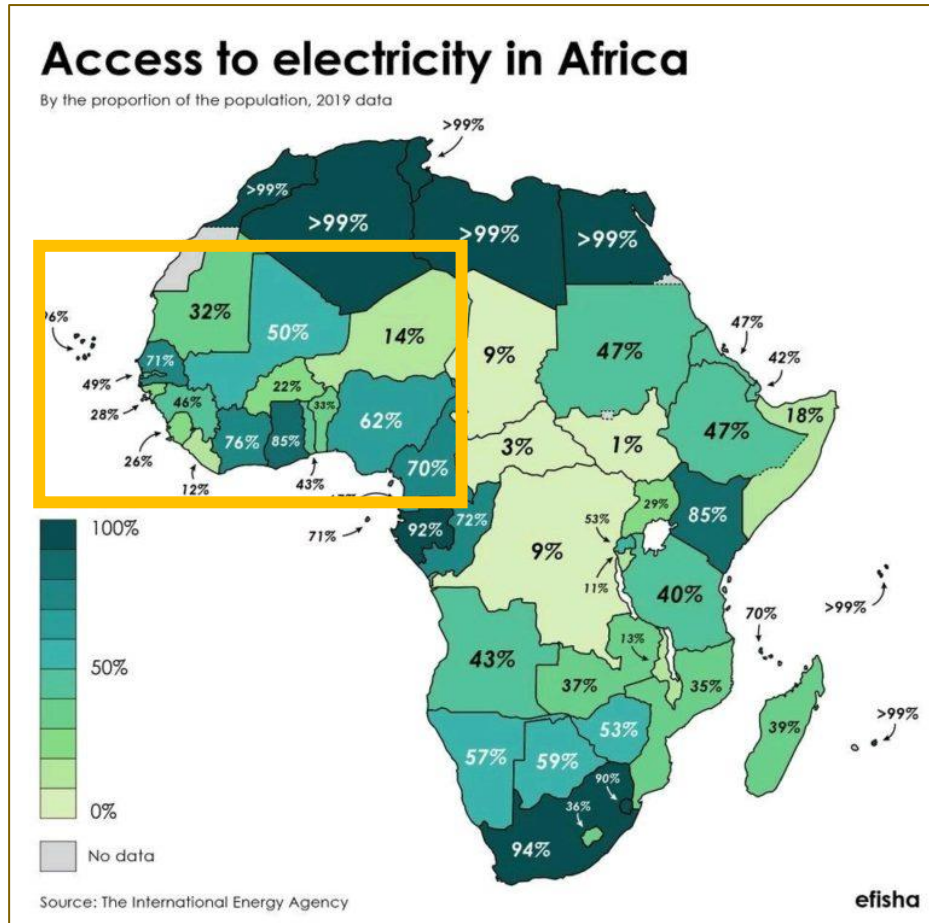


West African solar forecast errors and their link with meteorological conditions: case study of the Zagtouli solar farm (Burkina-Faso)

Léo Clauzel¹, Sandrine Anquetin¹, Christophe Lavaysse¹, Guillaume Tremoy², Damien Raynaud²

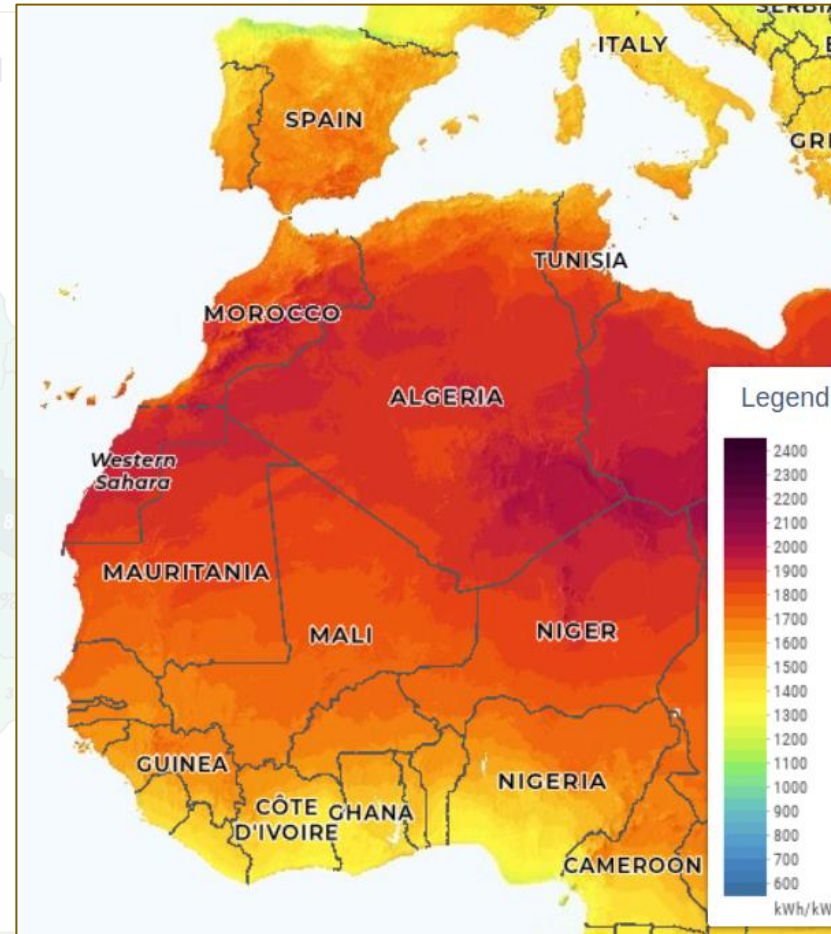
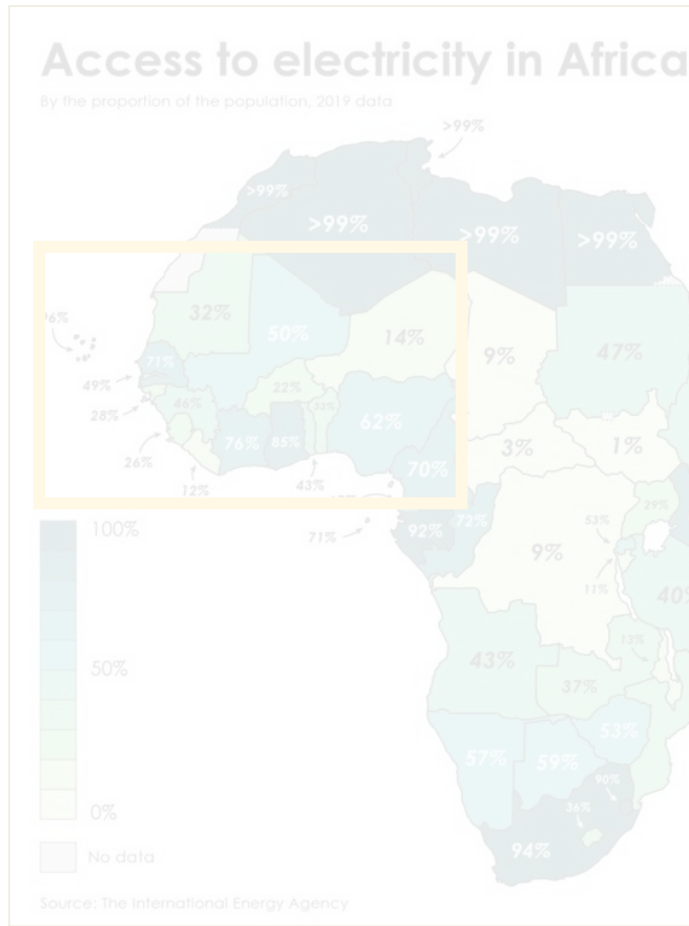
¹Institut des Géosciences de l'Environnement, Grenoble, France

