



## **3rd SFERA Summer School**

### **Solar resource forecasting**

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**B. Pulvermüller, Solar Millennium AG, Erlangen; M. Hoyer, D. Glumm, Flagsol GmbH**

**B. Kraas, CSPServices (former Solar Millennium AG)**

**J. Remund, Meteotest**





# Outline

- Forecast requirements
- Basic meteorological terminology and forecast methodologies
- Numerical weather prediction – verification measures, spatial error correlation
- Do we need aerosol forecasts?
- Possibilities of ground measurements?
- Satellite-based nowcasting
- Some thoughts on the economic value of forecasts (only market participation case)
  
- Credits to
  - EC – Global Monitoring for Environment and Security (GMES)  
    MACC Atmosphere core service preparations project  
    ENDORSE (Energy downstream services) project
  - ESA – Integrated Application Program – CSP-FoSyS project
  - EC DG ENV – EnerGEO project



This talk concentrates on

- Direct normal irradiance forecasting
- Focus on hourly resolution
- Less than hourly resolved forecasting
  
- No solar power output forecast  
= fct (power plant state, plant technology)
- No other meteo parameters  
(wind gust, temperature, wind)
  
- This overview concentrates on principles
- It does not provide many information on  
typical accuracies (look into the additional slides)



# Use cases in grid integration studies

## Transmission System operator (TSO) point of view

- grid stabilisation
- 15 minutes interval
- up to 1-2 hour forecast needed

## Power plant operator point of view (especially for CSP)

### Grid access license

To be considered as a predictable power source  
("producción gestionable"):

- 10% accuracy for 24 hour forecast needed
- 5% accuracy for 6 hour forecast needed
- Predictability test ("prueba de gestionabilidad")

### Plant operation

Optimization of plant operation using forecast systems:

- Power Block Control and security warnings  
(15-180 minutes)
- Plant operation planning (24-48 hours)
- Maintenance planning (1-7 days)

### Market participation

Participation on the Spanish electricity market (OMEL):

- Day ahead market (obligatory)
- 6 intraday market sessions
- False programming results in penalties

# Forecast parameters

Importance of meteorological parameters that have to be forecasted for a CSP Power Production prevision

## Parameter:

- Direct solar radiation
- Wind gust
- Temperature
- Wind speed

## Importance:

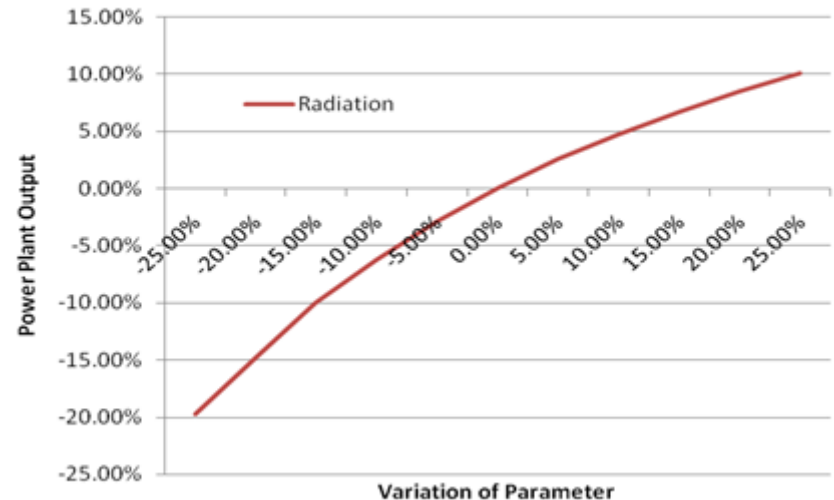


Based on 1 year measurements  
at the Andasol site

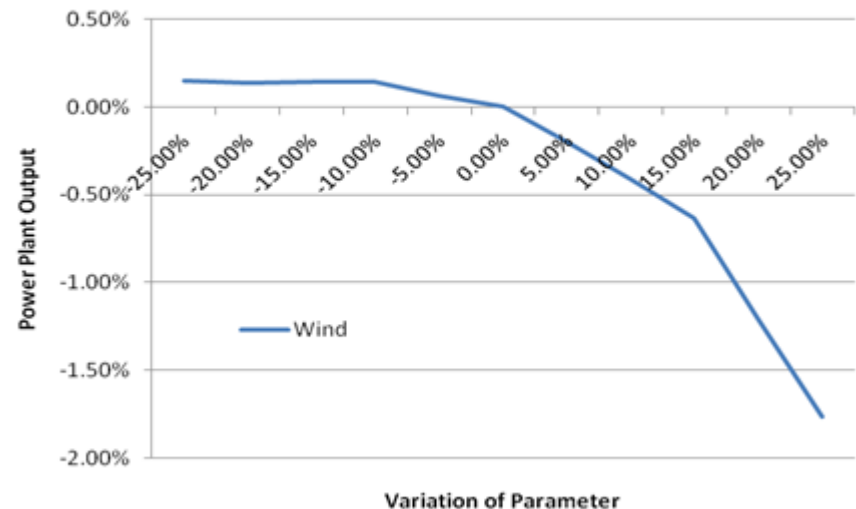


Deutsches Zentrum  
für Luft- und Raumfahrt e.V.  
in der Helmholtz-Gemeinschaft

## Influence of Forecast Parameters



## Influence of Forecast Parameters



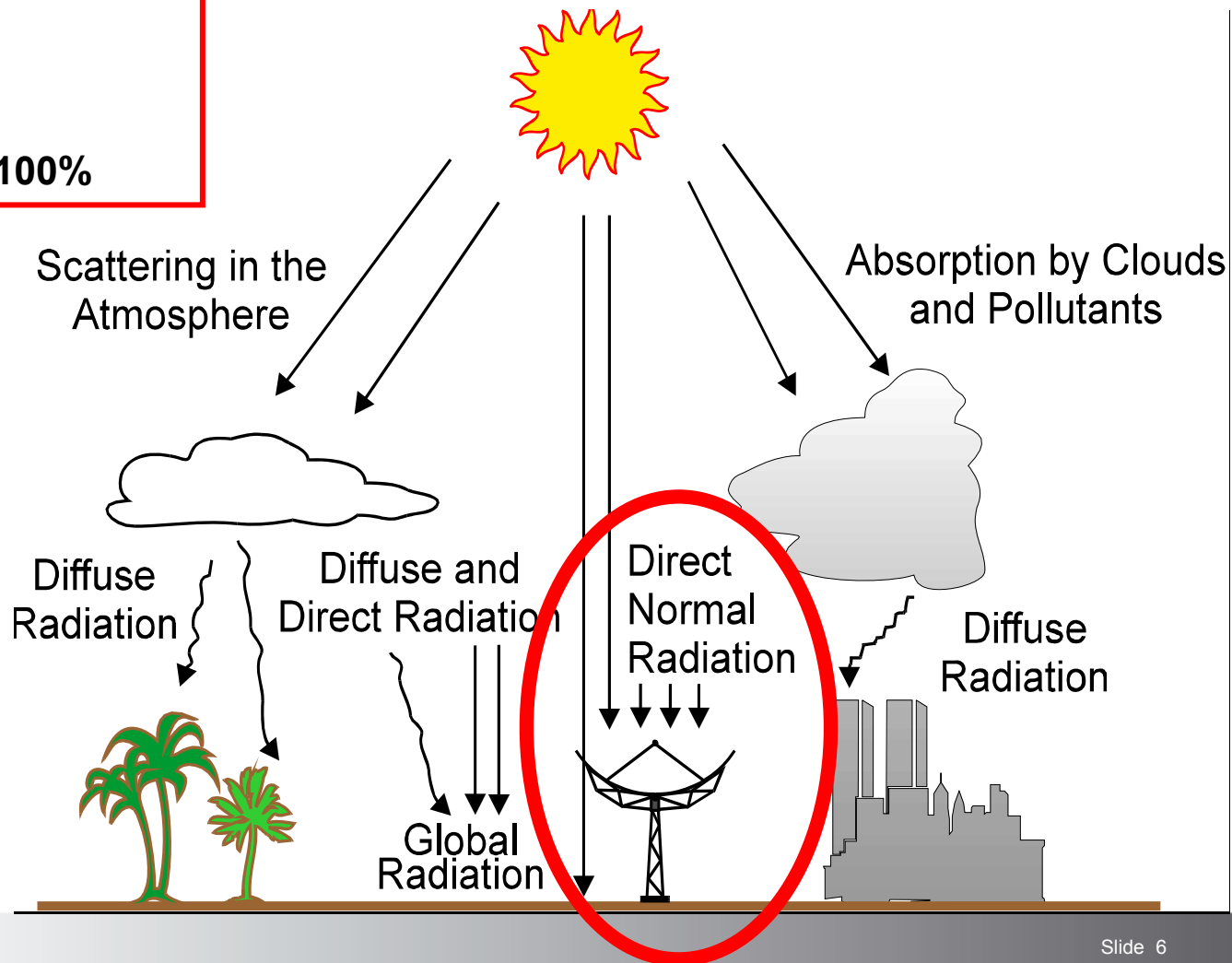


## Reduction of direct irradiance

**Clouds up to 100%**

**Water vapour 10-15%**

**aerosols 20-25%, up to 100%**



Hindcast

historical forecasts  
good for assessments/development

Nowcast

= up to 3 hours

Shortterm forecast

= up to 6 hours

Day ahead forecast

= up to 48 hours

Medium range forecast

= up to 10 days

Climate scenarios

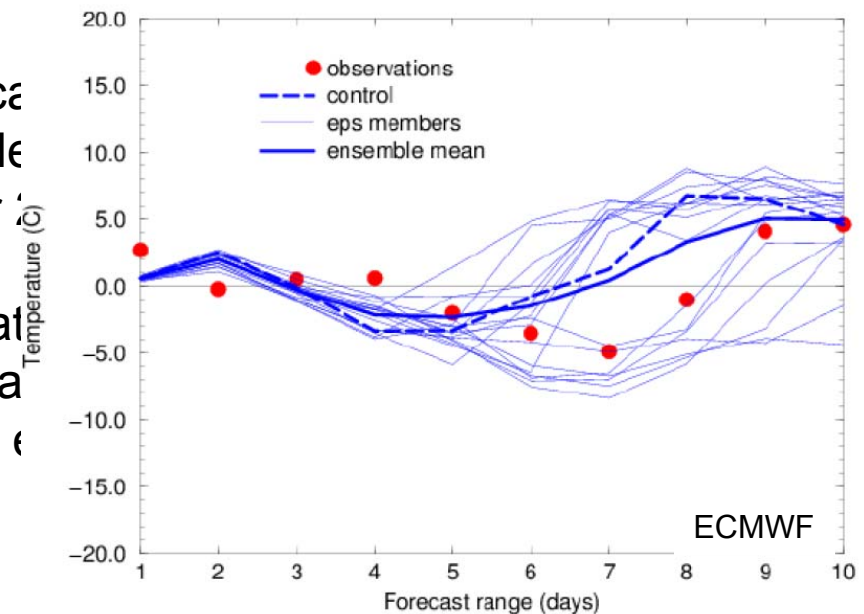
NOT a forecast  
a scenario development  
e.g. 2050 or 2100

Deterministic forecast

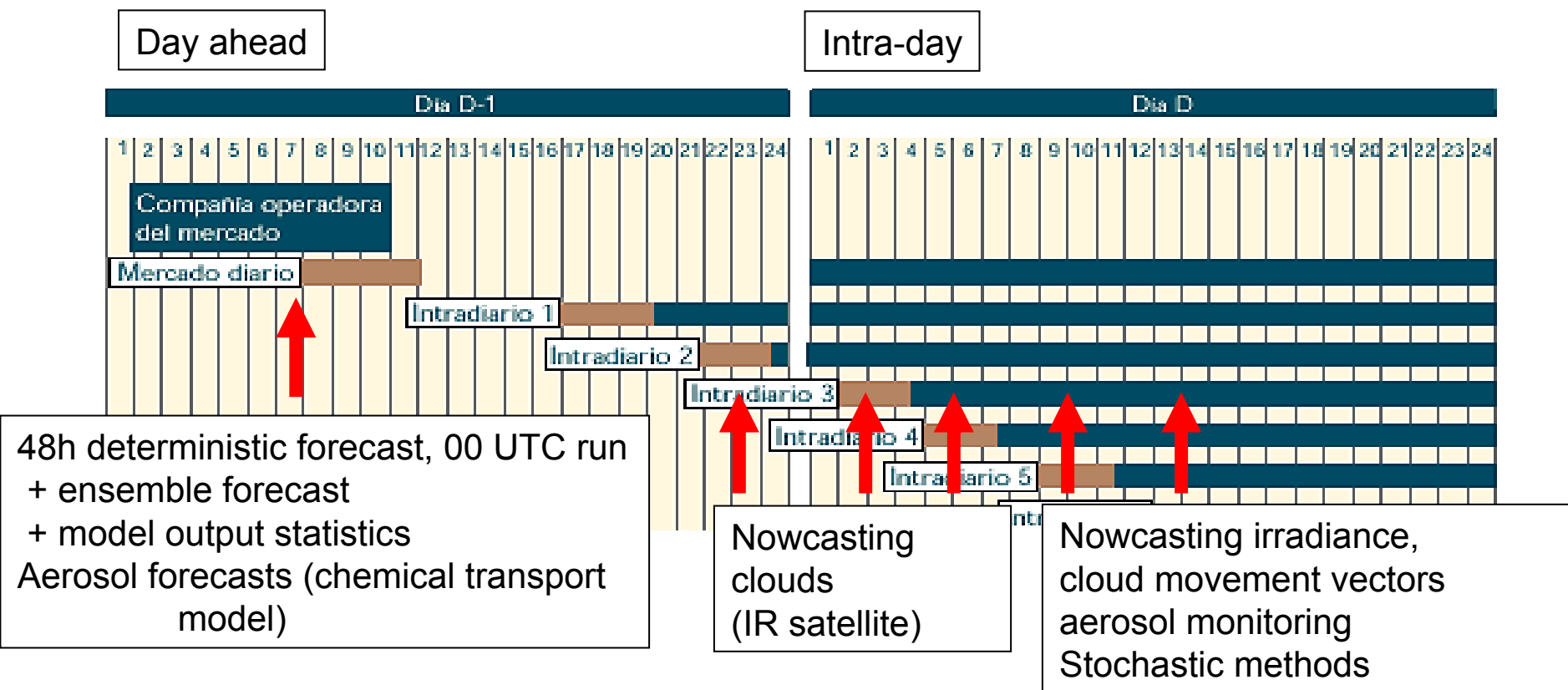
a single weather forecast

Ensemble forecast

30 or 50 weather forecasts  
probabilistic forecast



# CSP operator point of view – example case Spain



Large open question:

Use and value of forecasts in plant operations besides day ahead and intraday markets .... Not treated so far...





