



From point to area:

assessing the representativeness of point observations

Matthias **Schwarz**^(*,1), Doris Folini⁽¹⁾, Maria Hakuba^(2,3), Martin Wild⁽¹⁾

matthias.schwarz@env.ethz.ch

Affiliations:

- (1) ETHZ Zürich, Institute for Atmospheric and Climate Science
- (2) Department of Atmospheric Sciences, Colorado State University
- (3) Jet Propulsion Laboratory, California Institute of Technology

Why representativeness?







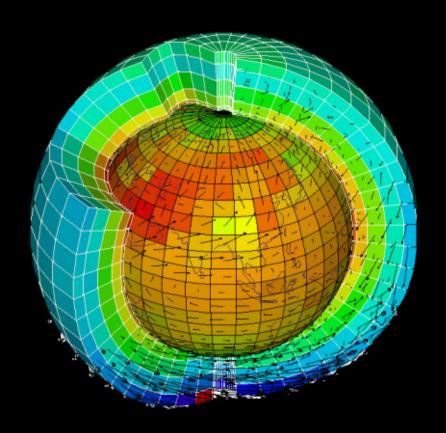


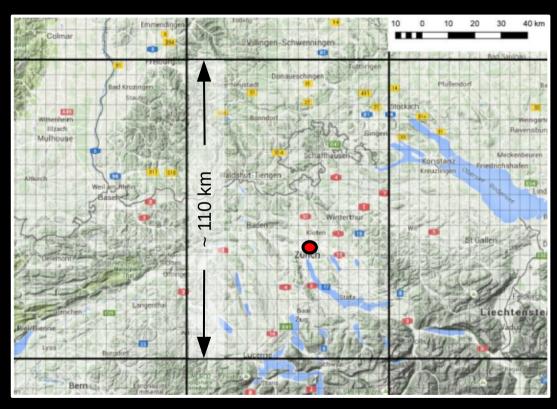




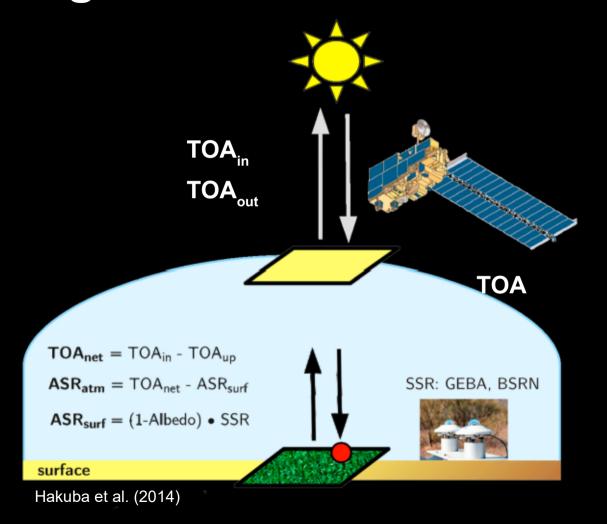


Why representativeness?





Combining TOA and surface observations



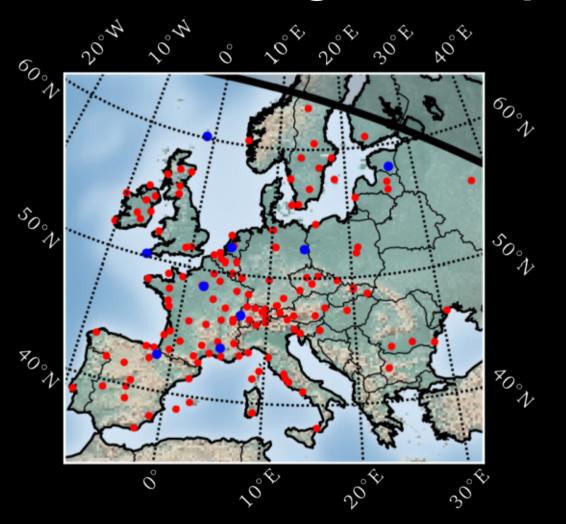
Is a single monthly mean surface solar radiation (SSR) time series representative for a 1° gridbox?



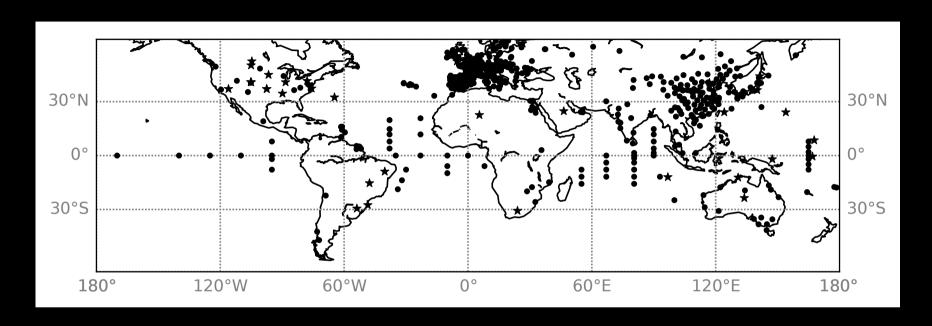
The Adjustic Applications of the Adjustic App