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- Access to electricity is a key challenge in West African development

Context – Study sites – Objective – Results – Conclusion - Perspectives



- Access to electricity is a key challenge in West African development

- Solar PV is a good candidate for answering the West African stakes

West African solar forecast errors and their link with meteorological conditions

Context – Study sites – Objective – Results – Conclusion - Perspectives

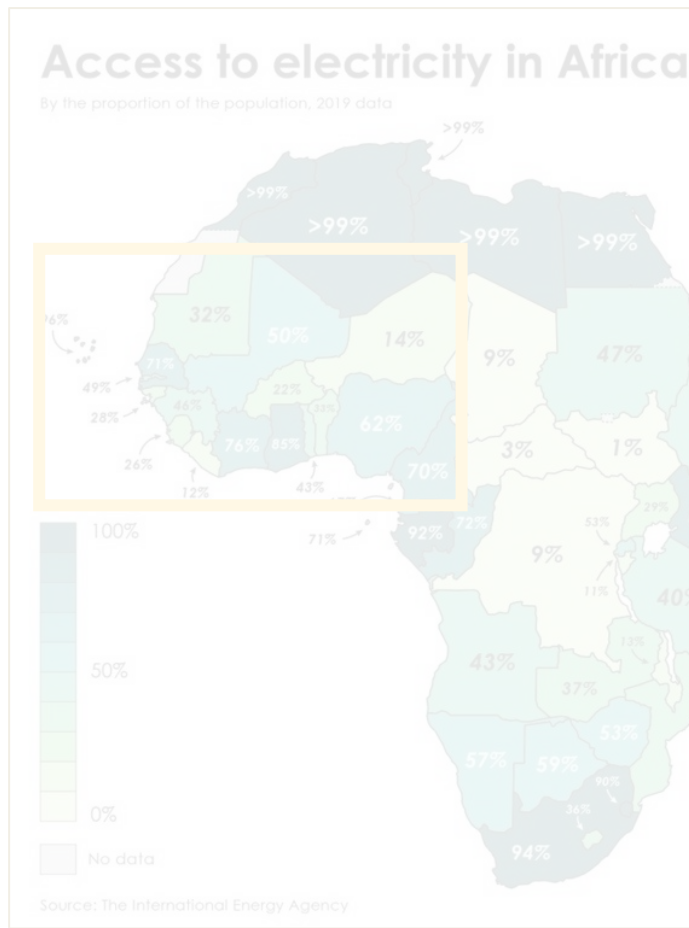
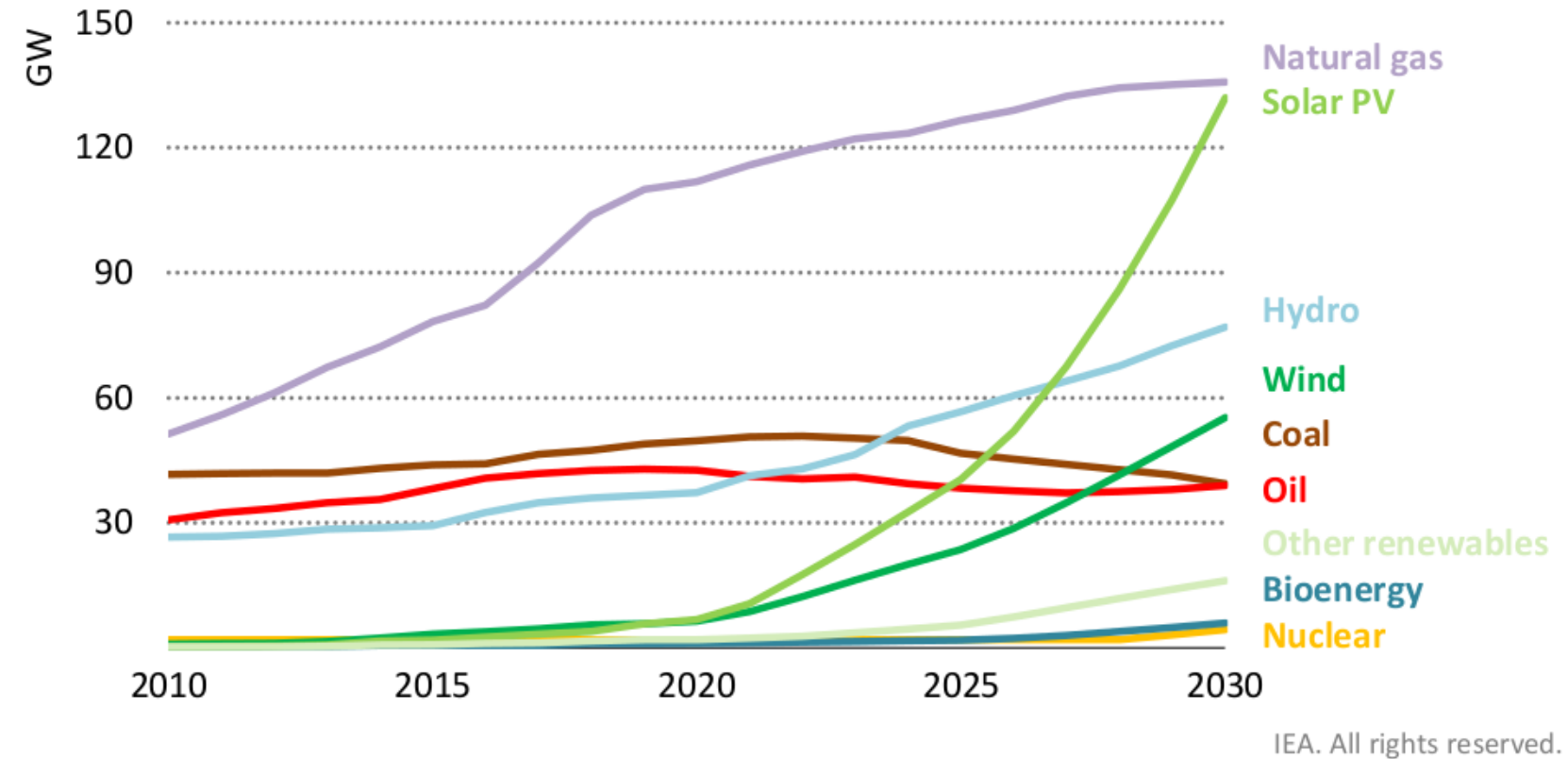


