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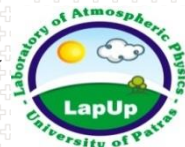


Validation of satellite solar irradiance estimates: separating clear-sky and cloud effect

L. Vuilleumier¹⁾, A. Meyer¹⁾, R. Stöckli¹⁾, S. Wilbert²⁾, L.F. Zarzalejo³⁾

1. Federal Office of Meteorology and Climatology, MeteoSwiss
2. German Aerospace Centre, Institute of Solar Research
3. Centre for Energy, Environment and Technological Research CIEMAT

15th BSRN Scientific Review and Workshop
19.07.2018 Boulder, USA





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Validation of satellite solar irradiance estimates: separating clear-sky and cloud effect

1. Introduction
2. Clear-sky estimates
3. All-sky estimates
4. Conclusions



Motivations

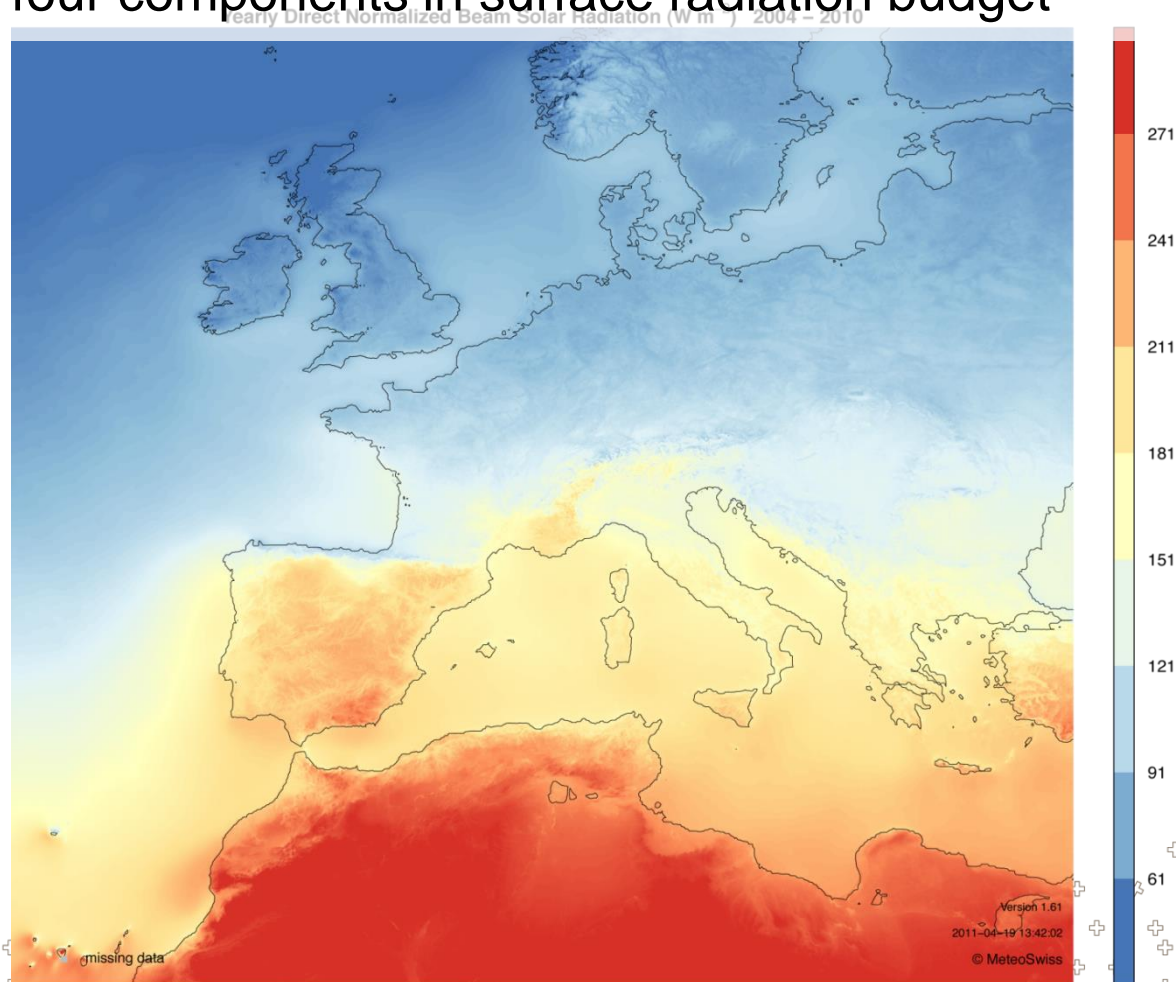
introduction

Solar radiation estimates obtained from satellite data:

1. Climate: one of four components in surface radiation budget

Yearly Direct Normalized
Beam Solar Radiation (Wm^{-2})
2004-2010
from Heliomont (Stöckli, 2013)
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Stöckli, R. (2013), The Heliomont
Surface Solar Radiation Processing,
Scientific Report MeteoSwiss, 93,
Editor: Federal Office of Meteorology
and Climatology MeteoSwiss



MeteoSwiss

Validation of satellite solar irradiance

15th BSRN meeting, 19.07.2018

L. Vuilleumier

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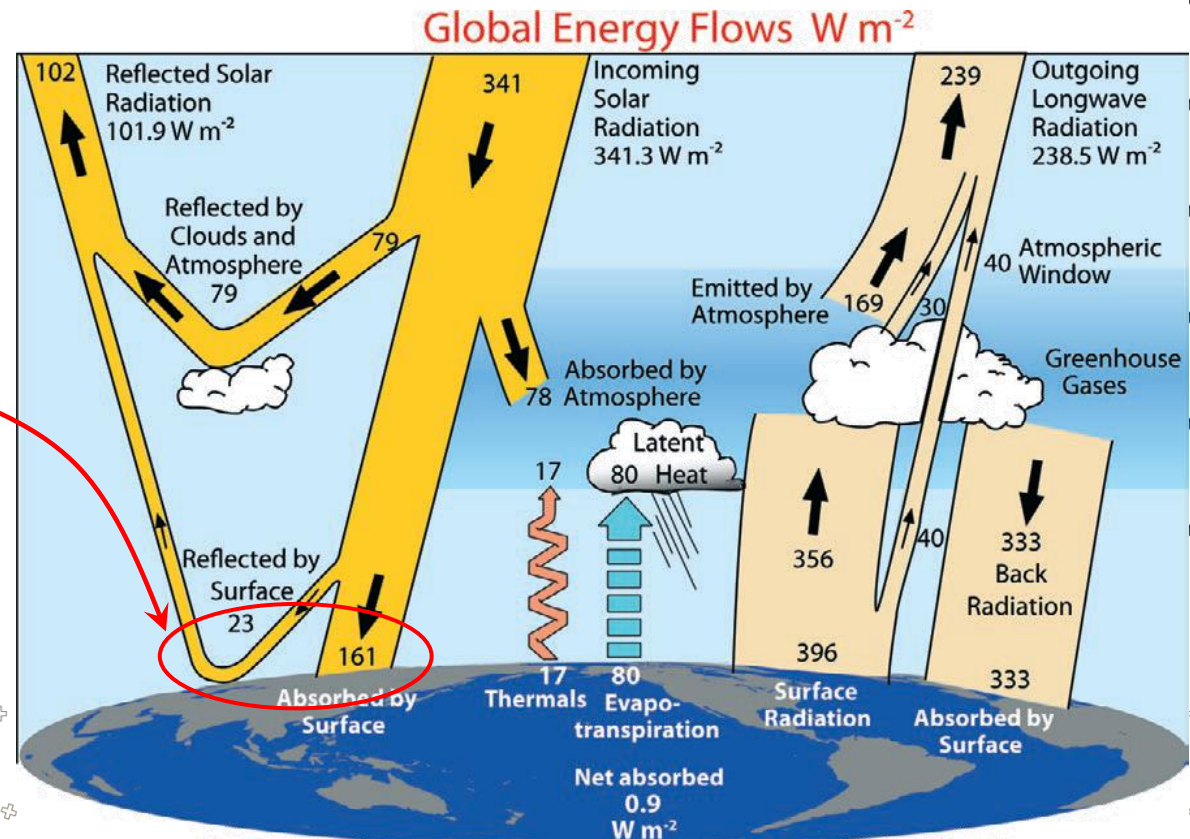
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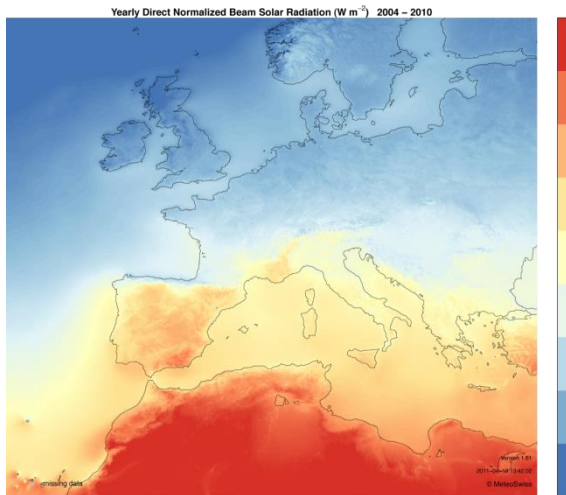


Motivations

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Solar radiation estimates obtained from satellite data:

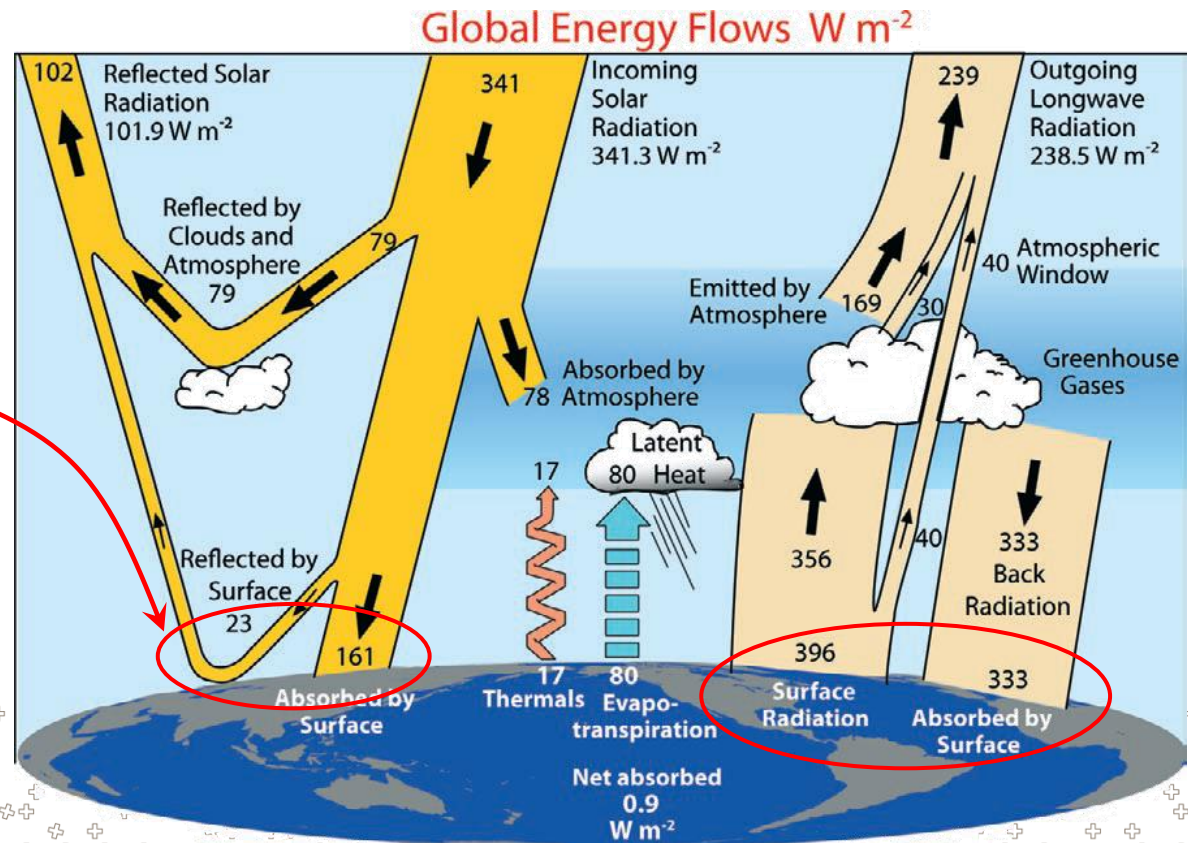
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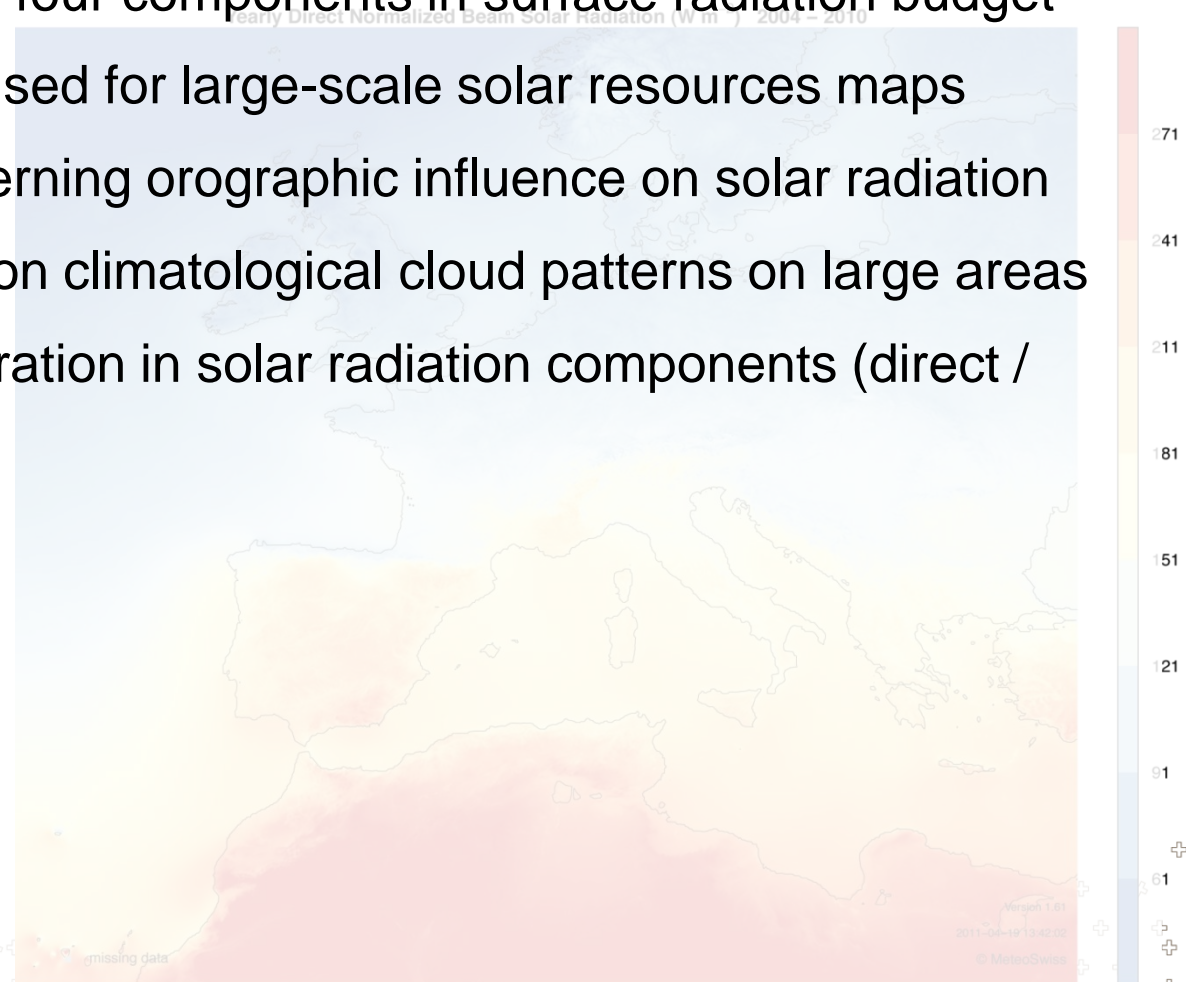