Observation with Pyranometer

Station coverage – Europe



Station coverage

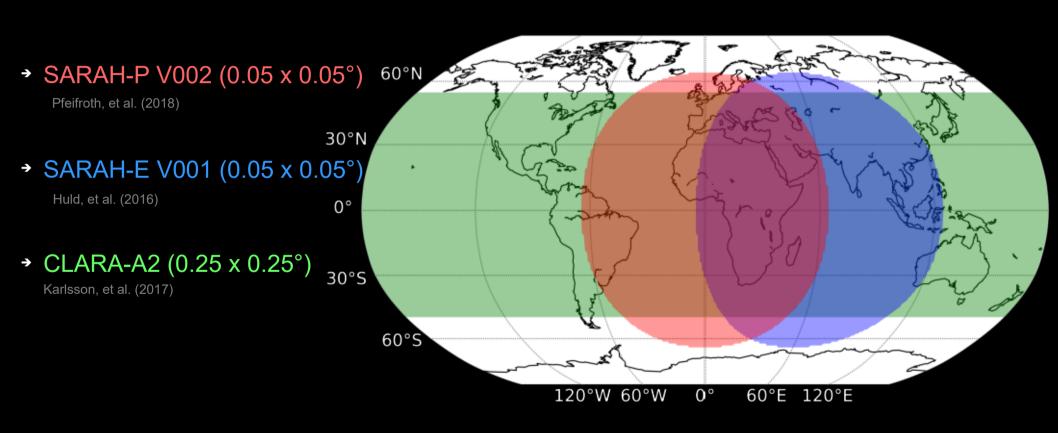


Too few stations available for global analysis

- → Using satellite derived surface radiation as surrogate for in-situ obs.
- → allows near global assessment of representativeness

CM-SAF's High Resolution Satellite-derived SSR Data

Global scale analysis with **monthly mean** satellite derived SSR from CM-SAF







From point to area:

assessing the spatiotemporal representativeness of monthly in-situ global radiation records from CM-SAF SIS data.

Matthias **Schwarz**^(*,1), Doris Folini⁽¹⁾, Maria Hakuba^(2,3), Martin Wild⁽¹⁾ matthias.schwarz@env.ethz.ch

Affiliations:

- (1) ETHZ Zürich, Institute for Atmospheric and Climate Science
- (2) Department of Atmospheric Sciences, Colorado State University
- (3) Jet Propulsion Laboratory, California Institute of Technology

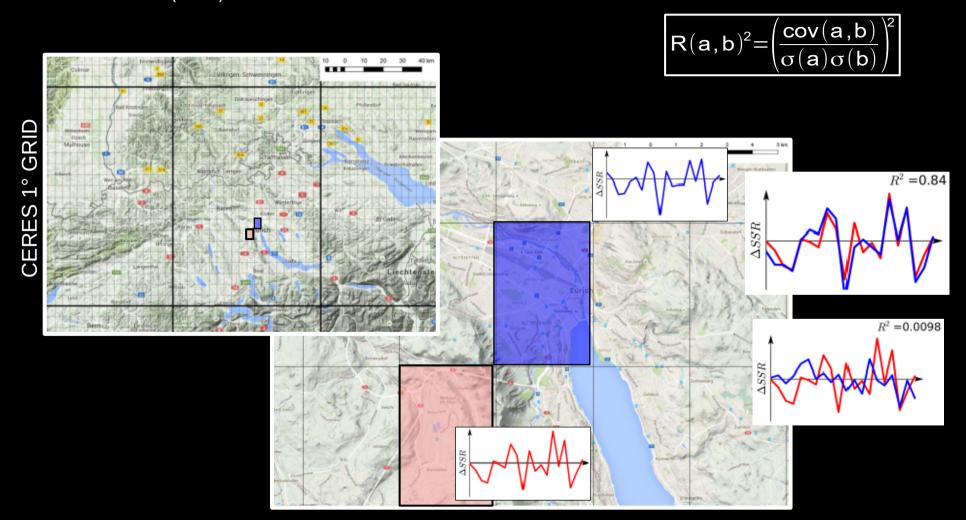
Three aspects of representativeness:

- I. Spatial correlations (R²)
- II. Spatial Sampling Biases (β)
- III.Spatial Sampling Errors (ε)