Figure 2.20 ▶ Installed power generation capacity by source in the SAS



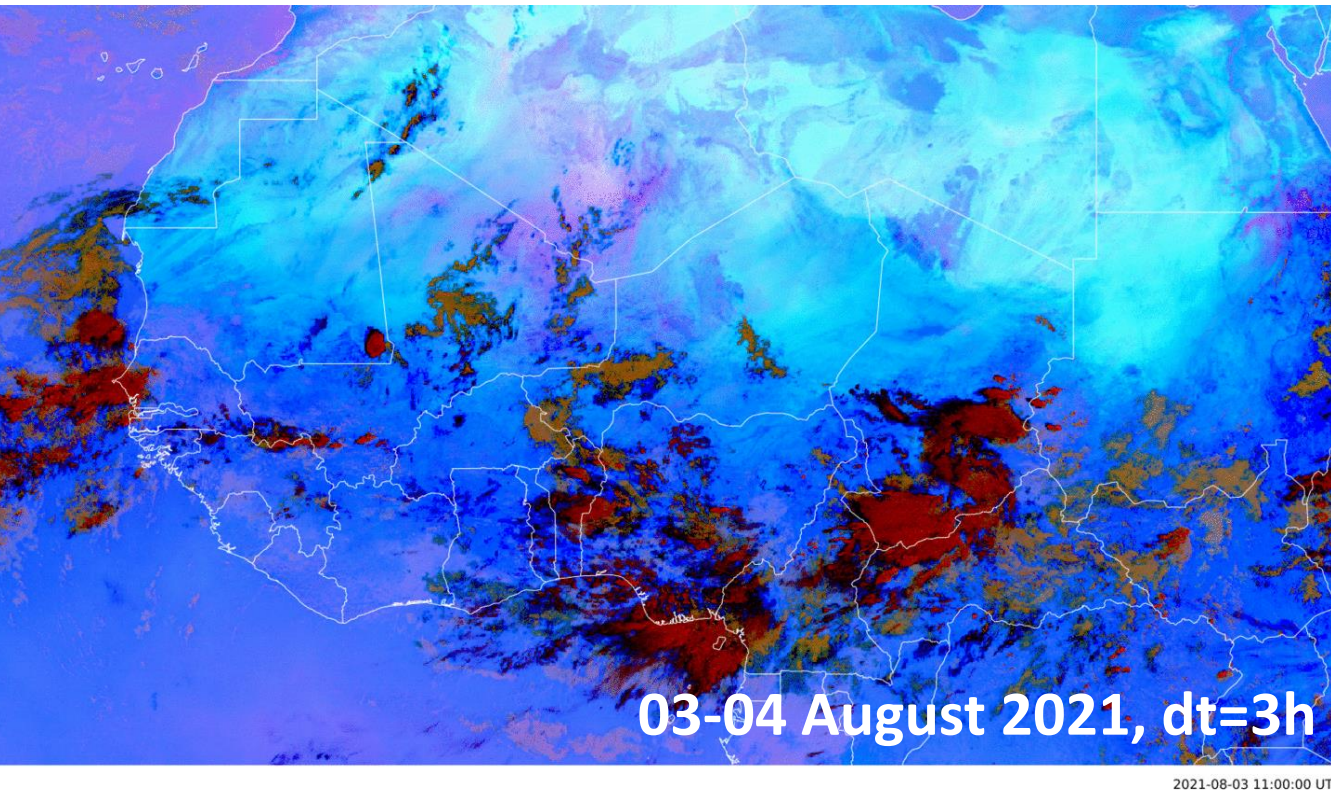
➤ Access to electricity is a key challenge in West African development

➤ Solar PV is a good candidate for answering the West African stakes

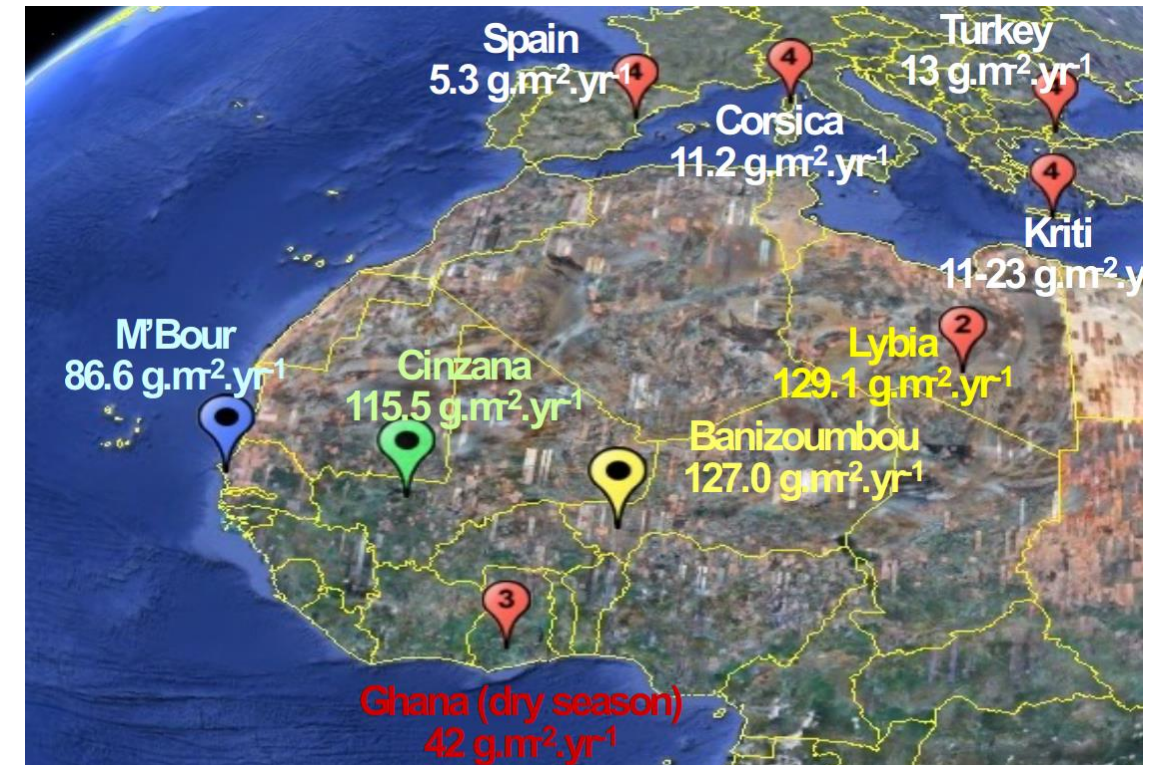
➤ Solar PV bring new constraints on the network => need for high-performance forecasting