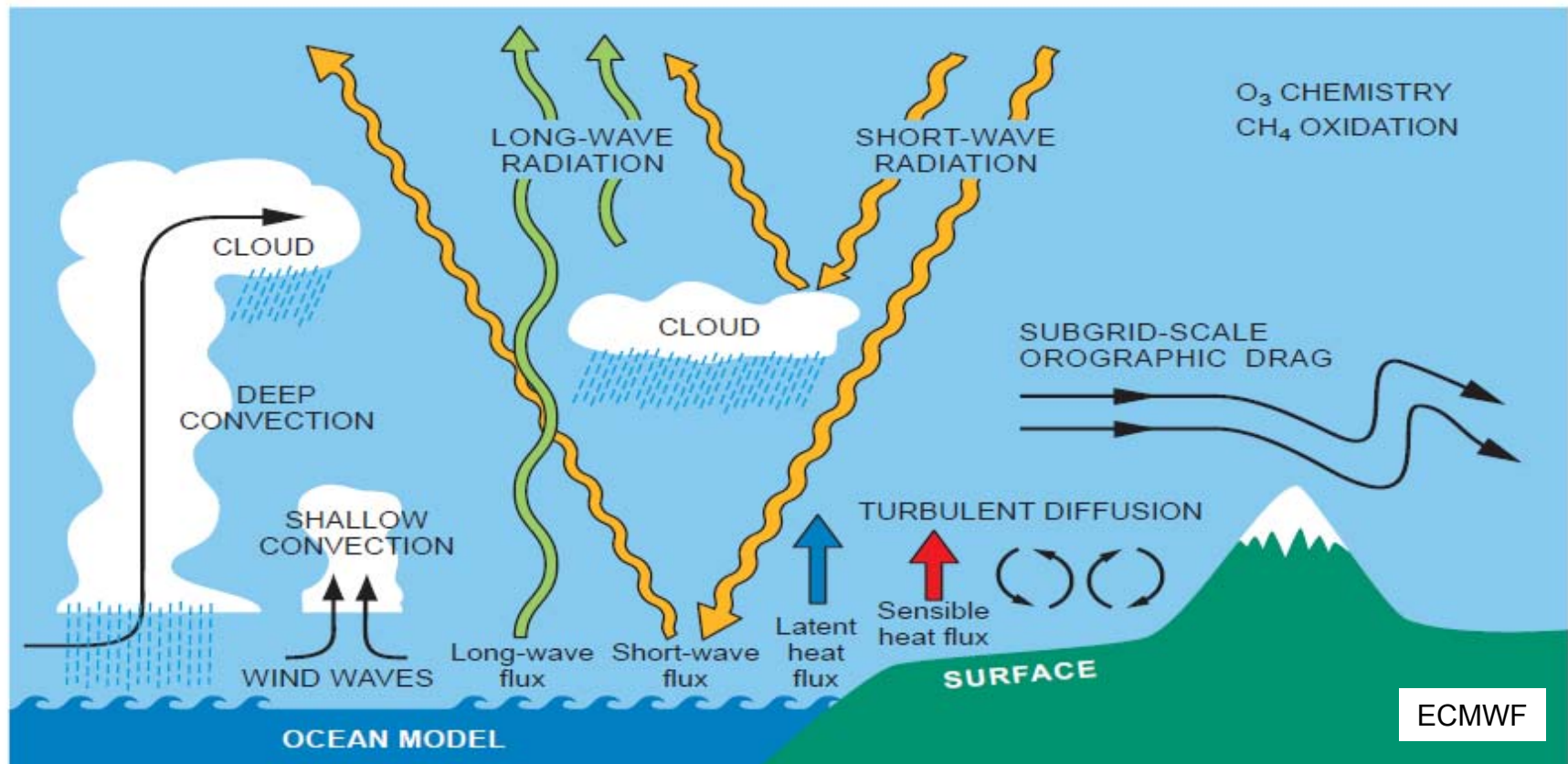
# Methods and results



# Physical methods

Apply physical principles (atmospheric dynamics, thermodynamics, radiative transfer, optical properties)

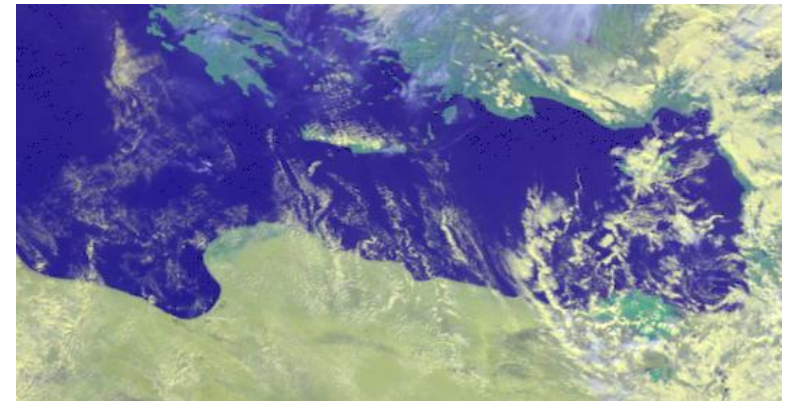
Numerical weather prediction



# Physical methods

## Satellites

optical measurements in visible and infrared spectrum  
optical properties like refractive indices  
radiative transfer modeling -> radiation



Forecast based on either NWP through data assimilation  
or on statistical procedures (next slide)



# Stochastic learning techniques

## Characteristics:

- rely on past data to train models, no/little physical assumptions
- assumption that future irradiation can be predicted based on historical patterns
- assume persistence in opacity, direction and velocity of clouds

## Methods

- persistence approach ,yesterday == tomorrow' (2 day sometimes needed)
- auto-regressive models
- artificial intelligence/neural network
- cloud motion vectors (cloud camera/satellites)

## Advantages

- Good for intra-hour or nowcasting
- cheap



# Characteristics of solar forecasting techniques

Technique	Sampling rate	Spatial resolution	Spatial extent	Maximum Suitable Forecast horizon	Application
Persistence	High	One point	One Point	Minutes	Baseline
Whole Sky Imagery	30 sec	10 to 100 meters	3-8 km radius	10s of minutes	Ramps, regulation
Geostationary satellite imagery	15 min	1 km	65°S – 65°N	5 hours	Load following
Numerical weather prediction (NWP)	1 hour	2 - 50 km	Worlwide	10 days	Unit commitment regional power prediction

Source: Remund, 2012



## Circum-solar taken into account?

- Typical textbook definition:

*the radiant flux collected by a surface normal to the direction of the Sun, within the extent of the solar disk only (half-angle  $\delta = 0.266^\circ$ )*

- Three “worlds” : three definitions of DNI!

- ▣ World of Radiative Models (e.g. 6S, libRadtran):  
Solar source is a Dirac

NWP,  
some SAT

- ▣ World of Physical Measurements: definition of DNI by the corresponding measurement device. From WMO CIMO Guide (2008):

*“Direct solar radiation is measured by means of pyrheliometers, the receiving surfaces of which are arranged to be normal to the solar direction. By means of apertures, only the radiation from the sun and a narrow annulus of sky is measured, the latter radiation component is*

Some  
stochastic  
methods

- ▣ sometimes referred to as circumsolar radiation or aureole radiation”

- ▣ World of Solar Energy Conversion: definition of DNI by the corresponding concentrating solar system (0.9 to  $5^\circ$ )



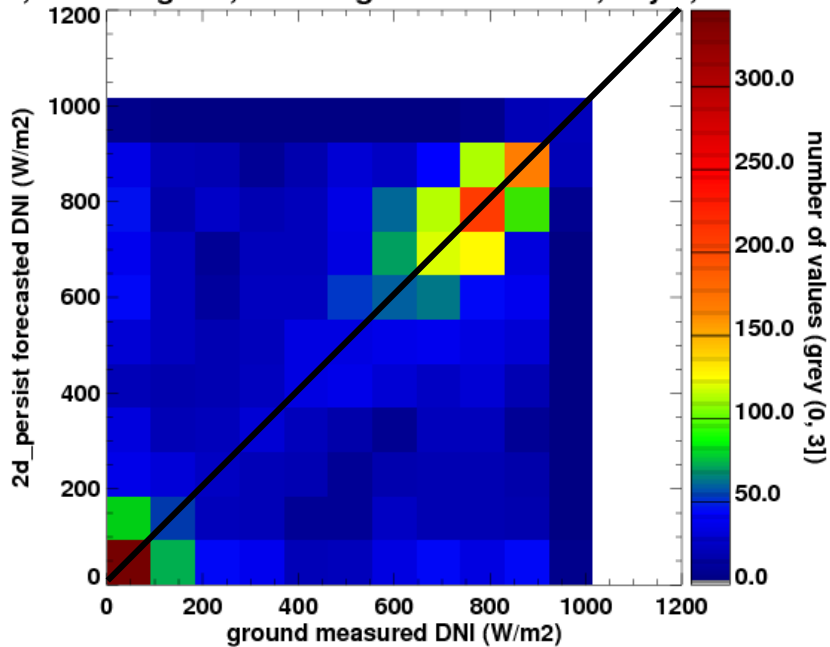


# Numerical weather prediction

- Model runs initiated two to four times per day (e.g. 0, 6, 12 and 18 UTC)
- initial conditions from satellite, in-situ, ground observations
- Pre-processed and interpolated to the 3D grid
- resolution of global NWP models is coarse (15 to 90 km)
- mesoscale or limited area models: limited geographical area with higher resolution (up to few km)
- Trouble with mesoscale models: Place clouds still at the wrong place and have RMSE worse than global ECMWF model.
- ECMWF = European Centre for Middle-Range Forecasts
- Since 2011 ECMWF provides also direct irradiance forecasts
- Most models provide only global -> statistical global to direct conversion needed

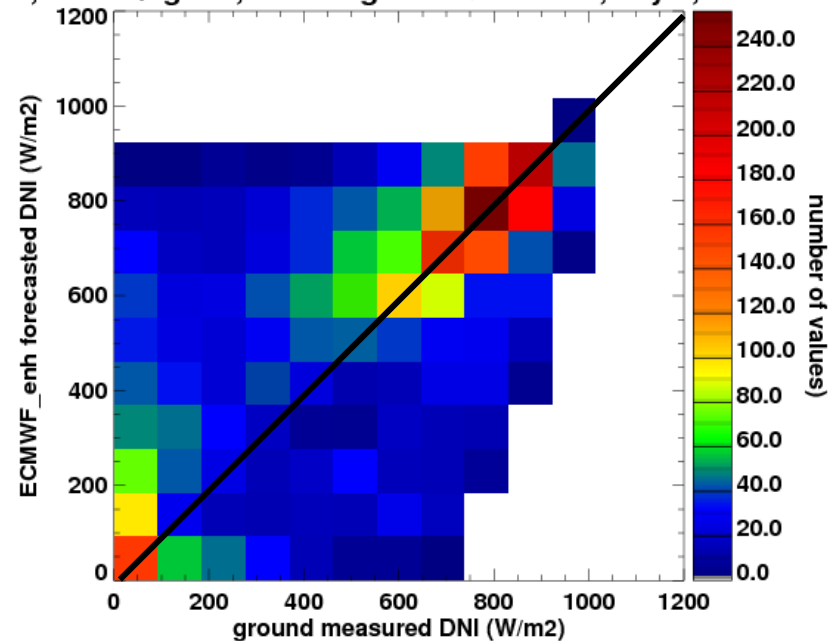
# Verification ECMWF/DLR : Scatterplots for hourly values

PSA, 2D histogram, FC and ground > 0 W/m2, day 2, N = 4191



2 day persistence,  
2005

PSA, 2D histogram, FC and ground > 0 W/m2, day 2, N = 4079



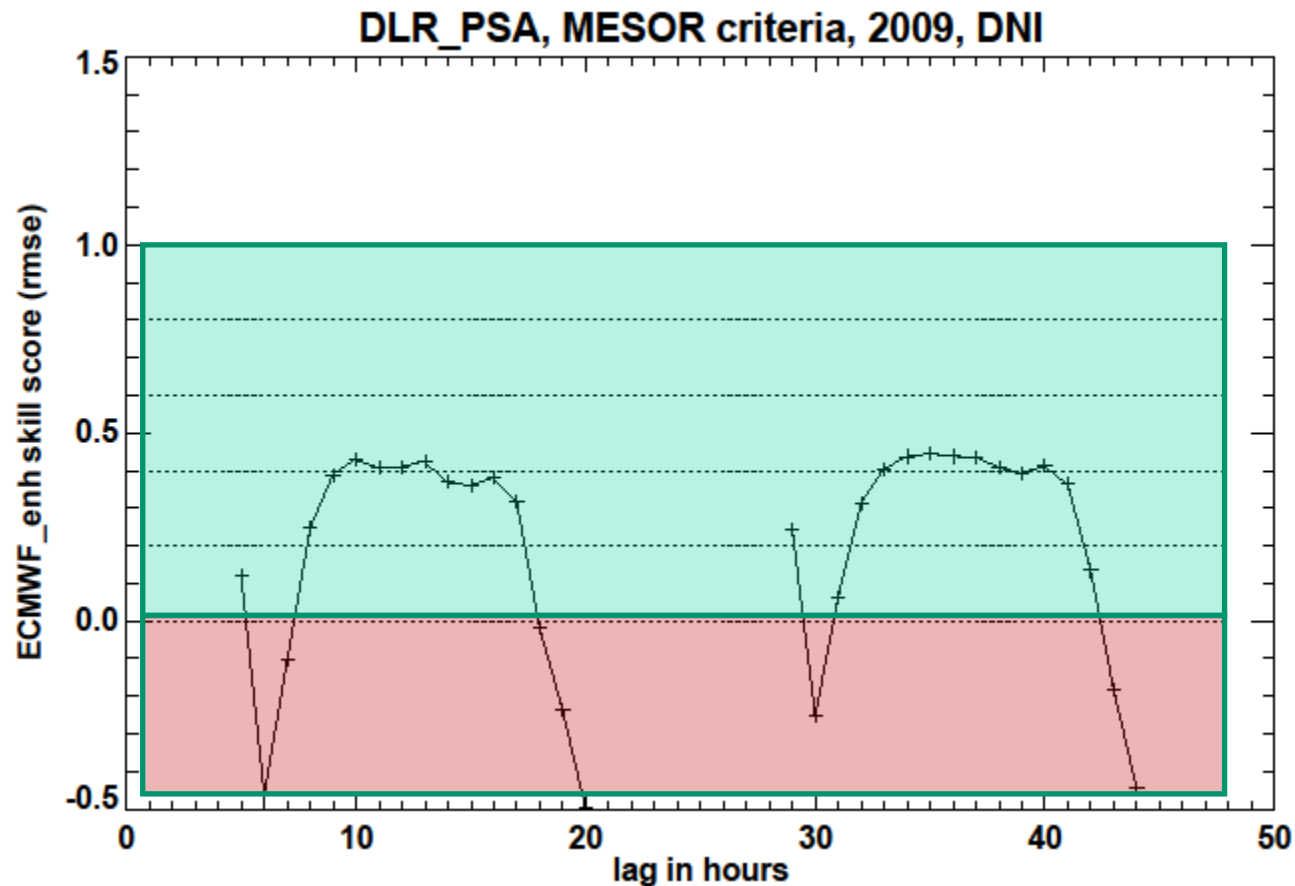
ECMWF/DLR,  
2005



# Verification ECMWF/DLR

$$\text{skill score} = \frac{\text{Metric}_{\text{reference}} - \text{Metric}_{\text{forecast}}}{\text{Metric}_{\text{reference}} - \text{Metric}_{\text{perfect forecast}}}$$

$$\text{Skill score} = (\text{RMSE}_{\text{fc}} - \text{RMSE}_{\text{pers}}) / \text{RMSE}_{\text{pers}}$$



