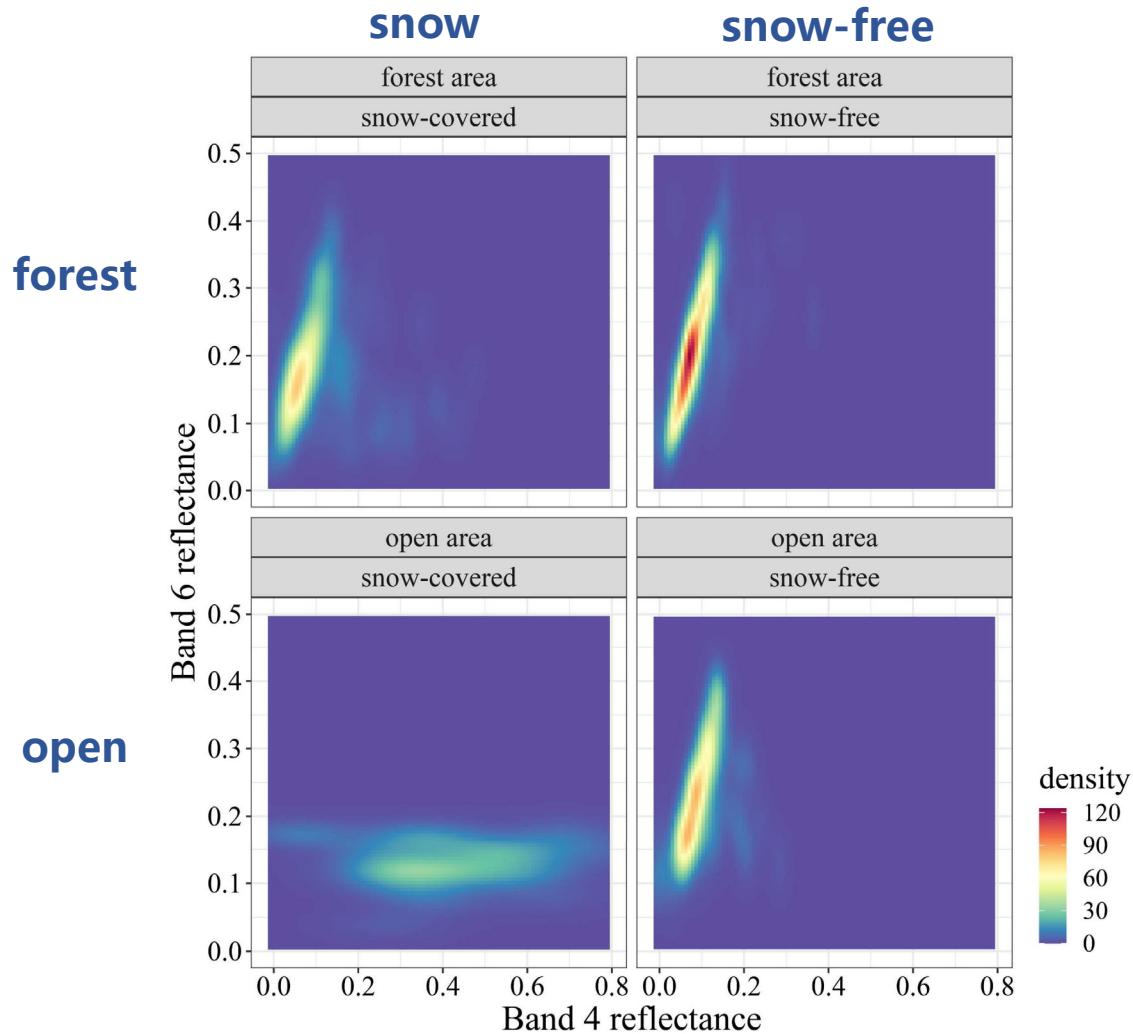
f) Evaluation with ground-truth images



Legend

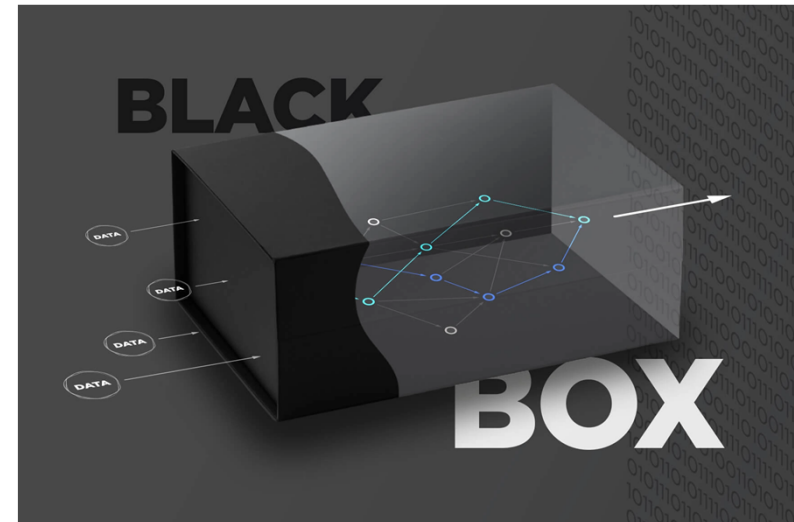


Discussion: Why machine learning?



$$NDSI = \frac{b4 - b6}{b4 + b6}$$

V/S



Conclusions

- Time-lapse photography is effective in monitoring forest snow process
