

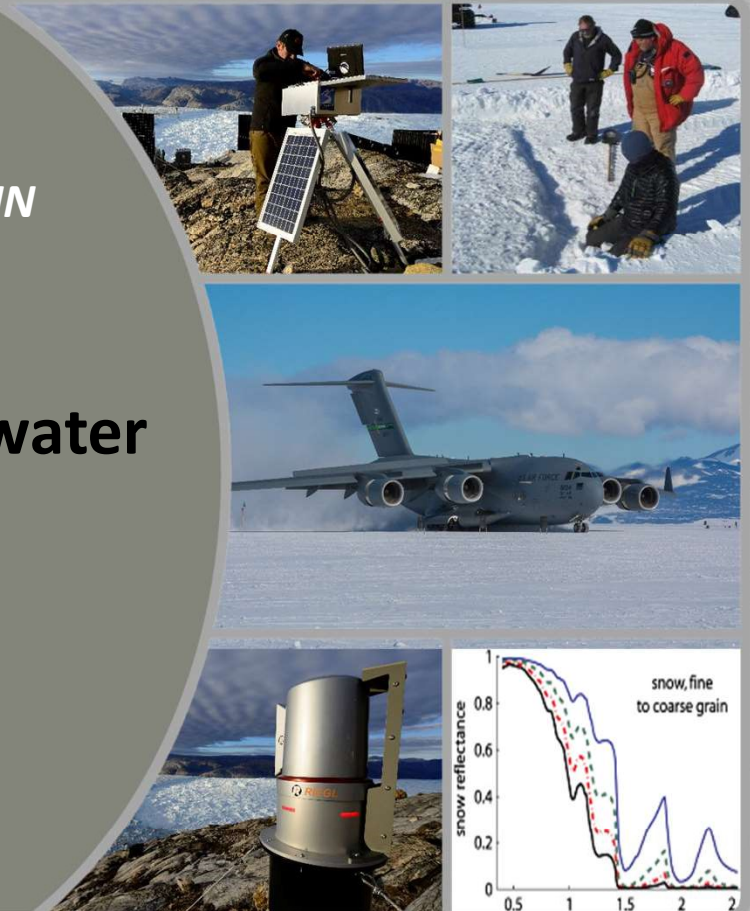
# *SUPPORT OF MISSION PLANNING AND OPERATIONS IN SNOW-IMPACTED REGIONS*

## Methods to robustly assess the snow water resource in remote mountains

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University of California, Santa Barbara  
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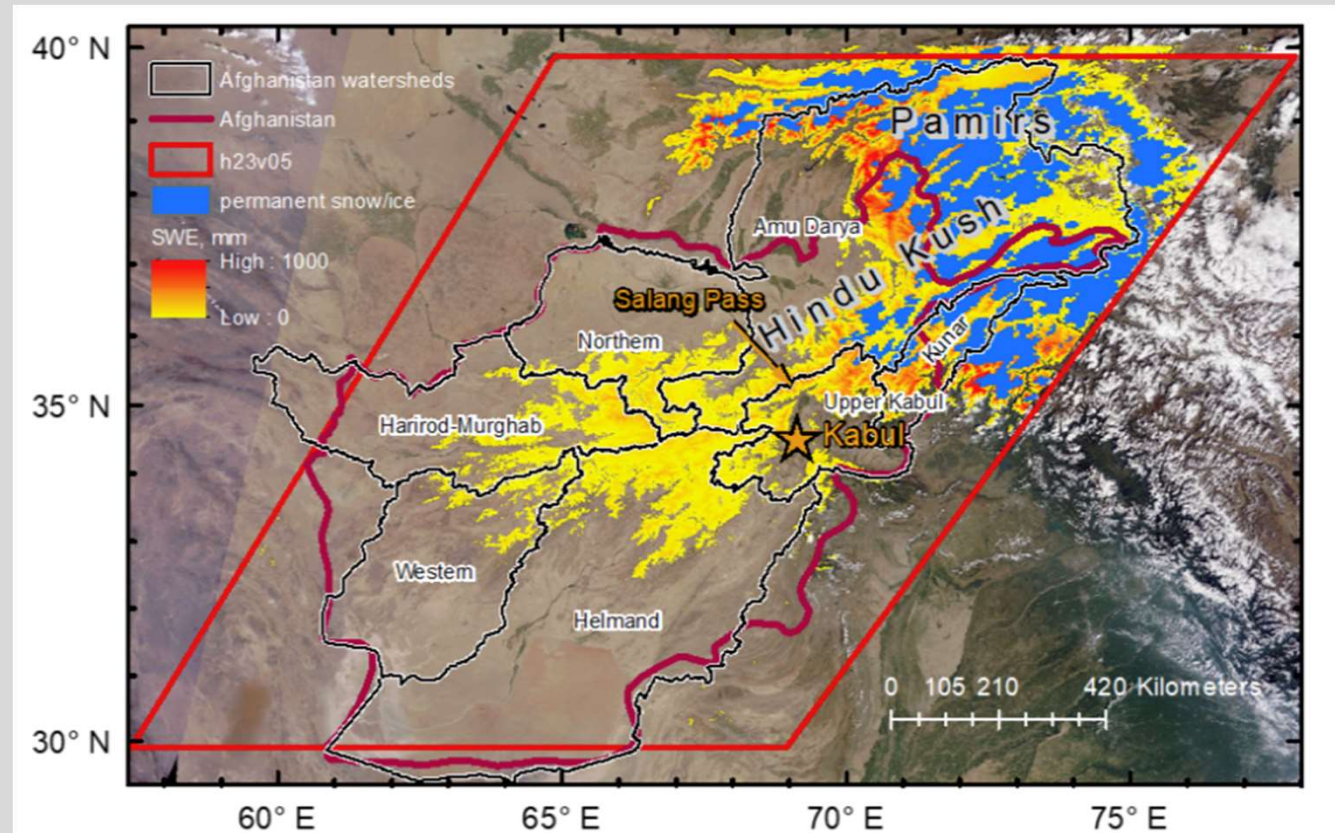