Bad morning  
due to  
3-hourly  
forecast

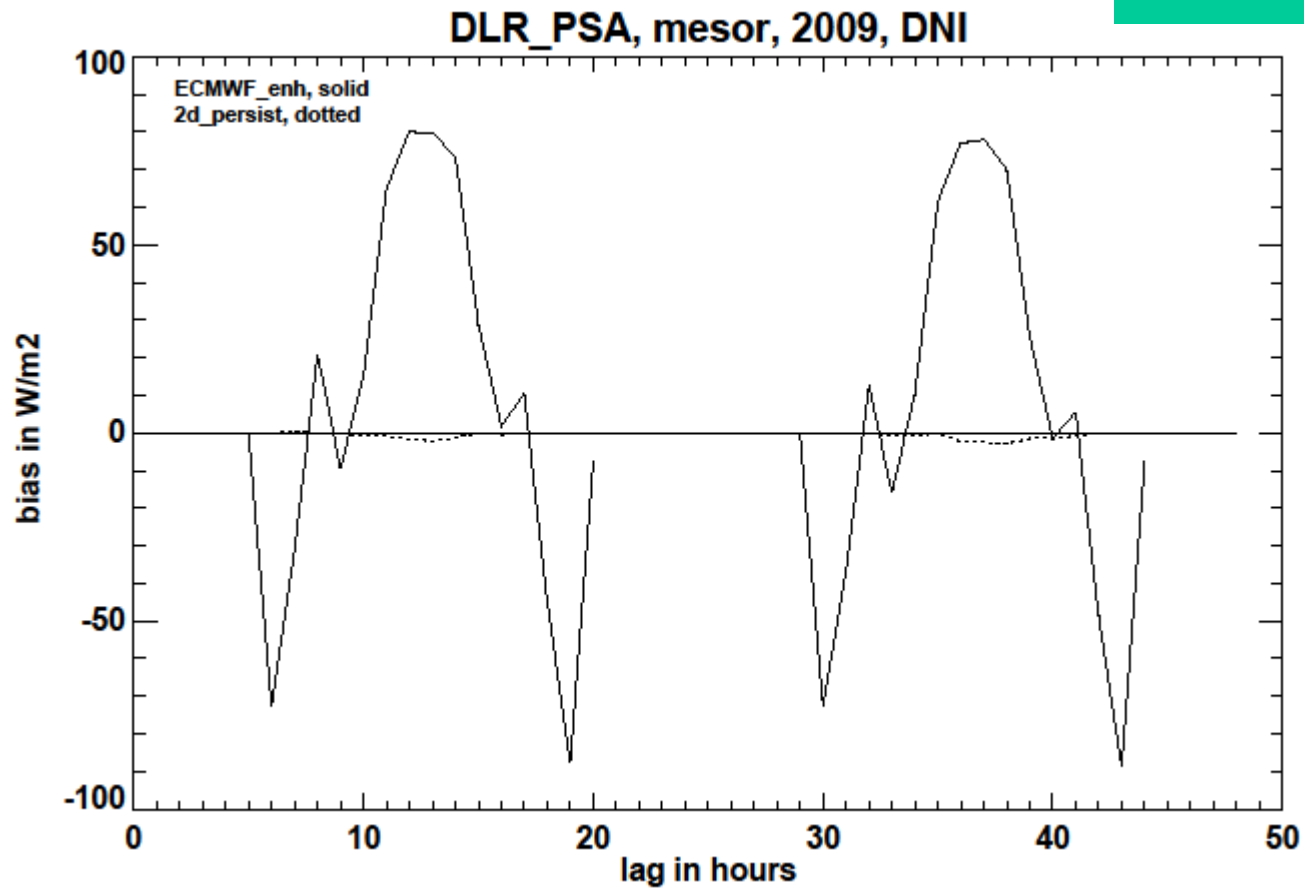
Better  
than  
persistence

worse

# Verification ECMWF/DLR

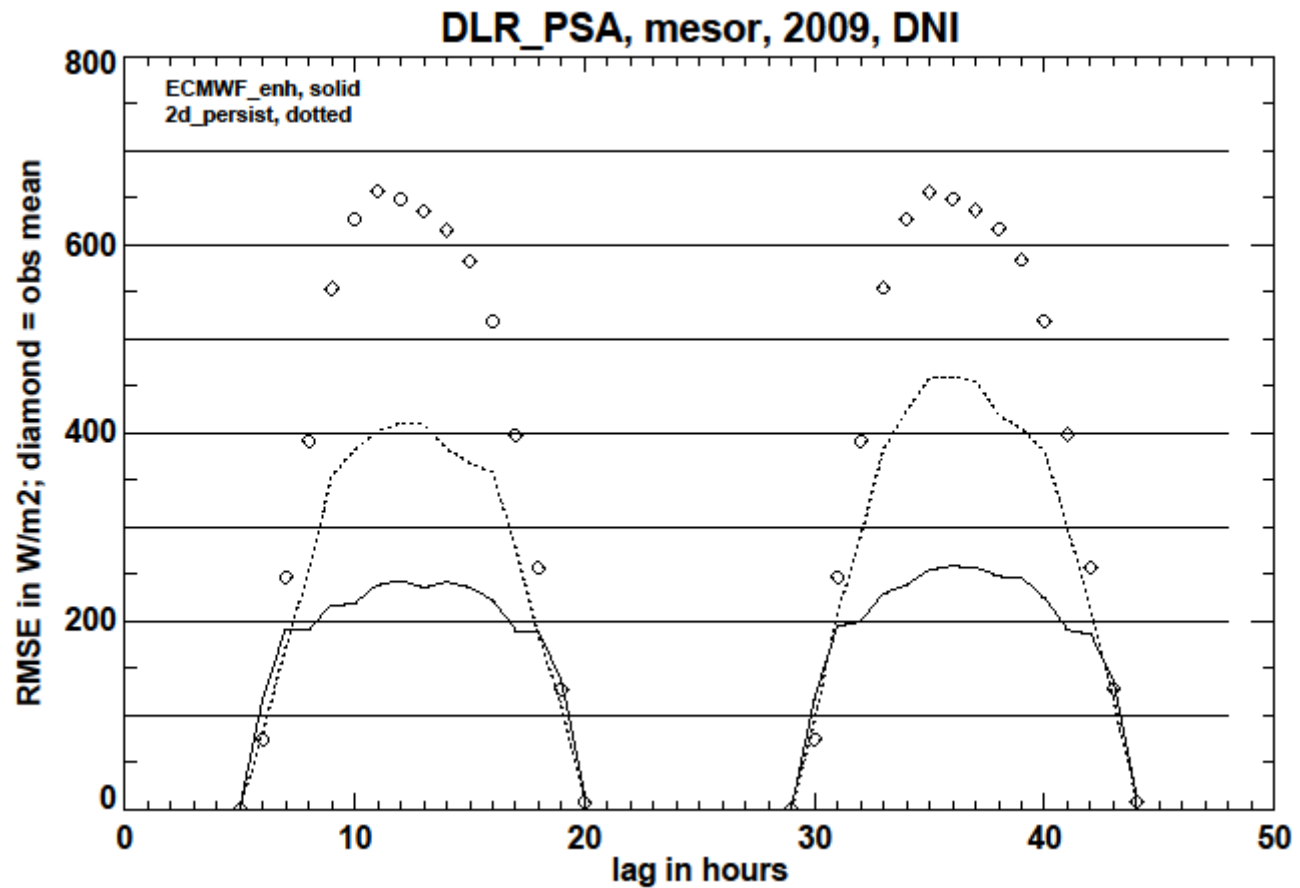
biases = fct (time of day)

under-  
estimation  
thin ice clouds



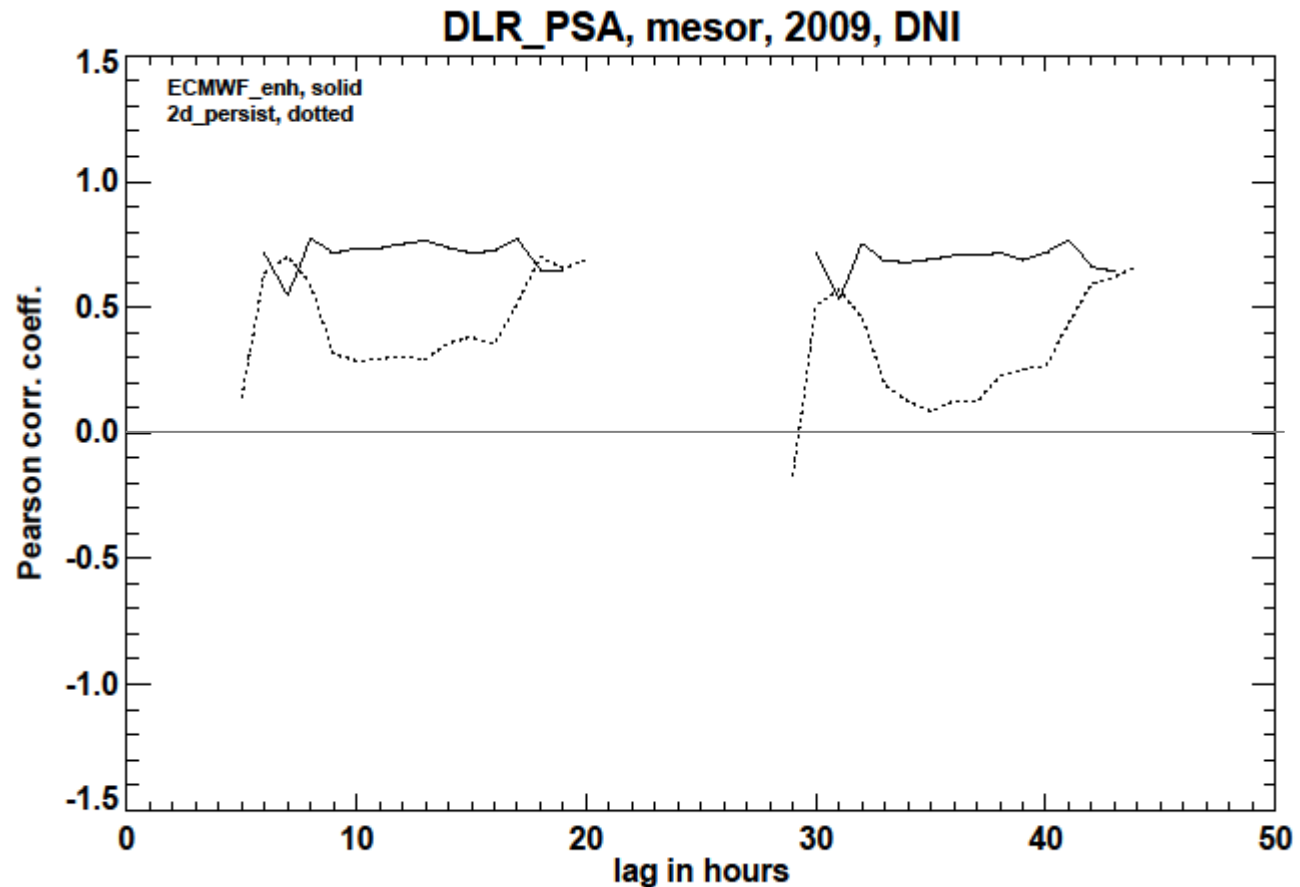
# Verification ECMWF/DLR

RMSE = fct (time of day)



# Verification ECMWF/DLR

Local correlation coeff. = fct (time of day)







# Verification ECMWF/DLR

Same parameters:

Biases, RMSE, rel RMSE, local correlation coeff.,  
skill scores

= fct (month of year)

Temporal lag errors (if any)






Perfect forecast capability for a renewable resource



No reserve power needed to cope with forecasting errors



Reduced costs from the grid operator point of view



Perfect forecast capability for a renewable resource – **not the case for solar**



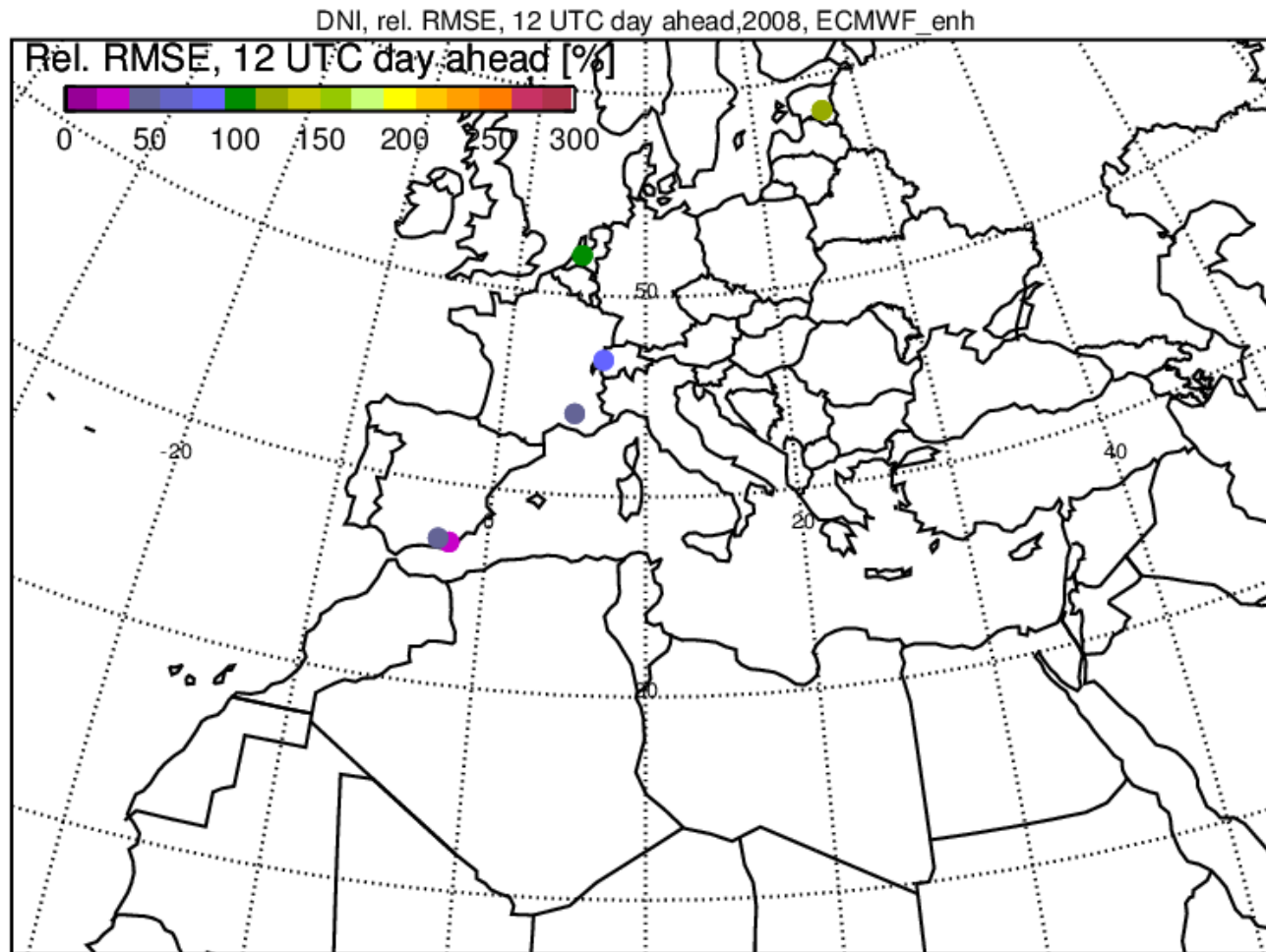
**reserve power needed** to cope with forecasting errors



