The Conditional Relationship Between Atmospheric River Moisture, Wind, and Precipitation in **Satellite Observations**



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Motivations

- Atmospheric Rivers (ARs) are elongated and narrow filaments of water vapor transport in the atmosphere with **convergence often associated with heavy** precipitation events
- Previous work on AR "flavors" mainly on the regional scale, but not global [2]
- There is a gap in an understanding of the relationships between AR moisture, wind, integrated vapor transport (IVT), and precipitation

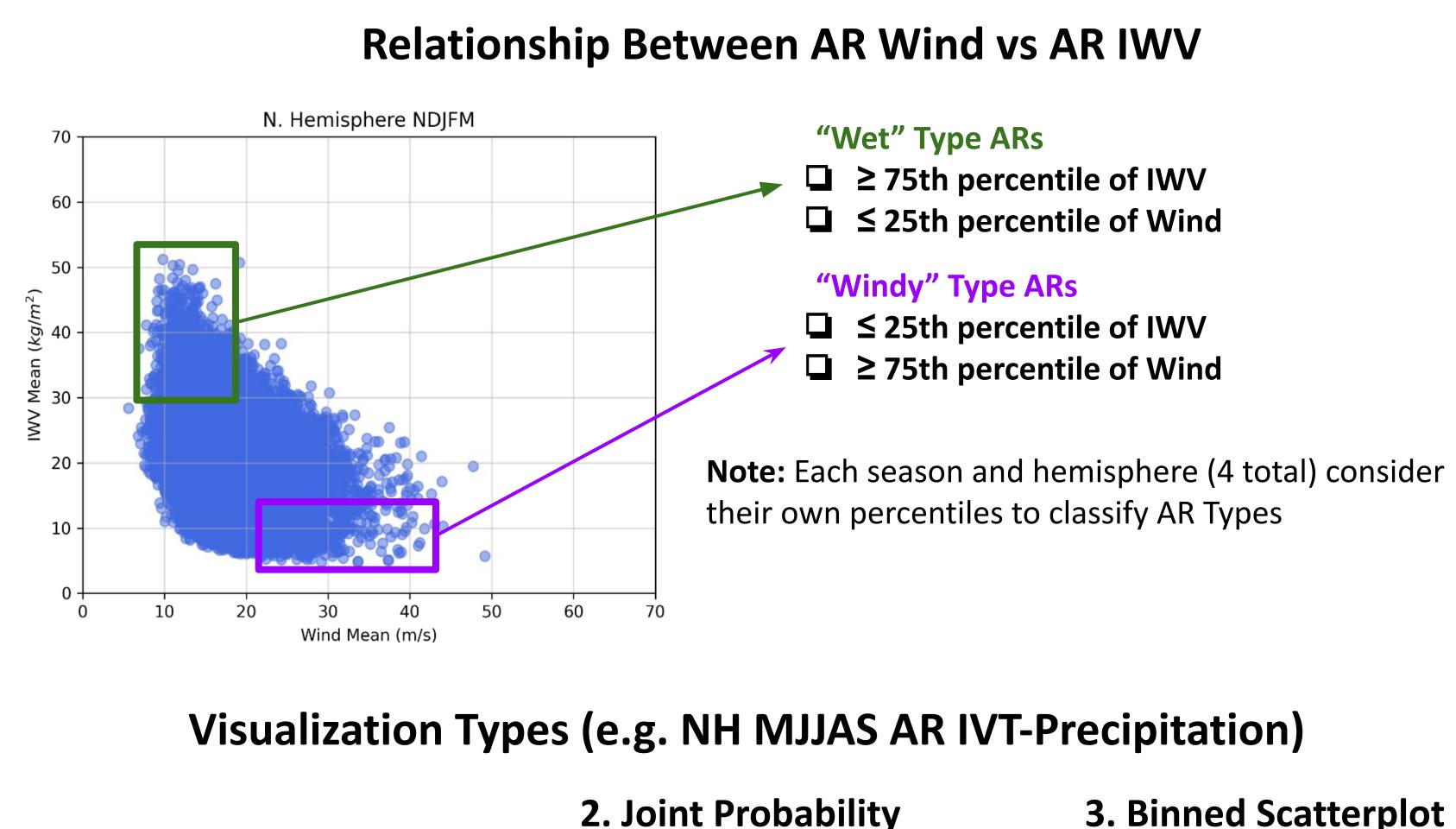
We analyze the AR moisture, wind, IVT-precipitation relationship by...

