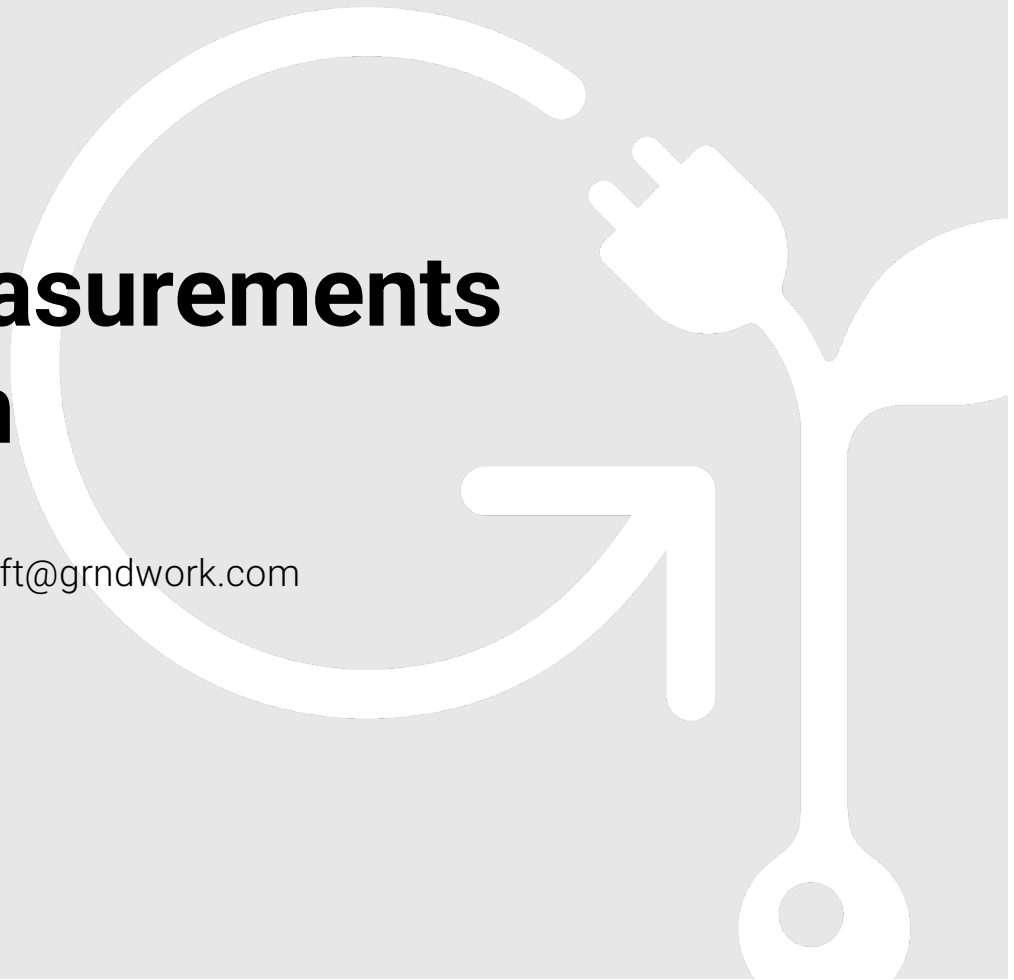




Solar Resource Measurements and Site Adaptation

PVPMC 2026

Author: Nate Croft, Director, Advisory Services, ncroft@grndwork.com



Site-Adaptation (aka Tuning, MCP, Bias Correction)

Short Term (1-2 yrs) high quality ground measurements
Long-term (20+ yrs) modeled resource data (satellite-based)

Use relationship between measured and satellite in the the concurrent period to adjust the full period of satellite data

Goals of Adaptation:

1. More Accurate Data – Reduce Error – Reduce EYA Bias
2. More certainty that the result is likely to be accurate – Reduce Uncertainty

This Session:

Test and compare methods
Test against data from GroundWork campaigns
Note: One-year tuning periods for this research



GroundWork Measured Data

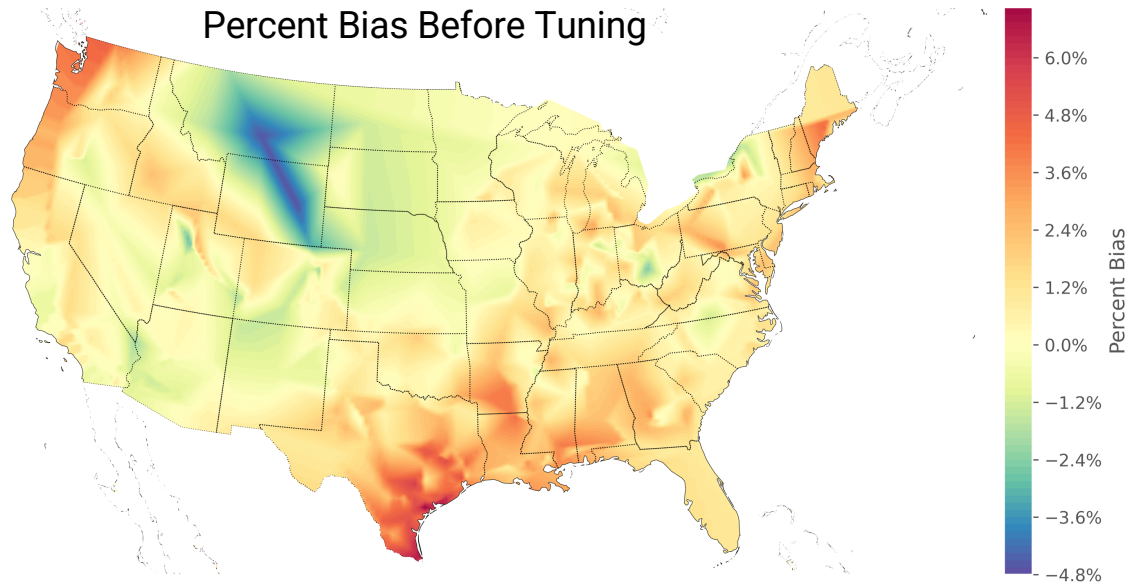
Dual Co-located GHI Class A Pyranometers
Weekly maintenance
Extensive QC process