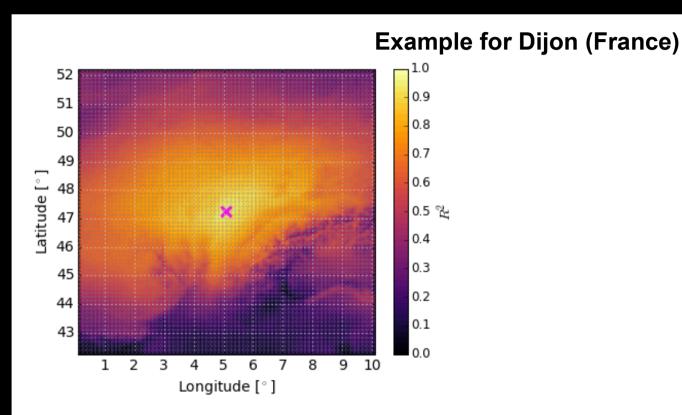


Spatial Correlations

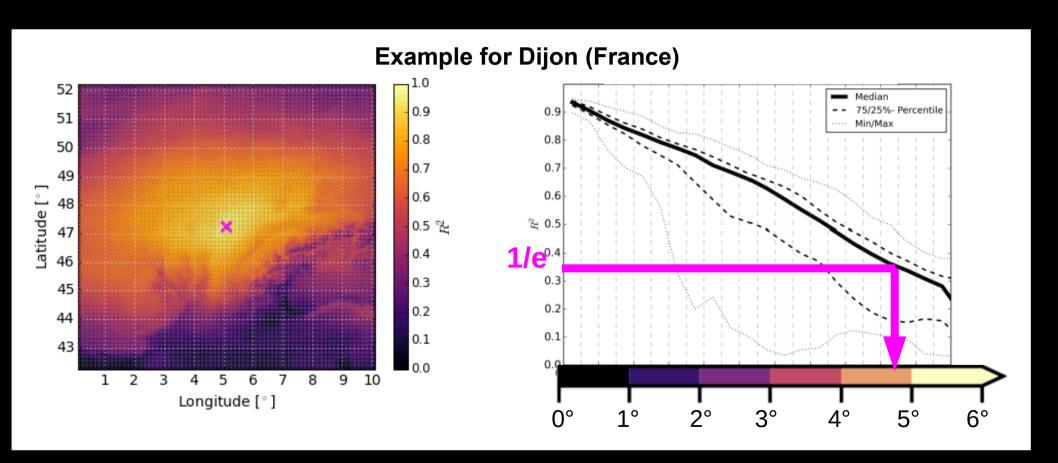
Schwarz et al. (2017)



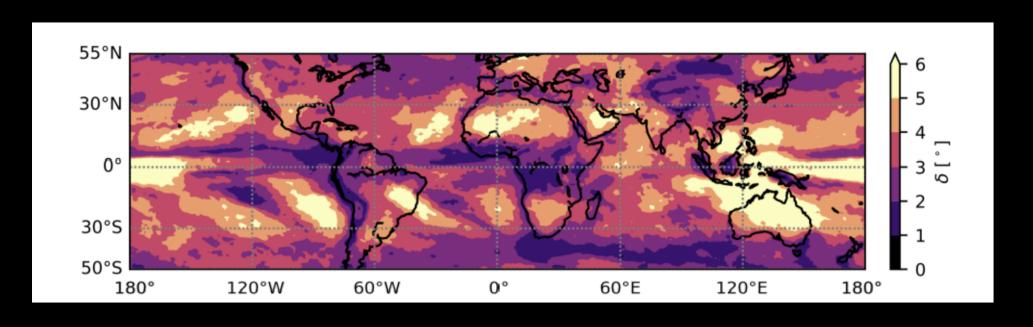
Spatial Correlations: Decorrelation Length (δ)



Spatial Correlations: Decorrelation Length (δ)



Spatial Correlations: Decorrelation Length (δ)



- Near-global (50S-55N) mean δ ≈ 3.4°
- Roughly
 - − \sim 2% of 1° boxes have average δ < 1°
 - − ~5% of 1° boxes have average δ < 2°

Combination of SSR from point observations with 1° gridded data is feasible in most regions!

Three aspects of representativeness:

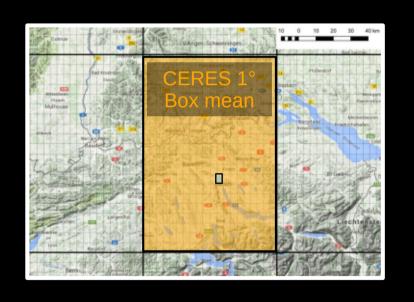
I. Spatial correlations (R²)

- Grid independent metric
- II.Spatial Sampling Biases (β)
- III.Spatial Sampling Errors (ε)