West African solar forecast errors and their link with meteorological conditions

Meteorological constraints on solar radiation availability



RGB composite, MeteoSat Second Generation

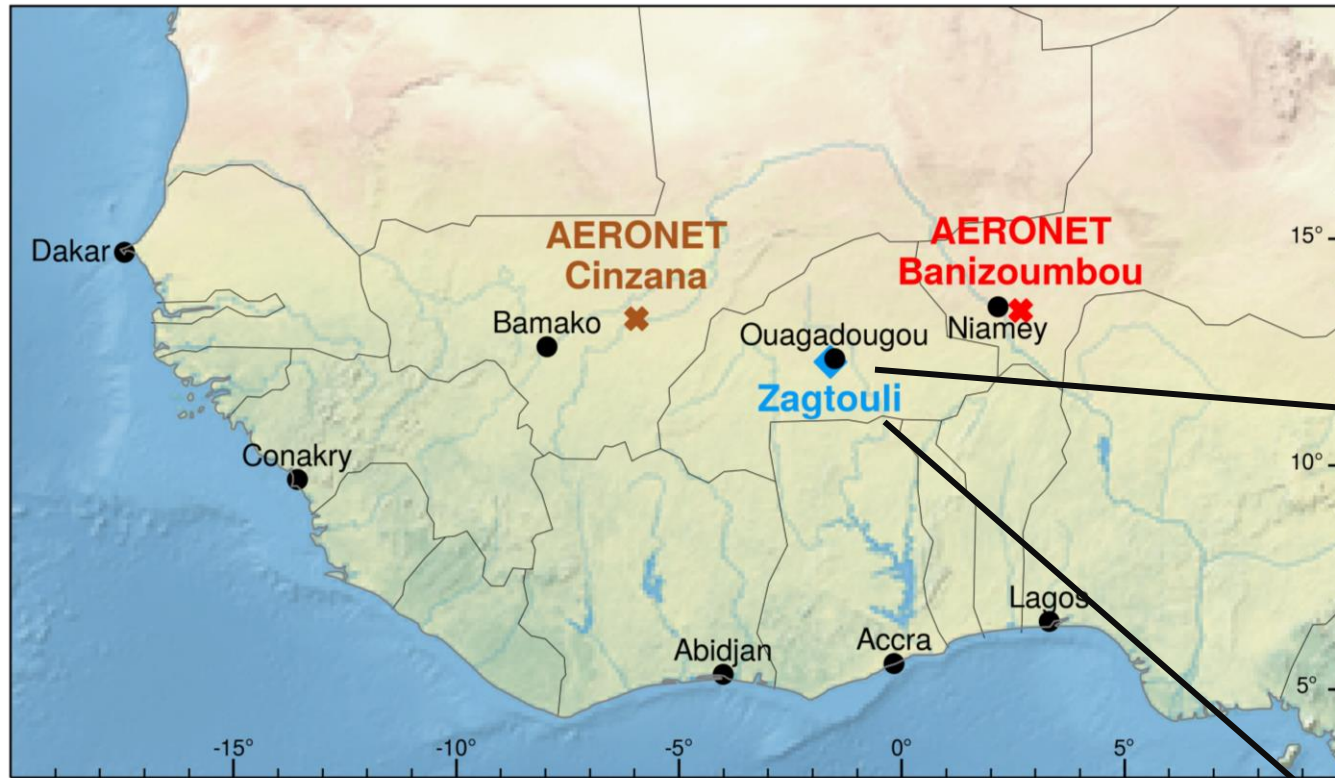


Dust deposit, INDAAF Network (Marticorena et al., 2017)

➤ Specific cloud dynamic (Mesoscale Convective System, ...)

➤ High dust load in the region

West African solar forecast errors and their link with meteorological conditions



Zagtouli solar farm (Burkina Faso)

- 34MWp
- 56GWh/year since 2018
- 4% of national electric production
- SONABEL (national electricity company)

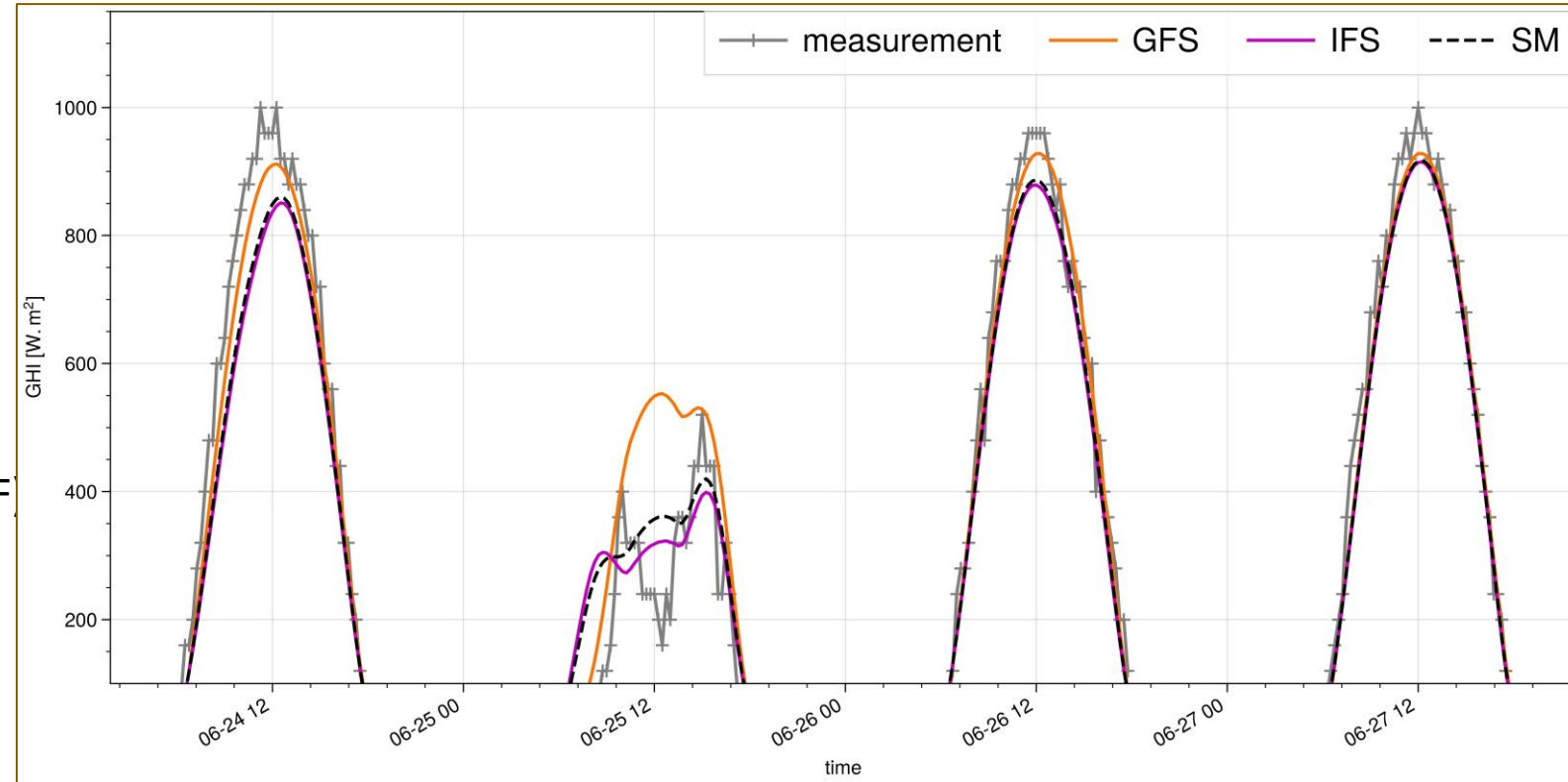


West African solar forecast errors and their link with meteorological conditions

Available data :