515 locations with more than 1 year of measurements
87 locations with more than 2 years of measurements

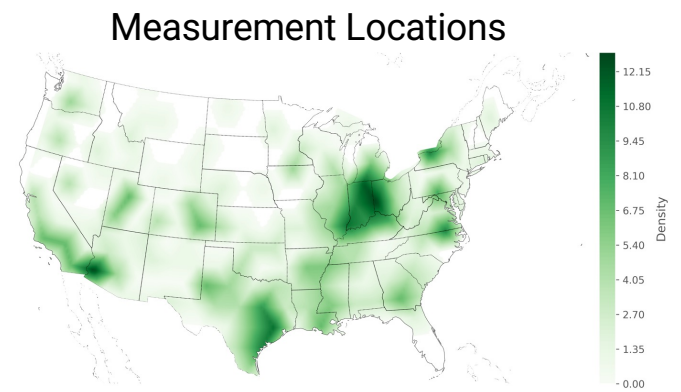
Data is from before 2025 and only accessed for this research use



Modeled Data has Error



NSRDB PSM v4
CONUS Range: -4.8% to +7.3%



515 Measurement Locations
1-year+ campaign length



Site-Adaptation Method Examples

Ratio Style Methods	Example
Linear / Regression Ratio (no intercept)	$Measured = m * Satellite$
Linear / Regression Slope and Intercept	$Measured = m * Satellite + b$
Sinusodial	$Error = Asin\left(\frac{2\pi}{365.25} (Dayofyear + B) + C\right) + D$

Parameter Style Methods	Example
Model Output Statistics ("MOS")	$Error = a_0Predictor_1 + a_1Predictor_2$
Polynomial	$Measured = p_1Sat^3 + p_2Sat^2 + p_3Sat$

CDF Style Method	Example
Cumulative Distribution Function*	$Error \sim CDF_M / CDF_S$

* CDF can be used in combination with other methods

Option: Split into groups and do unique adjustments to each. E.g. Clearsky/Non-Clearsky, Seasonal

Full List of Tested Combinations:

- All_Ratio
- Clearsky_Ratio
- Seasonal_Ratio
- CS/Seas_Ratio
- All_Poly
- Clearsky_Poly
- Seasonal_Poly
- CS/Seas_Poly
- All_MOS
- Clearsky_MOS
- Seasonal_MOS
- CS/Seas_MOS
- Daily_Ratio w/Int
- Ratio w/Int
- Clearsky_Ratio w/Int
- Seasonal_Ratio w/Int
- CS/Seas_Ratio w/Int
- Seasonal_Sine
- CS/Seas_Sine
- All_CDF
- Clearsky_CDF
- Seasonal_CDF
- CS_Seas_CDF
- CS/Seas_Ratio+CDF
- CS/Seas_Sine+CDF
- CS/Seas_Poly+CDF
- All_MOS+CDF
- CS/Seas_MOS+CDF



Testing Method

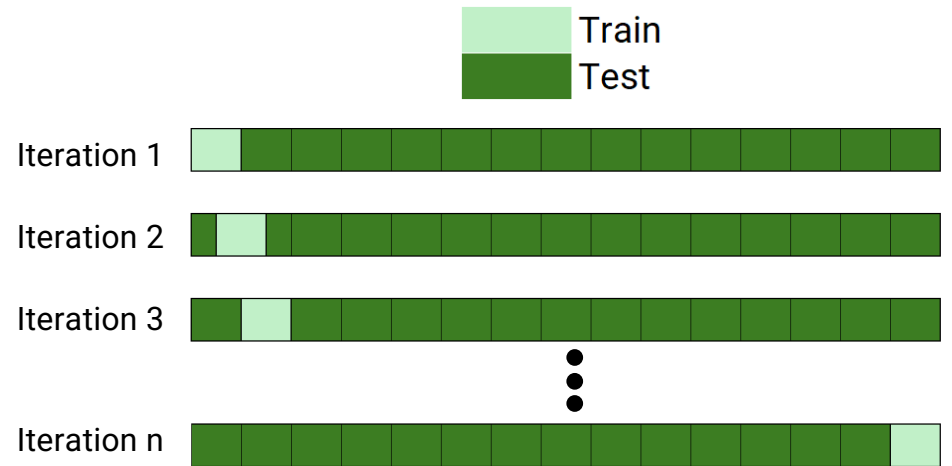
Long data sets (~2000-present) of measured data
NSRDB PSM v4 Satellite-based data

Train on 1 year, Test against rest of dataset
Iterate through multiple options for the test year

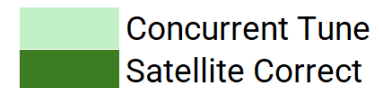
Outputs

Resulting Bias of Tuned data vs the Measured Data
Spread of possible outcomes of the method (Uncertainty)