Motivations

introduction

Solar radiation estimates obtained from satellite data:

1. Climate: one of four components in surface radiation budget
2. Solar energy: used for large-scale solar resources maps
 - Details concerning orographic influence on solar radiation
 - High resolution climatological cloud patterns on large areas
 - Explicit separation in solar radiation components (direct / diffuse)



Motivations

introduction

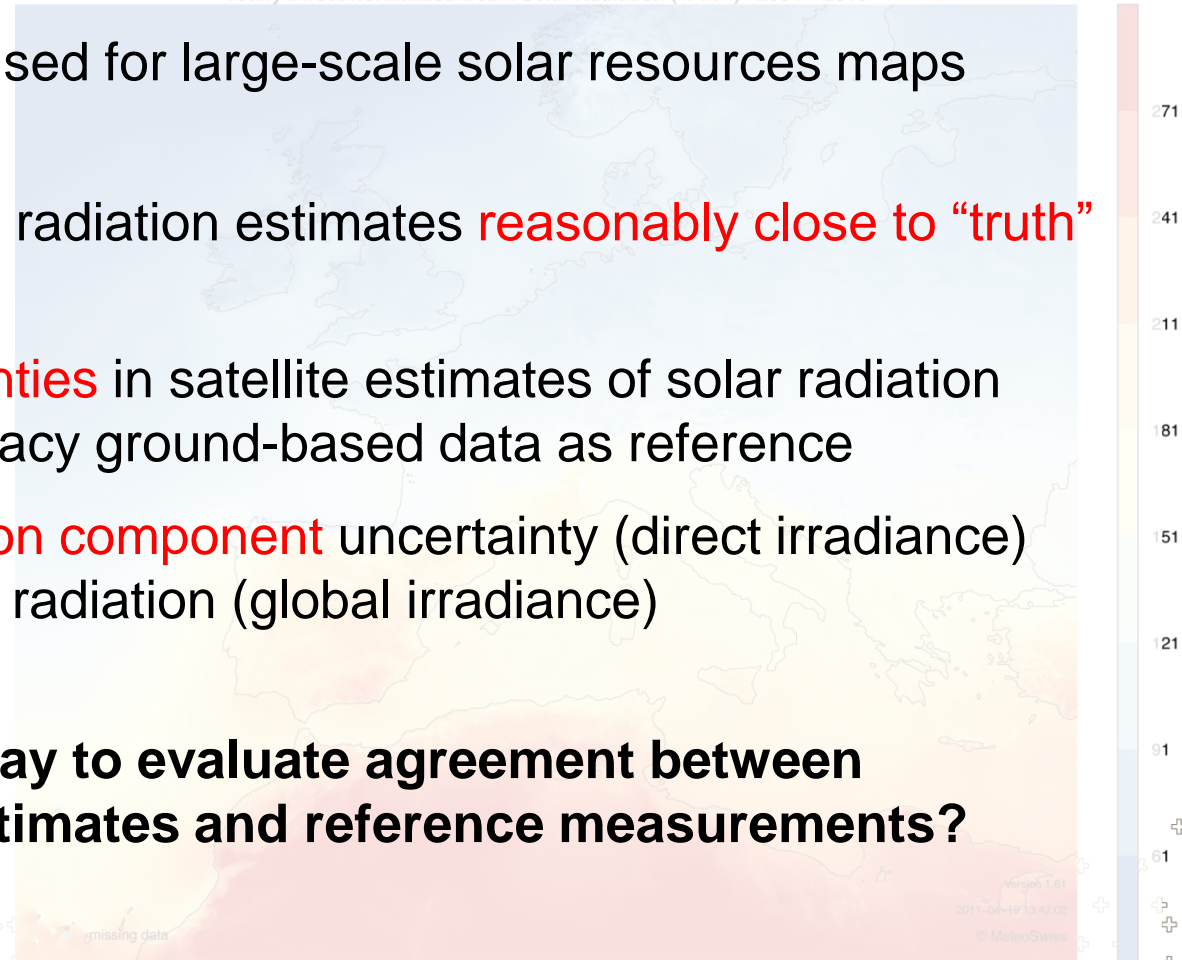
Solar radiation estimates obtained from satellite data:

1. Climate: one of four components in surface radiation budget
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Useful only if satellite radiation estimates **reasonably close to “truth”**

- Assess **uncertainties** in satellite estimates of solar radiation using high accuracy ground-based data as reference
- Look into **radiation component** uncertainty (direct irradiance) instead of global radiation (global irradiance)

Best way to evaluate agreement between satellite estimates and reference measurements?



Background

introduction

General scheme for radiation estimates
both sides should be tested separately

Clear-sky estimates

Radiative transfer model (RTM \rightarrow LUT)

Non-trivial input:

- **Aerosol** (AOD, SSA, α , ...)
- Surface Albedo (satellite)
- Water vapor column (ECMWF)
- O3 (ECMWF)



Cloud modification factor

Cloud macrophysical information
(e.g., satellite VIS & IR imagery)

Typically:

- Cloud cover
 - Cloud brightness
 - Cloud depth
- } ~ COD



physics understood
as good as input

physics limited in model
empirical and quite crude

Background

introduction

Solar zenith angle ✓

Clear-sky estimates

Radiative transfer model (RTM → LUT)

Non-trivial input:

- **Aerosol** (AOD, SSA, α , ...)
- Surface Albedo (satellite)
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- O₃ (ECMWF)



Altitude ✓
Cloud modification factor

Cloud macrophysical information
(e.g., satellite VIS & IR imagery)

Typically

Aerosols ← various DB ✓

- Cloud cover

- Cloud brightness

- Cloud depth

} ~ COD

Surface albedo ← sat ✓

H₂O + O₃ ECMWF ← sat ✓

Background

introduction

Cloud modification factor can be verified only if clear-sky in good agreement

Cloud modification factor

Cloud macrophysical information (e.g., satellite VIS & IR imagery)

Typically:

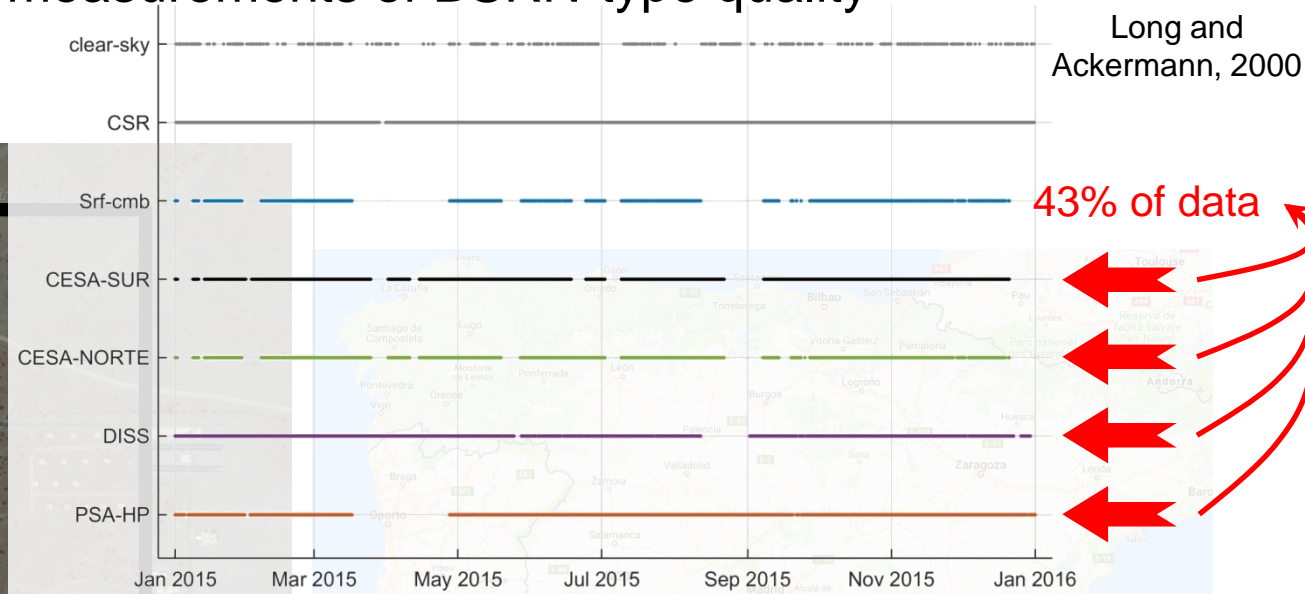
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Data & data availability

data

PSA allows using a combination of 4 pyrhelimeters for full year (2015)
DNI measurements spread on about 1 MSG SEVIRI pixel (visible)

1-min measurements of BSRN-type quality

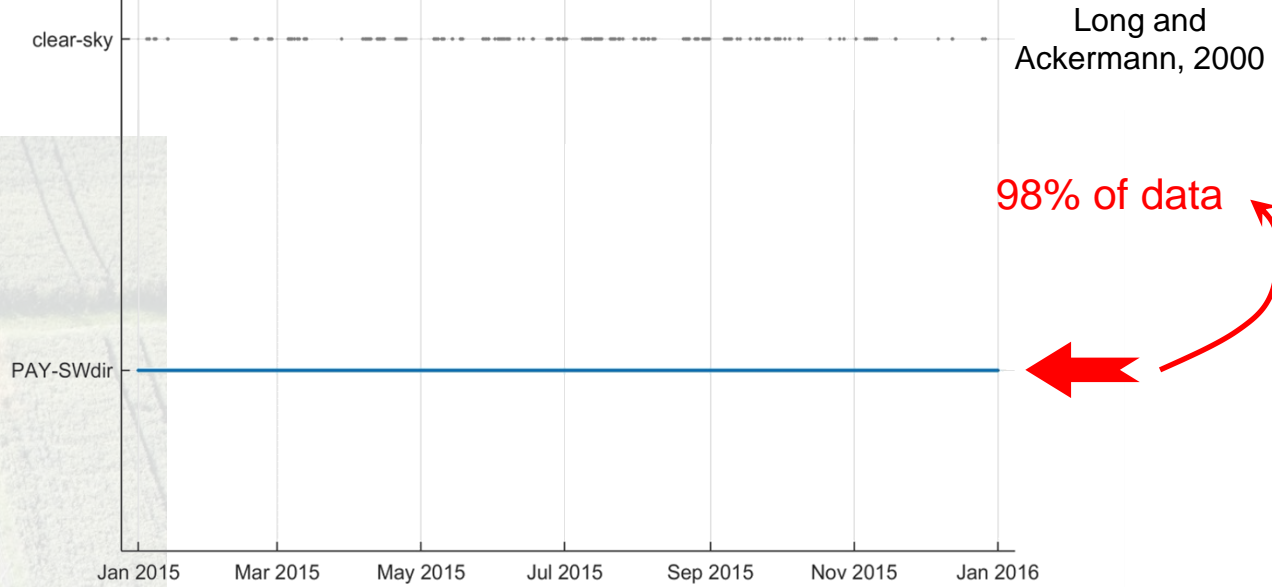


Data & data availability

data

Payerne allows confirmation of PSA results with data from station with climatic conditions significantly different from arid Spain desert

1-min BSRN measurement

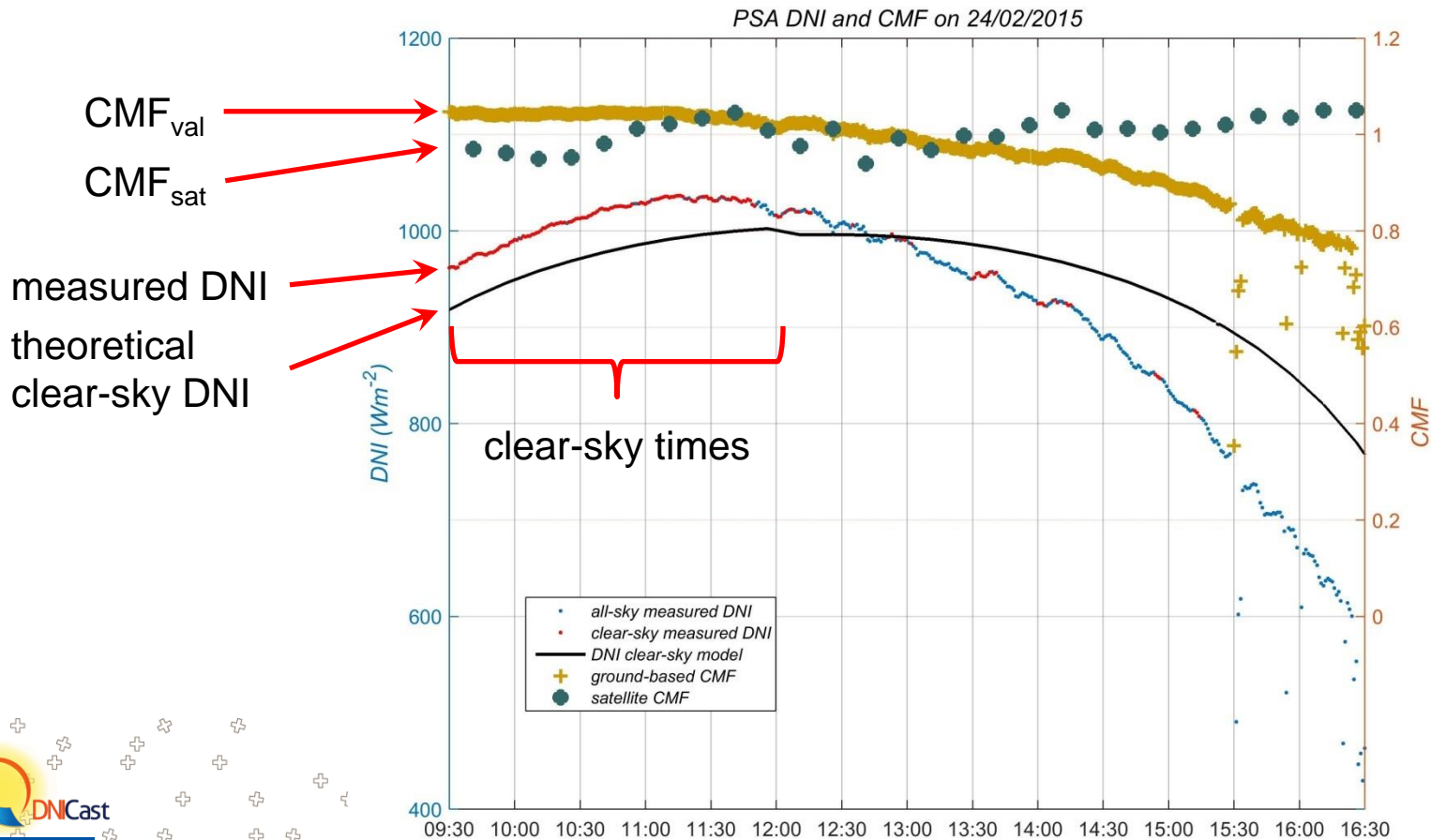


Radiation and CMF

data

Satellite information \rightarrow CMF_{sat} compared to $CMF_{val} = \frac{I_{meas}}{I_{cs}^{th}}$