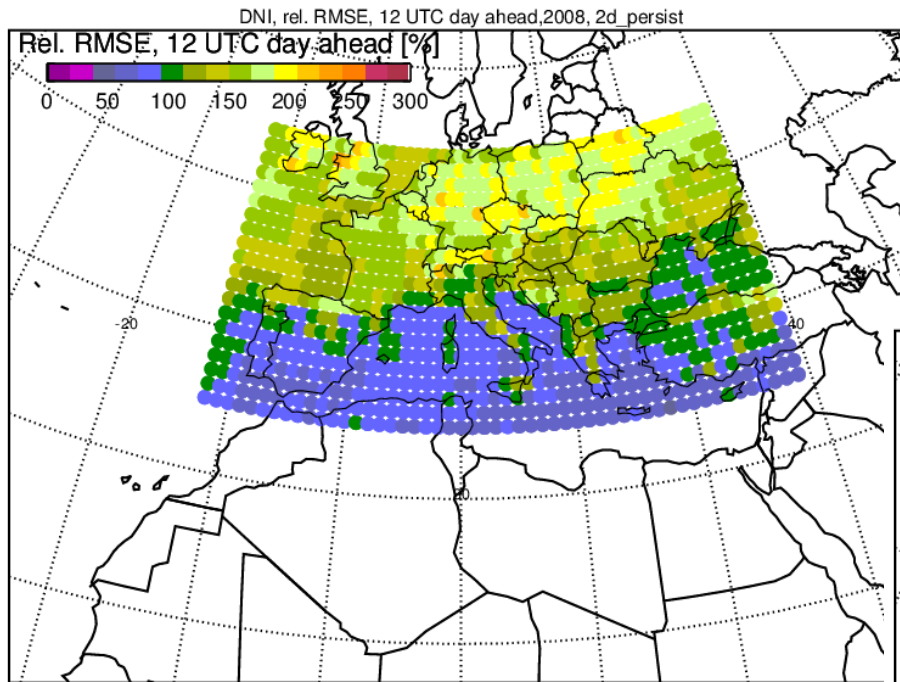
**Costs need to be taken into account** from the grid operator point of view

**Is there any spatial variability in these costs?**

# Go from a single location to a list of validation sites – European BSRN sites

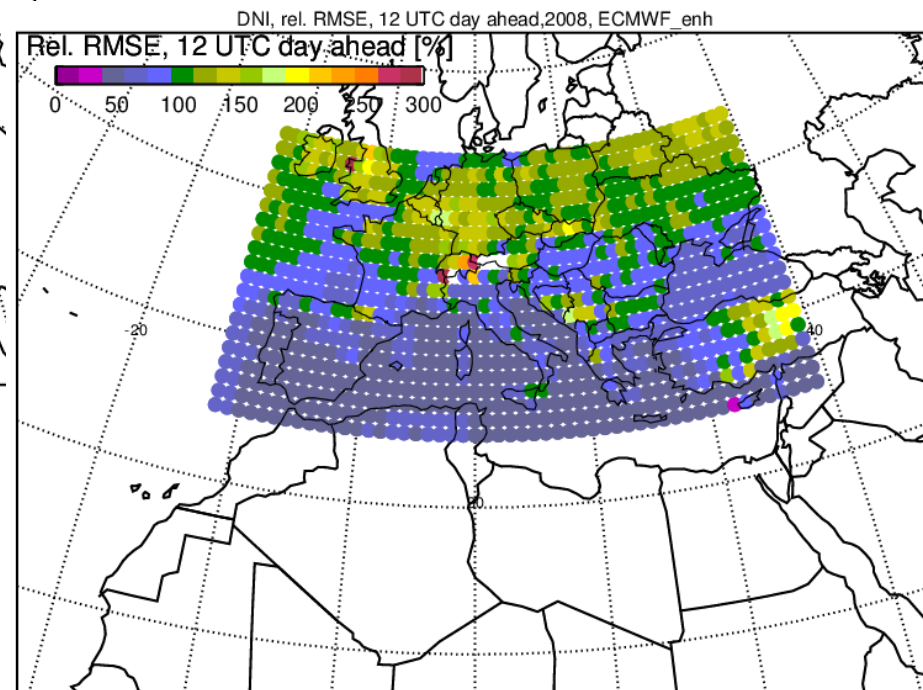


# Assessment based on satellite measurements as ground truth

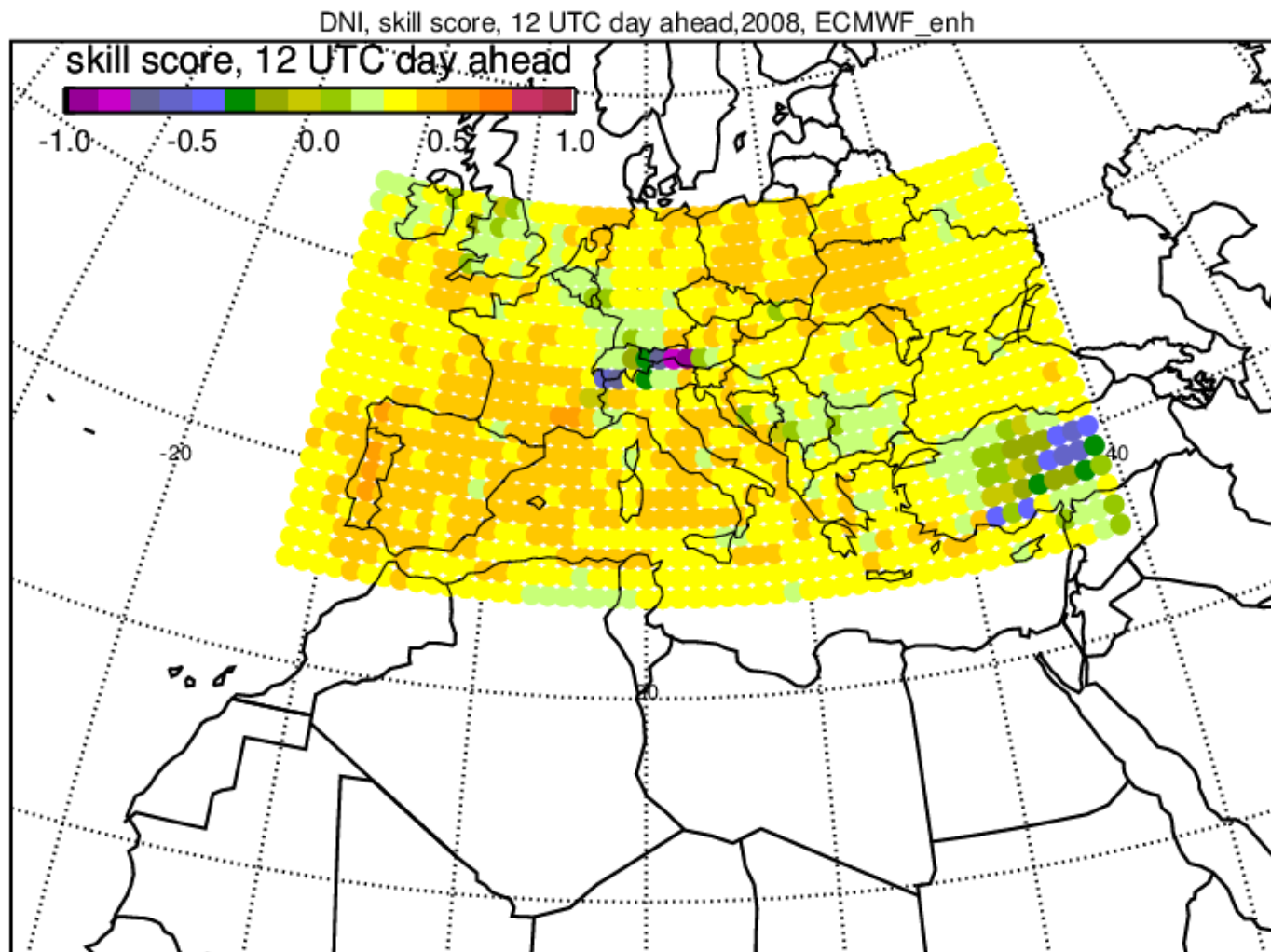


2 day persistence  
Rel. RMSE, 2008

ECMWF



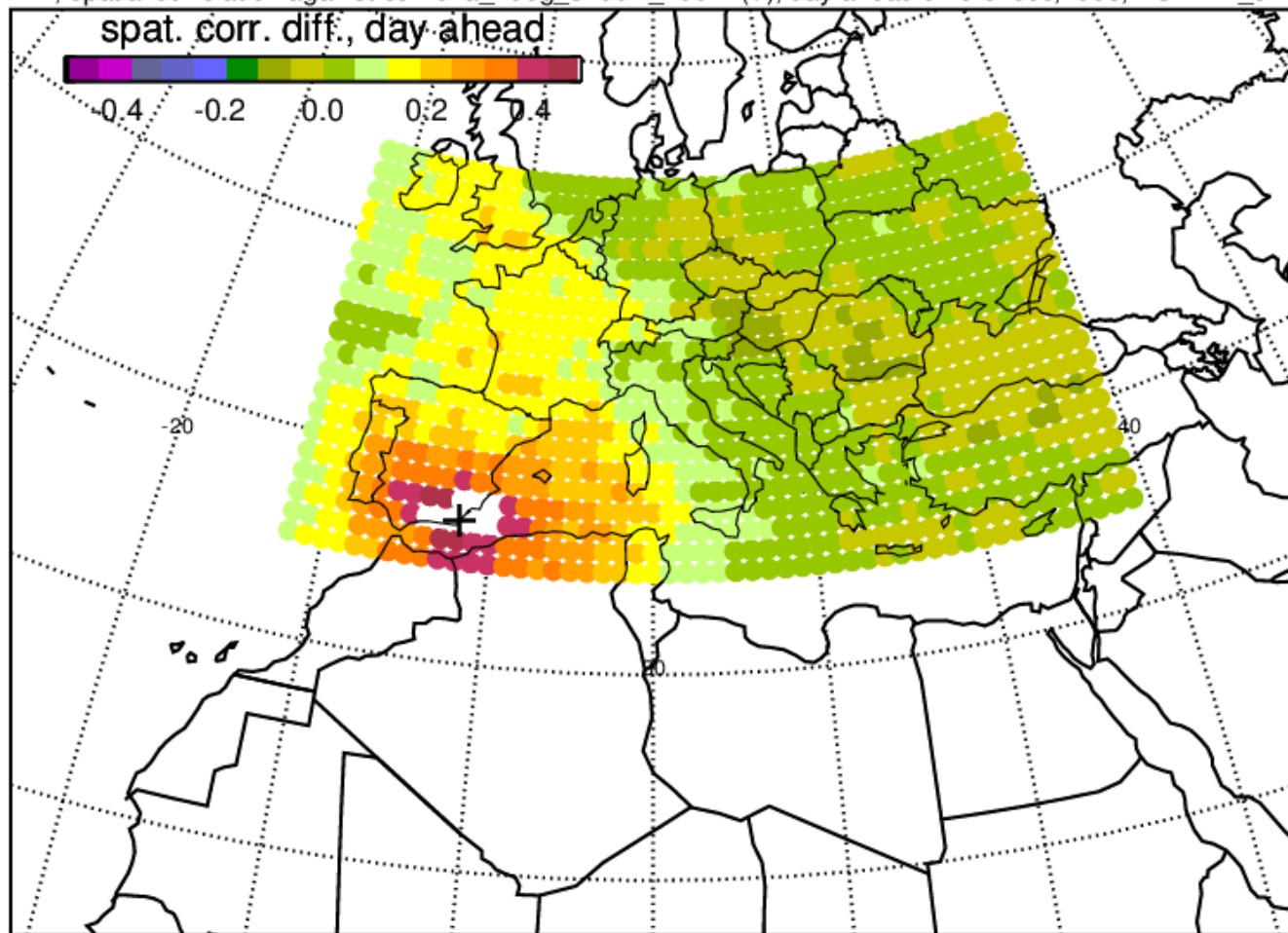
# ECMWF skill score, 12 UTC, 2008





# ECMWF, error spatial correlation vs. cross, 2008

DNI, spatial correlation against eumena\_1deg\_3700N\_200W (+), day ahead differences, 2008, ECMWF\_enh