- 1. Developing a **global, percentile-based classification of AR types** in satellite observations
- 2. Using a **binned scatter approach** to quantify the sensitivity of mean AR precipitation to changes in other AR metrics
- 3. Identifying spatial patterns of precipitation sensitivity and whether there is a dependence on AR type

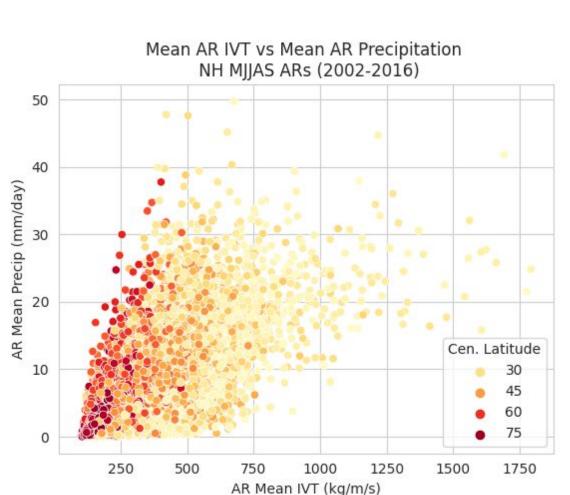
Data & Methodology

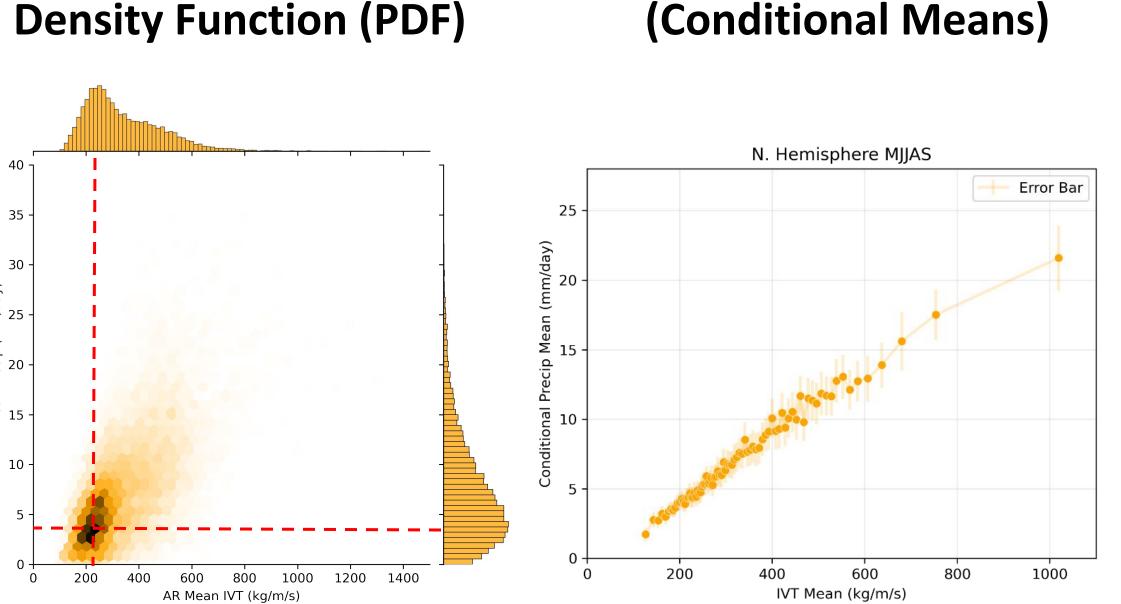
- **Satellite-based** data over 2002–2016 case study period U Version 6 of the AIRS/Aqua L3 Daily Standard Physical Retrieval (AIRS-AMSU) at a spatial
- resolution of 1.0° x 1.0° (Teixeira et al., 2013) U Version 6 the GPM IMERG Final Precipitation L3 1 Day (IMERG) at a spatial resolution of 0.1° x 0.1° (Huffman et al., 2019)
- AR detection algorithm based on **GIVT-threshold** developed for satellite data [3,5]
- AR geometry (shape), AR centroids, integrated water vapor (IWV), integrated vapor transport (IVT), geostrophic wind, and precipitation on **daily temporal scale**
- Ginned scatter divides the independent variable (x) into equally sized bins and find the **conditional mean** of the dependent variable (y) within each bin **[1,6]**
- \Box Precipitation threshold of $\geq 0.01 \text{ mm/day}$ to remove non-precipitating ARs Grid point **linear regression** using regular statistics (*R*, *R*², 95% confidence)
- intervals)