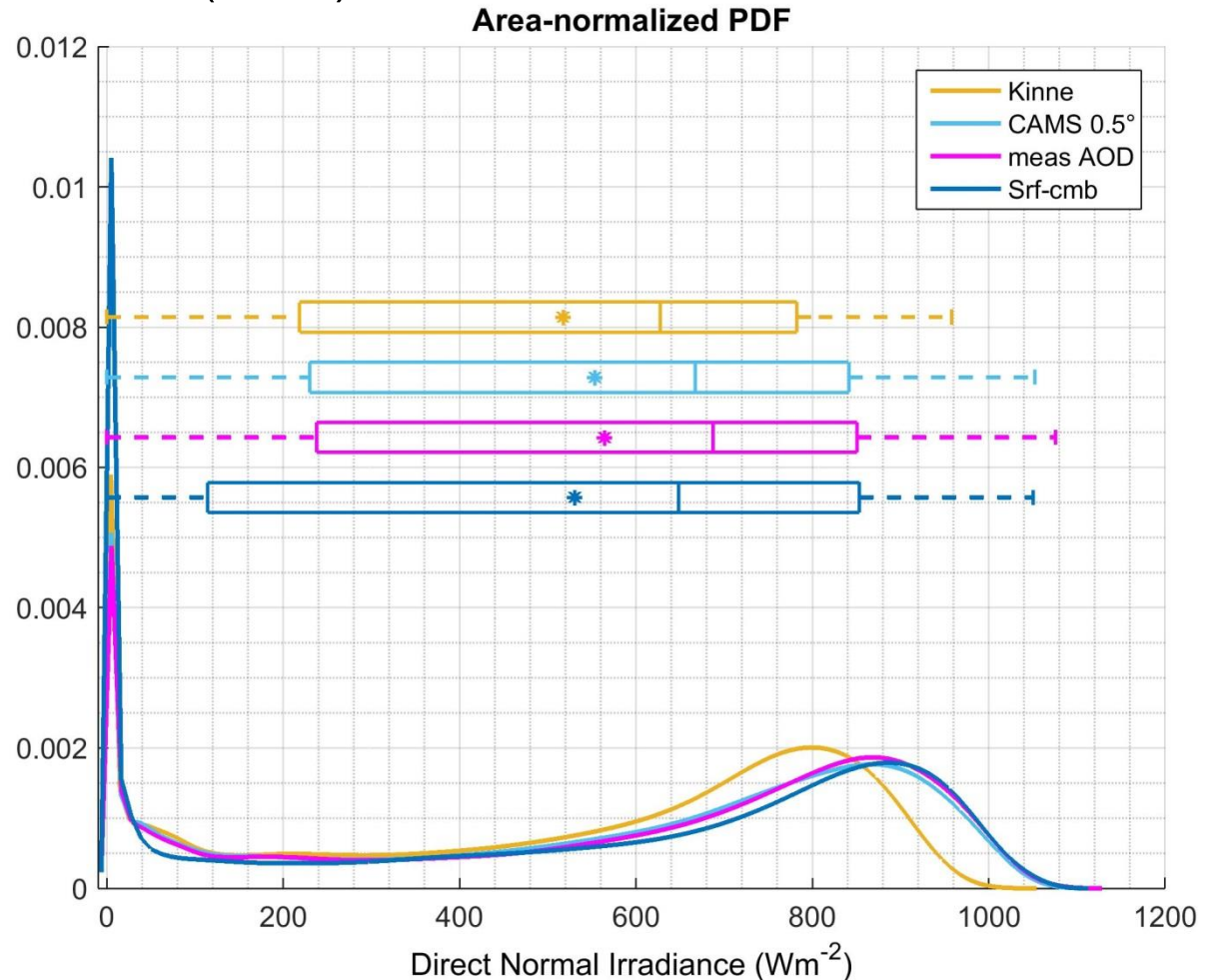
At clear-sky times radiation theoretical prediction can be compared with data.



Probability density function

all-sky estimates

- Using averages or medians to compare ground-based all-sky DNI measurements to satellite corresponding estimates suggests aerosol climatology by Kinne et al. (2006) is best



Probability density function

clear-sky estimates

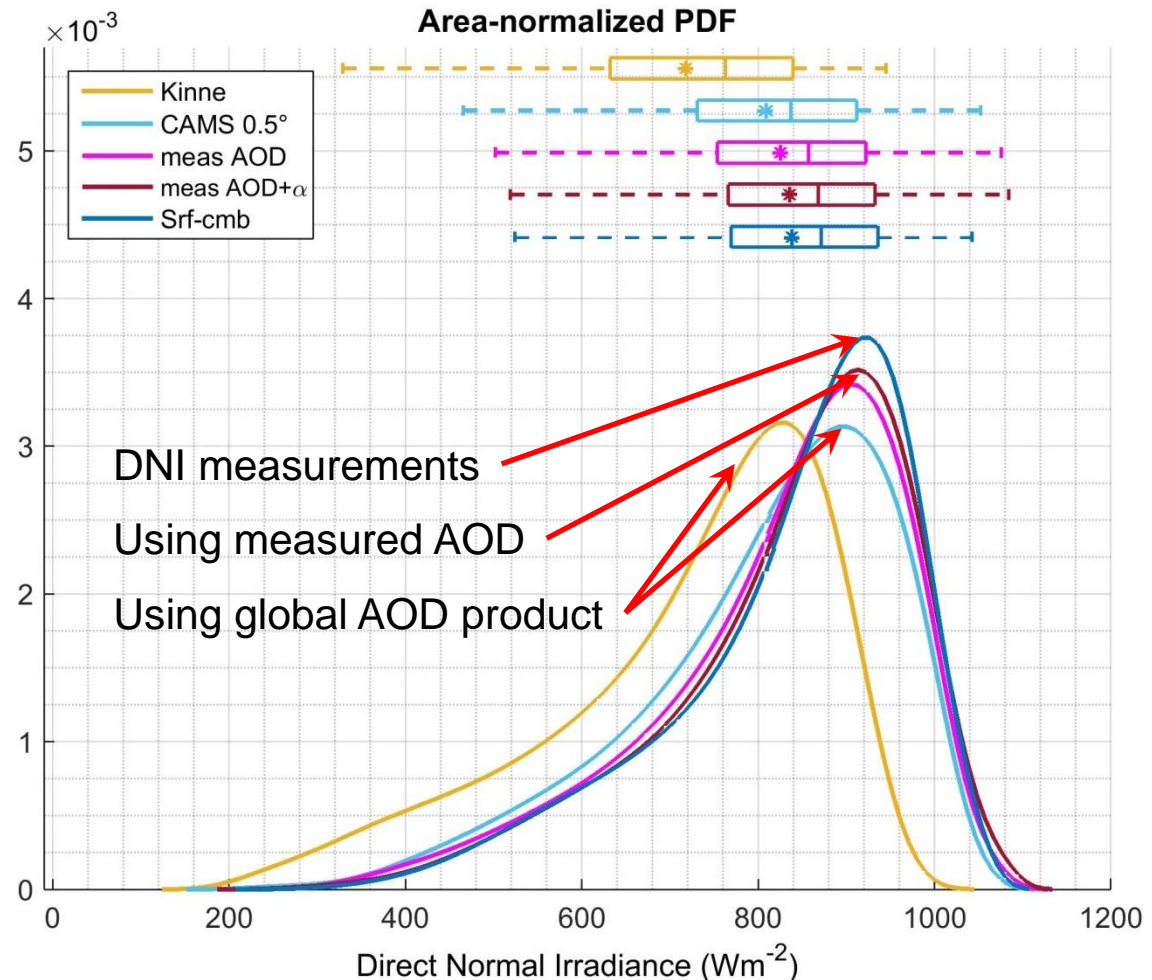
Similar PDF for all ground-based measurements → same for their average

Estimates using **measured AOD**: extremely close to measurements

Estimates using **global AOD products**:

- Kinne et al. (2006) climatology: strong underestimation
- CAMS AOD product: small underestimation (too many high AOD events)

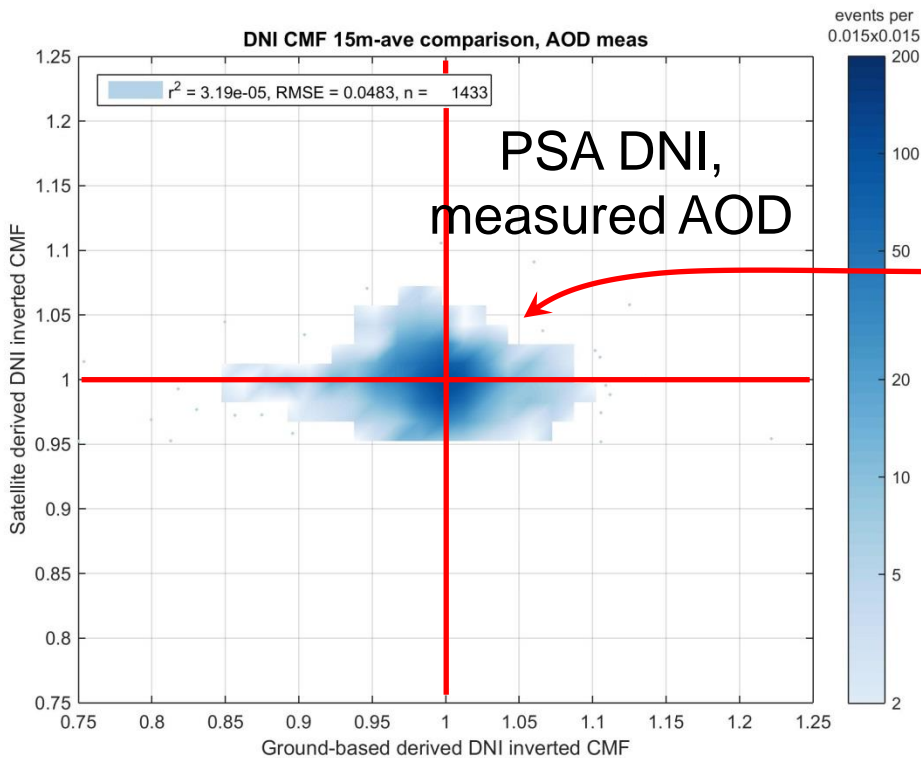
Kinne, S., et al. (2006): An AeroCom initial assessment – optical properties in aerosol component modules of global models, *Atmos. Chem. Phys.*, **6**, 1815–1834, doi:10.5194/acp-6-1815-2006.



Inverse CMF distributions

clear-sky estimates

Validating CMF for clear-sky



For **clear-sky**, one expects CMF_{sat} vs CMF_{val} to be a blob centered on **(1,1)**