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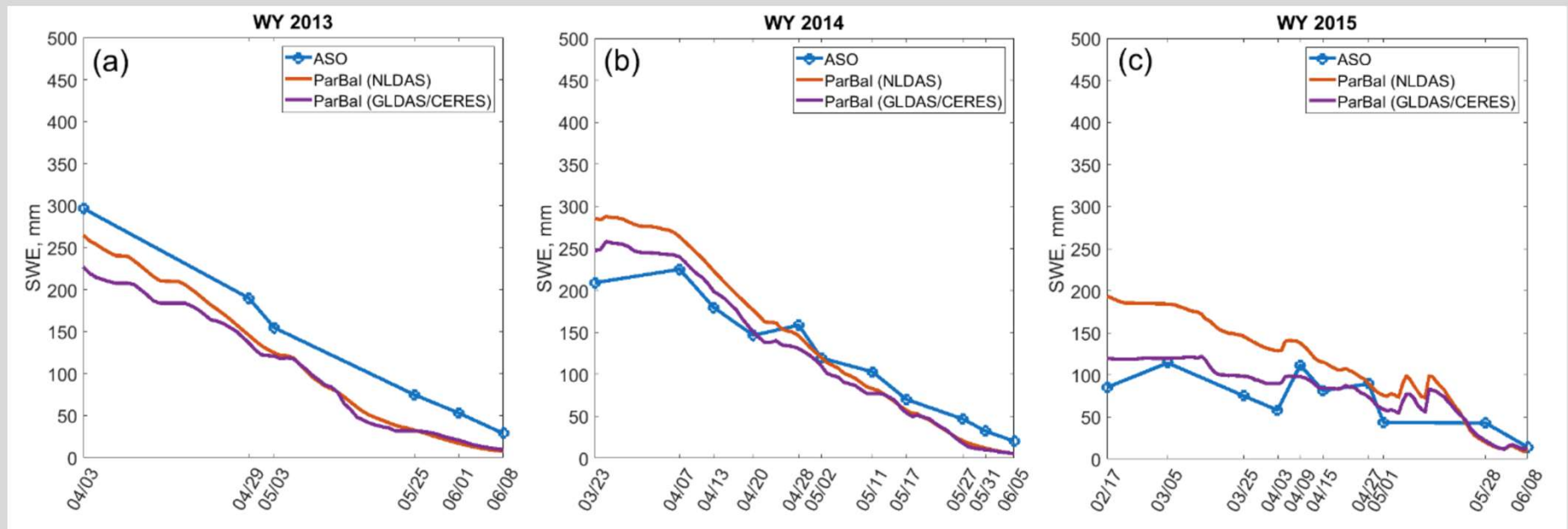
## Reconstruction of snow water equivalent (SWE)