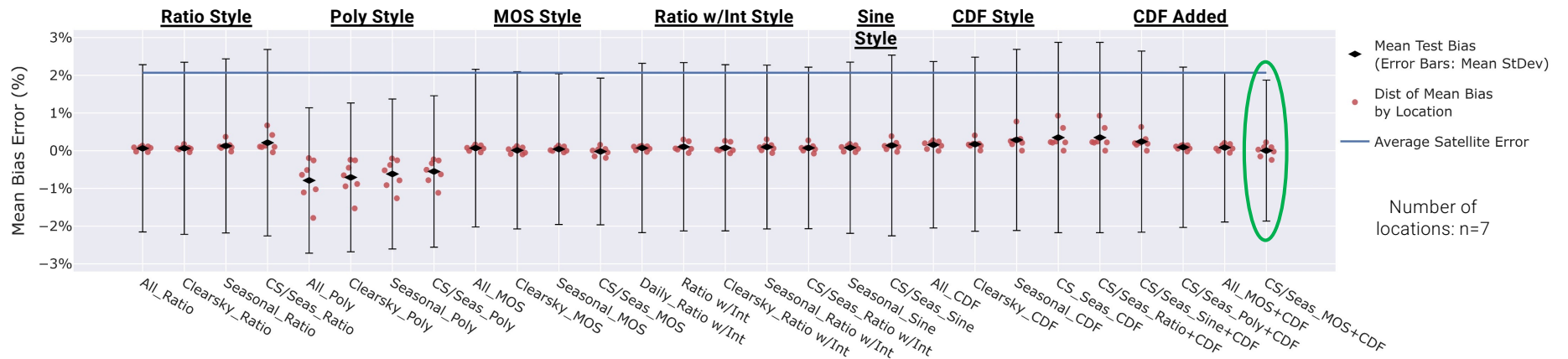
Allows a direct comparison of methods



Approximates a tuning scenario with 1 year of measurements



Full Period SURFRAD Data Test Results



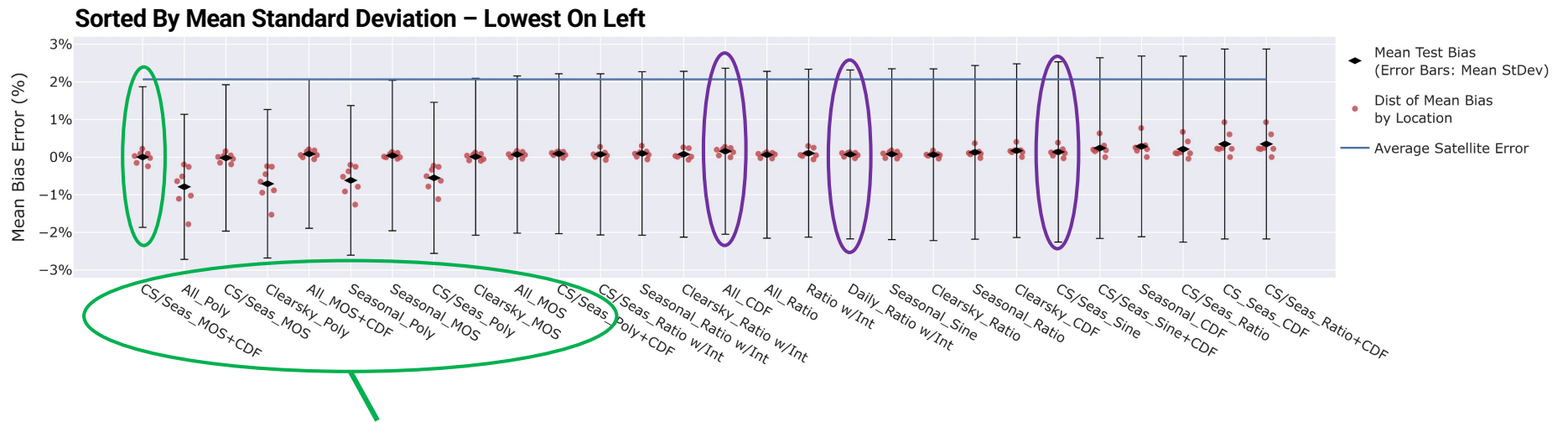
7 SURFRAD locations vs NSRDB PSM v4 **Markers: Mean Bias Error of tests** **Error Bars: Standard Deviation of tests**

Average Satellite error is ~+2%
 All methods correct that error on average
 Most, but not all, get close to 0% bias on average
 Mean standard deviation of tests is about 2%

One method has lowest uncertainty across all locations and very low bias:
Clearsky, Seasonal, Model Output Statistics (MOS), plus CDF correction



Full Period SURFRAD Data Tests Lowest Variability



Parameter Style Methods

Parameter Style Methods	Example
Model Output Statistics ("MOS")	$Error = a_0Predictor_1 + a_1Predictor_2$
Polynomial	$Measured = p_1Sat^3 + p_2Sat^2 + p_3Sat$

Parameter methods have lower variability

One method has lowest uncertainty across all locations and very low bias: **Clearsky, Seasonal, MOS, plus CDF correction**



Model Output Statistics

MOS is based on the relationship of error

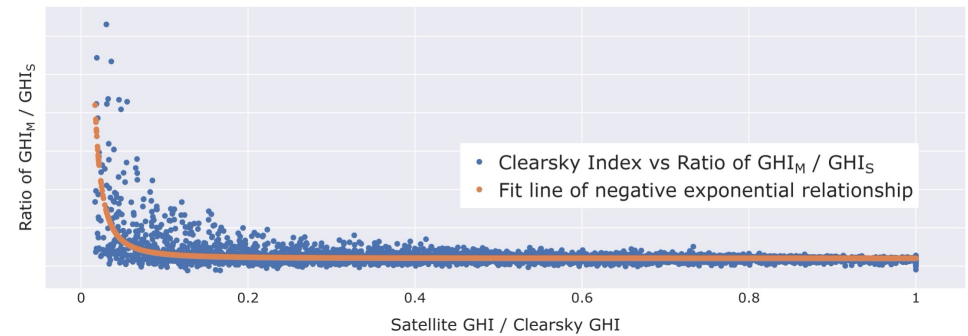
In our version:

We identified a negative exponential relationship
 Ratio of Measured GHI / Satellite GHI ~ Clearsky Index⁻²

$$\frac{GHI_M}{GHI_S} = a_0 CI^{-2} + a_1 GHI_S$$

Plus a CDF correction to reshape the final adjusted data

Error vs Clearsky Index with fitted relationship

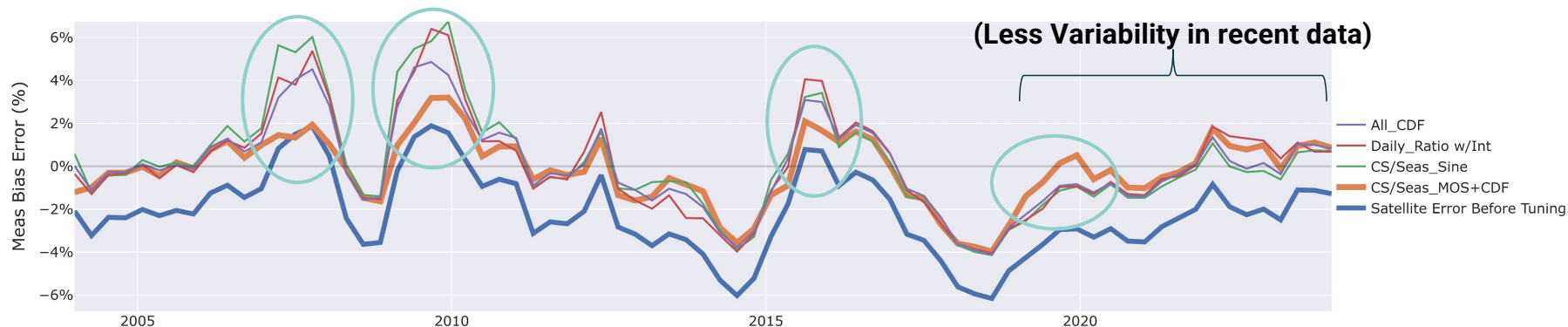


Satellite and Tuned Data vs Measured



Results of One Location Across Time for MOS, CDF, Ratio, and Sinusoidal

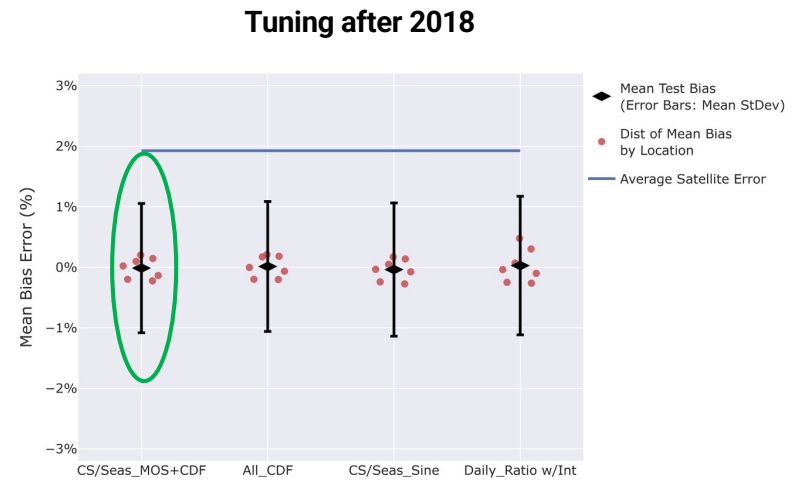
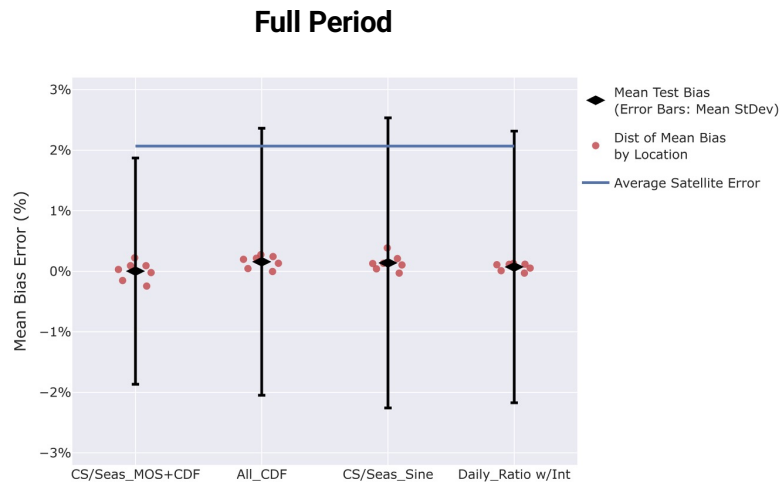
The tuning methods are subject to error in the training period *relative* to the error in the rest of the data
The MOS method shows some robustness to atypically high errors, and does not carry the full error through



These days we are only tuning on recent data, testing recent data is what should be relevant to our tuning outcomes



Comparison of Full Period Results with Results after 2018



Standard Deviation is approximately half using more recent data

MOS statistics from tests:

Mean standard deviation (across tests, temporal): 1.1%

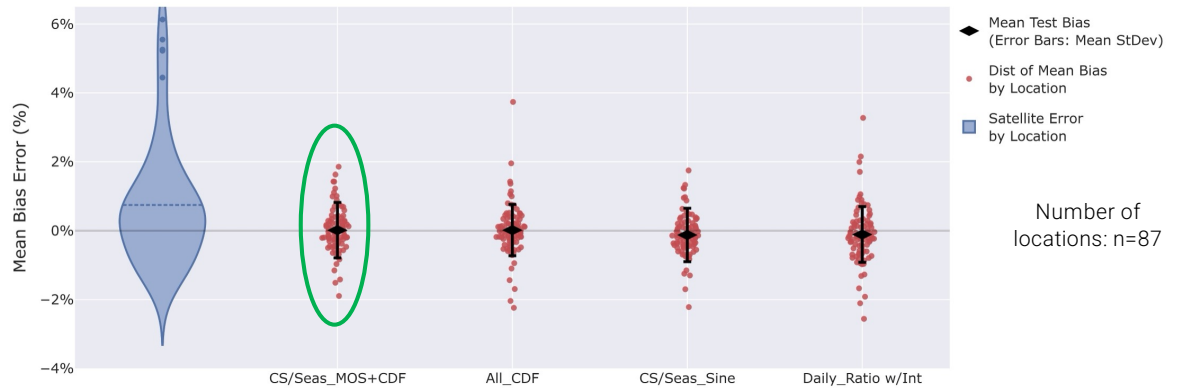
Standard deviation of means (across locations, spatial): 0.2%



Measured Data from Ground Stations

87 Locations with 2+ years of data

Satellite Errors shown with test results from
 Model Output Statistics (MOS)
 Cumulative Distribution Function (CDF)
 Sinusoidal
 Daily Ratio with Intercept