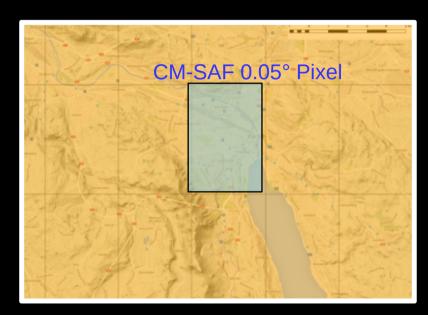
Grid dependent metrics



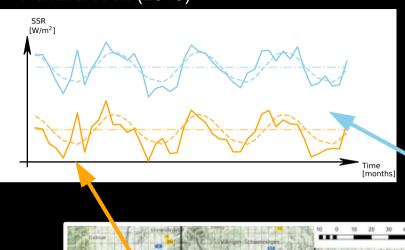
Grid dependent metrics

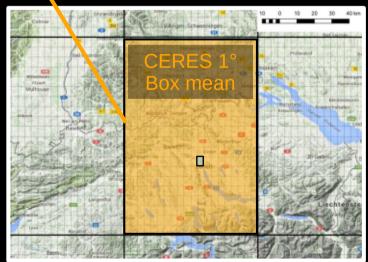


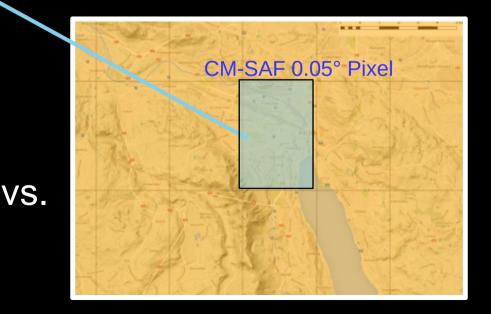
VS.



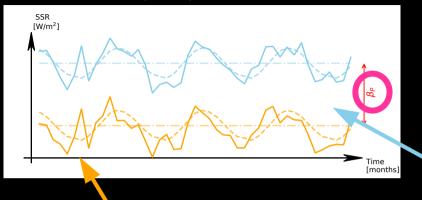
Spatial Sampling Bias (β) Hakuba et al. (2013)



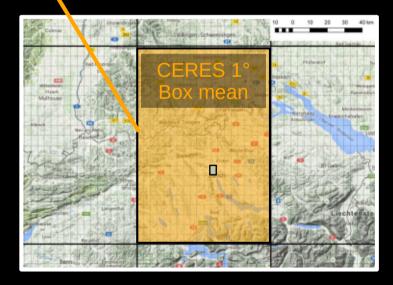




Spatial Sampling Bias (β) Hakuba et al. (2013)



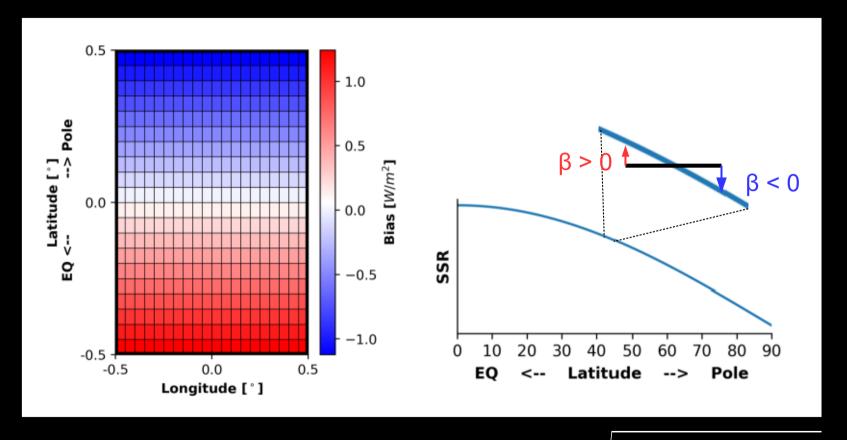
$$\beta_{P} = \overline{SSR_{P}} - \overline{SSR_{B}}$$



VS.