- Using remotely-sensed data on radiation and temperature, we build the snowpack in reverse with energy balance components to create historical maps of SWE
- Advantage: Can be done in areas with no ground-based measurements
- Limitation: Can only do this retrospectively after snow melts out



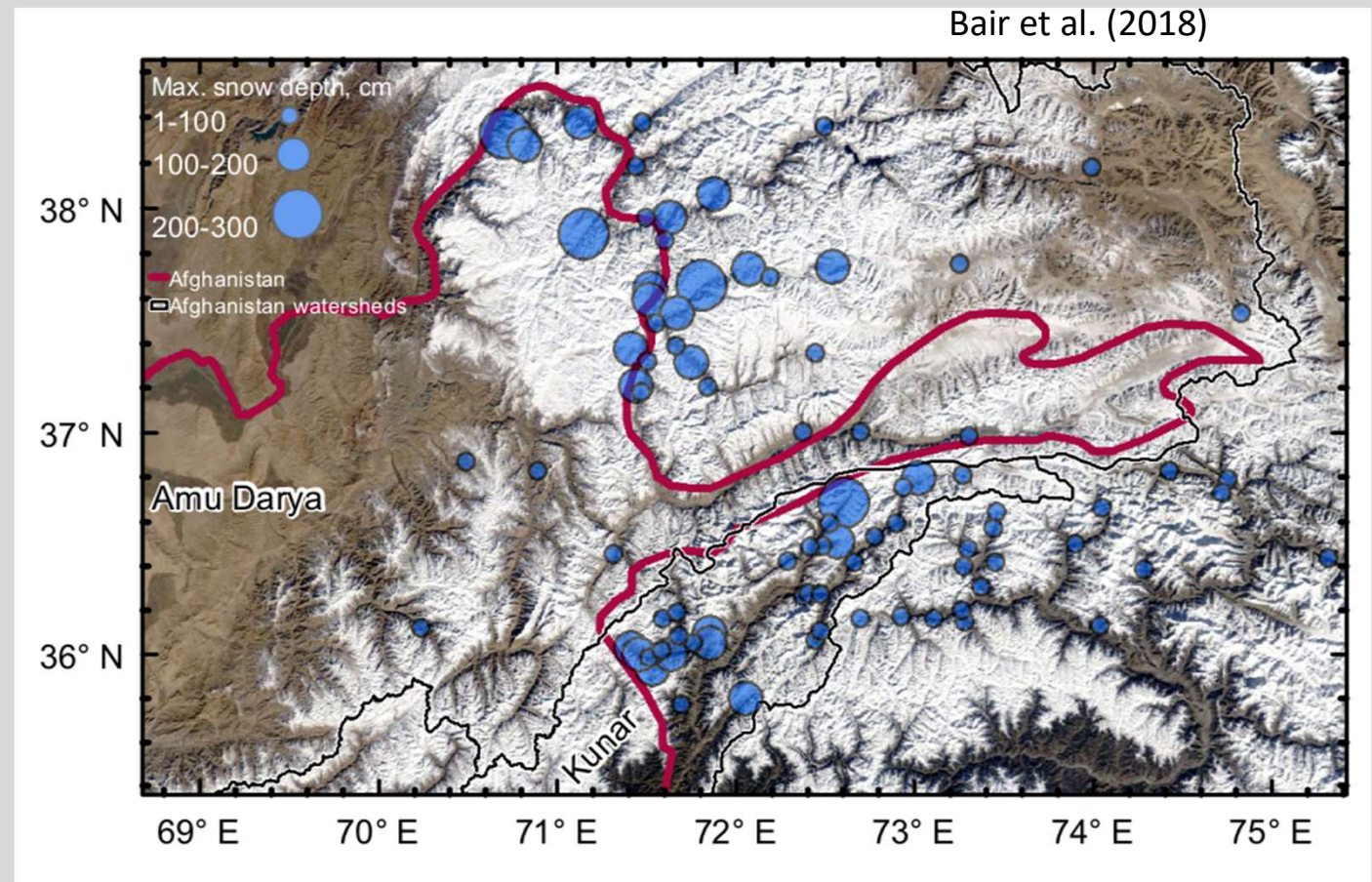
## SWE Reconstruction works well in the Sierra

- 22% Mean Absolute Error using GLDAS/CERES and MODSCAG/MODDRFS forcings in the upper Tuolumne for 2013-2015.