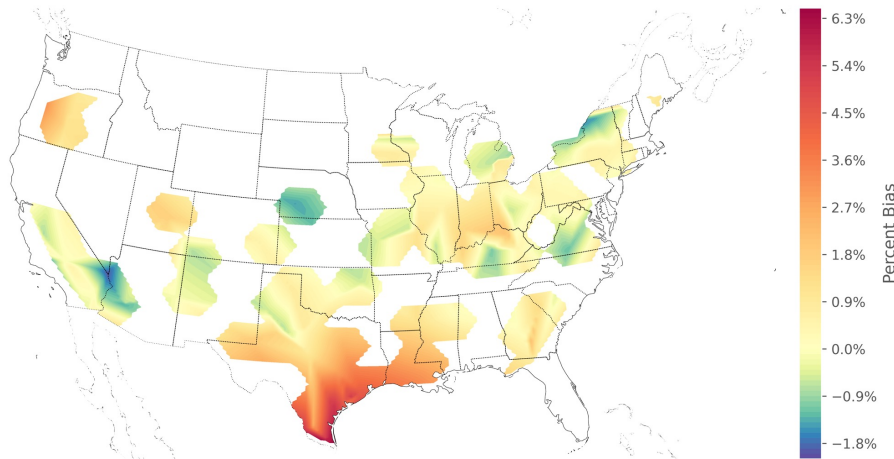
MOS method is among the best performers:
Near 0 bias (6th lowest)
Low standard deviation

MOS statistics from tests:
Mean standard deviation (across tests, temporal): 0.8%
Standard deviation of means (across locations, spatial): 0.7%



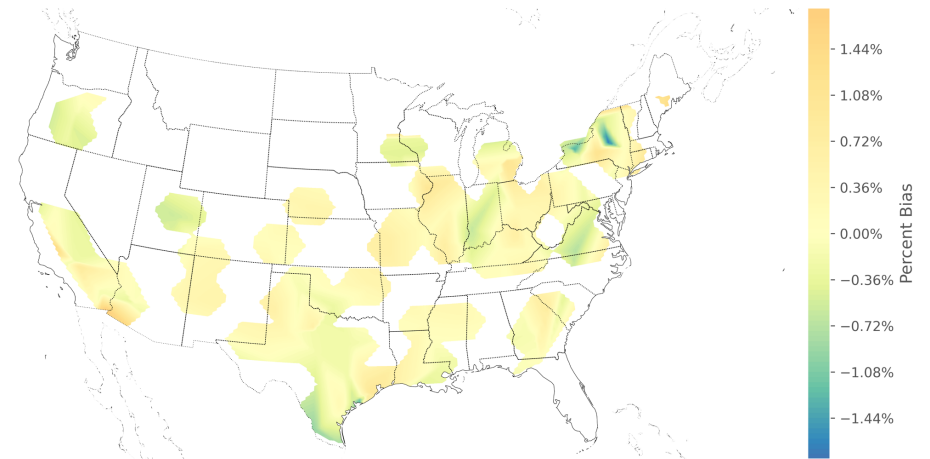
Comparison of Before and After Tuning with MOS by Location

Satellite Bias Before Tuning



NSRDB PSM v4
CONUS Range: -2.02% to +6.58%

Site-Adapted Bias After Tuning



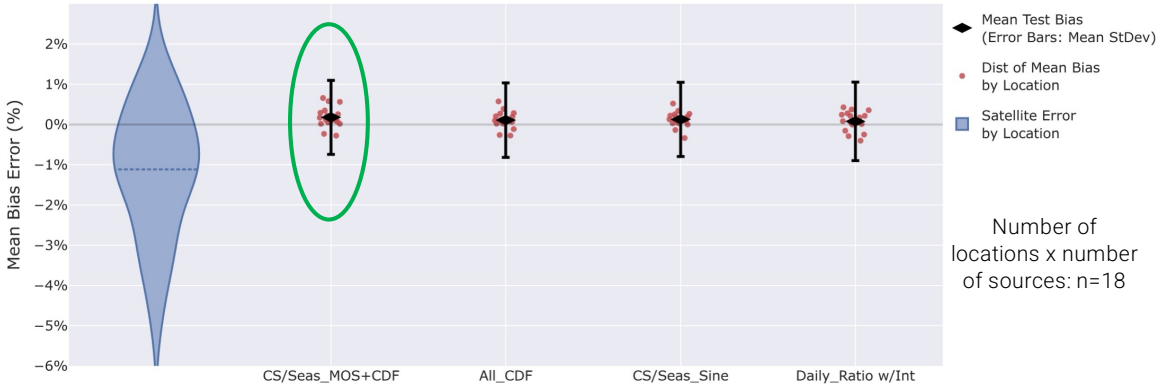
Adapted with CS/Seas MOS + CDF
CONUS Range: -1.89% to +1.85%



Does it work with the paid sources?

3 paid providers
 6 GroundWork station locations
 24.5 years of measured data (within the last 3-7 years)

Satellite Errors shown with test results from Model Output Statistics (MOS)
 Cumulative Distribution Function (CDF)
 Sinusoidal
 Daily Ratio with Intercept