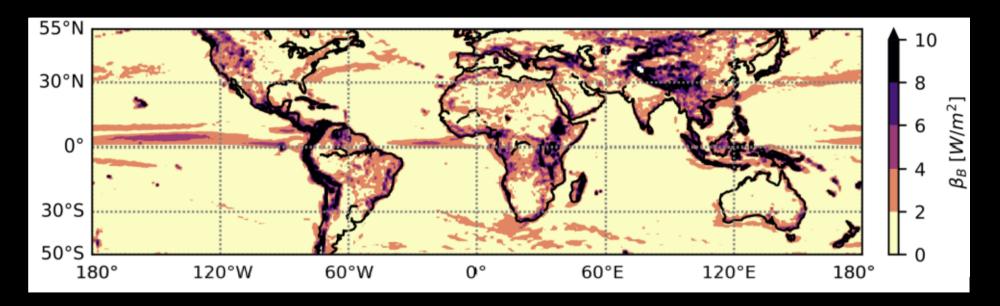


Spatial Sampling Biases – Pixel Based



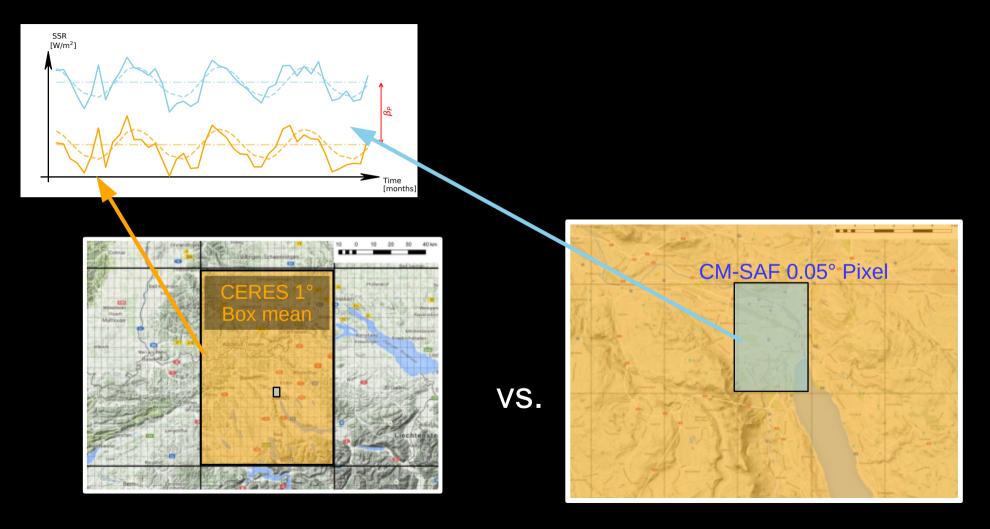
$$\beta_{\mathsf{B}} = \sqrt{\frac{1}{\mathsf{N}}} \sum_{\mathsf{N}} (\beta_{\mathsf{P}} - \overline{\beta_{\mathsf{P}}})$$

Spatial Sampling Biases – Box Aggregated

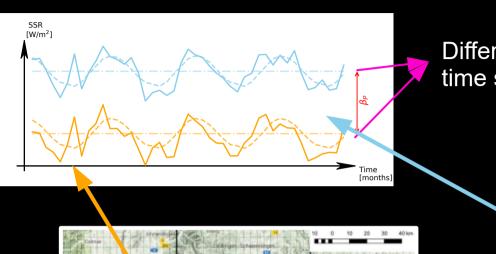


- Near-global (50S-55N) β_R ≈1.4 W/m²
- Magnitudes of biases vary across regions
- Bias of station depends on position within 1° box
- Biases can be corrected (if known)
- (biases have annual cycle)

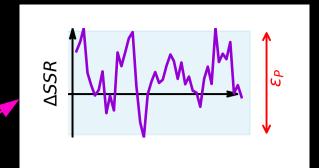
Spatial Sampling Error (ε)



Spatial Sampling Error (ε)

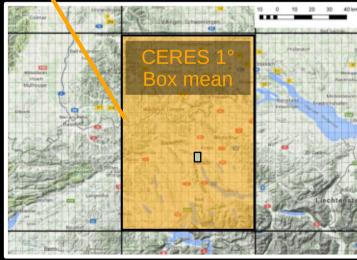


Difference / time series

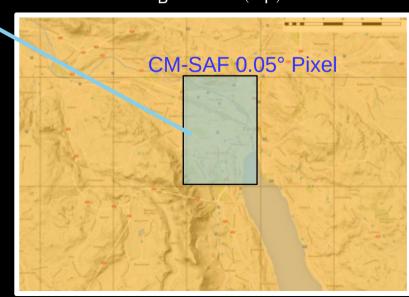


$$\epsilon_{P} = P^{95}(|SSR'_{P}(t) - SSR'_{B}(t)|)$$

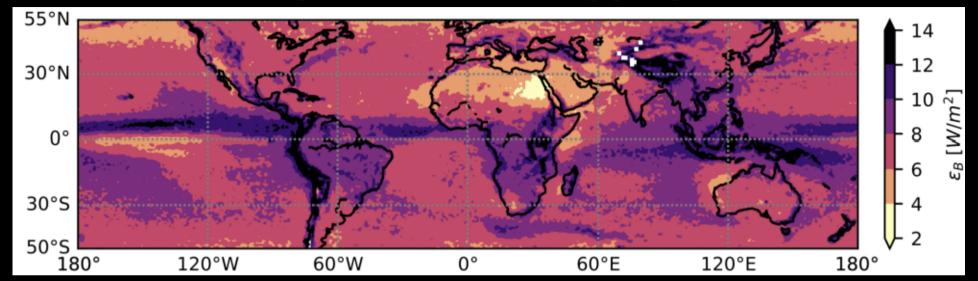
$$\epsilon_{\mathrm{B}}{=}\mathsf{P}^{68.2}(\epsilon_{\mathrm{P}})$$



VS.



Spatial Sampling Errors – Box Aggregated



- Global mean (50S-55N) $\varepsilon_{\rm B} \approx 7.5 \text{ W/m}^2$
- Errors are calculated form individually deseasonalized time series
 - → implicit bias correction

- Without bias correction errors are 10-15% higher
- Errors for other grids:
 - 0.5°x 0.5° grid ~ 30% smaller
 - 2.5° x 2.5° grid ~ 60% larger

Three aspects of representativeness:

I. Spatial correlations (R²)

Grid independent metrics

II.Spatial Sampling Biases (β)