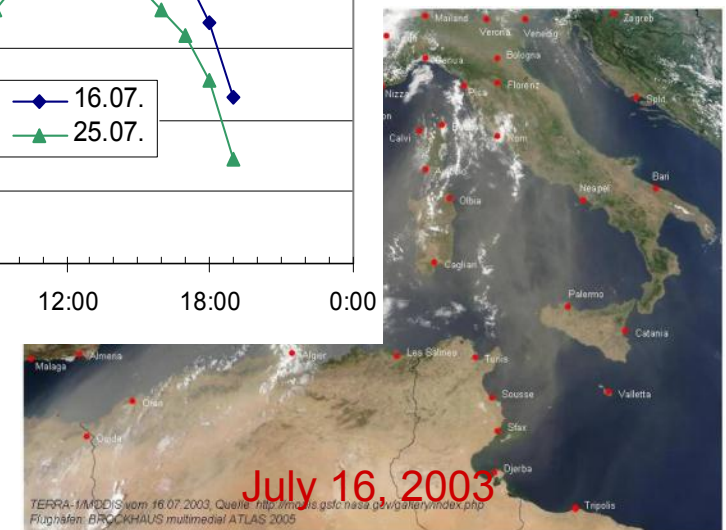
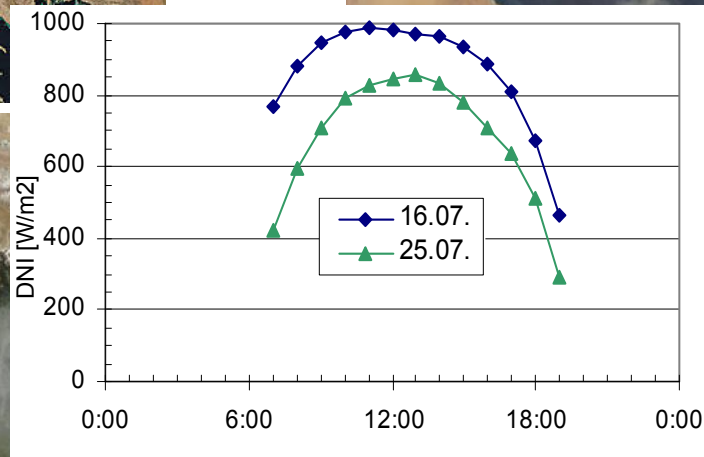
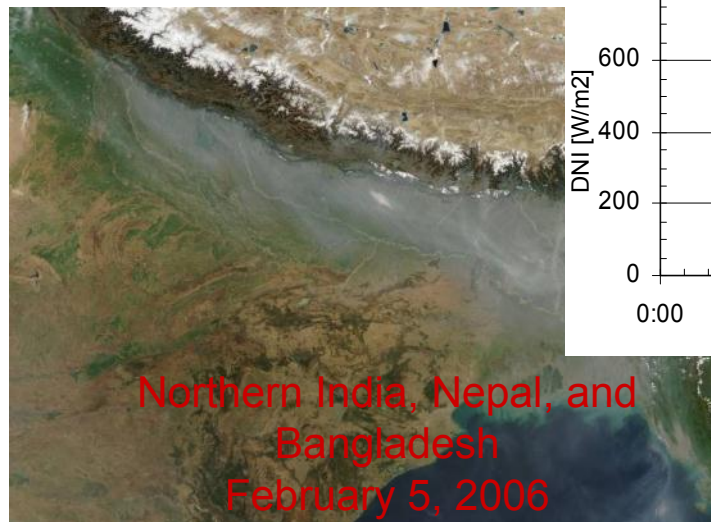


# Model output statics (MOS)

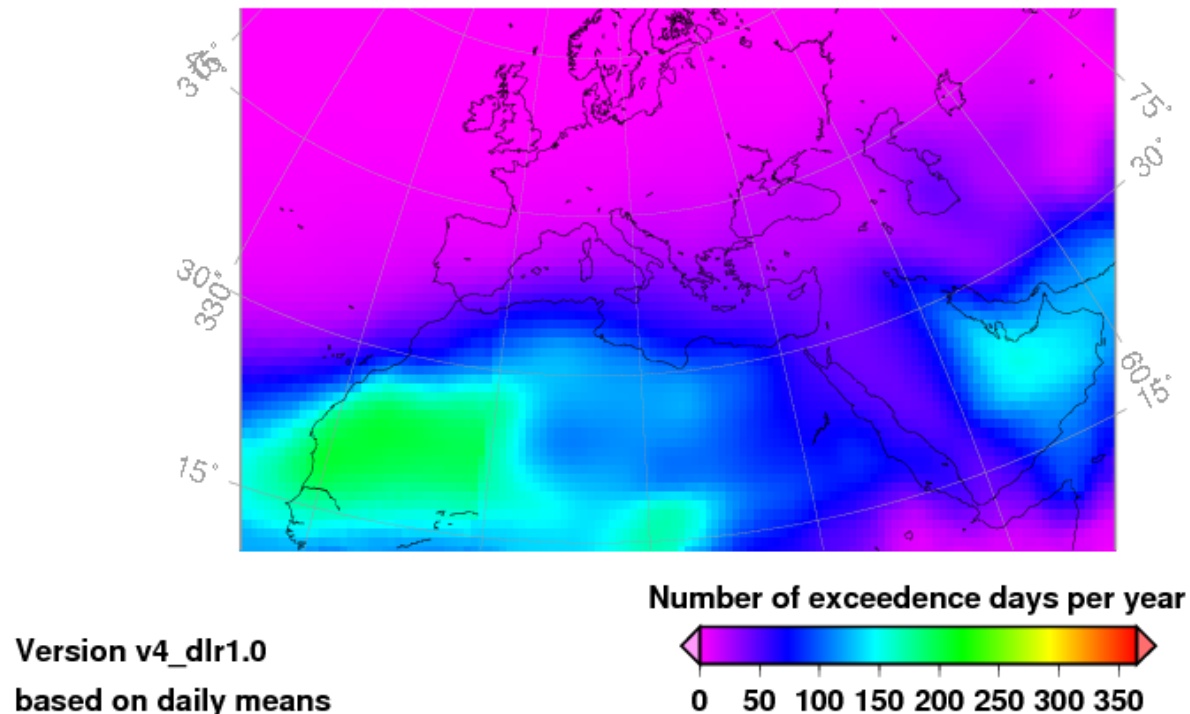
- It does:
  - Spatio-temporal interpolation and smoothing (from grid)
  - Local adaptation + bias correction
  - Multi-NWP-model combination
- General problems
  - Update for each NWP model change  
(happens frequently, monitoring needed)
  - Update for each location needed
- Specific problems for DNI
  - Generate DNI out of cloud coverage, global irradiance, other parameters
  - Most input models provide no information about aerosols
  - Temporal interpolation from global 3-hourly to hourly
- Finding for Spain: ECMWF is still performing better than existing commercial MOS systems

# Aerosole

typical: - 25% direct irradiance  
- 5% global irradiance



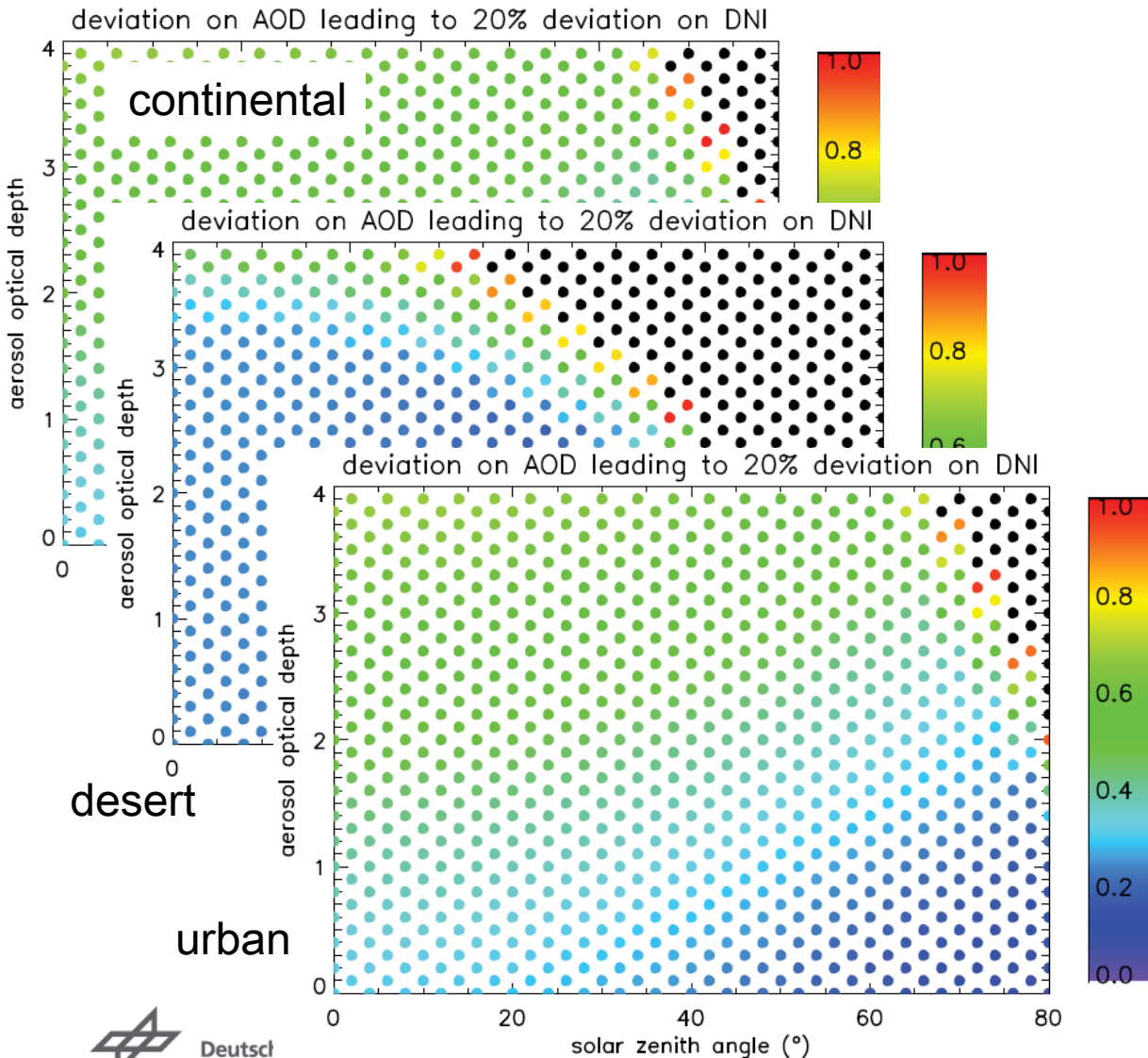
# Exceedance days per year – direct irradiance – desert dust only



Number of days with a direct irradiance extinction due to dust **above 30%**  
based on a 1983 -2007 DLR/MATCH aerosol model for the EUMENA region



# User requirements for aerosol accuracy



Basic input for CSP:  
**DNI (Direct Normal Irradiance)**

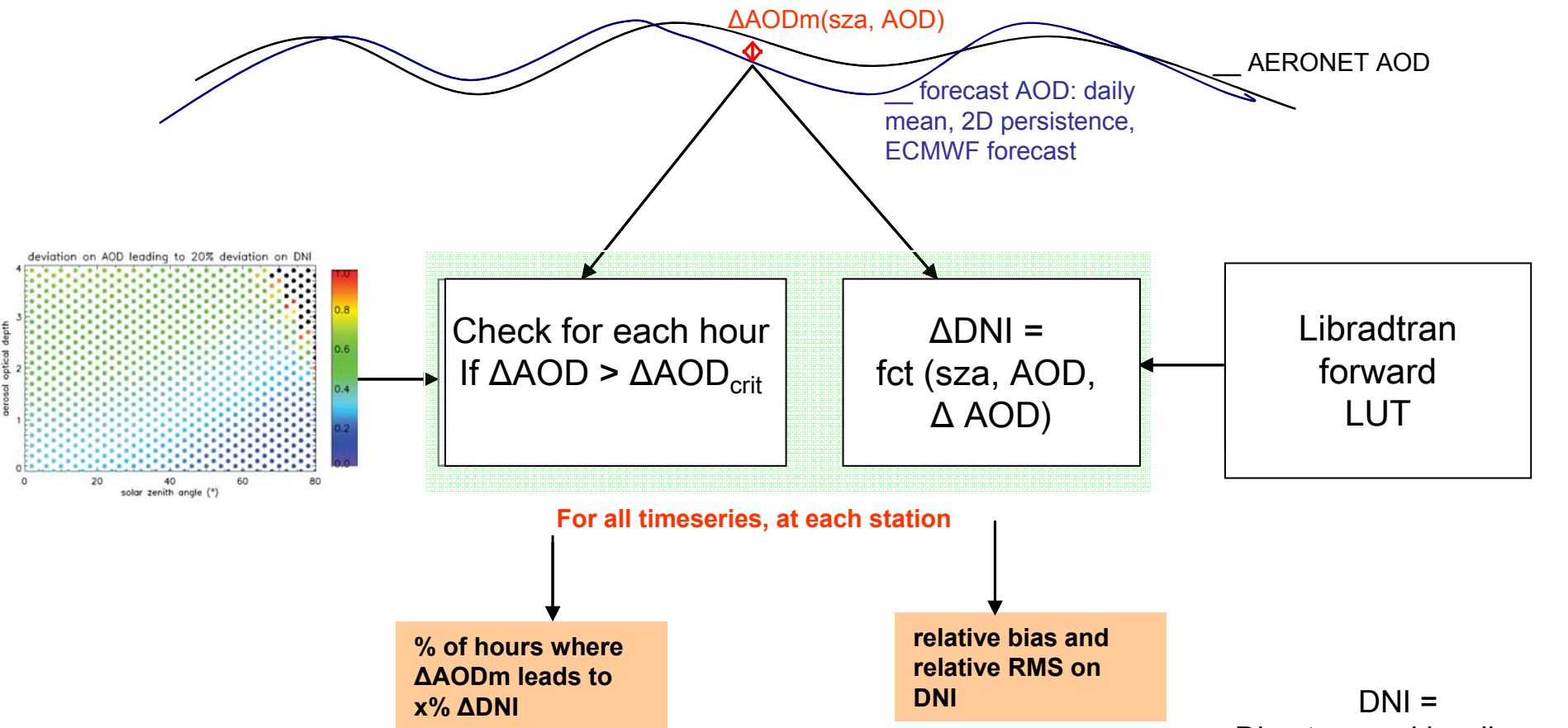
each AOD and each SZA,  
minimum deviation on AOD  
leading to **x% deviation on DNI**