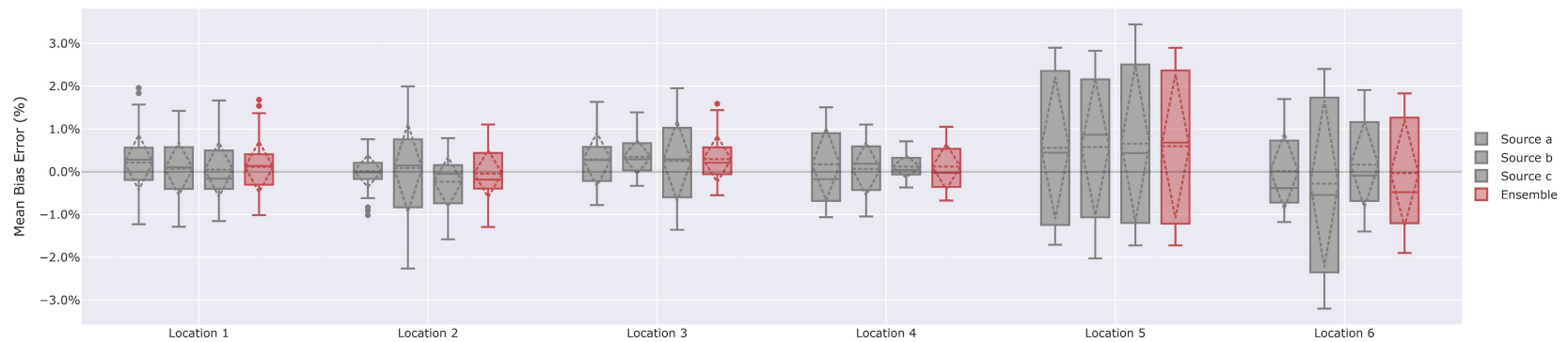
All methods achieve near 0% mean bias and mean standard deviation of about 0.9%

MOS statistics from tests:
Mean standard deviation (across tests, temporal): 0.9%
Standard deviation of means (across locations and sources): 0.3%



How Else Can We Improve Results – Ensemble Approach

Any one source can have error at times
We can mitigate that by tuning and merging multiple sources into the final target

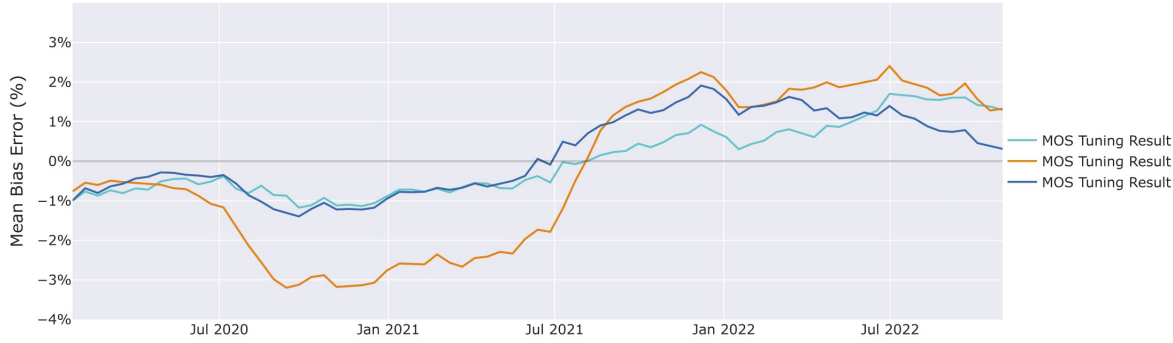


Uncertainty of individual tests: 0.95%
Uncertainty of Avg of tests: 0.88%

1. Tune All, find annual average GHI
2. Create a Typical "Tuned" Year (TTY?) using one complete dataset that targets the average GHI
3. Resulting P50 GHI with low uncertainty



Ensemble Tuning Example



Shows the tuning result in the train periods
Three Different Sources
Same Tuning Method: CS/Seas MOS +CDF



**Taking the average of tunings
results in less variability in the
outcome**



Uncertainty of GHI Values and Conclusions

Estimated Uncertainties of GHI, MOS Tuning Method (k=1)

SURFRAD Tune Full Period with NSRDB (7 Locations, ~25 years): 1.87%

SURFRAD Tune since 2018 with NSRDB (7 Locations, ~25 years): 1.08%

GR Stations with NSRDB (87 locations, ~2 - 7 years): 1.03%

GR Stations with Paid Sources (6 locations, ~3.5 - 7 years): 0.95%

GR Stations with Paid Sources, Ensemble (6 locations, ~3.5 - 7 years): 0.88%

Relevant statistics with large samples:

Across Years, Temporal Variability
(Mean standard deviation, SURFRAD tests)

1.1%

Across Locations, Spatial Variability
(Standard deviation of means, GroundWork Stations)

0.7%

Model Output Statistics (“MOS”) is a viable site adaptation method for low uncertainty and low bias error

Site-Adaptation works with about 1% - 1.25% uncertainty

Combining multiple sources in tuning reduces uncertainty



References

- Alfi, James, Alex Kubiniec, Ganesh Mani, James Christopherson, Yiping He, and Juan Bosch. "Importance of Input Data and Uncertainty Associated with Tuning Satellite to Ground Solar Irradiation." *2016 IEEE 43rd Photovoltaic Specialists Conference (PVSC)*, June 2016, 0301–5. <https://doi.org/10.1109/PVSC.2016.7749598>.
- Cebecauer, Tomas, and Marcel Suri. "Site-Adaptation of Satellite-Based DNI and GHI Time Series: Overview and SolarGIS Approach." *AIP Conference Proceedings* 1734, no. 1 (2016): 150002. <https://doi.org/10.1063/1.4949234>.
- Kankiewicz, Adam, John Dise, Elynn Wu, and Richard Perez. "Solar 2014: Reducing Solar Project Uncertainty with an Optimized Resource Assessment Tuning Methodology." In *American Solar Energy Society Annual Conference*, 2014. <https://research.asrc.albany.edu/people/faculty/perez/2014/K.pdf>.
- Mieslinger, Theresa, Felix Ament, Kaushal Chhatbar, and Richard Meyer. "A New Method for Fusion of Measured and Model-Derived Solar Radiation Time-Series." *Energy Procedia*, Proceedings of the 2nd International Conference on Solar Heating and Cooling for Buildings and Industry (SHC 2013), vol. 48 (January 2014): 1617–26. <https://doi.org/10.1016/j.egypro.2014.02.182>.
- Miller, Annalise, and Nate Croft. "What to Consider for Improved Accuracy and Reduced Uncertainty in Resource Measurement Campaigns." Poster. 2024 18th PV Performance Modeling Workshop, Salt Lake City, Utah, USA, May 7, 2023. https://www.sandia.gov/app/uploads/sites/243/dlm_uploads/2024/05/Miller_PVPMC2024_v3.pdf.
- Polo, J., S. Wilbert, J. A. Ruiz-Arias, et al. "Preliminary Survey on Site-Adaptation Techniques for Satellite-Derived and Reanalysis Solar Radiation Datasets." *Solar Energy* 132 (July 2016): 25–37. <https://doi.org/10.1016/j.solener.2016.03.001>.
- Tafur, Lucila D., Renn Darawali, and Halley Darling. "Expansion of Common Measure-Correlate-Predict Analysis Considerations." 2023 16th PV Performance Modeling Workshop, Salt Lake City, Utah USA, May 9, 2023. https://www.sandia.gov/app/uploads/sites/243/dlm_uploads/2023/06/UL_Solutions_PVPMC_MCP_2023-05-04_Tafur.pdf.
- Tahir, Zia ul Rehman, Muhammad Asim, Muhammad Azhar, Ghulam Moeenuddin, and Muhammad Farooq. "Correcting Solar Radiation from Reanalysis and Analysis Datasets with Systematic and Seasonal Variations." *Case Studies in Thermal Engineering* 25 (June 2021): 100933. <https://doi.org/10.1016/j.csite.2021.100933>.