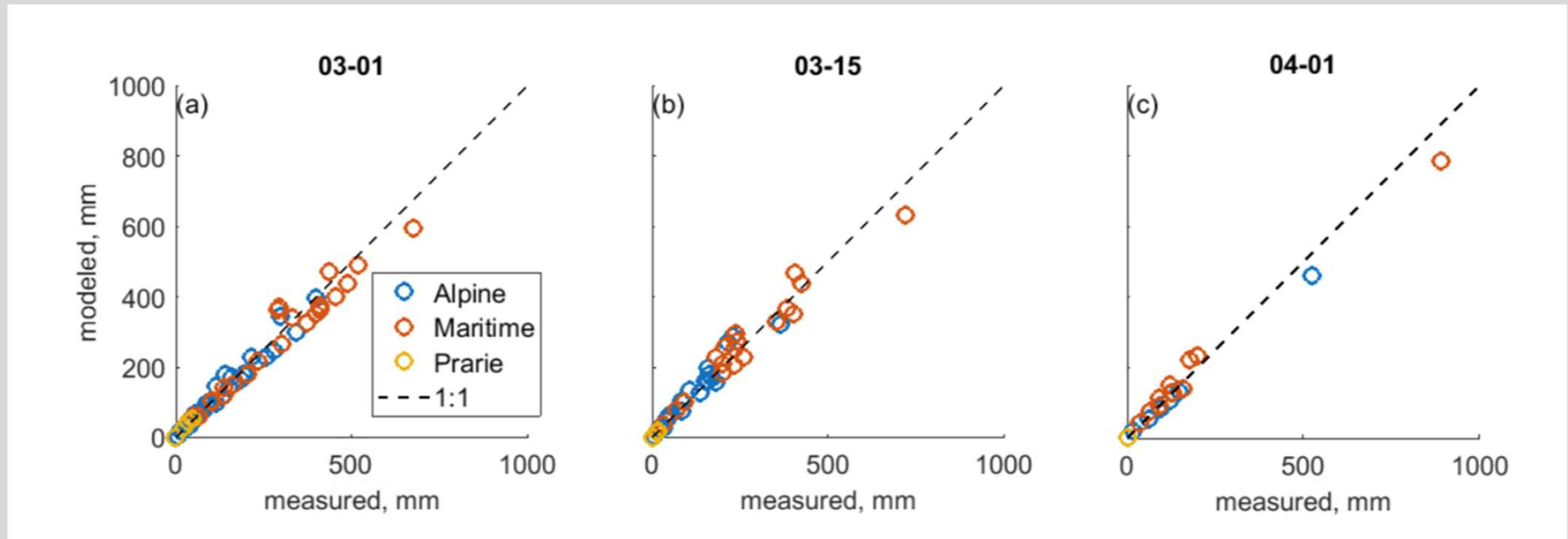


## Snow measurements in Hindu Kush supported by Aga Khan Agency for Humanity

- It has been difficult to get recent snow measurements for validation in remote, inhospitable regions
- The Aga Khan Agency for Humanity (AKAH) provided daily snow measurement from 88 stations for model validation (Chabot and Kaba, 2016)