done for **5%, 10%, 20%**

Minimum DNI : 10 W/m<sup>2</sup>



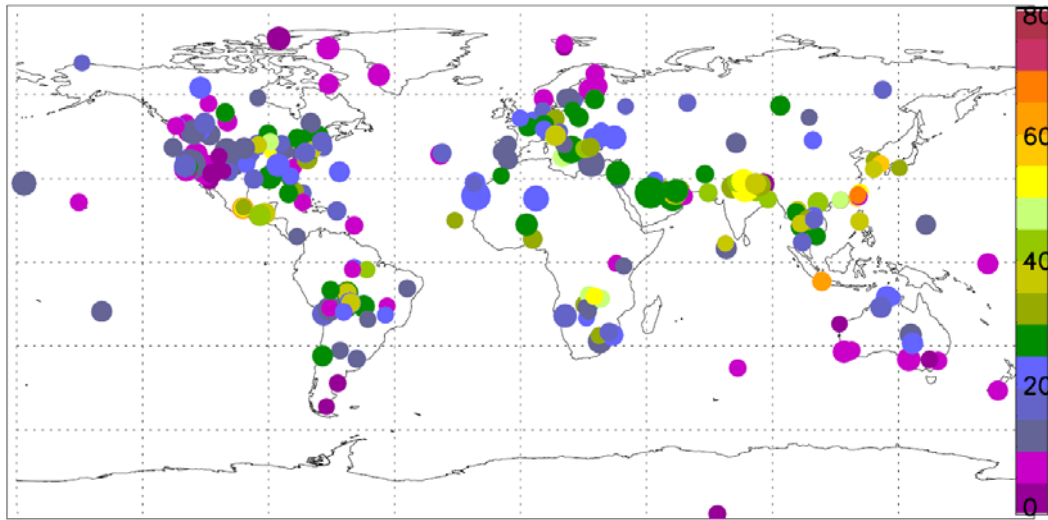
# User specific validation approach



DNI =  
Direct normal irradiance

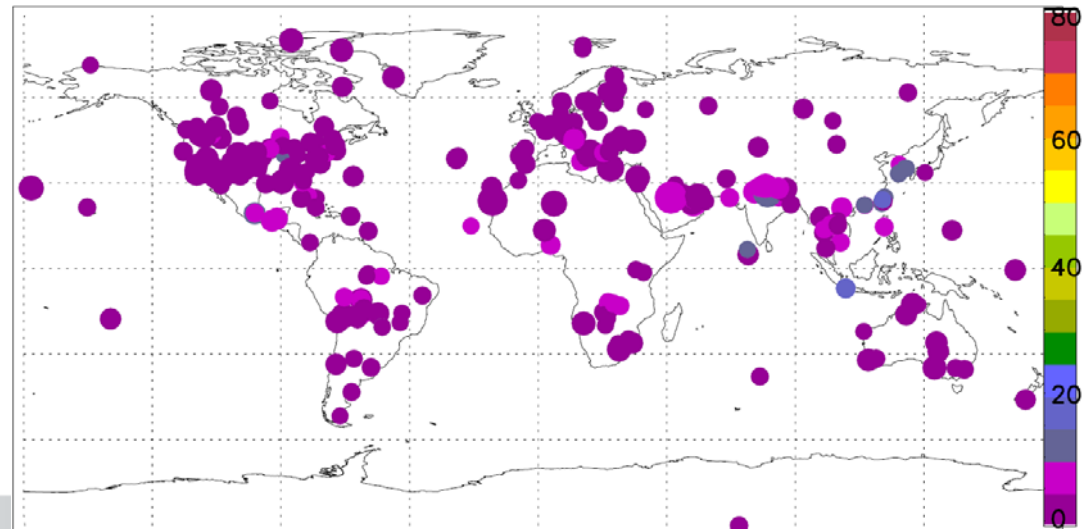


# Does intra-day variation matter in our case?

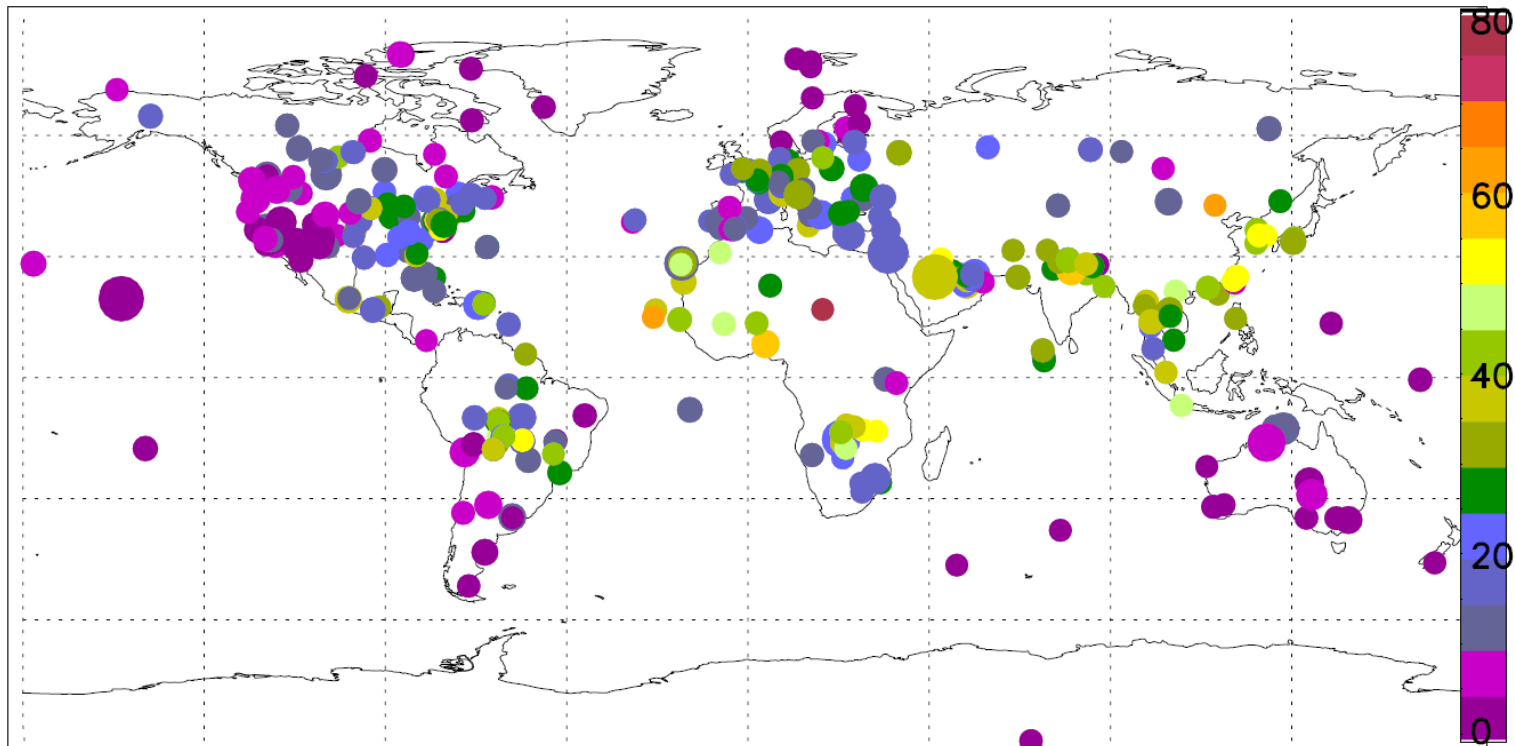


% of hours where the deviation  
on DNI due to intra-day variability  
is higher than 5%

% of hours where the deviation  
on DNI due to intra-day variability  
is higher than 20%



# Is a 2 day persistence forecast sufficient?

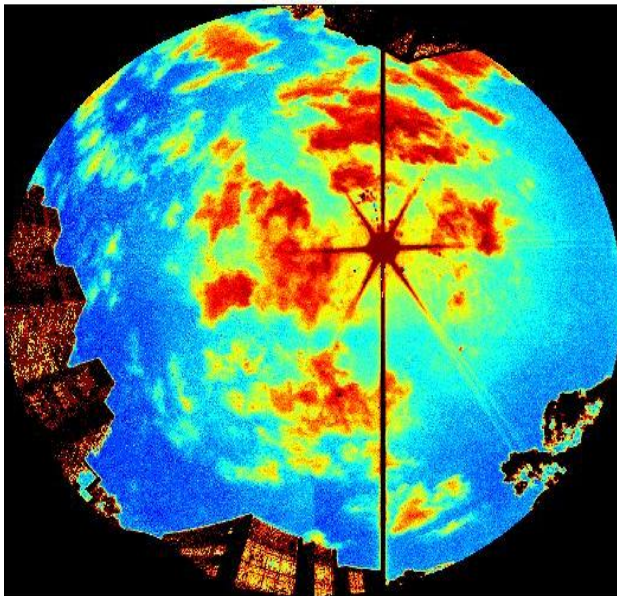


% of hours where the deviation  
on DNI due 2 day persistence  
forecasting is higher than 20%

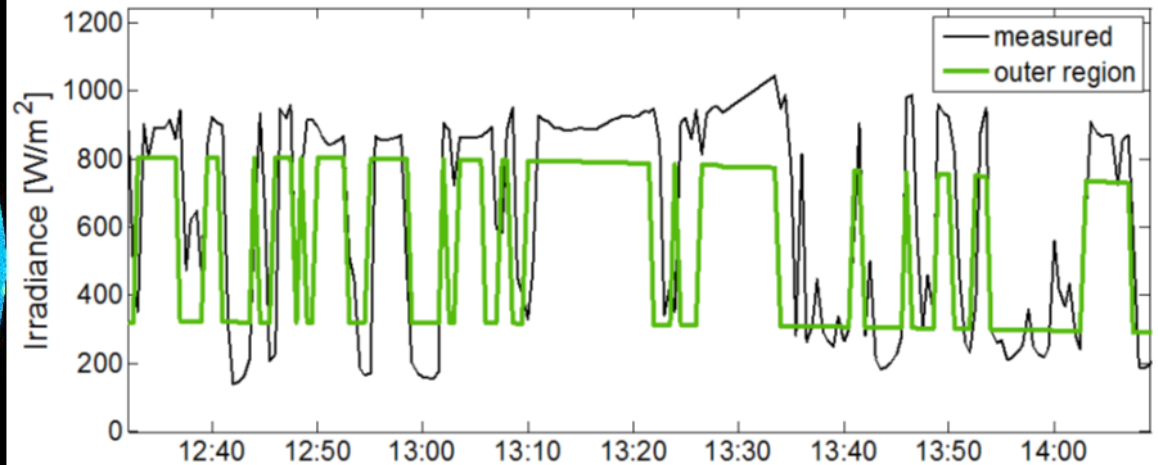


## Cloud camera, total sky imagery

- very short-term (minutes ahead) predictions of future cloud patterns
- does not account for cloud development and dissipation
- limited to the field of view of the sky imager
- may use multiple imagers at different locations
- actual look-ahead time depends upon the cloud velocity and cloud height
  - low and fast clouds the forecast horizon may only be 3 minutes
  - high and slow clouds it may be over 30 minutes
  - **generally horizons between 5 to 20 minutes are typical.**
- multiple height cloud layers have different motion vectors
  - **extremely important for DNI**



- Ratio of red and blue channel intensity
- a partly cloudy day
- UC San Diego solar forecasting sky imager
- high RBR are classified as cloudy



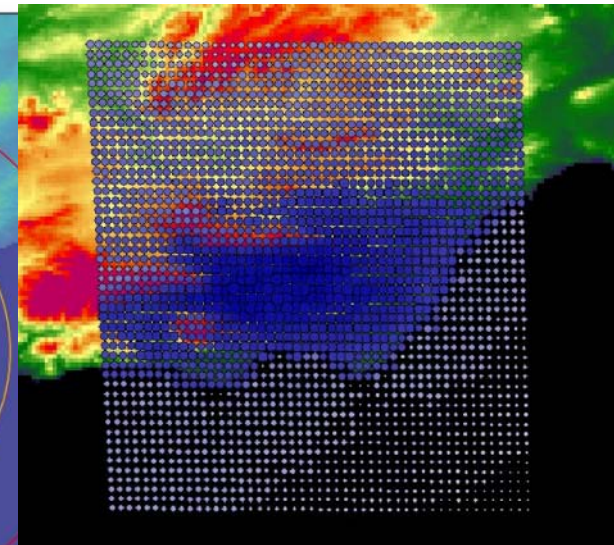
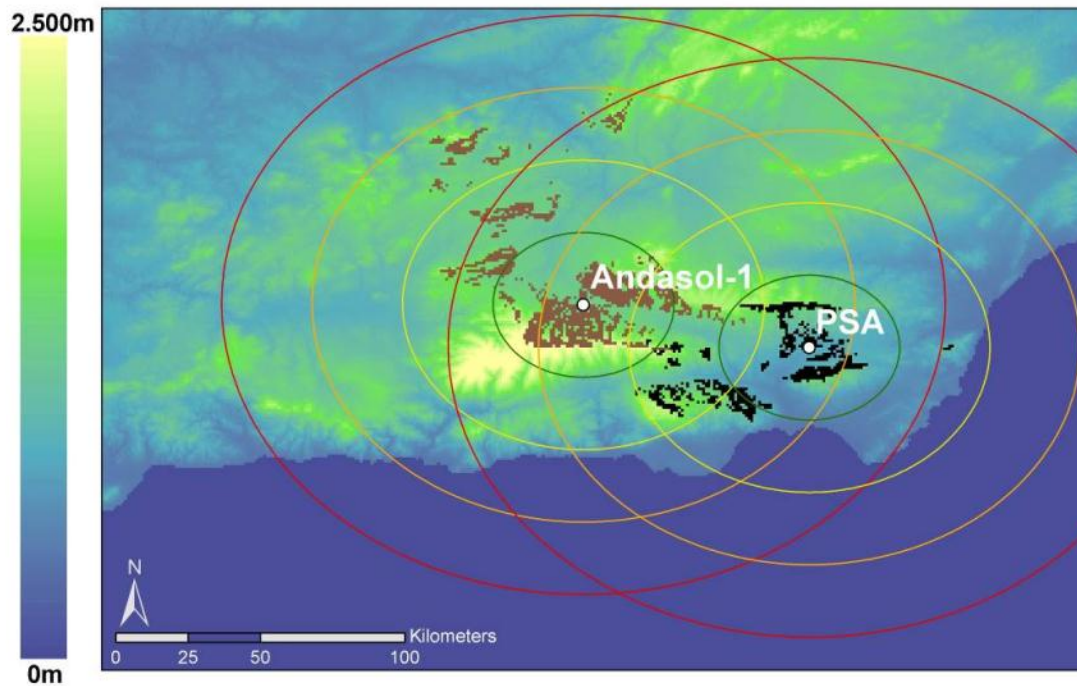
- 30 second GHI ramp forecast
- against 1 sec GHI measurements
- day with cumulus clouds

Mathiesen, Kleissl, 2012

No experience with DNI  
available so far...