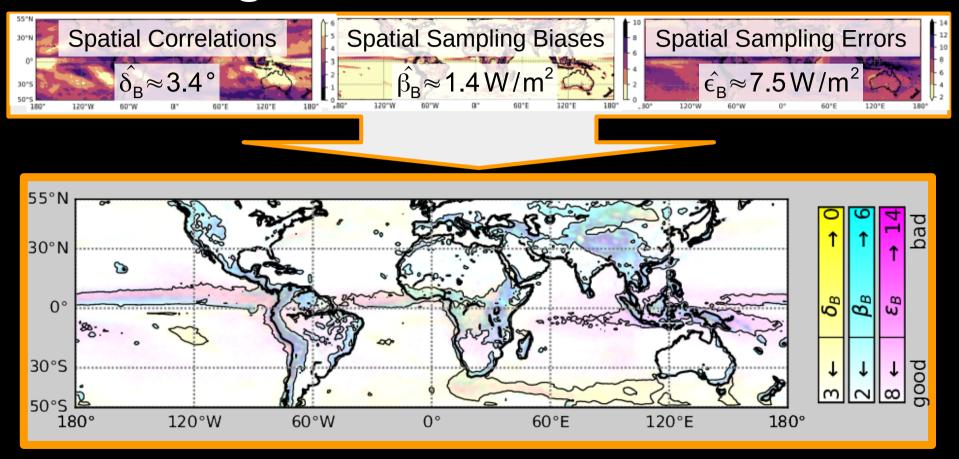
III.Spatial Sampling Errors (ε)

Grid dependent metrics

Let's combine all metrics.....



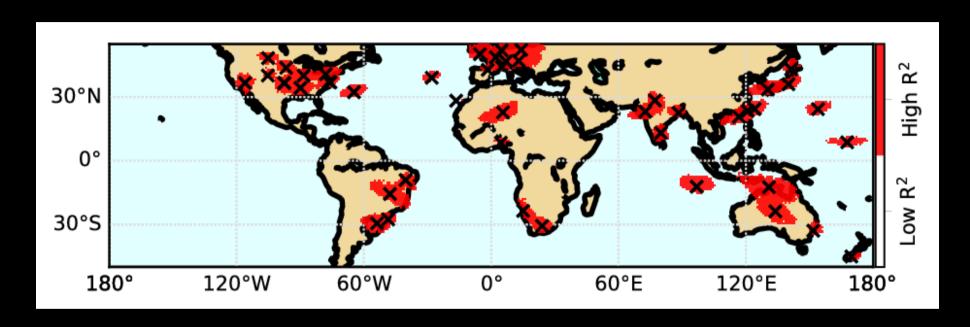
Combining the metrics



→ Different metrics limit representativeness in different regions

Case Study: Direct sampling capacity of the Baseline Surface Radiation Network (monthly mean SSR)

Case Study BSRN



The 47 BSRN stations inside domain can together directly ($R^2 > 1/e$) sample

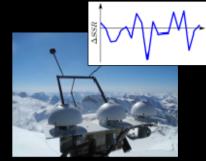
- 16% of the domains land pixels
- 7% of the domains total pixels

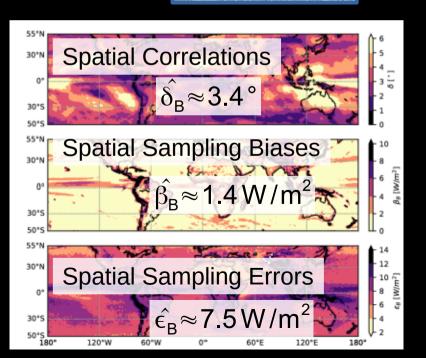
$$\begin{split} \left<\delta\right>_{\text{BSRN}} &\approx 3.5^{\circ} \\ \left<\beta_{\text{B}}\right>_{\text{BSRN}} &\approx 3.7 \text{ W/m}^2 \qquad \left<|\beta_{\text{P}}|\right>_{\text{BSRN}} \approx 2.9 \text{ W/m}^2 \\ \left<\epsilon_{\text{B}}\right>_{\text{BSRN}} &\approx 8.6 \text{ W/m}^2 \qquad \left<\epsilon_{\text{P}}\right>_{\text{BSRN}} \approx 8.9 \text{ W/m}^2 \end{split}$$

Thave Sense Bosts Bost Style Bost

Synthesis







- Combining point and (1°) gridded data is possible in most regions
- Grid specific bias correction is advisable
- Combined uncertainty (1° grid):

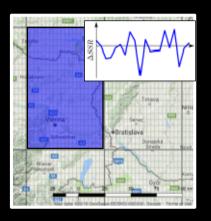
Measurement uncertainty

+ spatial sampling error (ε)

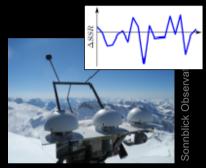
Total uncertainty ~40-50% higher than measurement uncertainty alone

- Large regional differences!
- Representativeness is limited in different regions due to different reasons!

