### Estimating wind direction with SWOT backscatter

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### Wind over lakes

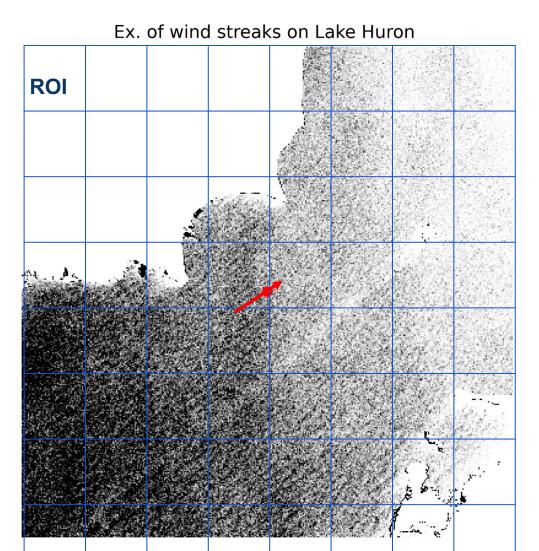
- Wind is an important driver of lake evaporation <sup>1</sup>.
- While wind can be extracted from reanalysis datasets, the resolution is often coarser than lake size.
- SWOT KaRIn backscatter (sigma0) observes wind-driven surface water roughness, and could be used to develop a wind speed model for lakes (ongoing work from Jessica Fayne's group)<sup>2</sup>.
- Wind direction is required to estimate wind speed from SAR <sup>3</sup>.



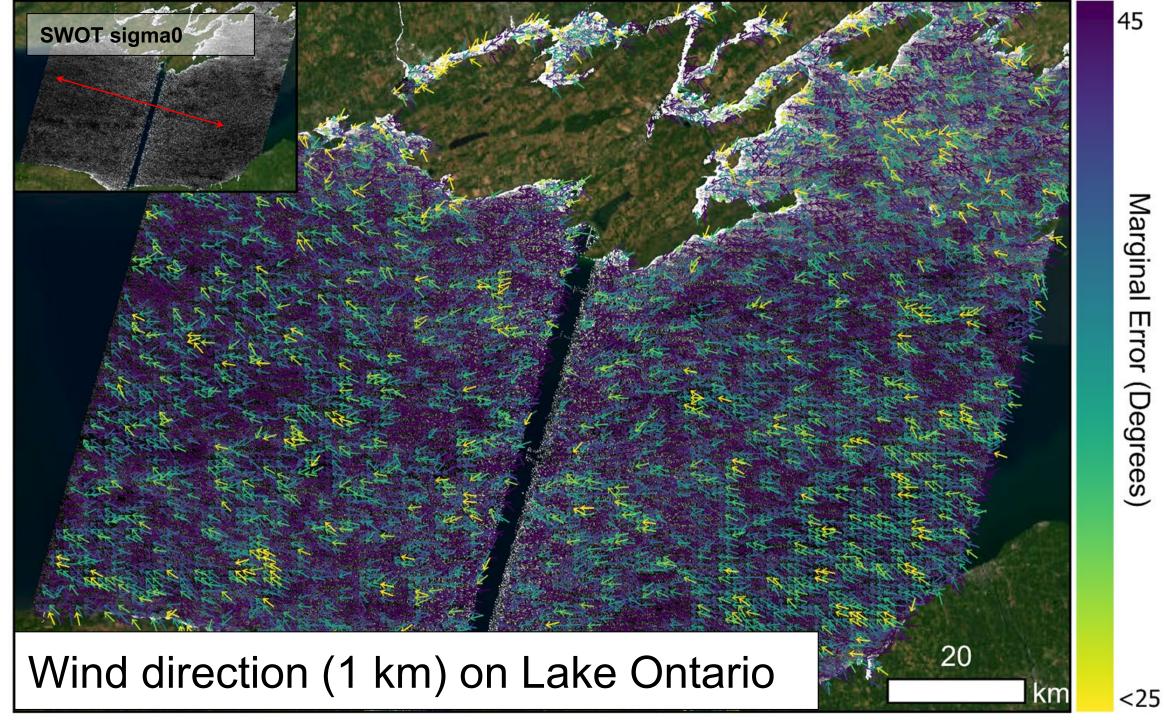
https://www.tahoedailytribune.com/news/wind-warmth-have-led-to-more-than-expected-evaporation-at-lake-tahoe/

## Use wind streaks to estimate wind direction from SWOT

- Estimate wind direction and marginal error (ME) from wind streaks using the Modified Local Gradient method <sup>3</sup> at 1 km resolution.
  - Subset the lake into ROIs of desired size (1 km)
  - Calculate local gradients within each ROI and use to estimate wind direction and ME
  - ME is a function of the local directions alignment within the ROI.
  - Smaller ME = more reliable wind direction estimate
- Test method using sigma0 from SWOT Level 2 Water Raster Image Data Product 2.0 (100m)