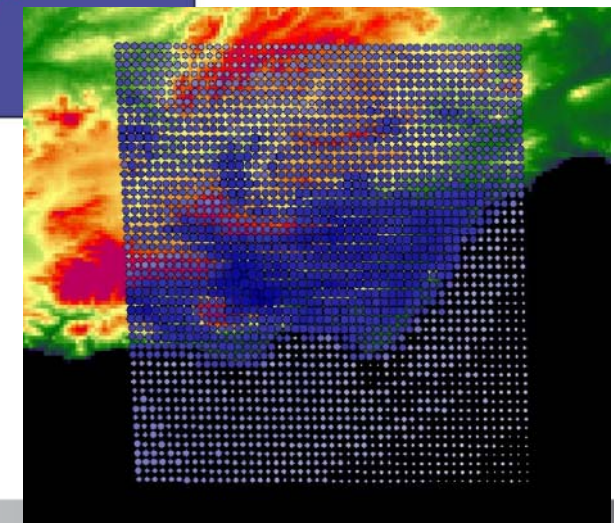


# Representativeness of nearby pixels



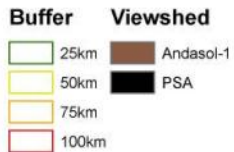
PSA  
-1 hours lag

scale from  
0.65  
to 0.73  
(42-53%  
explained)



PSA  
-2 hours lag

scale from  
0.48  
to 0.55  
(23-30%  
explained)



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in der Helmholtz-Gemeinschaft




Sun  
point  
information

cloud camera  
20 min forecast horizon  
high resolution

satellite  
several hours  
forecast  
type of cloud  
less spatial resolution







# Satellite based information about typical weather conditions at the power plant site



# Cloud compactness indicators

Cologne, D

overcast

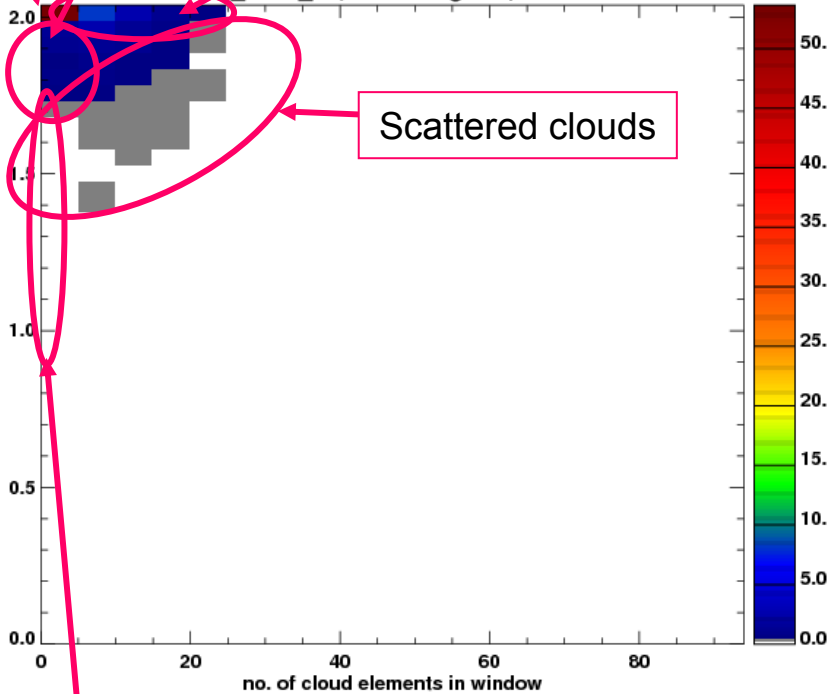
Few large clouds

Broken clouds

Scattered clouds

Koeln\_2005\_49, 2D histogram, N = 8045

Fractal box dimension



isolated clouds

No of cloud elements in a 49x49 window

Almeria, E

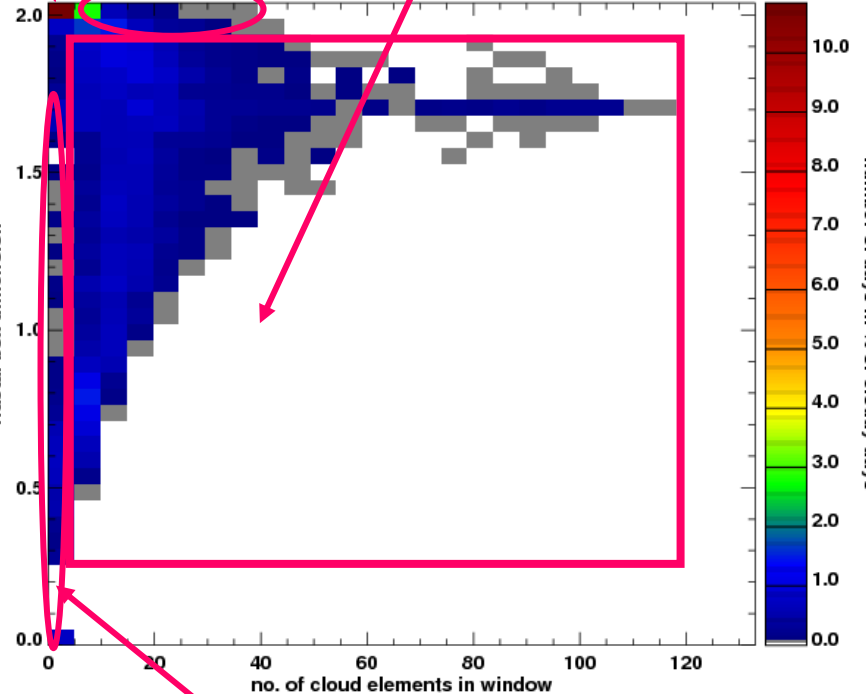
overcast

Broken clouds

Scattered clouds

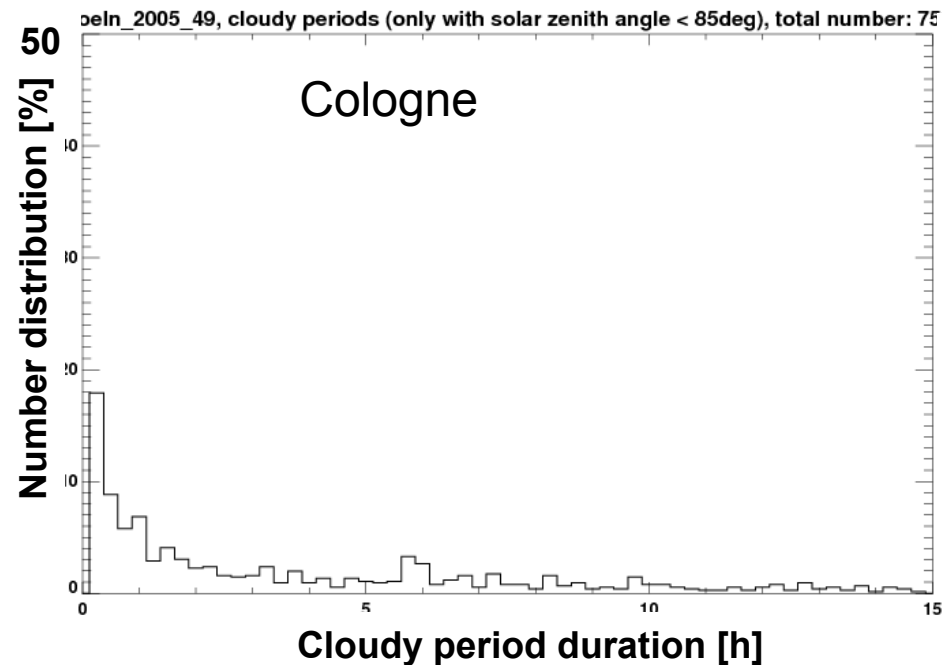
Almeria\_2005\_49, 2D histogram, N = 10681

fractal box dimension

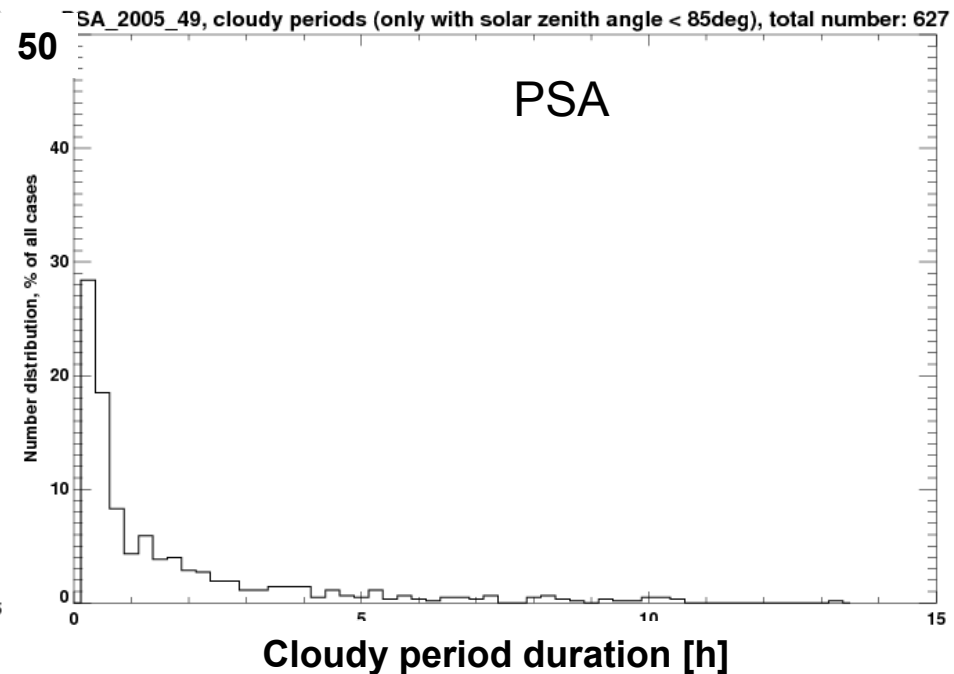


isolated clouds

# Cloud period duration statistics



18%  $\leq 15$  min  
39%  $< 1$  h  
90%  $< 9.75$  h  
98%  $< 13.5$  h



28%  $\leq 15$  min  
59%  $< 1$  h  
90%  $< 4.5$  h  
98%  $< 9.25$  h

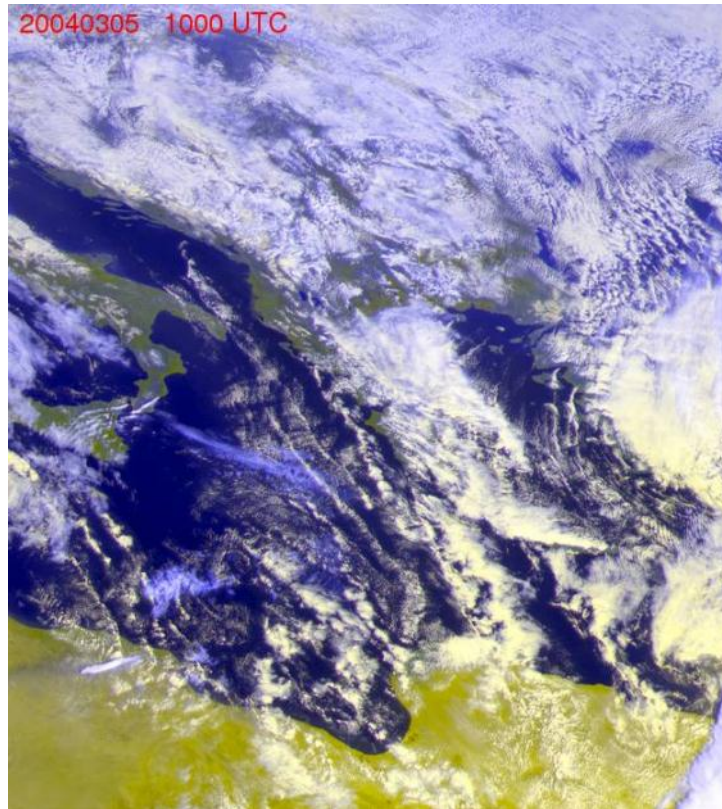


# Nowcasting based on satellite measurements



# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
HRV channel for Greece (source DLR-IPA), basis for short-term forecasts**



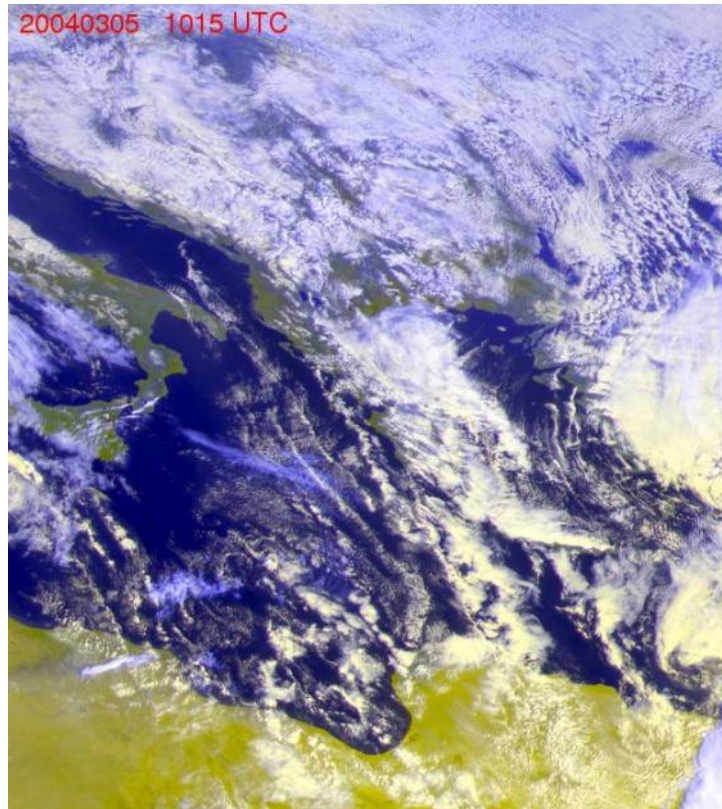
**cloud motion vectors**

**1/8**



# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
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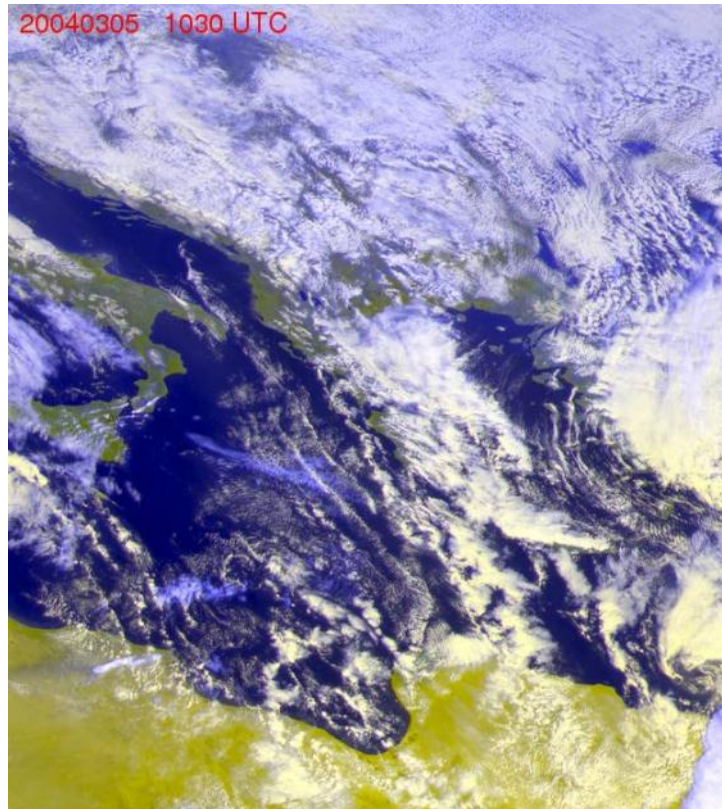
**cloud motion vectors**