Summary







- Correcting β is suggested!
- Uncertainty increases (pyranometer + ε)
- Large regional differences







Schwarz, M., Folini, D., Hakuba, M.Z., Wild, M., 2018. From Point to Area: Worldwide Assessment of the Representativeness of Monthly Surface Solar Radiation Records. J. Geophys. Res. Atmos. 123, 13,857-13,874. https://doi.org/10.1029/2018JD029169

Schwarz, M., Folini, D., Hakuba, M.Z., Wild, M., 2017. **Spatial Representativeness of Surface-Measured Variations of Downward Solar Radiation**. J. Geophys. Res. Atmos. 122, 2017JD027261. https://doi.org/10.1002/2017JD02726

Appendix

Pixel as Surrogate for Point Observation

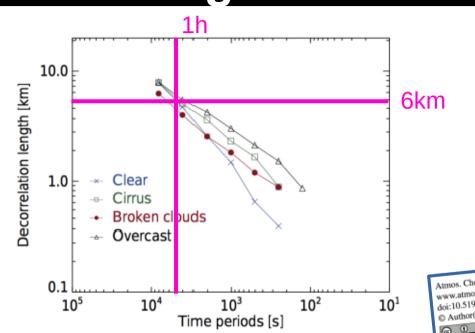


Figure 7. Decorrelation lengths a (in km), determined as e-fo time of the spatial correlation function, and its dependence o time period of variations.

Atmos. Chem. Phys., 17, 3317-3338, 2017 www.atmos-chem-phys.net/17/3317/2017/ doi:10.5194/acp-17-3317-2017 Author(s) 2017. CC Attribution 3.0 License.



Multiresolution analysis of the spatiotemporal variability in global radiation observed by a dense network of 99 pyranometers

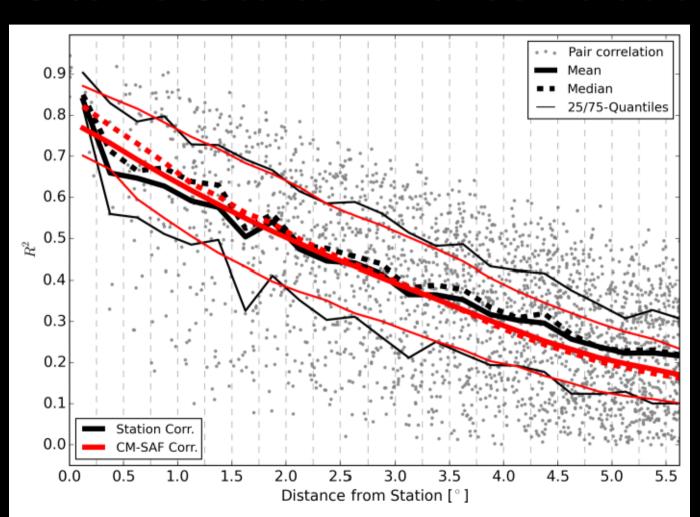
Bomidi Lakshmi Madhavan^{1,6}, Hartwig Deneke¹, Jonas Witthuhn¹, and Andreas Macke¹

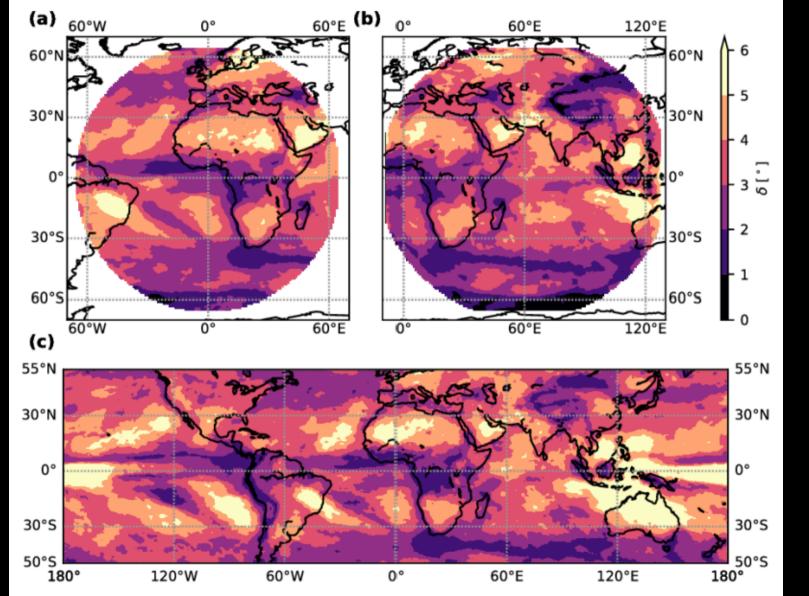
¹Leibniz-Institute for Tropospheric Research (TROPOS), Permoserstraße 15, 04318 Leipzig, Germany