AR Types & Visualization



1. Simple Scatterplot

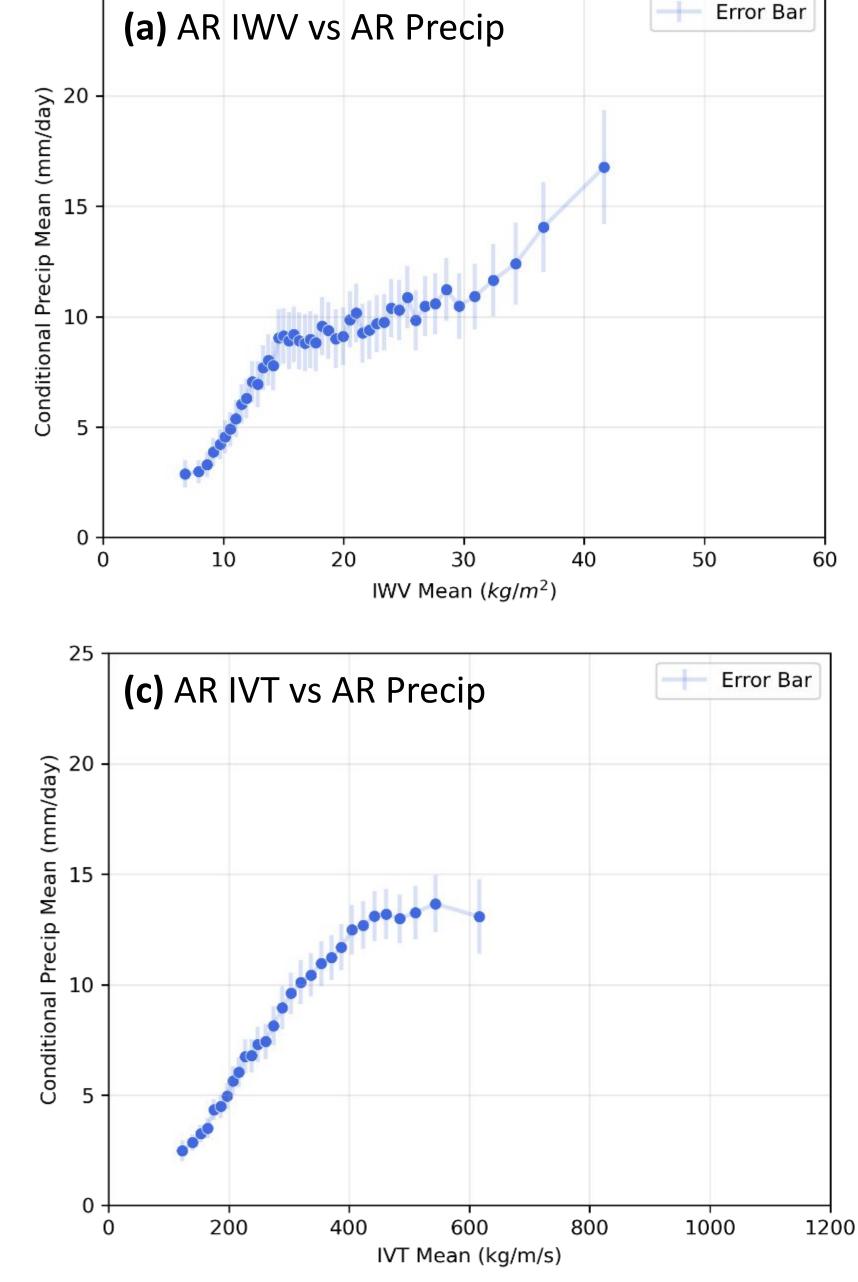




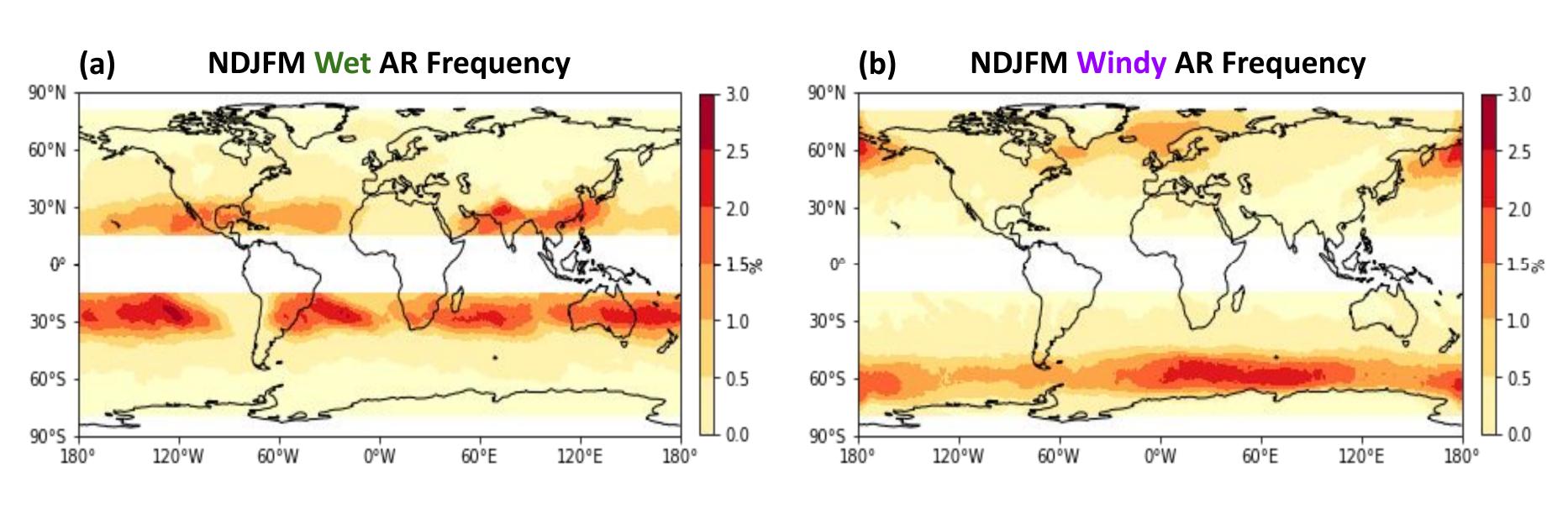
3. Binned Scatterplot (Conditional Means)

Conditional Relationships

Conditional Means of AR Precipitation, NH NDJFM ARs (2002 – 2016)



Frequency of Wet and Windy ARs



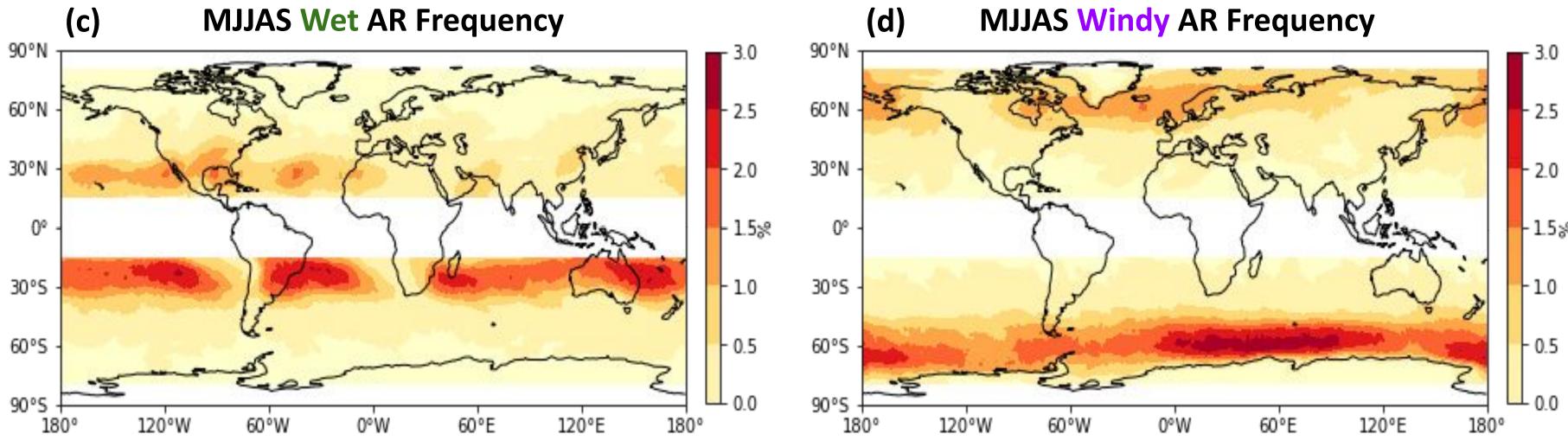


Figure 2. Satellite-derived global frequency of AR types for November–March (NDJFM) and May–September (MJJAS) interannual seasons. We define AR frequency as the fraction of (daily) timesteps a grid point experiences AR conditions, as determined by the GIVT-threshold algorithm. **Panels (a) and (c)** show **Wet** AR frequency. **Panels (b) and (d)** show **Windy** AR frequency.