- **Pyranometer GHI measurements :**
15min – local
- **Global Forecast System (GFS, NCEP) :**
3h – 25km
- **Integrated Forecast System (IFS, ECMWF)**
1h – 9km
- **SteadyMet (SM, Steadysun) :**
Aggregation

=> Study period = 2021-2022

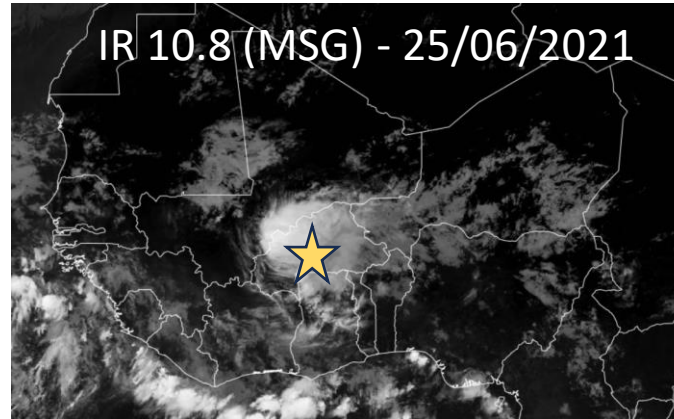


Zagtoui GHI day-ahead forecasts and measurement – 24 to 27 June 2021

Example of solar forecast defects and associated meteorological conditions

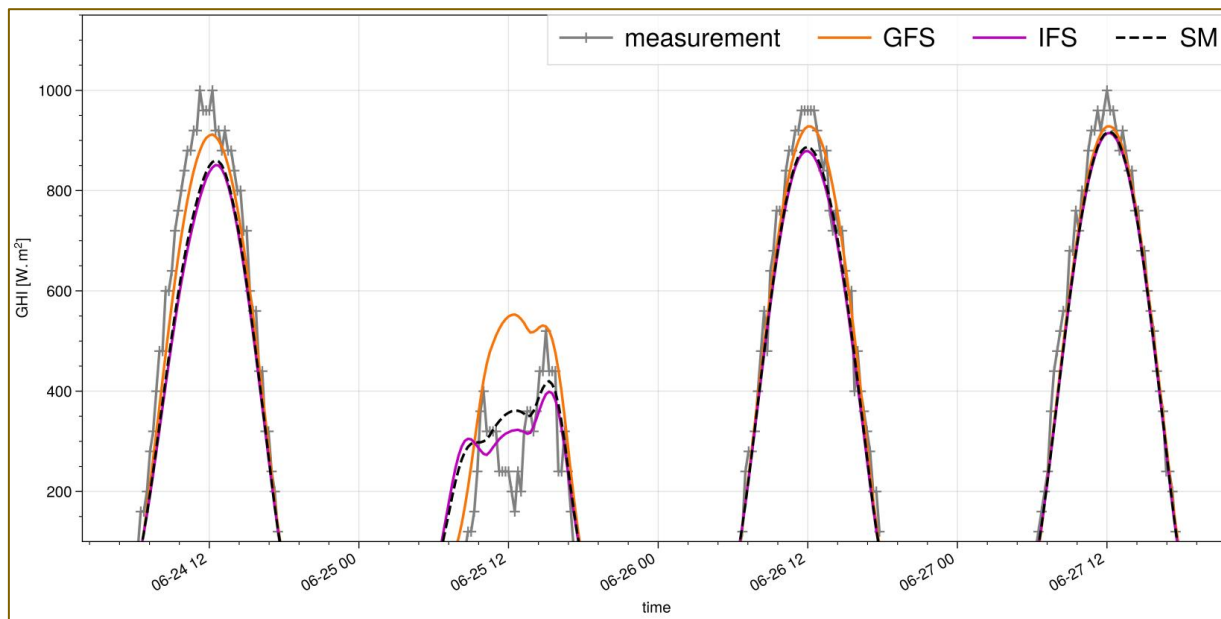
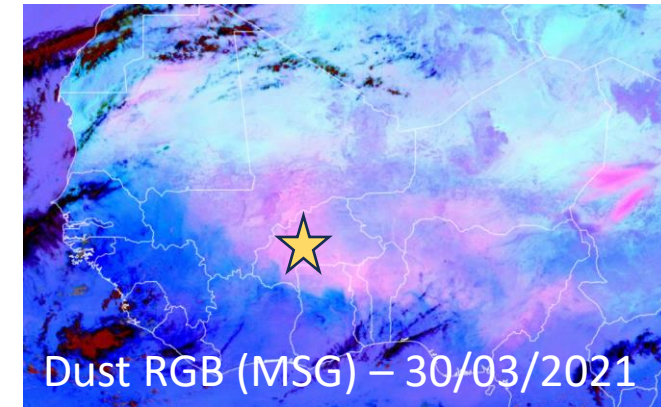
Case 1 :

- Cloud event
- Mesoscale Convective System



Case 2 :

- Dust event
- Emission from Bodele depression

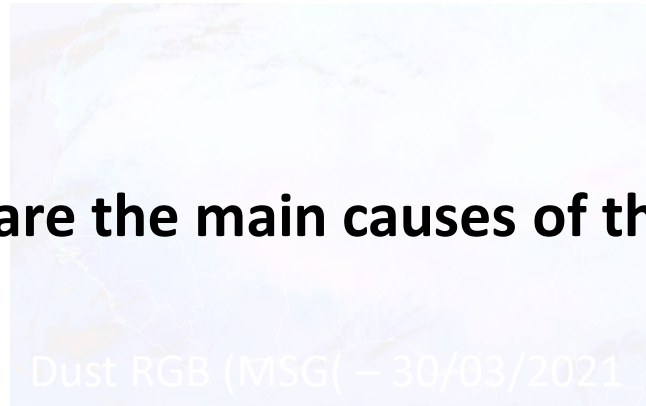


West African solar forecast errors and their link with meteorological conditions

Solar forecast defects and associated meteorological conditions

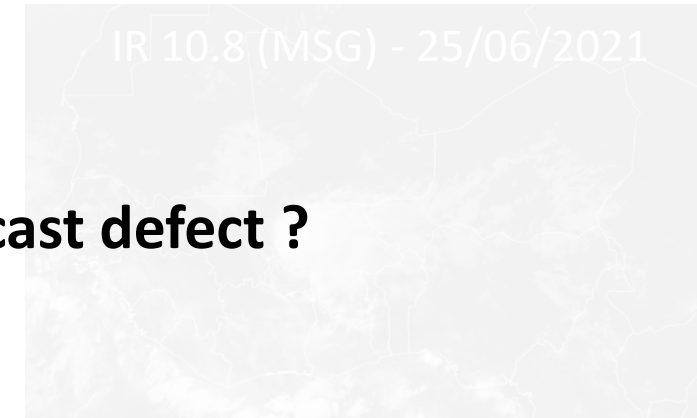
Case 1 :