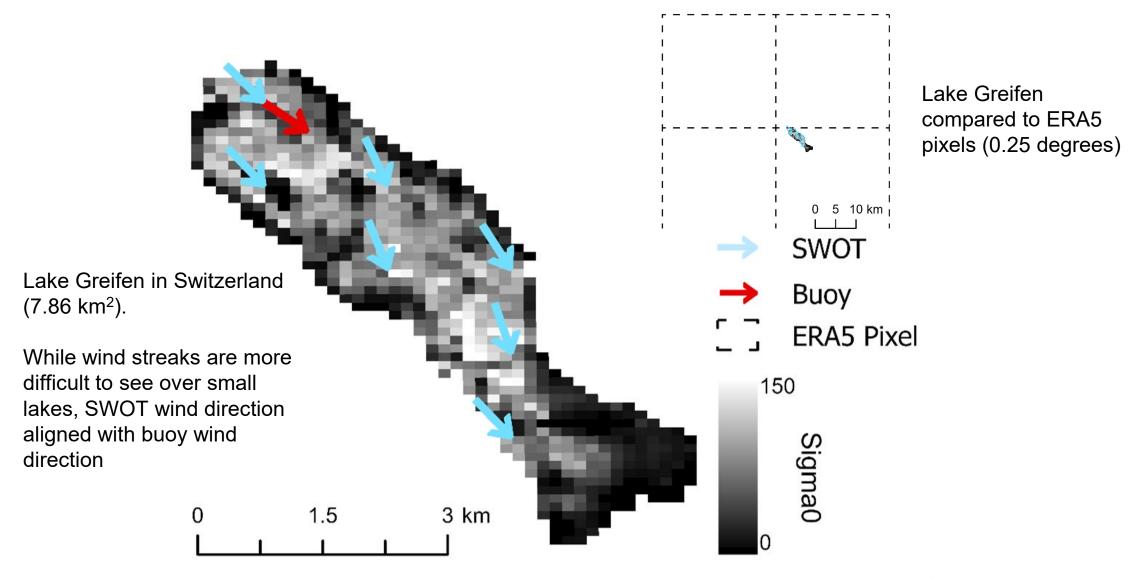


Example of wind streaks oriented NE-SW visible in Sentinel-1 radar backscatter over Lake Huron. The red arrow represents buoy wind direction at 56 degrees, aligning with the wind streaks.



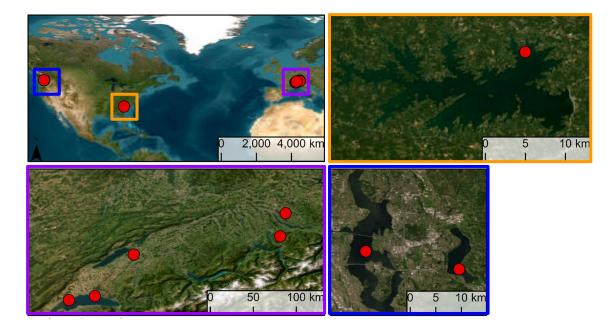
Marginal Error (Degrees)

# SWOT estimated wind direction at much higher resolution than global reanalysis dataset, ERA5



### Wind direction validation using over-water buoys

Buoy locations in US and Switzerland (N=8, 7-58 km<sup>2</sup>)



Compare performance stats of SWOT with ERA5 (180 degree ambiguity)

Image Subset	Number of images	LG-Mod MAE (degrees)	ERA5 MAE (degrees)
SWOT	70	37.72	39.58
SWOT (ME < 40)	32	36.48	40.99
SWOT (ME < 30)	24	34.78	40.32
SWOT (ME < 20)	10	28.13	50.38

- Wind direction estimated using SWOT had lower error compared to ERA5
- SWOT wind direction error improved after discarding estimates with high marginal error

### Preliminary takeaways

- Wind direction from SWOT was more accurate and higher resolution compared to frequently used global ERA5 dataset.
- Expand analysis to include additional buoys for more robust validation.

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#### Citations

- Zhao, B., Huntington, J., Pearson, C., Zhao, G., Ott, T., Zhu, J., Weinberg, A., Holman, K. D., Zhang, S., Anderson, R., Strickler, M., Cotter, J., Fernando, N., Nowak, K., & Gao, H. (2024). Developing a General Daily Lake Evaporation Model and Demonstrating Its Application in the State of Texas. *Water Resources Research*, 60(3), e2023WR036181. https://doi.org/10.1029/2023WR036181
- Fayne, J. V., & Smith, L. C. (2023). How Does Wind Influence Near-Nadir and Low-Incidence Ka-Band Radar Backscatter and Coherence from Small Inland Water Bodies? Remote Sensing, 15(13), Article 13. https://doi.org/10.3390/rs15133361
- Rana, F. M., Adamo, M., Pasquariello, G., De Carolis, G., & Morelli, S. (2015). LG-Mod: A Modified Local Gradient (LG) Method to Retrieve SAR Sea Surface Wind Directions in Marine Coastal Areas. Journal of Sensors, 2016, e9565208. <u>https://doi.org/10.1155/2016/9565208</u>