- Machine learning-based algorithm can obviously detect more forest snow than NDSI-based approach
- Forest fraction and aspect can affect the algorithm accuracy by changing illumination

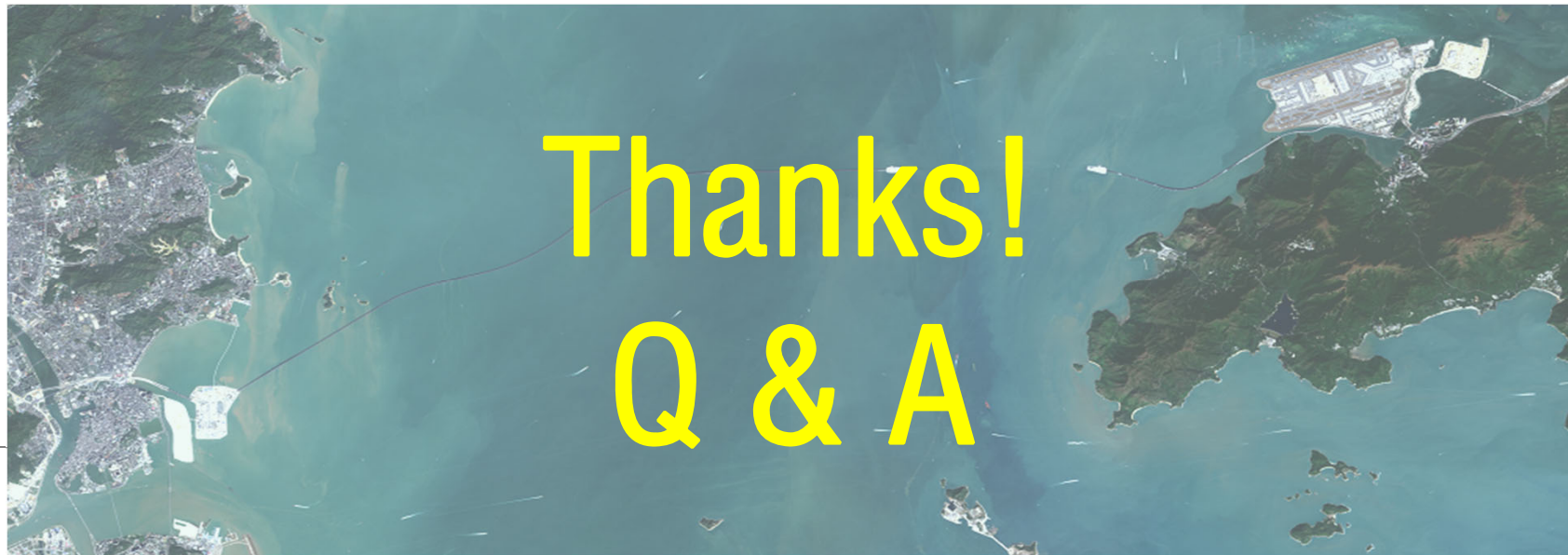
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