- Dust event
- Emission from Bodele depression



Case 2 :

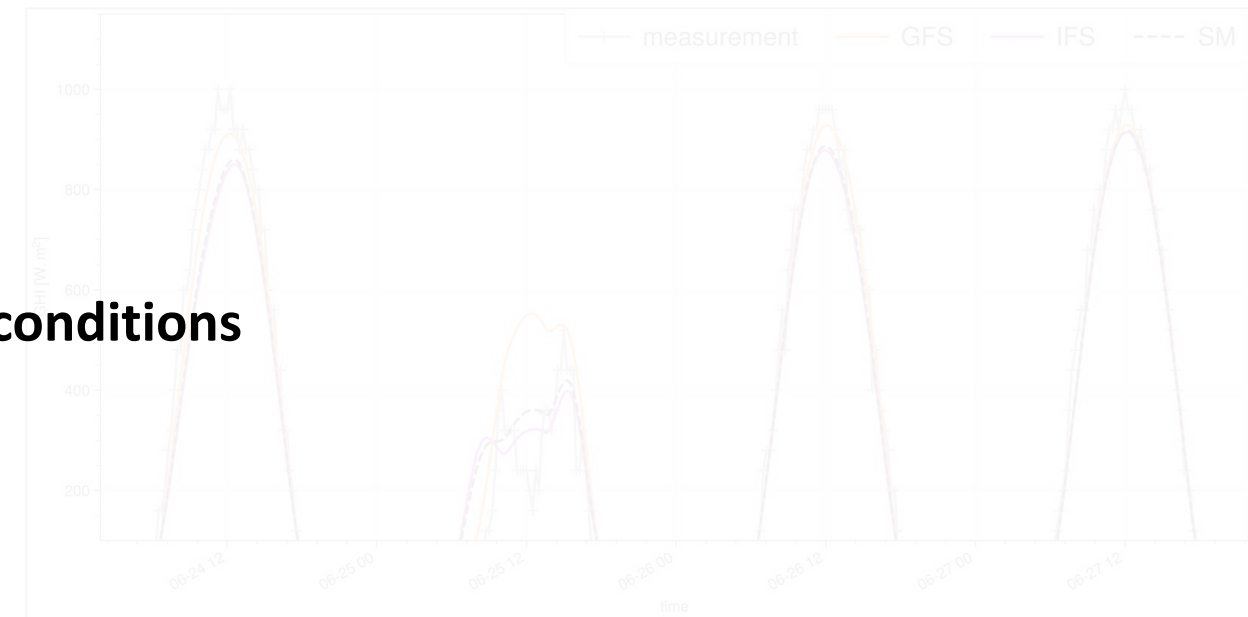
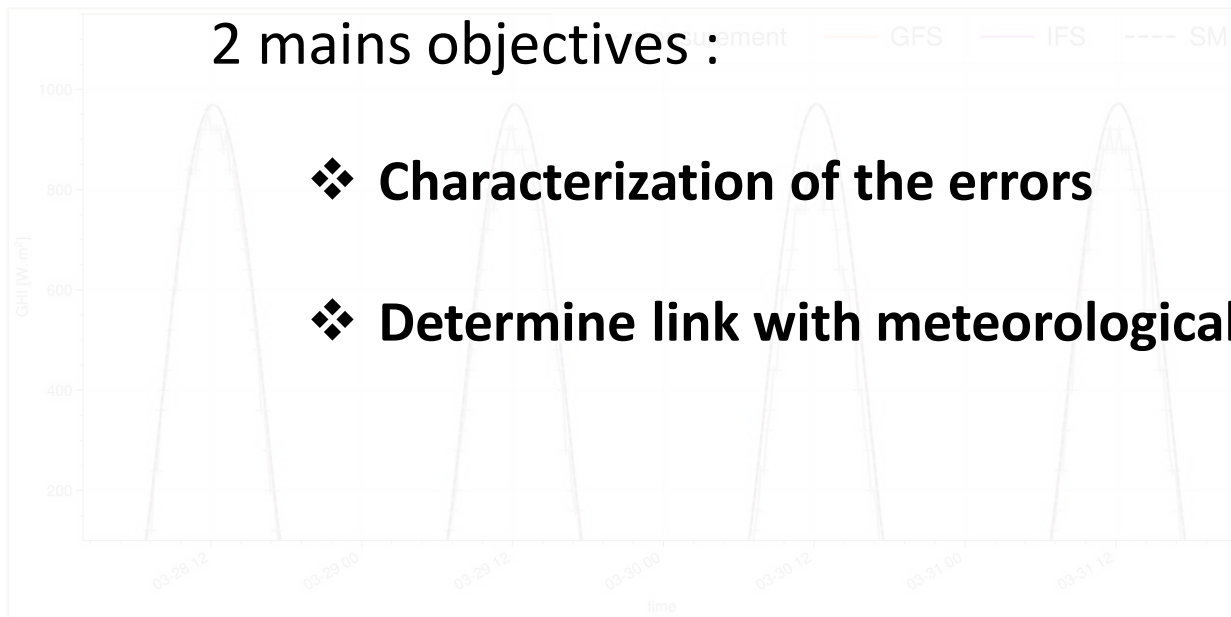
- Cloud event
- Convective System



What are the main causes of the solar radiation forecast defect ?