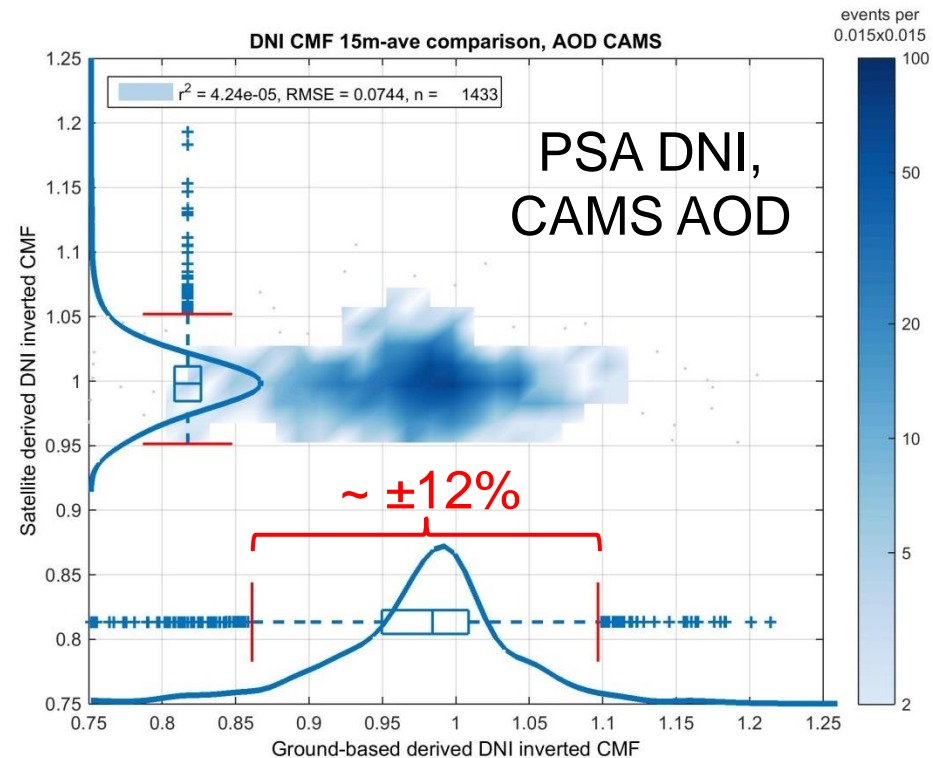
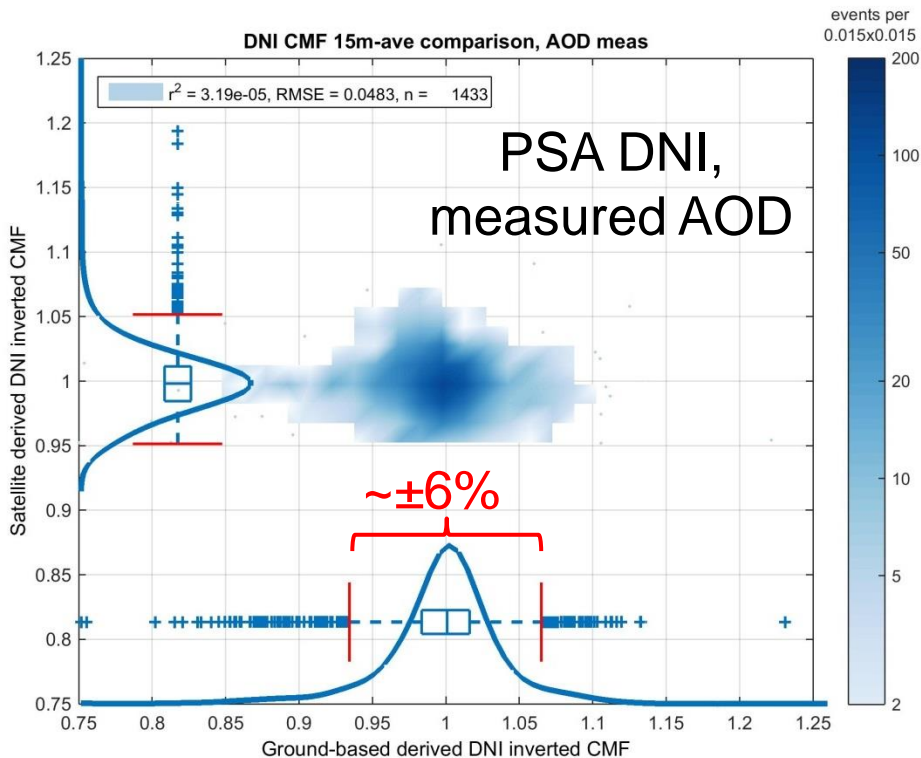
Inverse CMF distributions

clear-sky estimates

Validating CMF for clear-sky (should be =1) gives information on uncertainty of clear-sky estimates → uncertainty in aerosol input

$$\frac{1}{\text{CMF}_{\text{CS}}^{\text{val}}} = f(I_{\text{CS}}^{\text{th}}, I_{\text{CS}}^{\text{meas}}) = \frac{I_{\text{CS}}^{\text{th}}}{I_{\text{CS}}^{\text{meas}}}$$

$$\sigma^2(f) = \underbrace{\frac{\sigma^2(I_{\text{CS}}^{\text{th}})}{(I_{\text{CS}}^{\text{meas}})^2}}_{\sim 1} + \left(\frac{I_{\text{CS}}^{\text{th}}}{I_{\text{CS}}^{\text{meas}}}\right)^2 \underbrace{\frac{\sigma^2(I_{\text{CS}}^{\text{meas}})}{(I_{\text{CS}}^{\text{meas}})^2}}_{(\sim 1.5\%)^2}$$



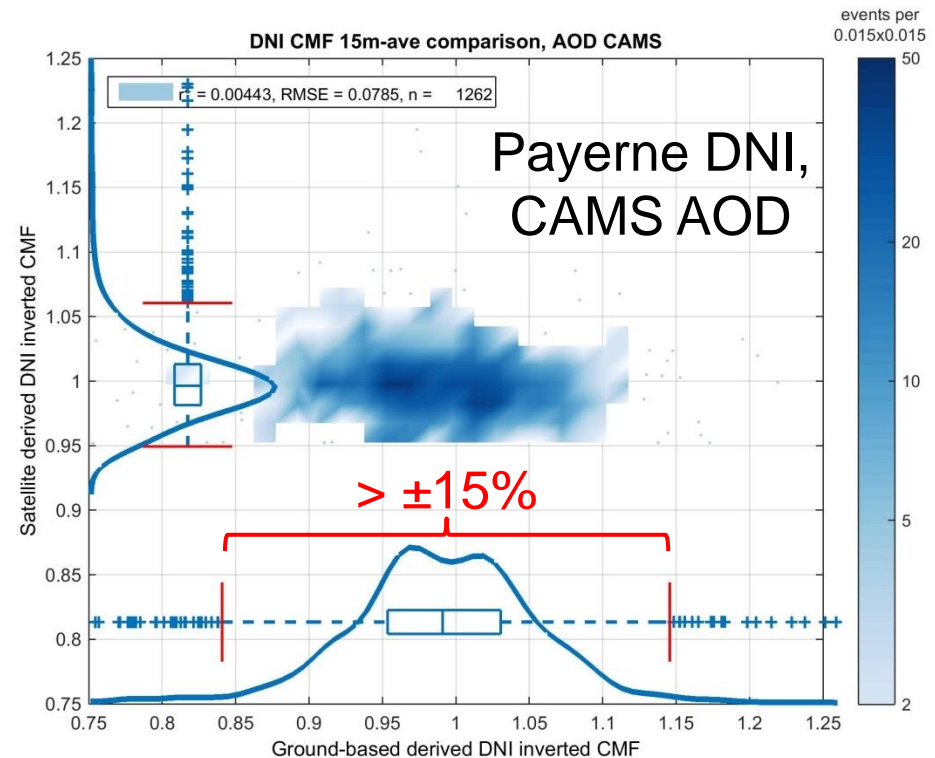
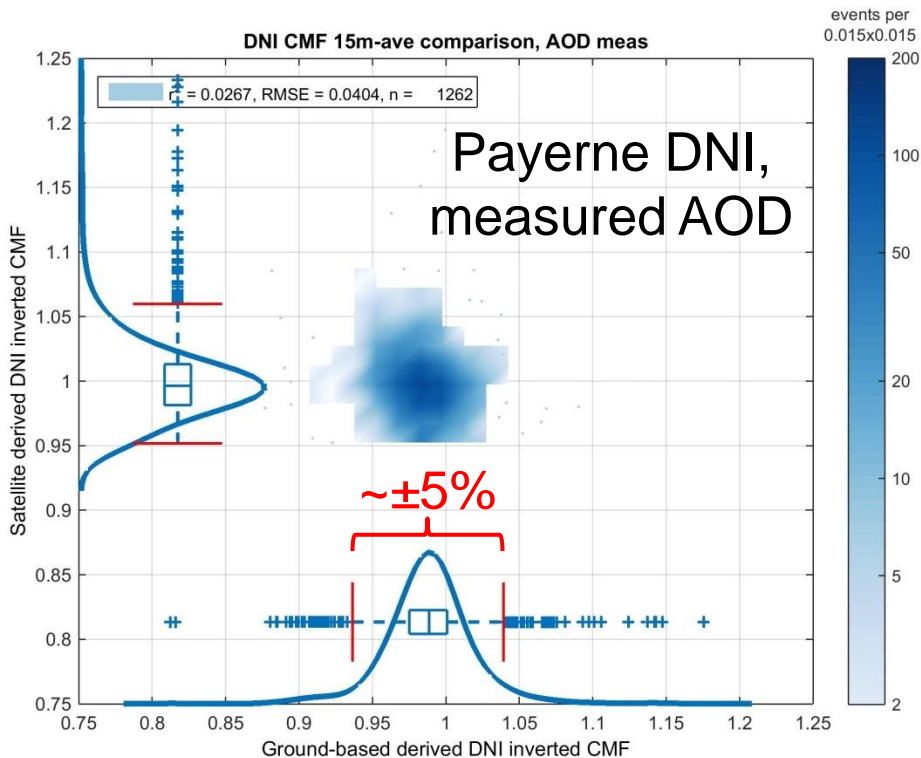
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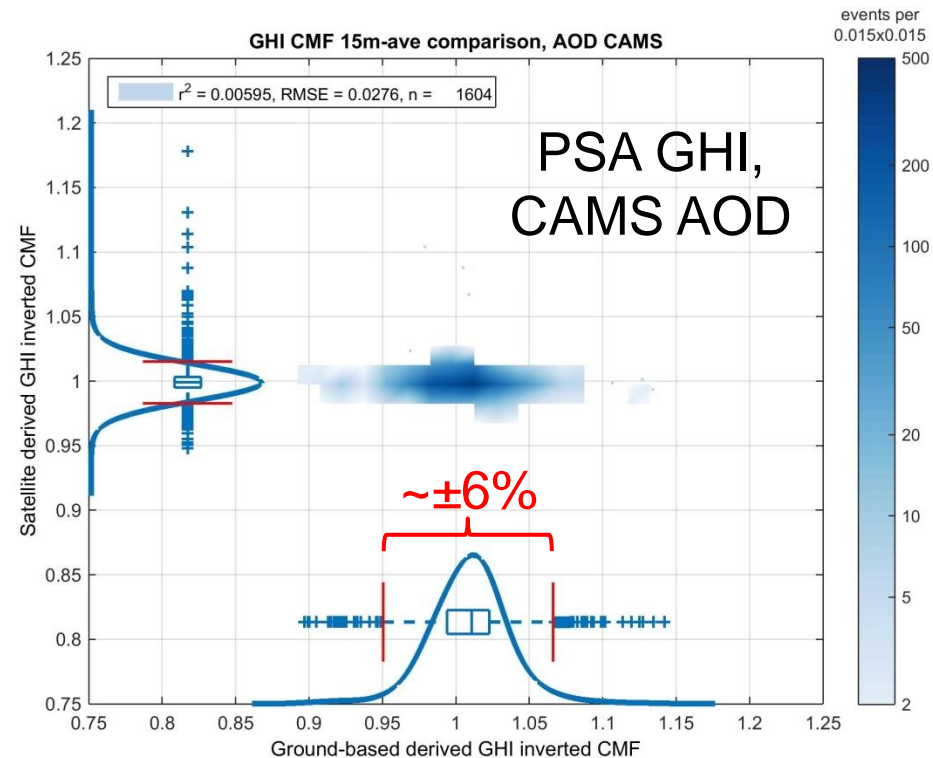
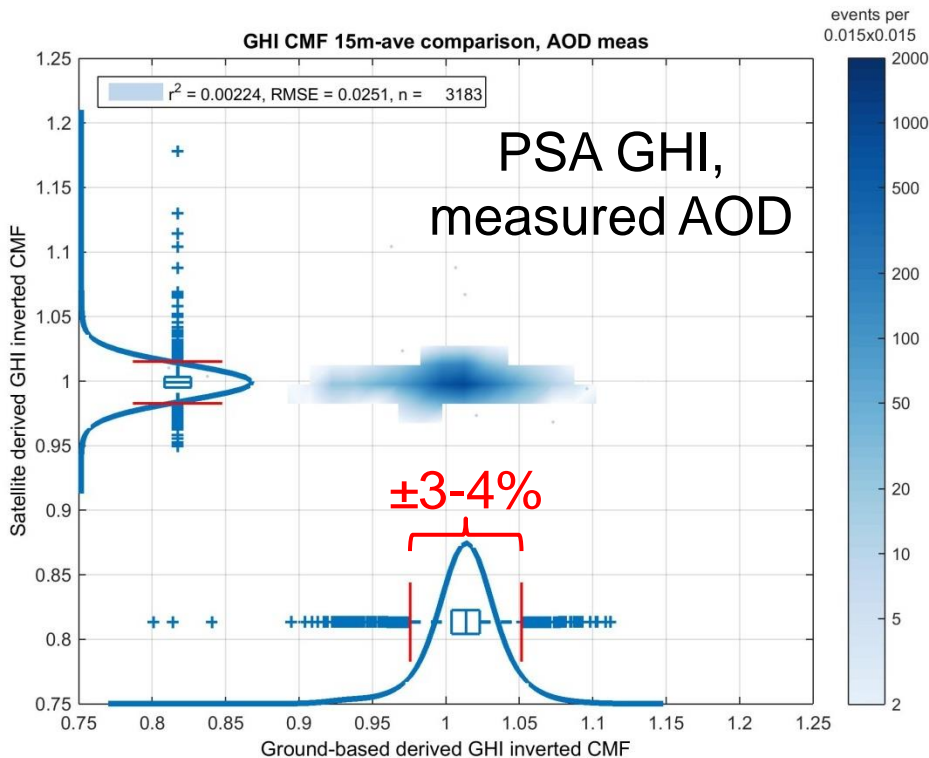


Inverse CMF distributions

clear-sky estimates

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Results for GHI are significantly better ← compensation between AOD effects on direct and diffuse component