References

[1] Cattaneo, M. D., Crump, R. K., Farrell, M. H., Feng, Y. (2021). On Binscatter. arXiv Economics.

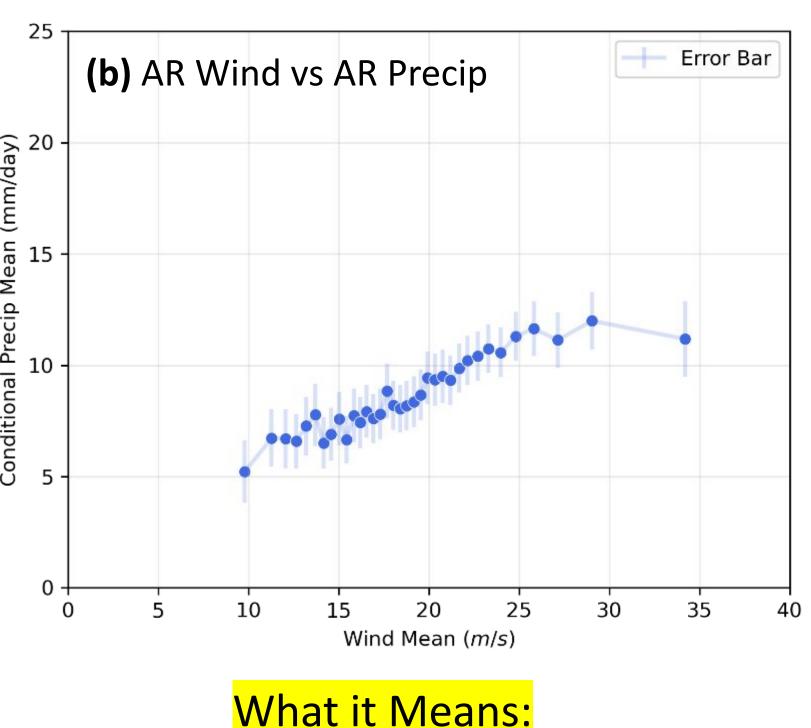
[2] Gonzales, K. R., Swain, D. L., Barnes, E. A., & Diffenbaugh, N. S. (2020). Moisture- versus wind-dominated flavors of atmospheric rivers. Geophysical Research Letters. 47, e2020GL090042.

[3] Guan, B., & Waliser, D. E. (2015). Detection of atmospheric rivers: Evaluation and application of an algorithm for global studies. Journal of Geophysical Research: Atmospheres, 120(24), 12514-12535.

[4] Guan, B., Waliser, D. E., & Ralph, F. M. (2023). Global application of the atmospheric river scale. Journal of Geophysical Research: Atmospheres, 128, e2022JD037180

[5] Ma, W., Chen, G., Guan, B., Shields, C. A., Tian, B., & Yanez, E. (2023). Evaluating the representations of atmospheric rivers and their associated precipitation in reanalyses with satellite observations. Journal of Geophysical Research: Atmospheres, 128, e2023JD038937

[6] Peters, O., Neelin, J. (2006). Critical phenomena in atmospheric precipitation. *Nature Phys* 2, 393–396.



With increasing moisture and wind values, AR precipitation also increases (varying sensitivities and small nonlinear deviations)

Figure 1. The satellite-derived conditional relationships between several AR metrics and mean precipitation, shown for Northern Hemisphere (NH) November–March (NDJFM) ARs. **Panels (a–c)** show the relationships for AR moisture (IWV), wind, and IVT, respectively. Error bars represent the 95% CI.