2 mains objectives :

- ❖ Characterization of the errors
- ❖ Determine link with meteorological conditions



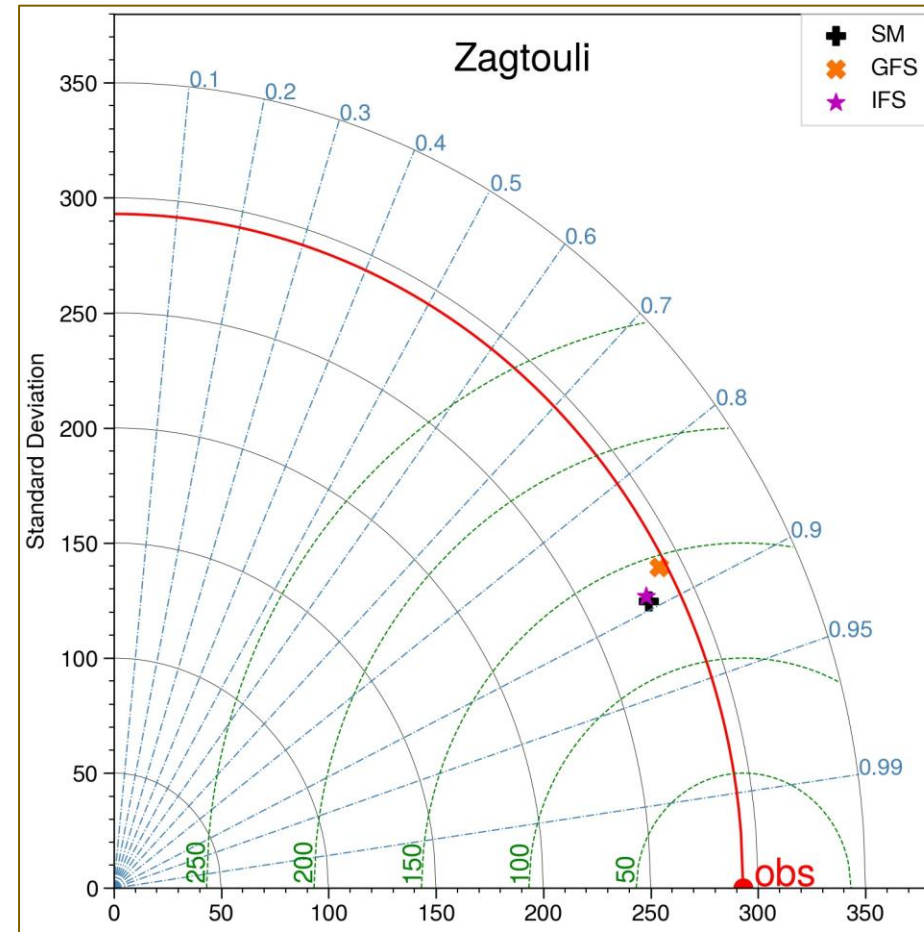
West African solar forecast errors and their link with meteorological conditions

1. Characterization of the errors

General statistics: comparison of 3 models

- GFS has the highest errors and lowest correlation
- SteadyMet performs slightly better than IFS
- GFS, IFS and SM have comparable performances

=> Advantage of aggregation
(multi-model, multi-member, multi-mesh)



Taylor diagram showing the evaluation of SteadyMet, GFS and IFS with ground GHI measurement

	MAE [W.m ²]
GFS	90.9
IFS	88.2
SM	86.8

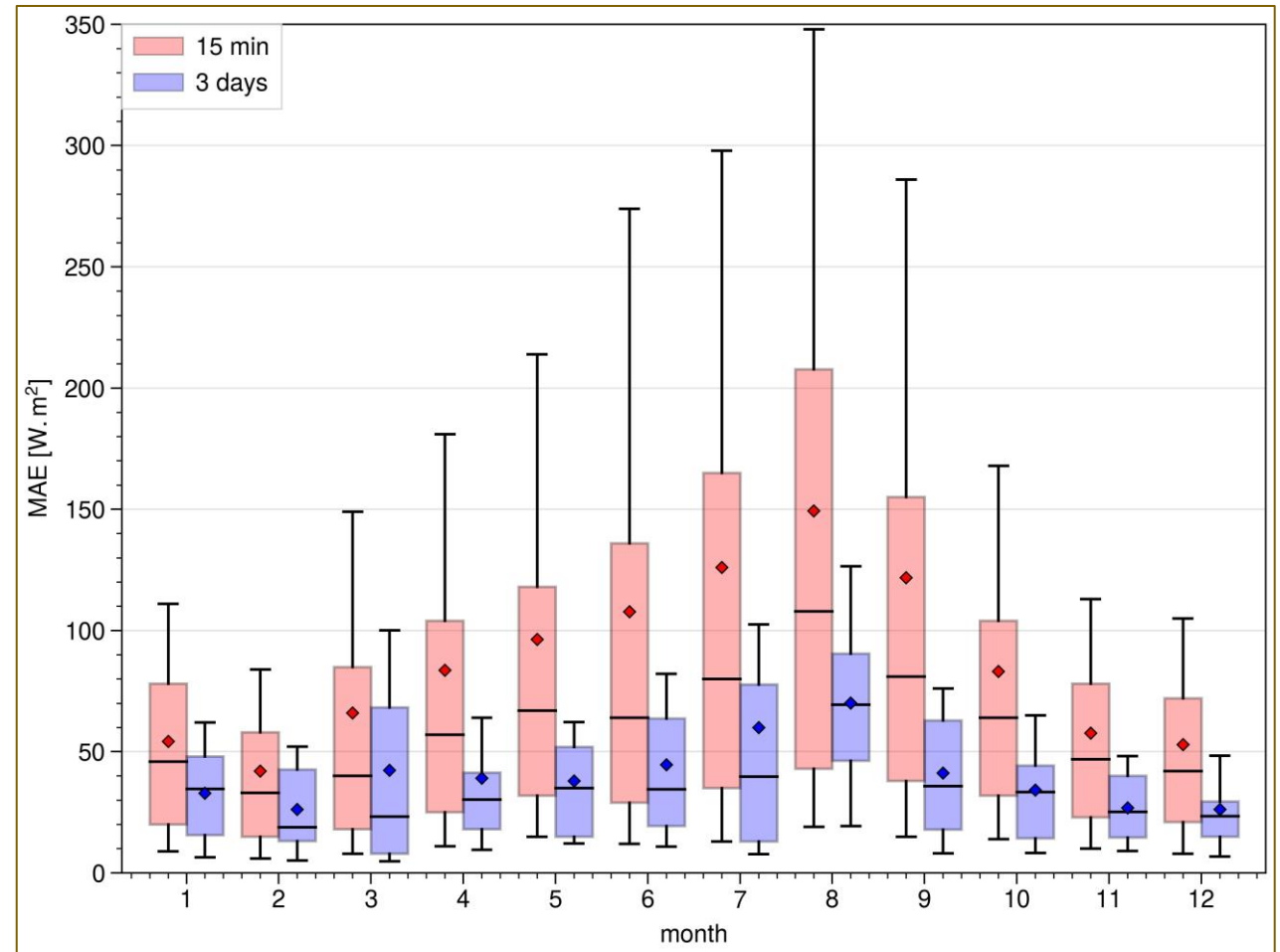
Mean daily MAE

1. Characterization of the errors

Error in function of the season

$$MAE_{dt} = \left| \left(\overline{measurement} \right)_{dt} - \left(\overline{forecast} \right)_{dt} \right|$$

- 15min-MAE : pic of mean value and dispersion in August (wet season)
- 3days-MAE :
 - ❑ Biggest pic of mean value in August and of dispersion in July (wet season)
 - ❑ Second pic of mean value and dispersion in March (dry season)



Monthly boxplot for MAE integrated on different time scale.