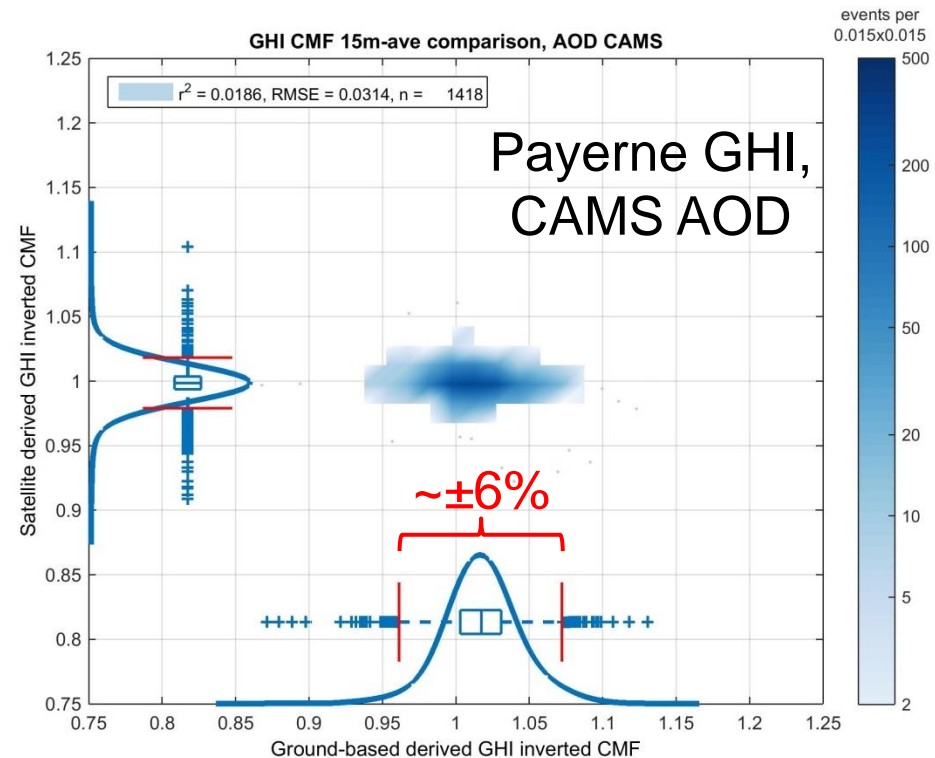
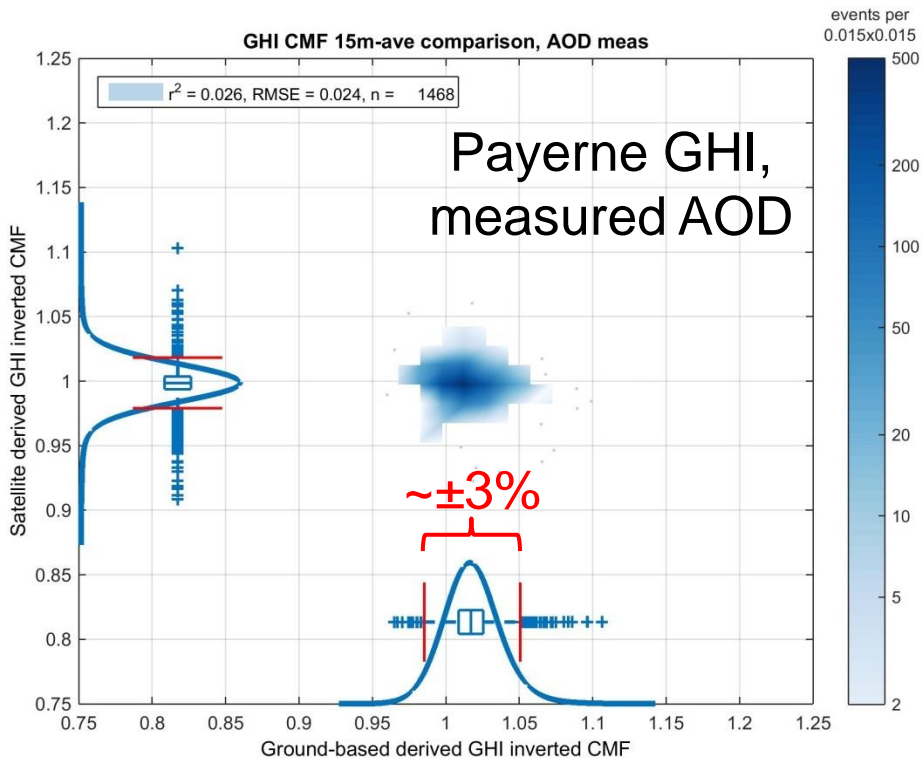


Inverse CMF distributions

clear-sky estimates

Validating CMF for clear-sky (should be =1) gives information on uncertainty of clear-sky estimates → uncertainty in aerosol input

Results for GHI are significantly better ← compensation between AOD effects on direct and diffuse component



Conclusions (intermediary, part 1)

Clear-sky situations:

- Model performing well with **main uncertainty** from **AOD input**
- Using measured AOD allows agreement within about 2–3 times measurement uncertainty
 - Measured AOD: **DNI** ~ -1% bias, uncertainty: ~ **±6%**
GHI ~ +1% bias, uncertainty: ~ **±3%**
- Using global AOD product
 - Kinne climatology → strong DNI underestimation (do not use)
 - CAMS AOD product: **DNI** ~ -3% bias, uncertainty: ~ **±15%**
GHI ~ +1% bias, uncertainty: ~ **±5%**
 - Unsolved issue with CAMS / MACC: **constant quality over time?**

Probability density function

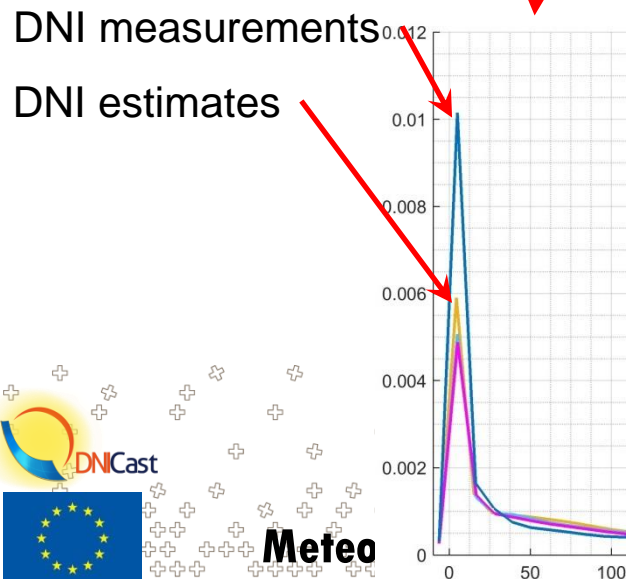
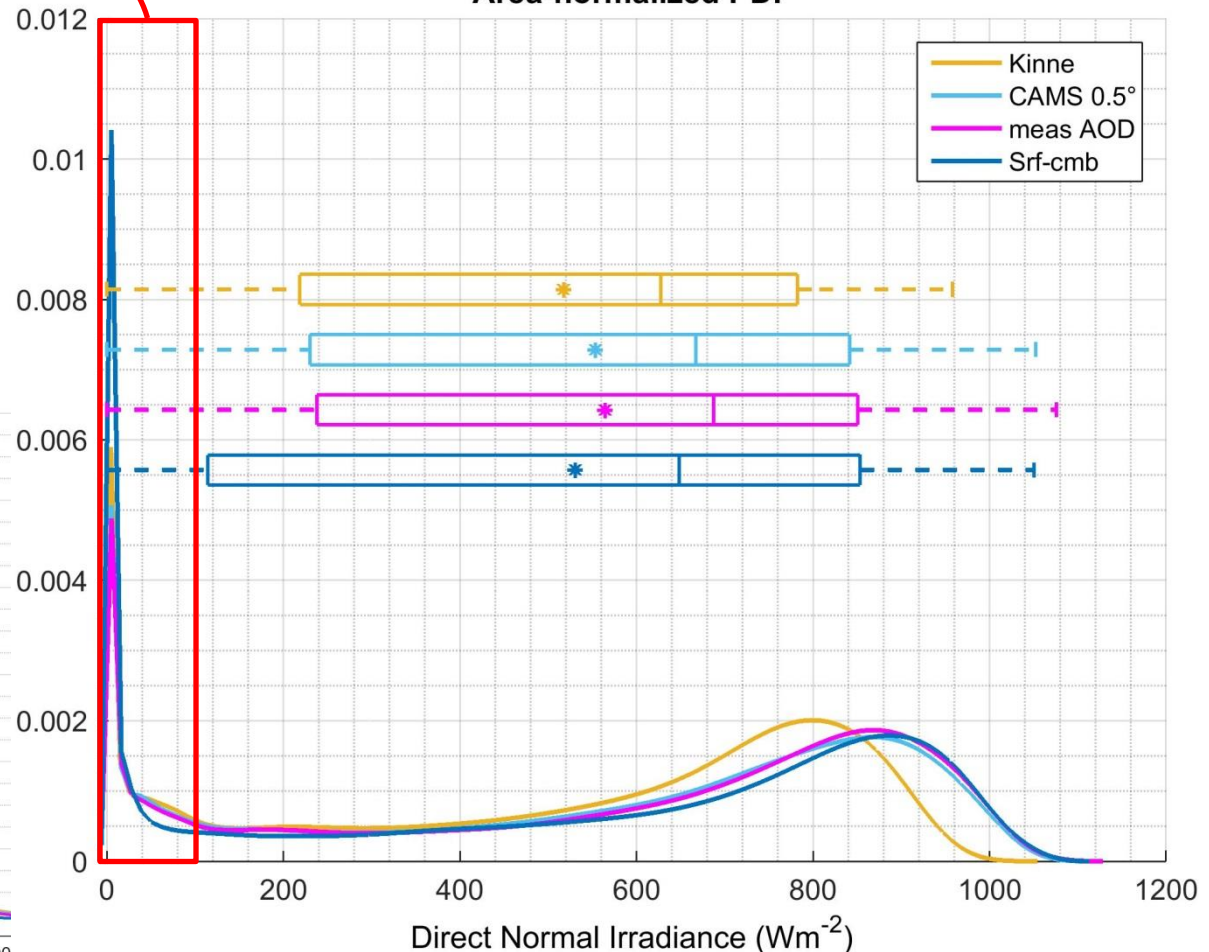
all-sky estimates

Similar PDF for all ground-based measurements → same for their average

Satellite estimates: **significant underestimation of (almost-)zero-DNI events**

- Strong effect on 1st quartile
- “Compensate” for Kinne AOD overestimation

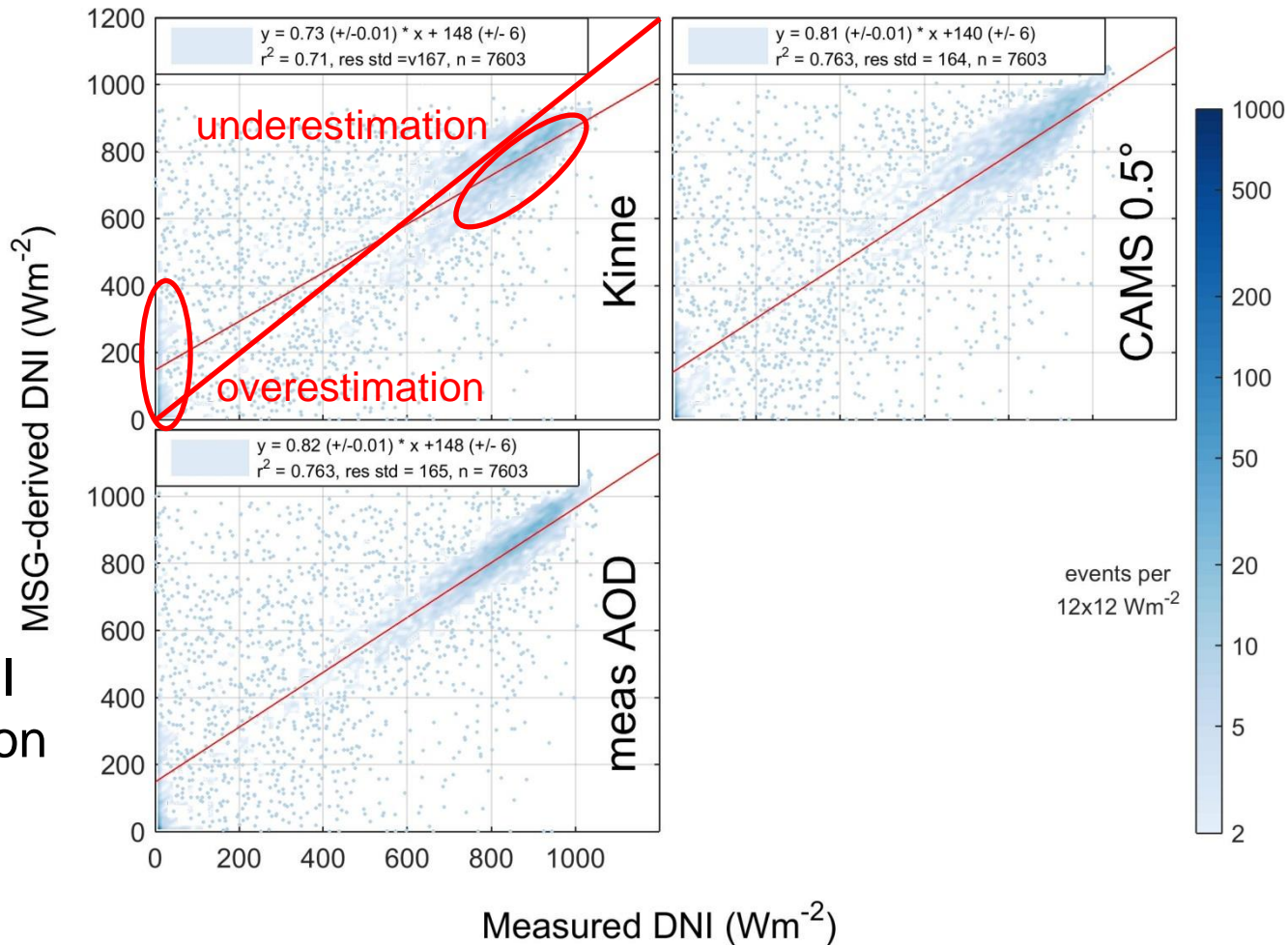
Area-normalized PDF



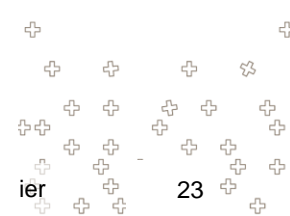
DNI comparison

all-sky estimates

- Kinne agrees better **on average**, but CAMS better **event by event**
- Significant scatter