GroundWork Advisory Services

Expert technical guidance for every stakeholder in renewable energy—whether you're financing projects, buying/selling, or building them.

Independent Engineering Services

Independent technical validation for lenders and investors across development, construction, and operations.

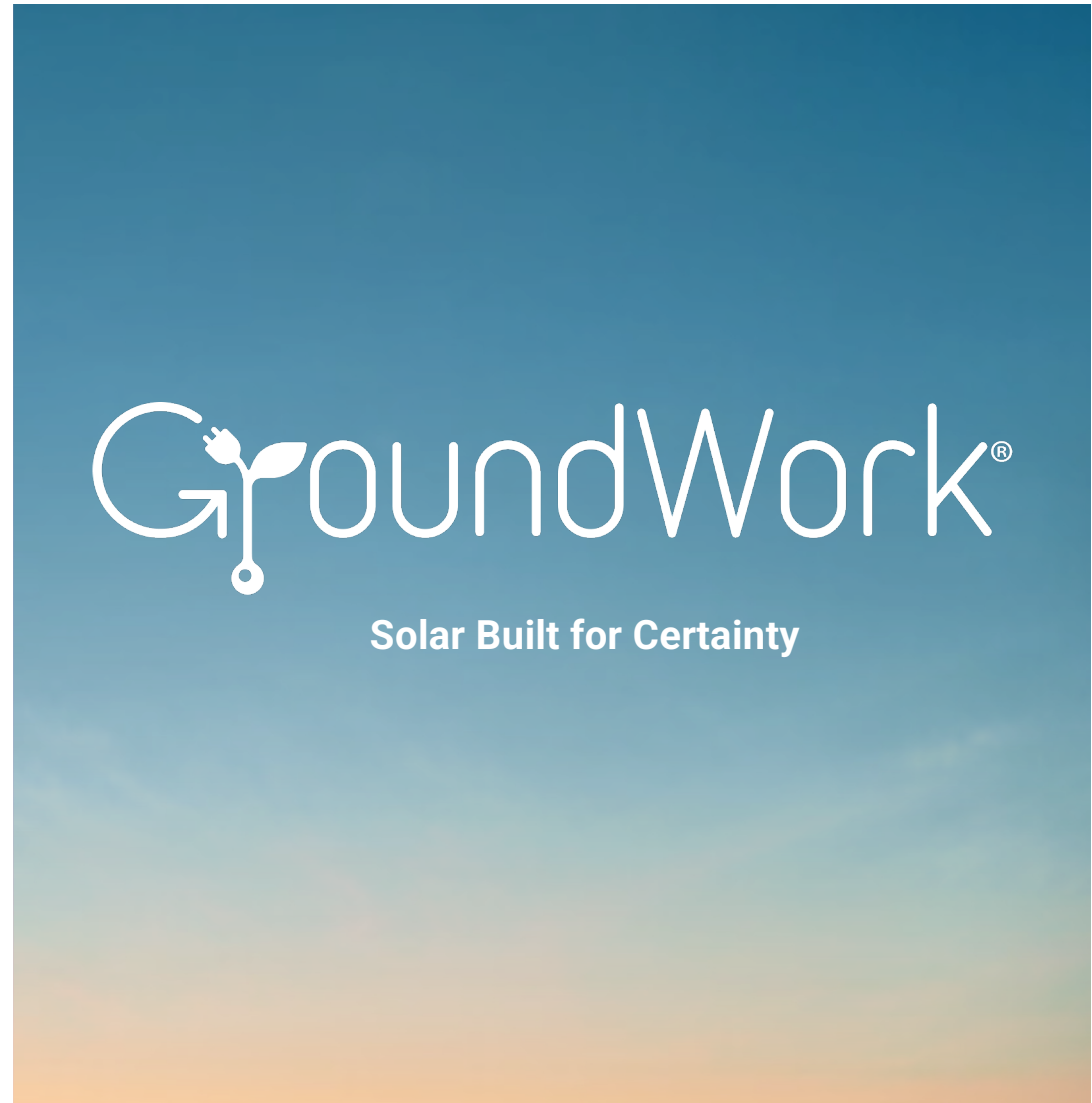
Owner's Engineering Services

Technical advocacy and decision support for asset owners and developers throughout the project lifecycle.

M&A Due Diligence Services

Performing limited technical due diligence in support of buy/sell transactions for assets at any stage.