2. Link with meteorological conditions

Meteorological data :

A) Aerosol indicators

- **AERONET AOD** = ground measurement, local, dt=5min (clear sky)
- **MODIS AOD** = satellite observation, dx=1°, dt=daily (clear sky)
- **CAMS Dust OD** = atmospheric reanalysis, dx=0.75°, dt=3h (all time)

=> **validation of CAMS** with AERONET and MODIS

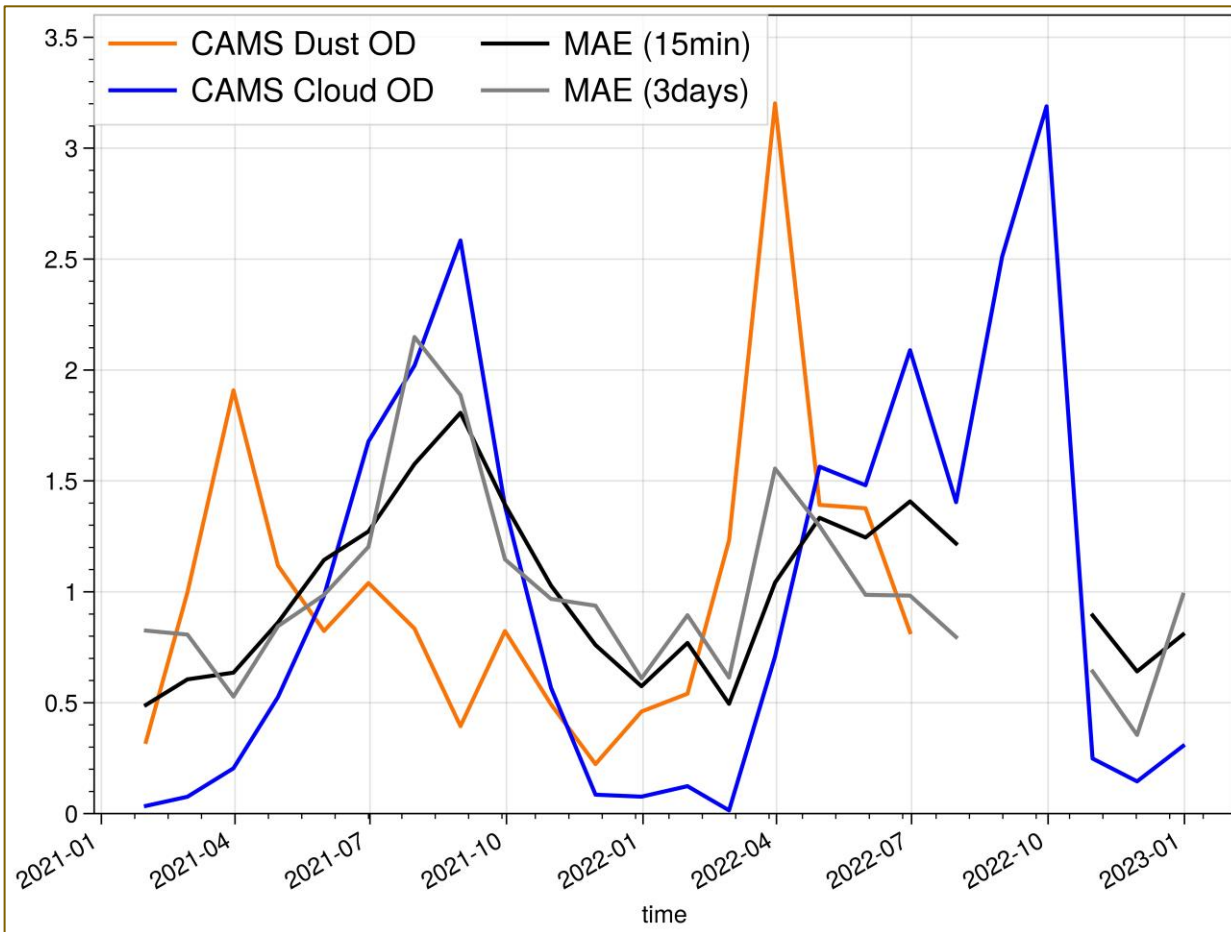
=> CAMS advantage : consistency, day and night, cloudy and clear sky days

B) Cloud indicators

CAMS solar radiation time-series :

- derived from Meteosat Second Generation satellite observation
- Cloud Optical Depth, dx=5km, dt=15min

2. Link with meteorological conditions



Normalized monthly mean of errors and meteorological indicators

Seasonality comparison

- 15min-MAE and Clouds both pic during the wet season (July-October)
- Dust pics in the late dry season (April)
- 3days-MAE shows a pic in the wet season (2021) and a pic in April 2022

=> Cloud may be the major cause of defect in solar forecast model

=> Depending on the time scale, dust may have a significant impact

2. Link with meteorological conditions

		Cloud Optical Depth (COD)	Dust Optical Depth (DOD)
MAE	3 hours	0.55	0.01
	1 day	0.61	0.04
	3 days	0.50	0.13

Table 1. Pearson correlation coefficient between MAE integrated for different timesteps with Dust and Clouds indicators

- Correlations are significantly higher for COD than for DOD
- Correlations are higher for 1day-MAE with COD and for 3days-MAE with DOD
 - Forecast errors are linked with clouds event

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		Cloud Optical Depth (COD)	Dust Optical Depth (DOD)	COD & DOD
MAE	3 hours	52%	28%	11%
	1 day	61%	27%	17%
	3 days	50%	31%	16%

Table 2. Co-occurrence between the 30% most critical errors and the 30% highest values of Cloud and Dust indicators, for different integration timesteps.

- Co-occurrences with Clouds are higher than with Dust
- Co-occurrences with Dust are not negligible, especially with 3days-MAE
- Above 10% of major errors occurred with Dust and Clouds
 - The Dust impact should not be underestimated (Aerosol-Cloud Interaction)

Main conclusions :

- i. The three models have similar overall performances, SteadyMet is slightly better (contribution of aggregation)
- ii. Errors are highest **during the wet season**, but when integrated over several days we also see a **secondary peak during the late dry season**
- iii. Correlations suggest that the presence of **clouds explains the most critical errors**, but co-occurrences show that the presence of **aerosols has a secondary impact**

Perspectives :

1. Improving the knowledge on the **impact of meteorological condition** :
==> Strengthen the measurement networks on site (AOD, cloud mask) to have a better view on the local conditions
2. Taking into account **dust deposition impact** on the solar PV production
==> Reduce the dust deposition impact by determining an optimize cleaning strategy
3. Improving the **solar forecast performances**
==> Development of a high-resolution forecast model that takes into account dust particles (WRF-CHIMERE)



NETWAT project :

See poster 48 : *West African Mineral Dust: a key in the NExus ClimaTe – WATer – Energy* (Sandrine Anquetin, Guillaume Tremoy), presented on Tuesday

Thanks for your attention !

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