PSA DNI comparison

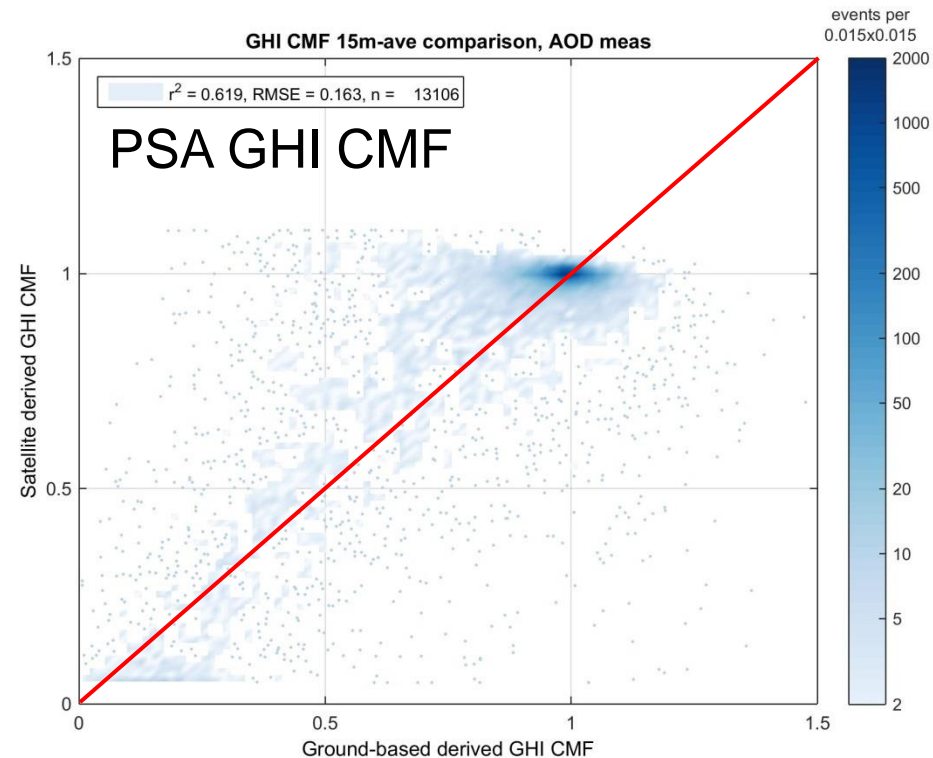
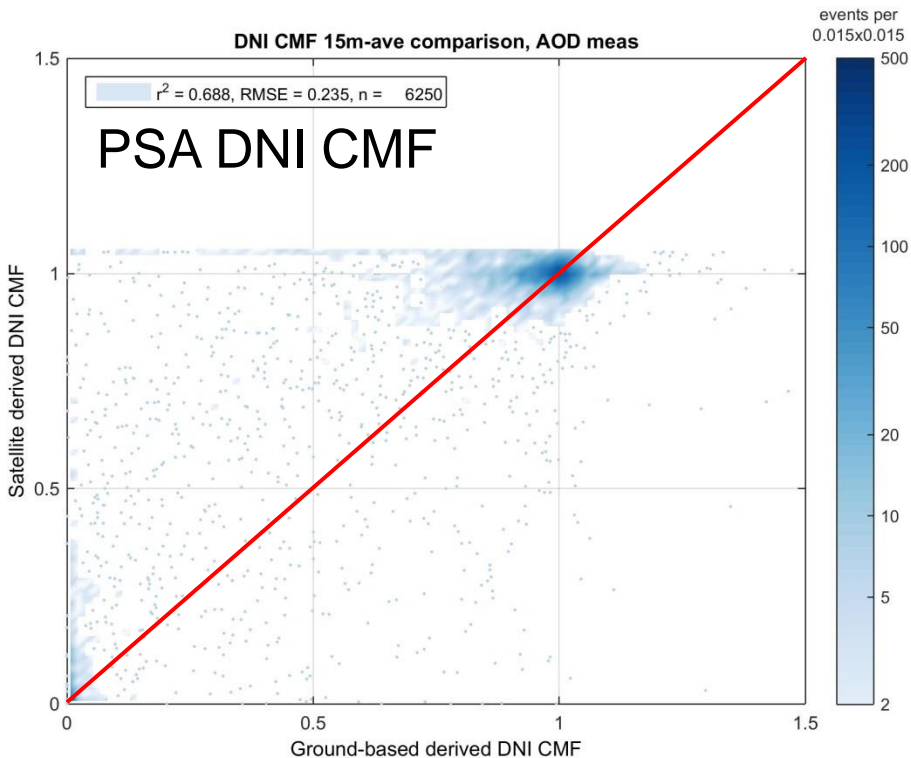


CMF comparisons

all-sky estimates

Satellite CMF determination for DNI & GHI is empirical with limited physical understanding

Issues in empirical algorithm going beyond spatial and temporal variability for both DNI and GHI (may be specific to HeliMont)

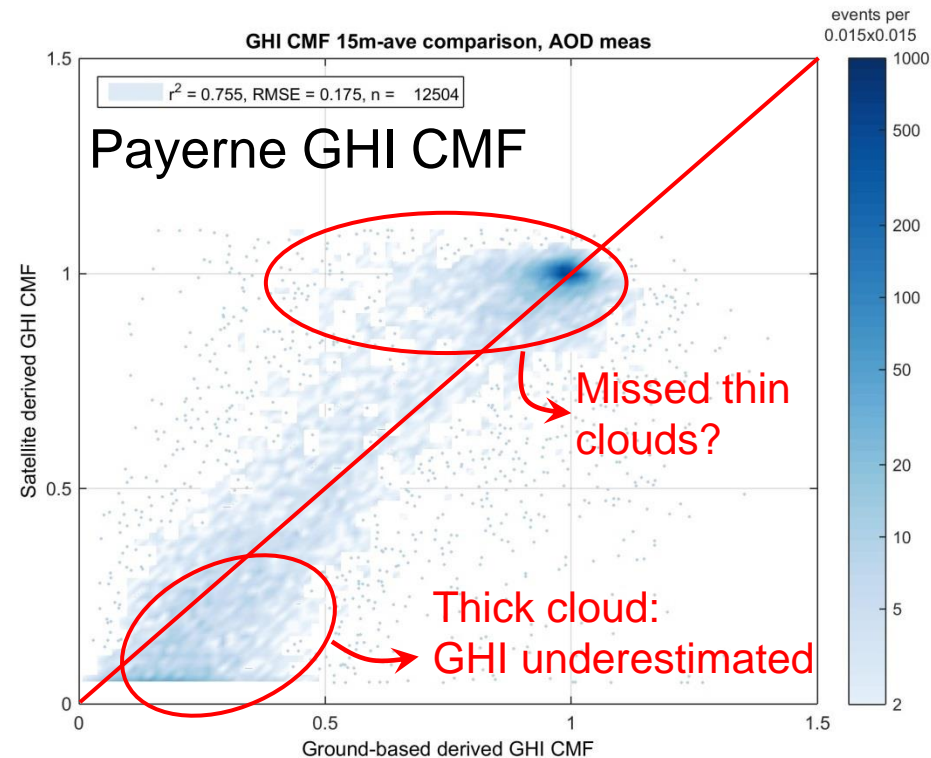
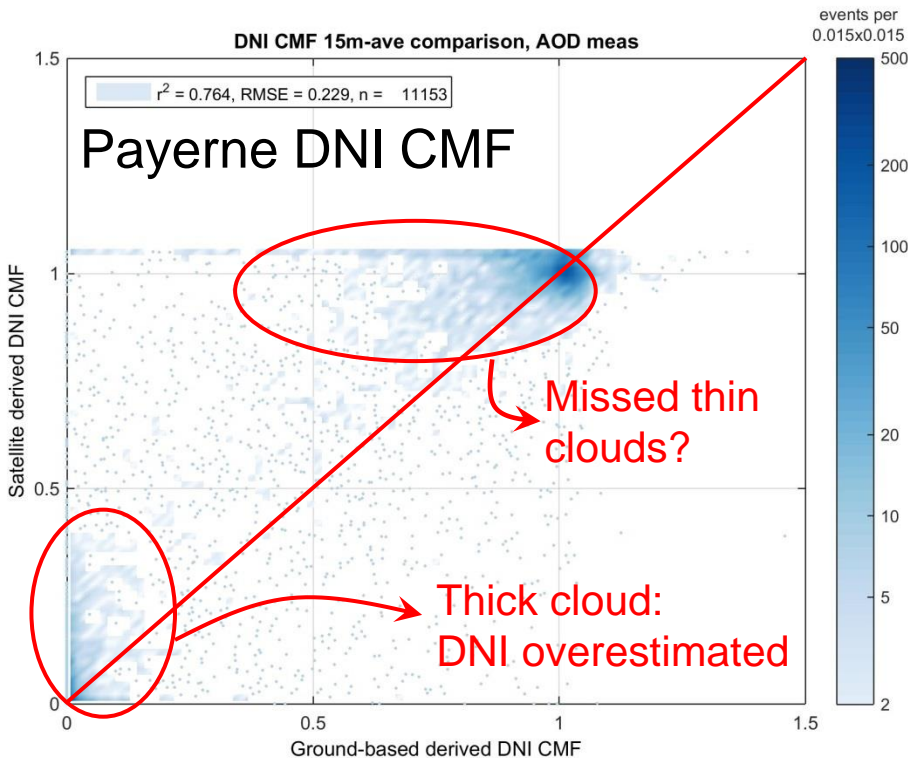


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Conclusions

All-sky situations:

- Issues remain to be solved in satellite-based CMF determination
 - Thick clouds: DNI overestimation / GHI underestimation
 - Thin clouds missed?
- Optimism: issues identified, roadmap for solving them can be set!

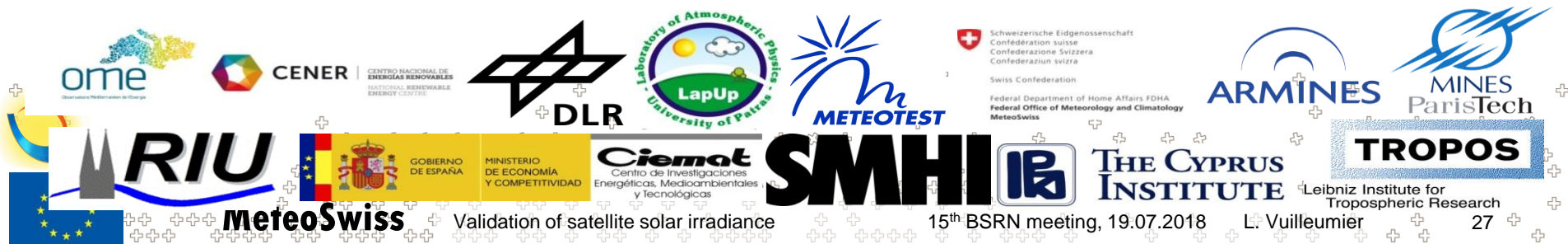
Separately validating clear-sky irradiance and cloud effect provides better understanding of satellite irradiance estimation performance

- Spatial / temporal averaging → reduction cloud-induced scatter in comparisons between area satellite estimates and point ground-based measurements, 1-hour averaging satisfactory over MSG pixel

Thank you

? / !

L. Vuilleumier, Meyer, A., R. Stöckli, S. Wilbert and L. F. Zarzalejo (2018). Validation of Direct Normal Irradiance from Meteosat Second Generation, manuscript in preparation.

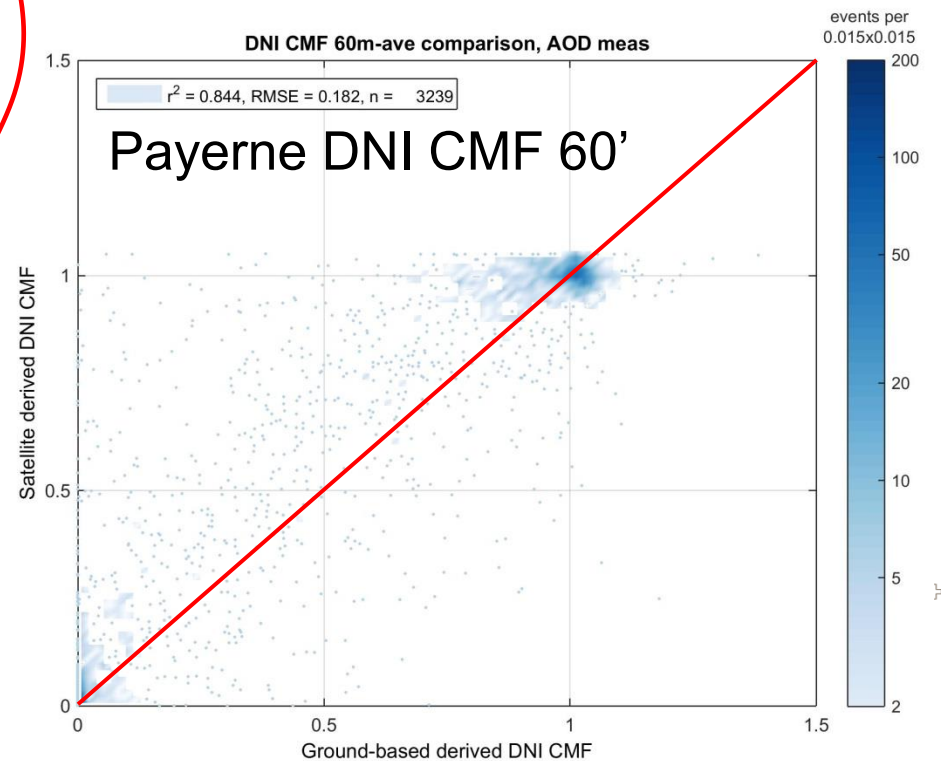
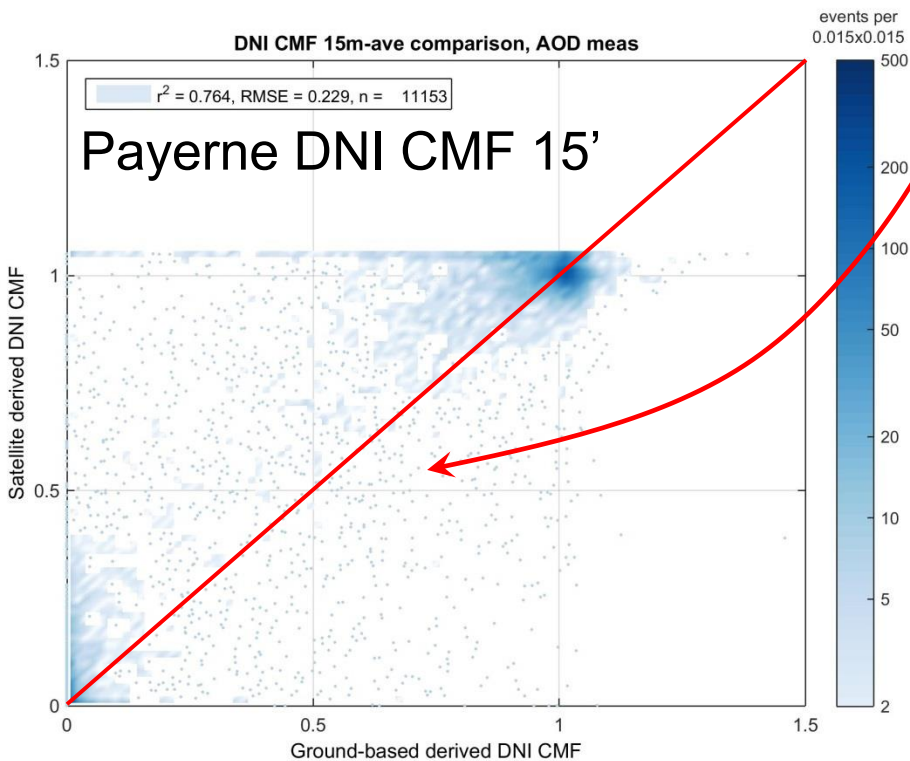