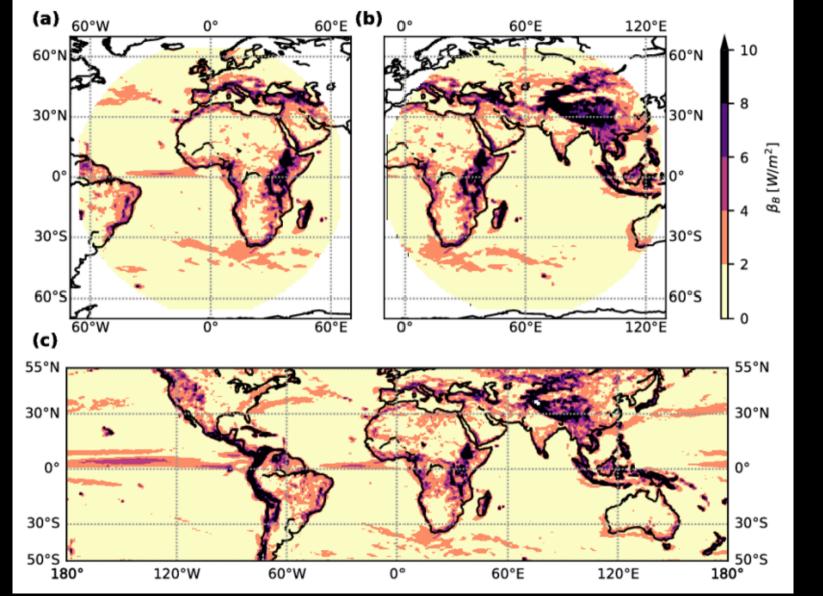
*now at: Department of Marine Sciences, Goa University, Goa 403 206, India

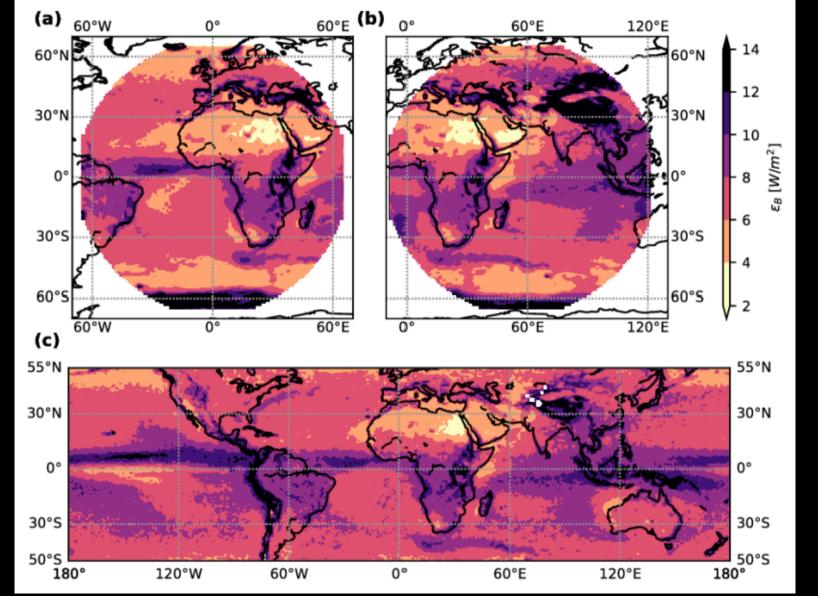
Correspondence to; Bomidi Lakshmi Madhavan (madhavan.bomidi@tropos.de, blmadhavan@gmail.com)

Site-to-Site vs Site-to-Pixel correlations

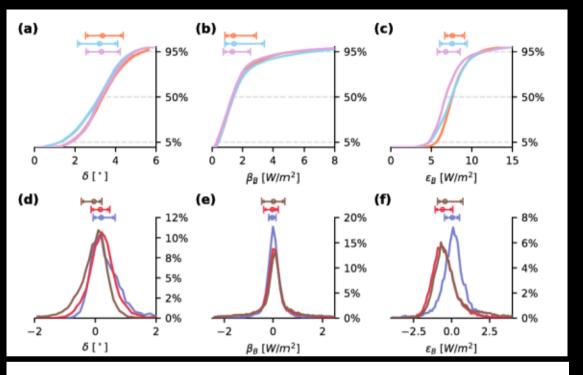








Comparison (CDFs and PDFs) of δ , β , and ϵ from SARAH, SARAH-E, and **CLARA**



	$\delta_B[^\circ]$	$\beta_B [\mathrm{Wm}^{-2}]$	$\epsilon_B [\mathrm{Wm}^{-2}]$
SARAH-P	3.31 (2.41, 4.36)	1.33 (0.57, 2.69)	6.96 (5.56, 9.06)
SARAH-E	3.21 (2.00, 4.23)	1.43 (0.66, 3.61)	7.75 (5.82, 9.95)
CLARA	3.36 (2.39, 4.50)	1.39 (0.67, 3.08)	7.58 (6.32, 9.39)
SARAH-P - SARAH-E	0.20 (-0.16, 0.74)	-0.03 (-0.28, 0.23)	0.05 (-0.63, 0.70)
SARAH-P - CLARA	0.17 (-0.22, 0.56)	-0.02 (-0.49, 0.33)	-0.41 (-1.08, 0.45)
SARAH-E - CLARA	-0.05 (-0.55, 0.30)	0.02 (-0.57, 0.59)	-0.23 (-0.91, 1.08)