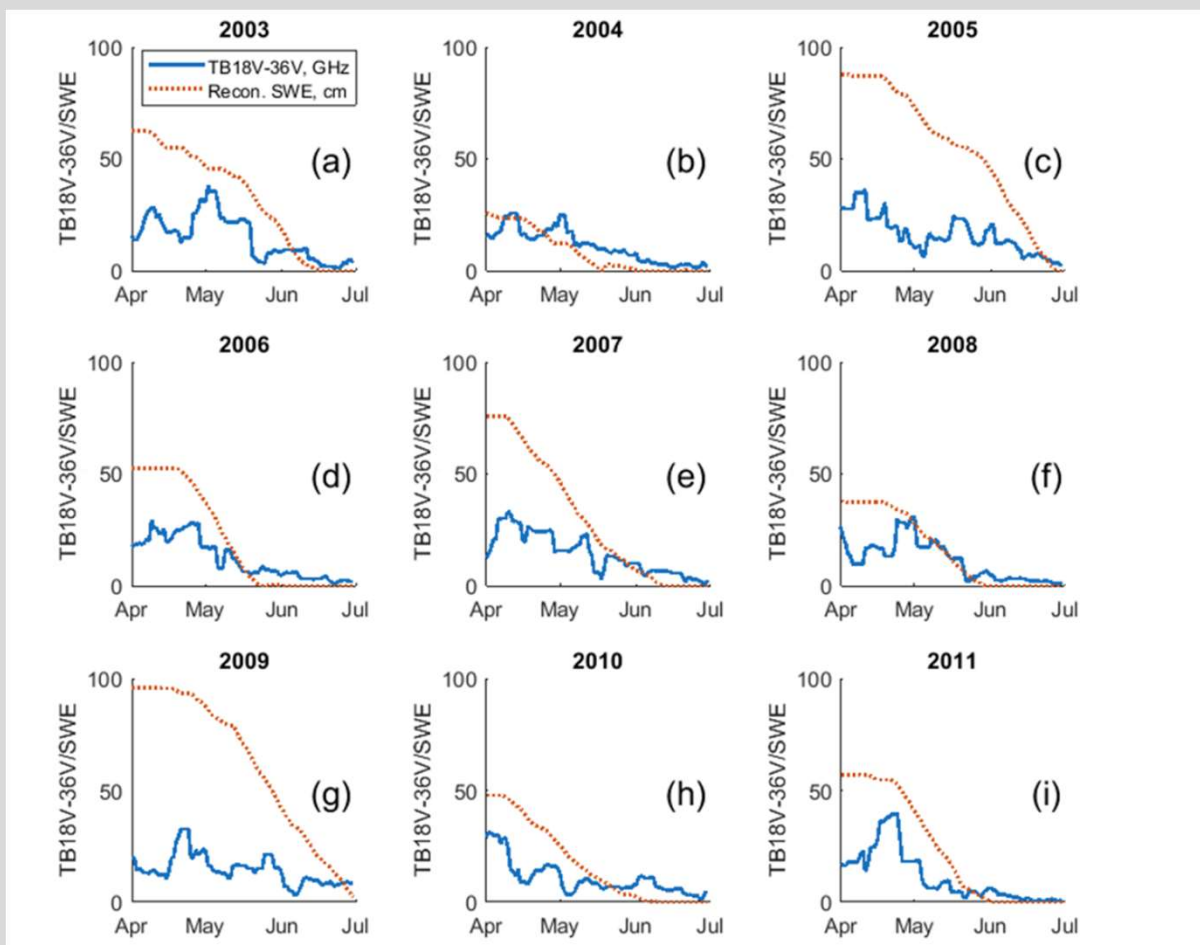


## SWE Reconstruction works well in Afghanistan (2017)



- Catch is that only snow depth is measured, so density had to be modeled based on snow climate (Sturm et al. 2011), yielding -12 to +26% uncertainty in SWE. Graphs above show the best-case given that uncertainty in density.