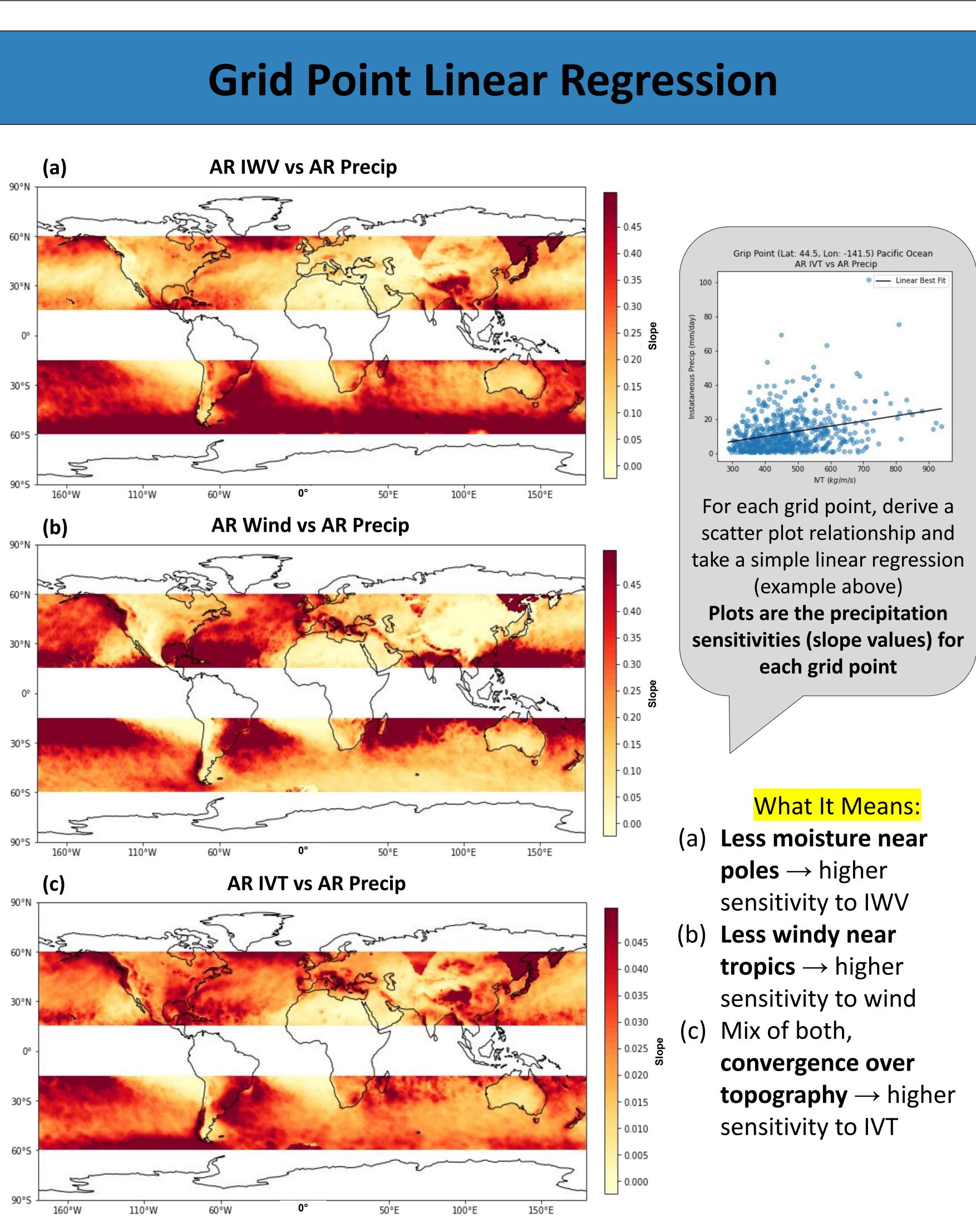


Figure 3. Global grid point linear regression values for several AR metrics (independent variable) versus mean precipitation (dependent variable). A higher slope value represents a higher precipitation sensitivity to the AR metric and vise versa. **Panels (a–c)** show the relationships for AR moisture (IWV), wind, and IVT, respectively.

Key points: ocean basin and season is most sensitive to moisture most sensitive to wind associated with nonlinearities **Potential future work:** expanding on regional case [4] IVT-precipitation relationship

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Summary

- Strong, positive conditional AR IVT-precipitation relationship, regardless of
- \Box High precipitation sensitivity to moisture near poles \rightarrow Windy AR precipitation
- \Box High precipitation sensitivity to wind near tropics \rightarrow Wet AR precipitation is
- □ Wet (Windy) ARs have a higher frequency near tropics (poles)
- Mixed regions of precipitation sensitivity to IVT in tropics and near topography
- (Not shown) AR IVT-precipitation relationship has a small dependence on AR type,
- Use the conditional AR IVT-precipitation relationship to define a global ranking scale,
- Incomporate reanalysis and climate model data to evaluate biases in the AR
- **Explore how AR precipitation sensitivities change under global warming** Acknowledgements: Bin Guan & Sudip Chakraborty