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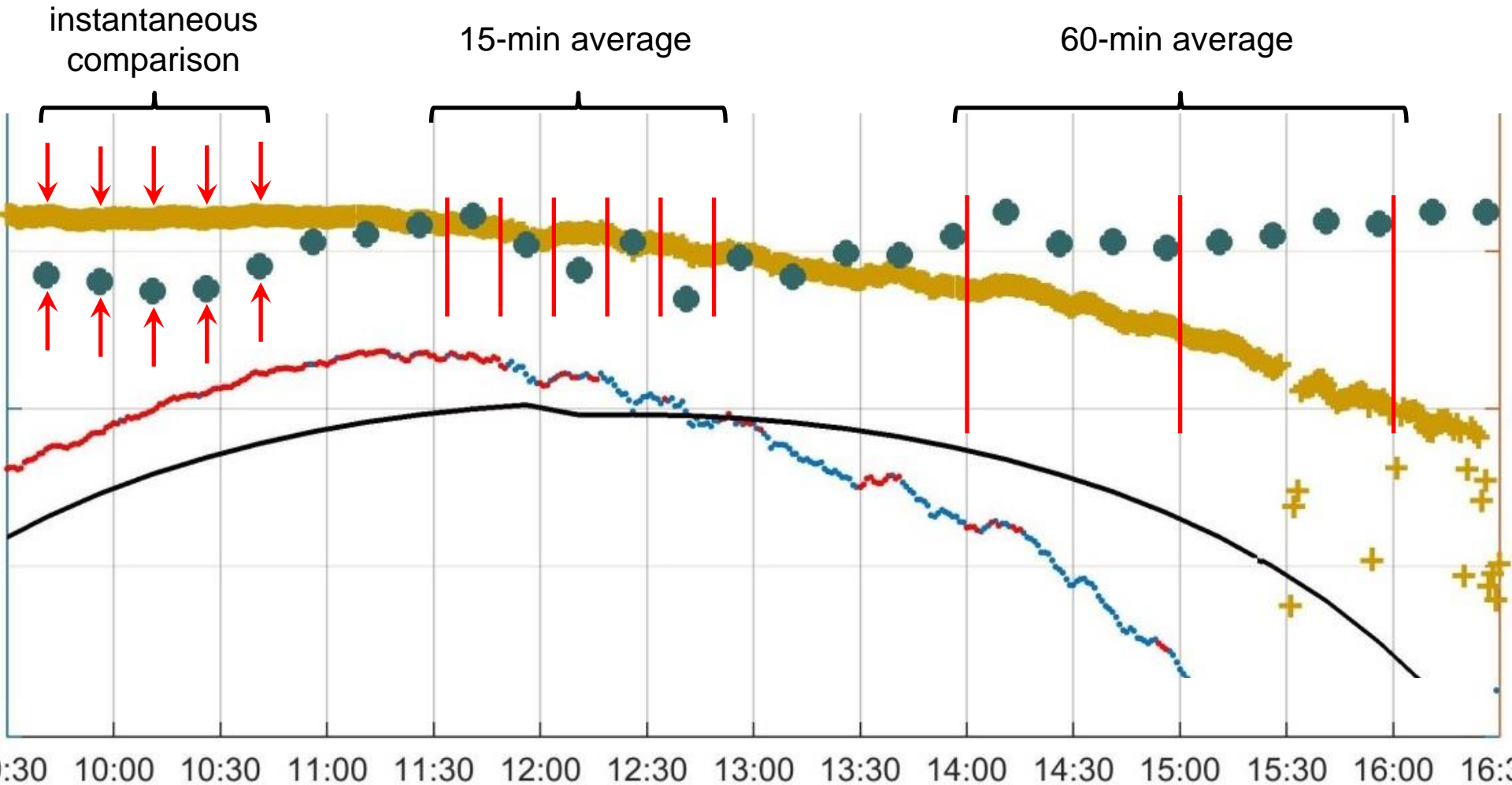
Have you notice there is some scatter?



DNI spatial and temporal averaging

all-sky estimates

Computing r^2 and RMSE of **CMF comparisons** for different temporal and spatial averaging (single vs. 4-instrument average at PSA)



DNI spatial and temporal averaging

all-sky estimates

- Temporal & spatial averaging **increase correlation** and **decrease RMSE**
- At **1-hour** temporal averaging, **very small difference** between PSA with and without spatial averaging

1-hour averaging satisfactory for point-to-area comparison in cloud effect

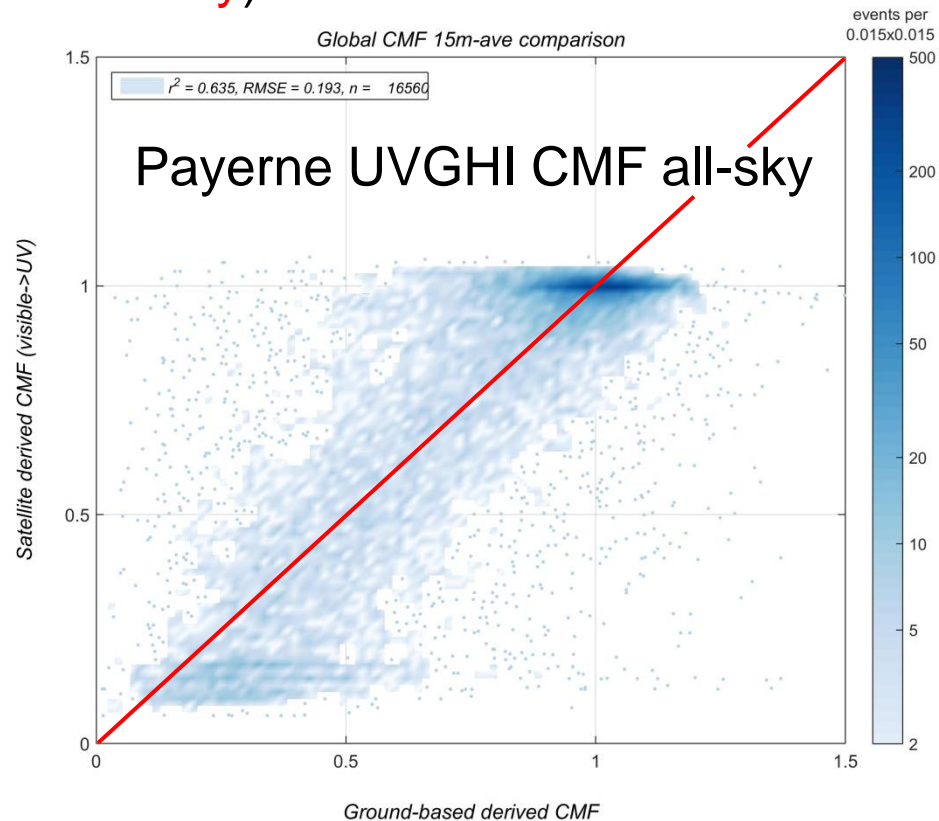
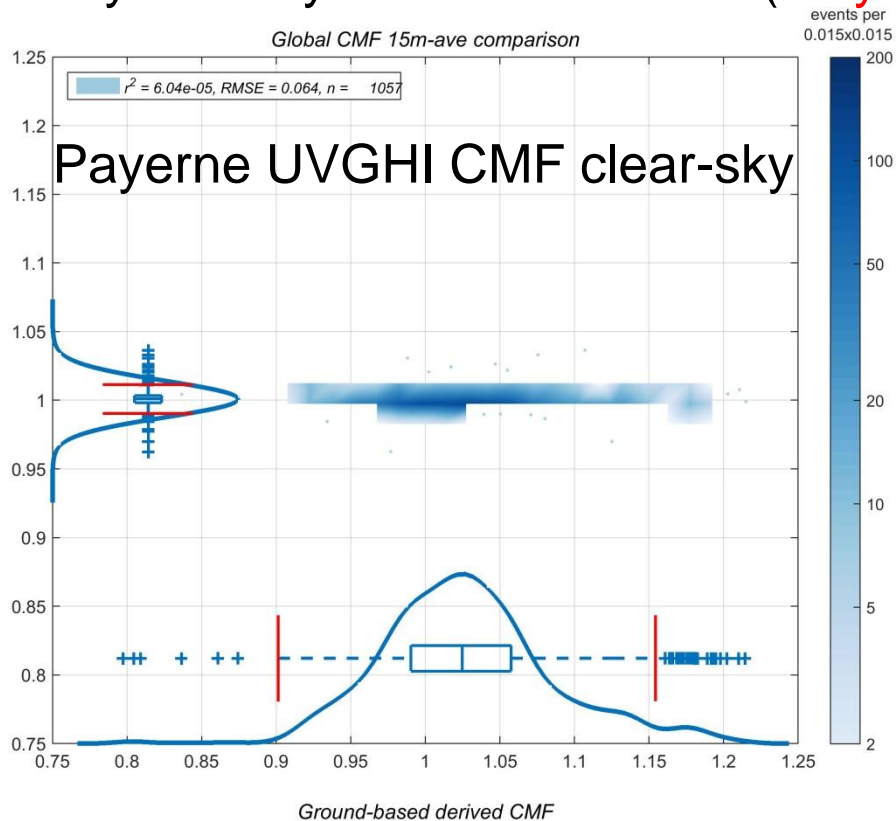
temporal averaging →

			instantaneous	15-min.	60-min.	
Correlation (r^2)	PSA	all instruments	0.667	0.708	0.789	nearly equal
		single instrument	0.640	0.699	0.788	↑ spatial averaging
	PAY	single instrument	0.718	0.764	0.844	increase
RMSE	PSA	all instruments	0.237	0.217	0.181	nearly equal
		single instrument	0.249	0.219	0.178	↑ spatial averaging
	PAY	single instrument	0.255	0.229	0.182	decrease

Outlook

- More stations → finding spatially homogenous?
- More years → stability over time
- More understanding → can we verify hypothesis on current issues

By the way how is it for UV ? (**very preliminary**)



Probability density function

clear-sky estimates

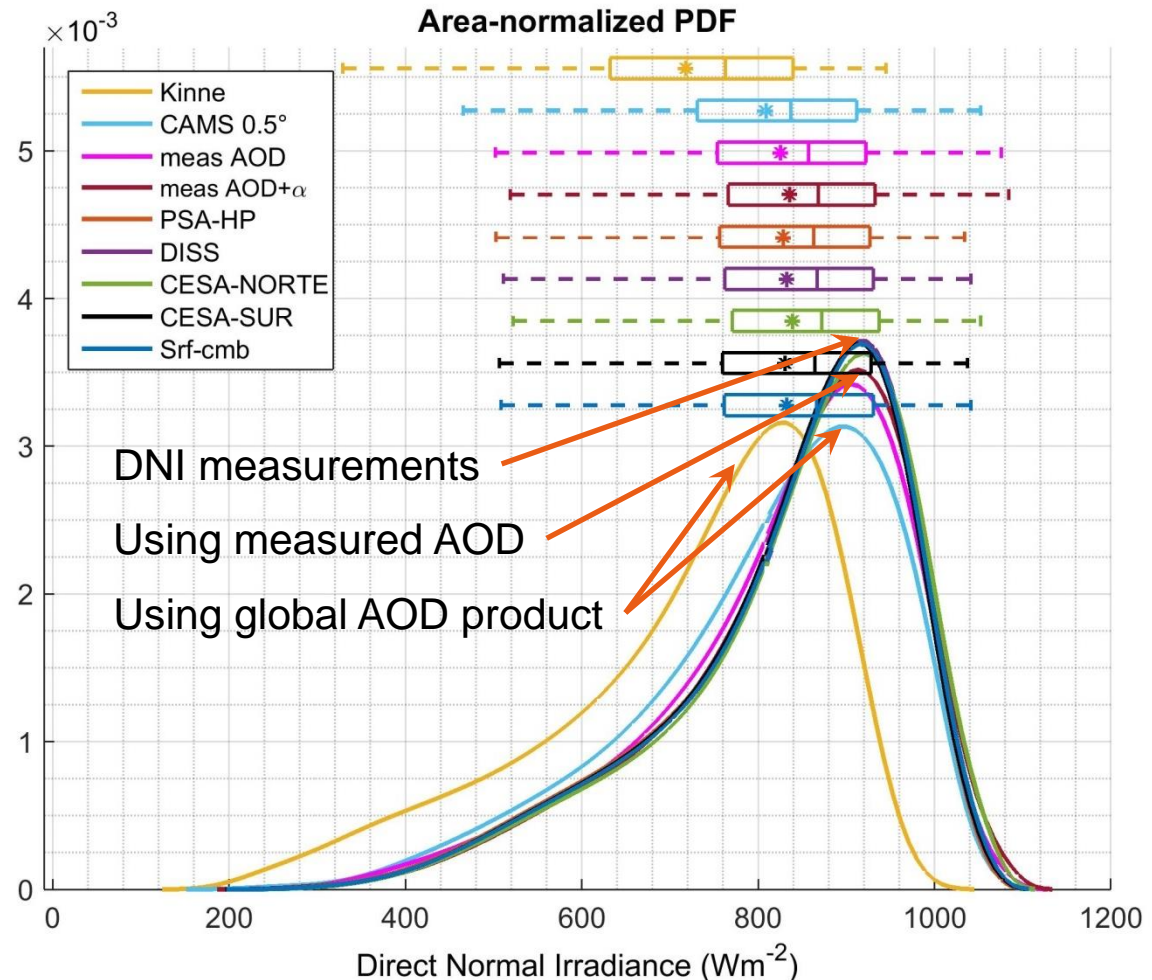
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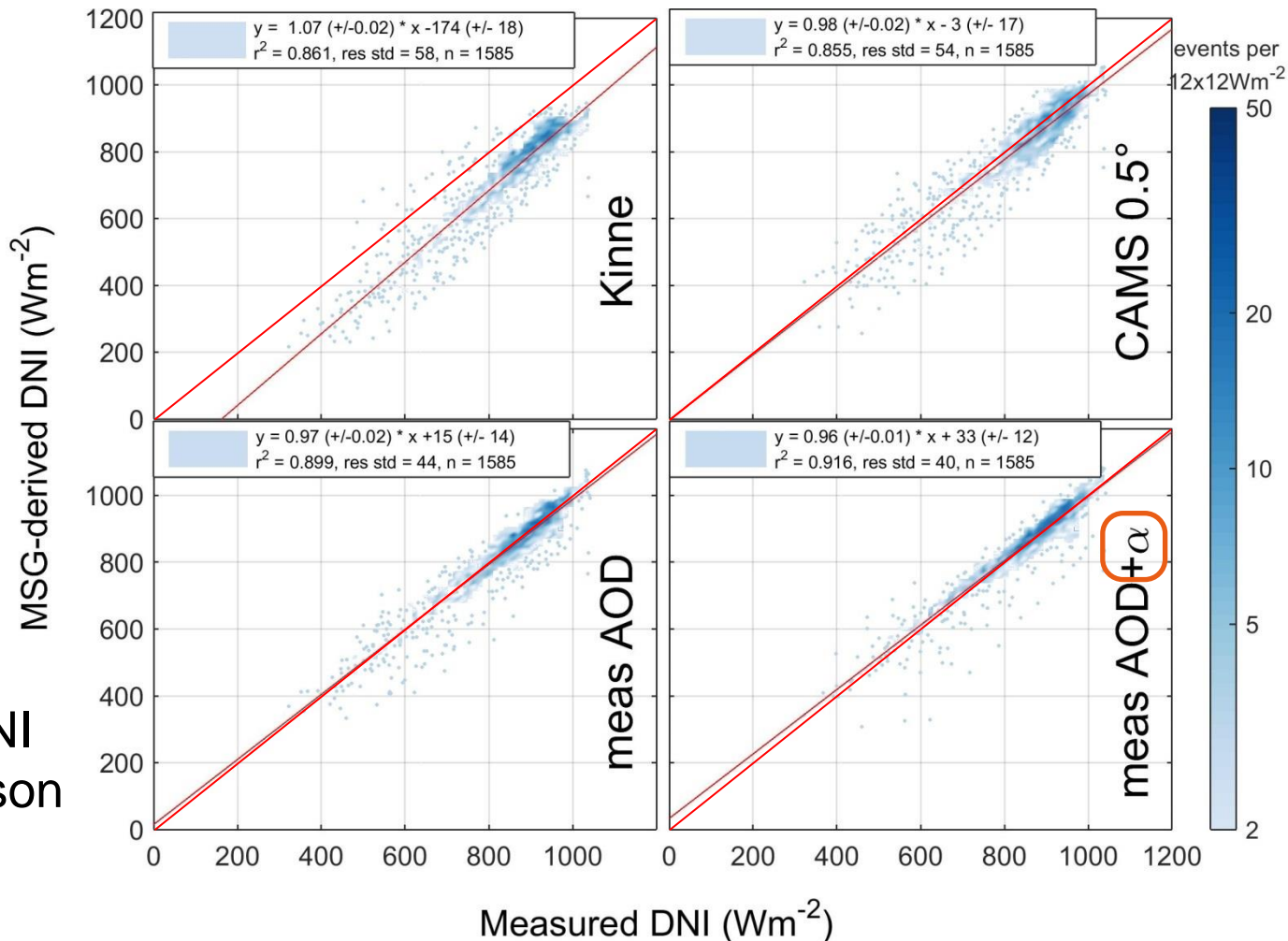
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DNI comparison

clear-sky estimates

The understanding of physics processes and the high measurement accuracy allow deep tests of aerosol influence on radiation:

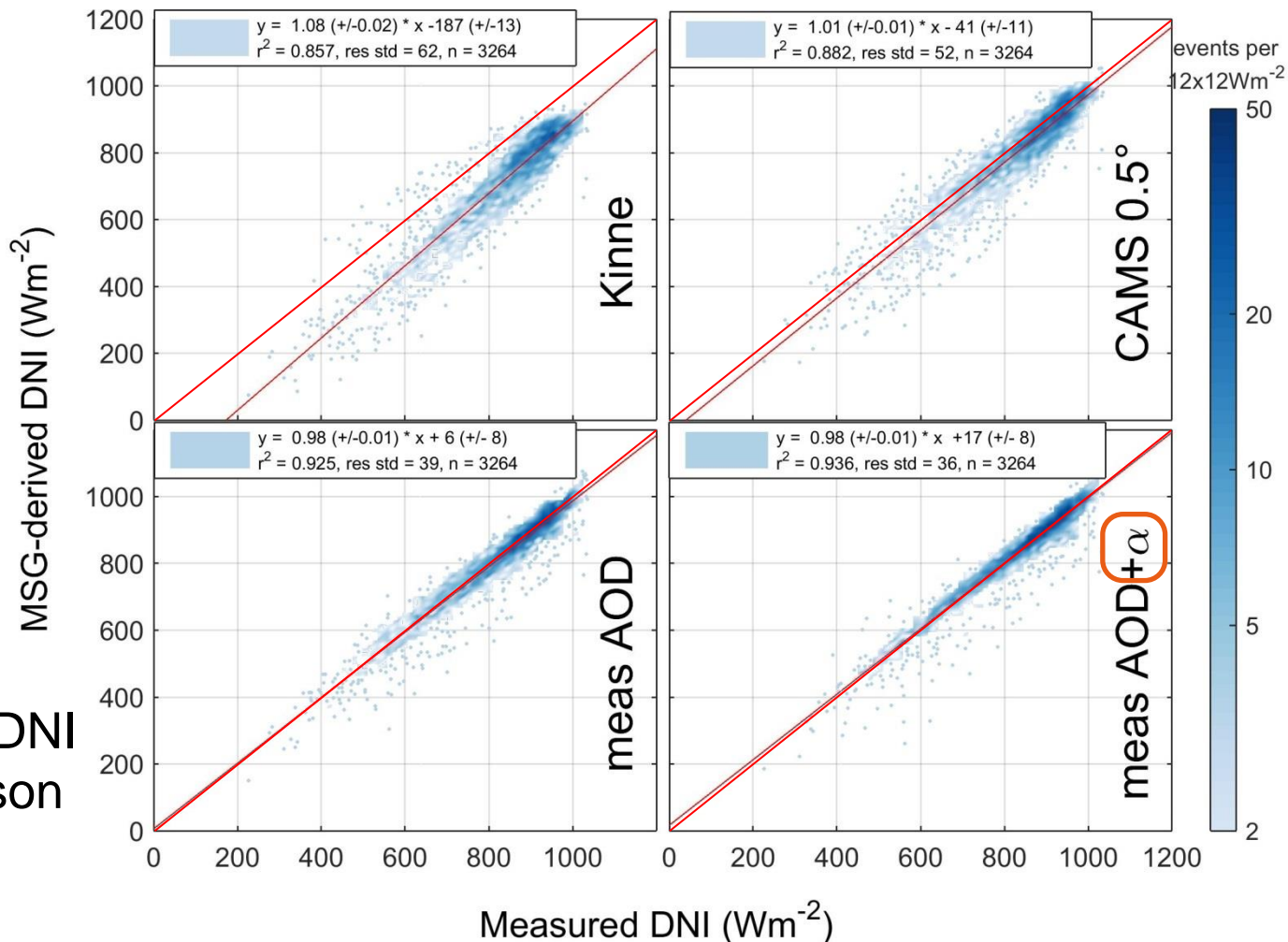


At limit: effect of AOD wavelength dependence

DNI comparison

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At limit: effect of AOD wavelength dependence

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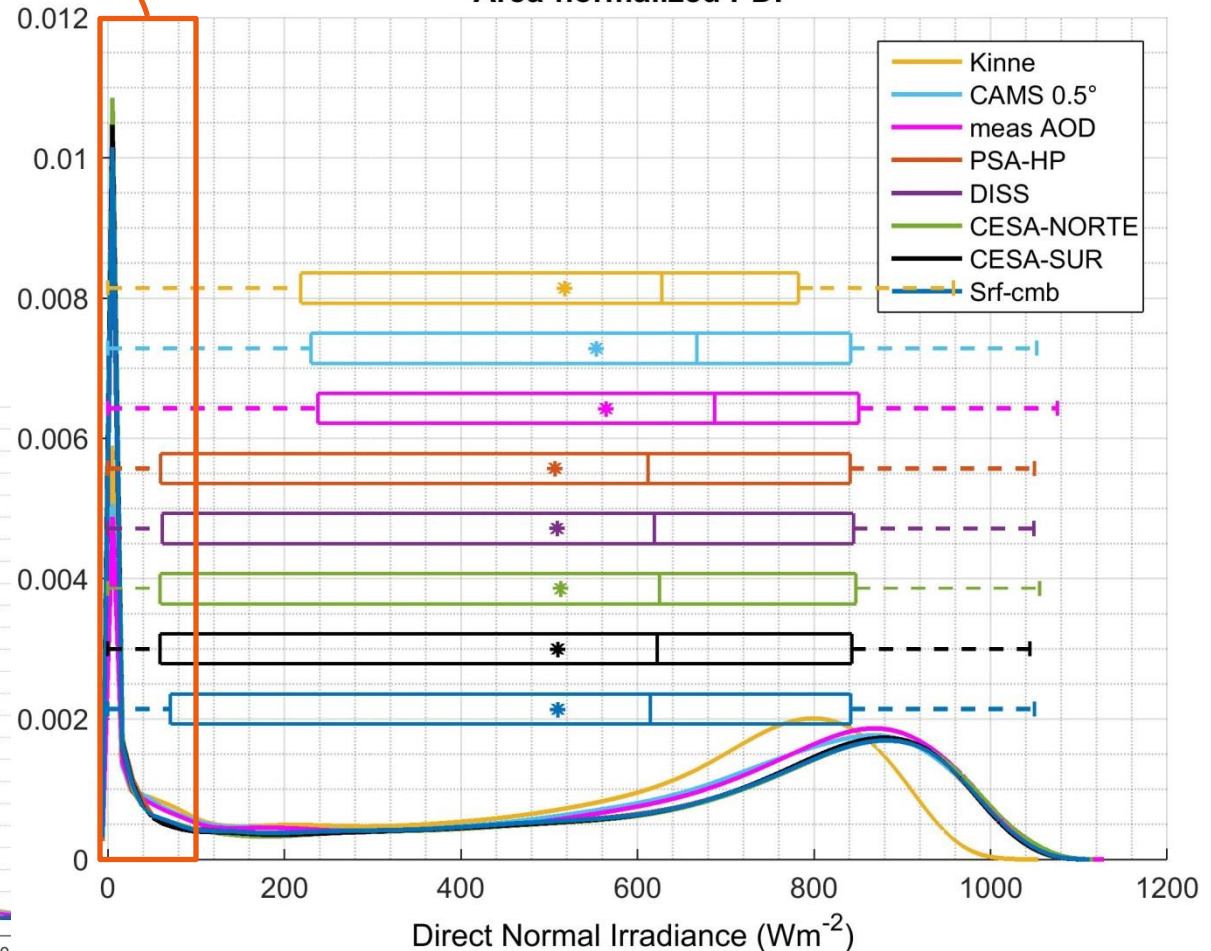
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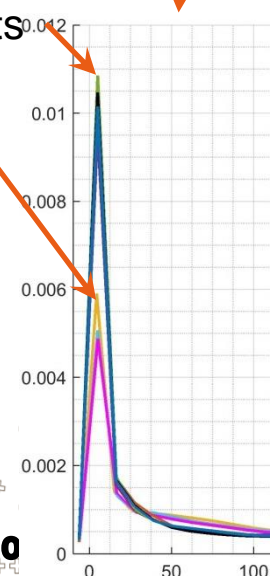
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Area-normalized PDF



DNI measurements

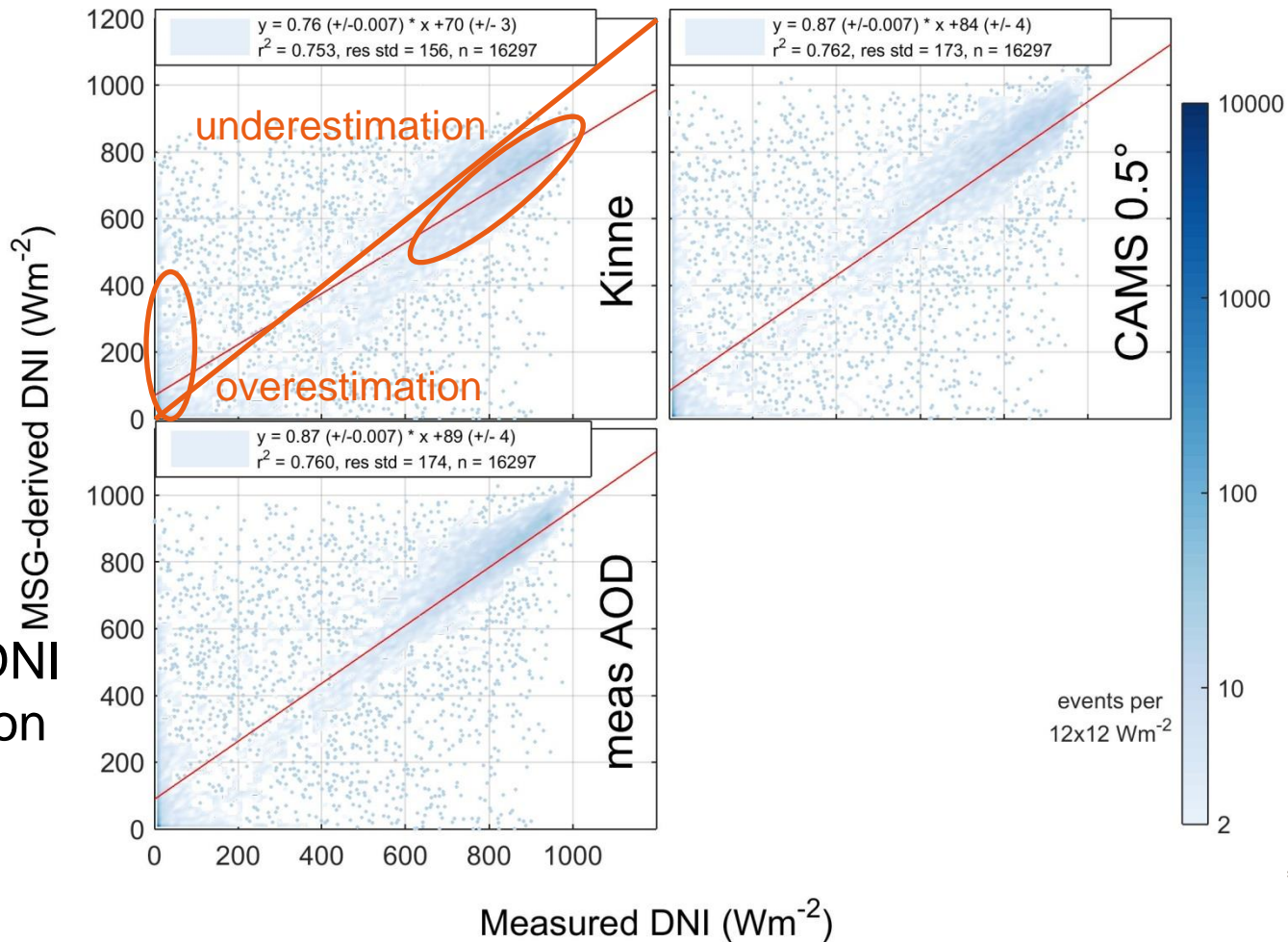
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Payerne DNI
comparison



DNI spatial and temporal averaging

all-sky estimates

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temporal averaging →

			instantaneous	15-min.	60-min.	
Correlation	PSA	all instruments (CSR corrected)	0.667 (0.648)	0.708 (0.688)	0.789 (0.767)	↑ spatial averaging increase
		single instrument (CSR corrected)	0.640 (0.625)	0.699 (0.677)	0.788 (0.764)	
	PAY	single instrument	0.718	0.764	0.844	
RMSE	PSA	all instruments (CSR corrected)	0.237 (0.255)	0.217 (0.235)	0.181 (0.202)	↑ spatial averaging decrease
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	PAY	single instrument	0.255	0.229	0.182	



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