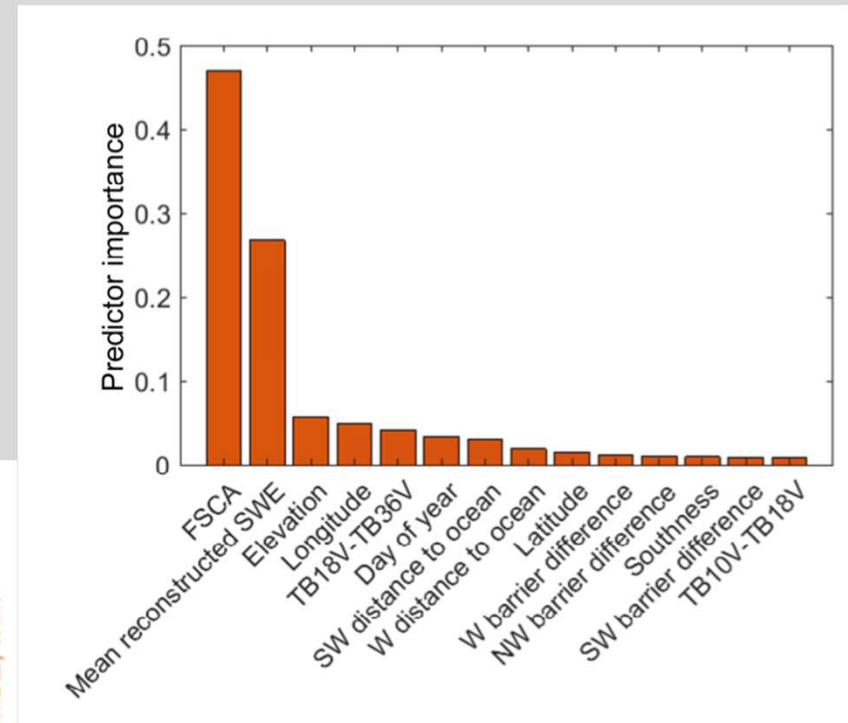
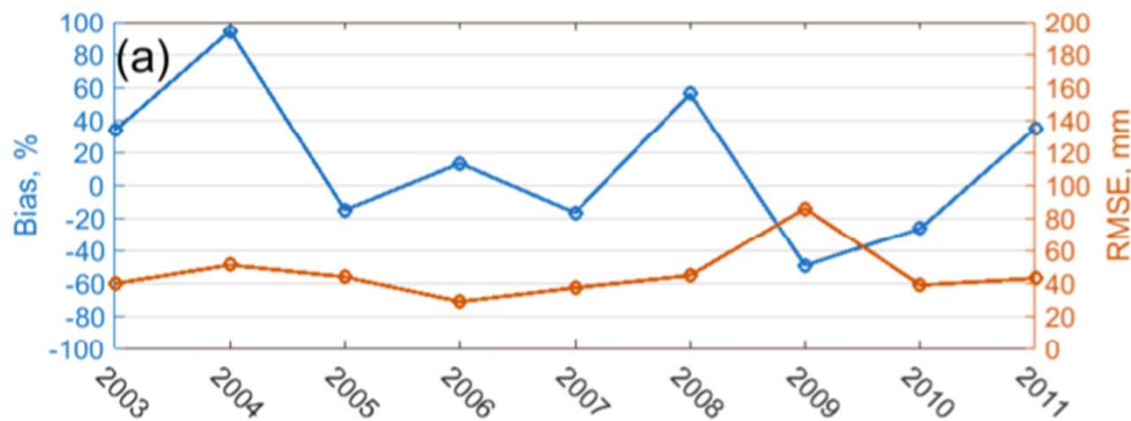
## Problems with passive microwave snow assessment in mountains



- Aggregated over large basins in Afghanistan, our reconstructed SWE values agree well with passive microwave estimates on April 1 (Daly et al. 2012)
- But on April 1, many areas are snow-free or have shallow snow cover, which masks errors in volume
- Passive microwave, even at enhanced (3 km) resolution, fails to capture the correct magnitude or rank for deep snow areas, such as Salang Pass (left)
- It also saturates at about SWE=200 mm

# SWE prediction in Afghanistan using machine learning

- Use reconstructed SWE to train machine learning models that use predictors available during the snow season
- Specifically, bagged trees (random forests) and neural networks were used
- Those models were used to predict seasonal SWE throughout Afghanistan
- 20% of training data (reconstructed SWE) was held out for validation
- Nash-Sutcliffe efficiency is 0.68 for all years, indicating substantial improvement over currently used forecasts from Air Force Weather



## Ongoing work, supported by U.S. Army

- 4 year effort through September 2020, now in Year 2
- We are at ARL 6 (System/subsystem model or prototype demonstration in a relevant environment). From the DOD *Defense Acquisition Guidebook*, we fit the ARL 6 example of “testing the prototype in a simulated operational environment”

Task Name	Description	Product/Deliverable	Dates
1) Improvements to ParBal and Reconstruction	e.g. improve $f_{SCA}$ , snow cloud discrimination, albedo, and radiative forcings	Code and peer reviewed publication on new spectral unmixing ( $f_{SCA}$ ) approach	ParBal code on GitLab; publication in 2019; ongoing development
2) Machine learning	SWE prediction using machine learning	Code and peer reviewed publication on machine learning	Publication just completed (Bair et al., 2018); code on GitLab
3) Improvement to passive microwave retrievals	e.g. enhanced resolution passive microwave products	Use of enhanced resolution passive microwave as a predictor in machine learning	Used in machine learning publication (Bair et al. 2018); ongoing development
4) Hydrologic modeling	Not pursuing due to funding constraints	N/A	N/A



## References

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D. Chabot photos of training for AKAH snow and avalanche observers