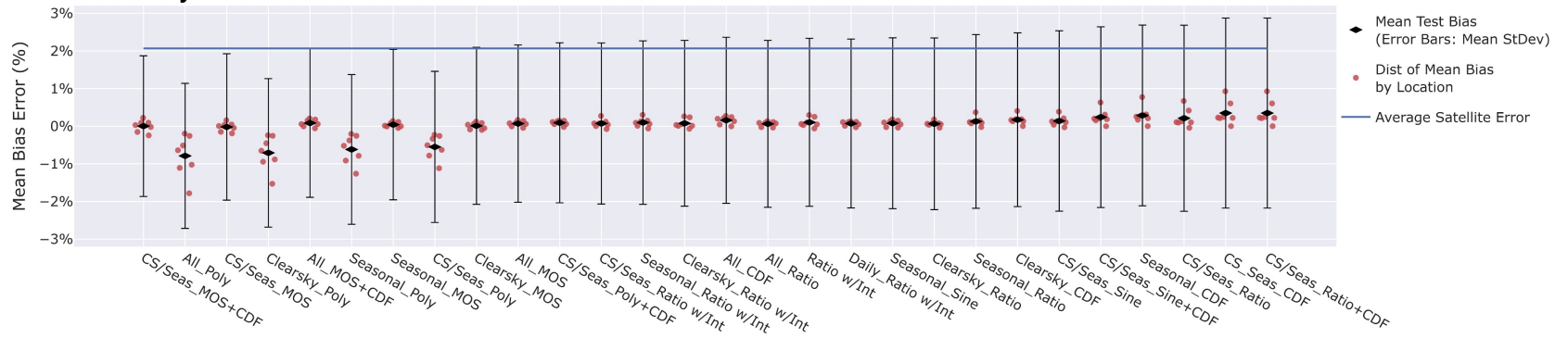
Thank You!
ncroft@grndwork.com



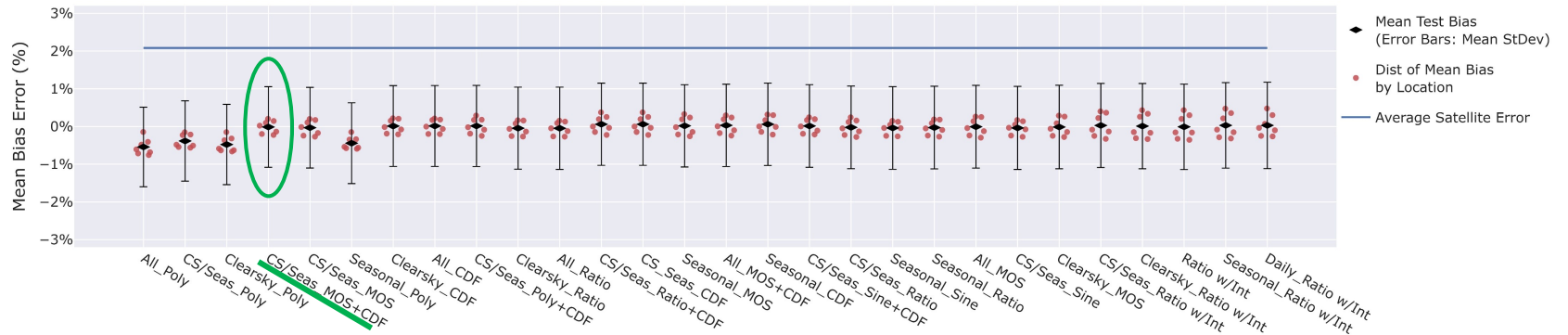
Comparison of Full Period Results with Results after 2018

Sorted By Mean Standard Deviation – Lowest On Left

Full Period



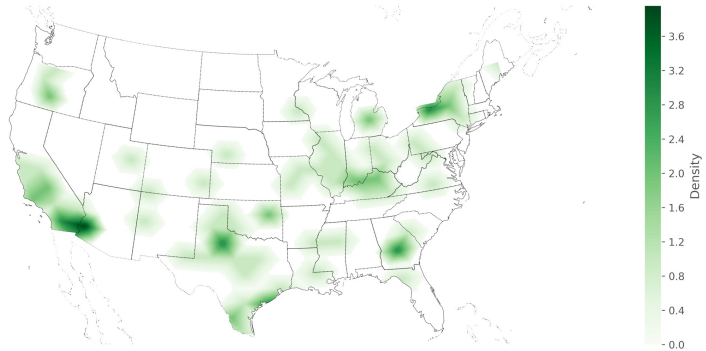
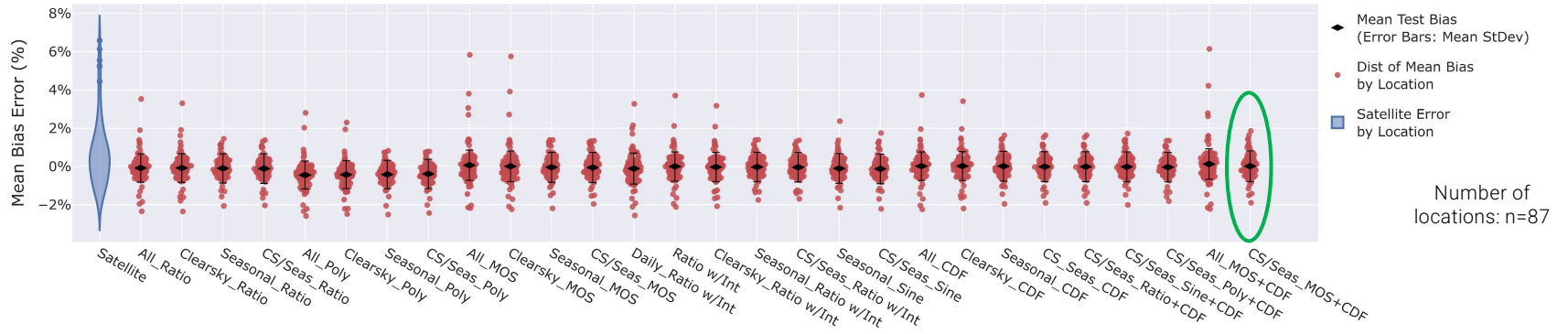
Tuning after 2018



Standard Deviation is approximately half on more recent data



Measured Data from Ground Stations



The MOS method is among the best performers

Near 0 bias (6th lowest)
Low standard deviation



Does it work with the paid sources?

3 paid providers
 6 GroundWork station locations
 24.5 years of measured data (within the last 3-7 years)