**2/8**



# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
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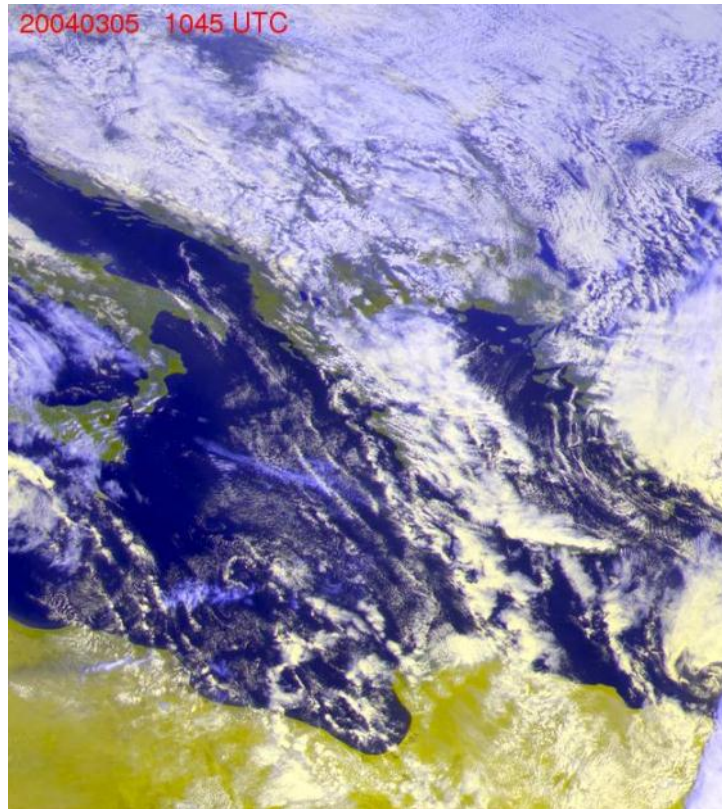


**cloud motion vectors**

**3/8**

# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
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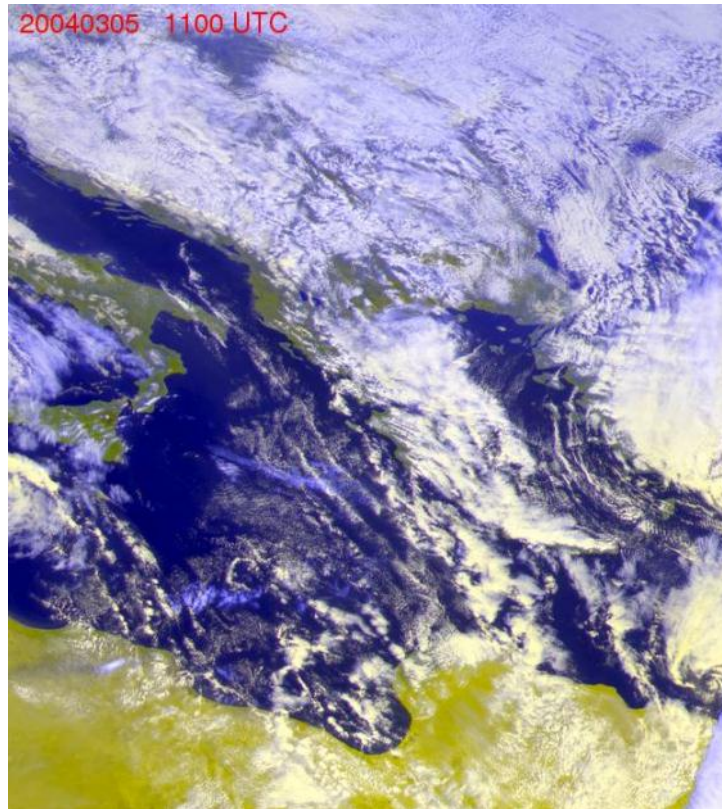


**cloud motion vectors**

**4/8**

# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
HRV channel for Greece (source DLR-IPA), basis for short-term forecasts**



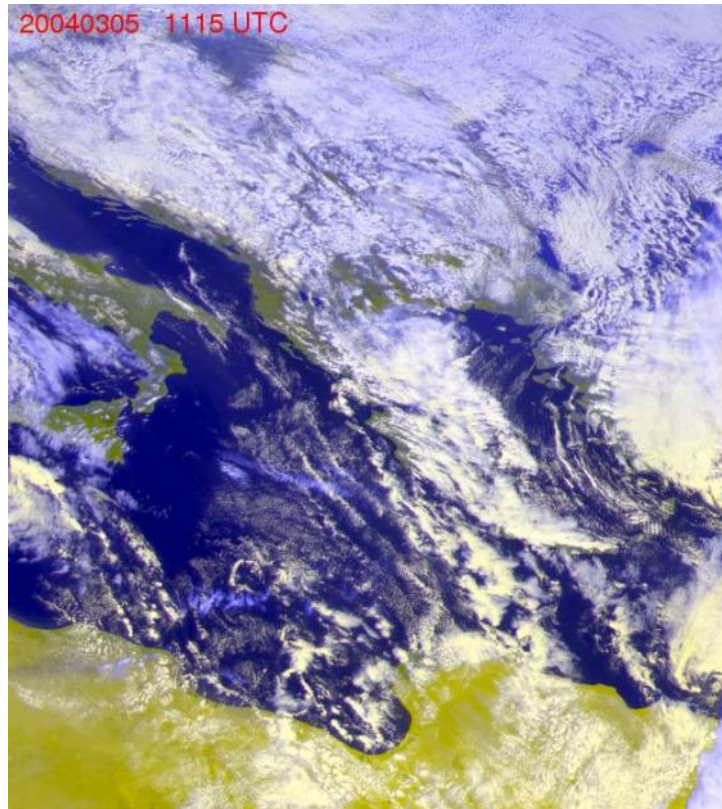
**cloud motion vectors**

**5/8**



# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
HRV channel for Greece (source DLR-IPA), basis for short-term forecasts**

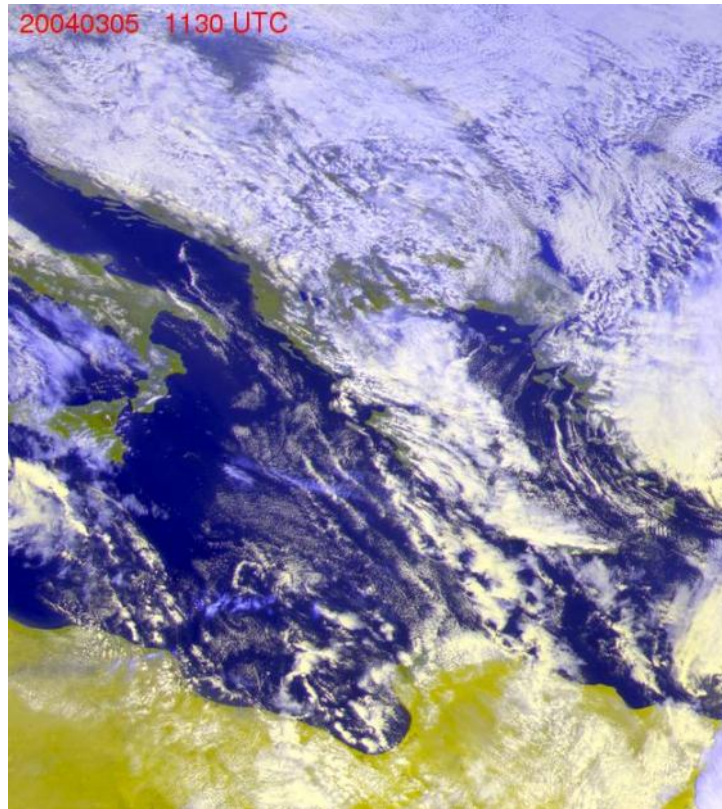


**cloud motion vectors**

**6/8**

# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
HRV channel for Greece (source DLR-IPA), basis for short-term forecasts**

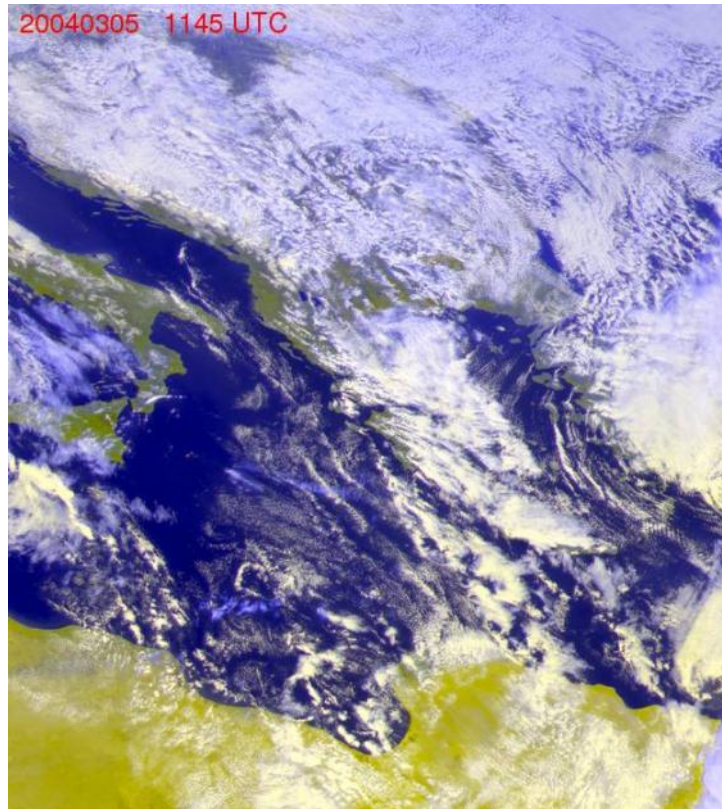


**cloud motion vectors**

**7/8**

# Nowcasting and short-term forecasting based on EO

**Example: cloud fields from Meteosat Second Generation  
HRV channel for Greece (source DLR-IPA), basis for short-term forecasts**



**cloud motion vectors**

**8/8**



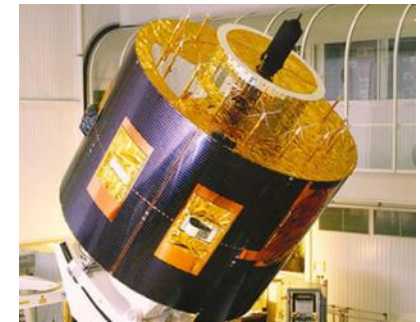
# METEOSAT SECOND GENERATION – What can be seen more?

- More spectral information allows assessment of physical cloud parameters
- 15 min resolution instead of 30 minutes
- Pixel size 3 km at nadir;  
4x5 or 5x6 km in Europe

## Upcoming Key Questions:

- How do I choose technical features of the power plant ?  
(Engineering)
- How do I operate my power plant? (Operation & Maintenance)

Channel no.		Characteristics of spectral band ( $\mu\text{m}$ )		
		$\lambda_{\text{cen}}$	$\lambda_{\text{min}}$	$\lambda_{\text{max}}$
1	VIS0.6	0.635	0.56	0.71
2	VIS0.8	0.81	0.74	0.88
3	NIR1.6	1.64	1.50	1.78
4	IR3.9	3.90	3.48	4.36
5	WV6.2	6.25	5.35	7.15
6	WV7.3	7.35	6.85	7.85
7	IR8.7	8.70	8.30	9.10
8	IR9.7	9.66	9.38	9.94
9	IR10.8	10.80	9.80	11.80
10	IR12.0	12.00	11.00	13.00
11	IR13.4	13.40	12.40	14.40
12	HRV	Broadband (about 0.4 – 1.1)		

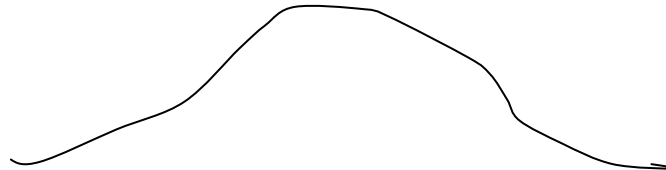


Copyright EUMETSAT



# Strategy to use nowcasting at the power plant

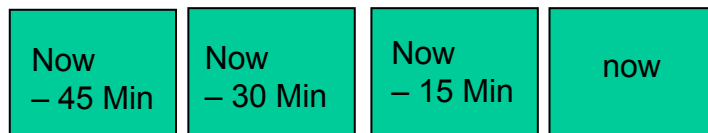
Type A



**48 h forecast made by weather service yesterday**

Type B

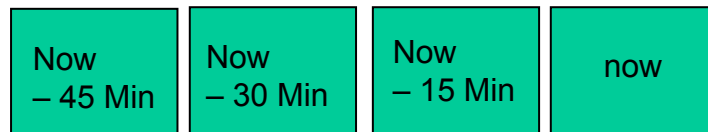
Satellite images



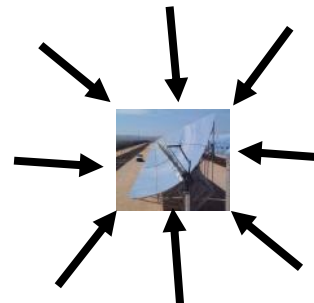
**Human** forecast with knowledge about upcoming cloud fields

Type C

Satellite images



now

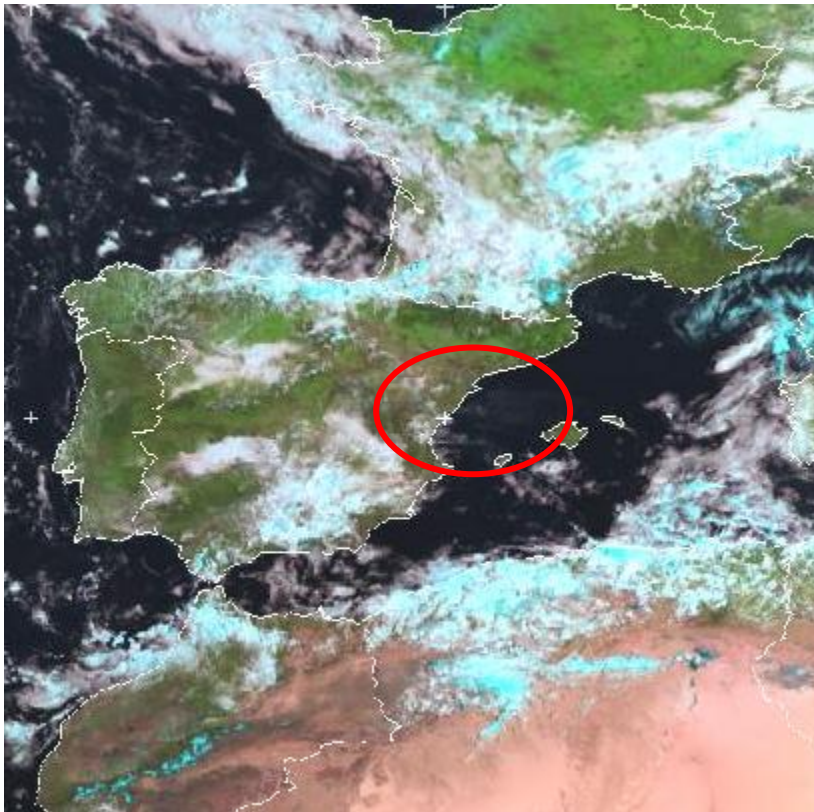


**Automatic 8h-nowcast** of upcoming **cloud** fields

**Automatic 8h-nowcast** of upcoming **DNI** values

## Photo-like quick look

Low clouds white, high clouds turquoise blue



Looks ,natural‘

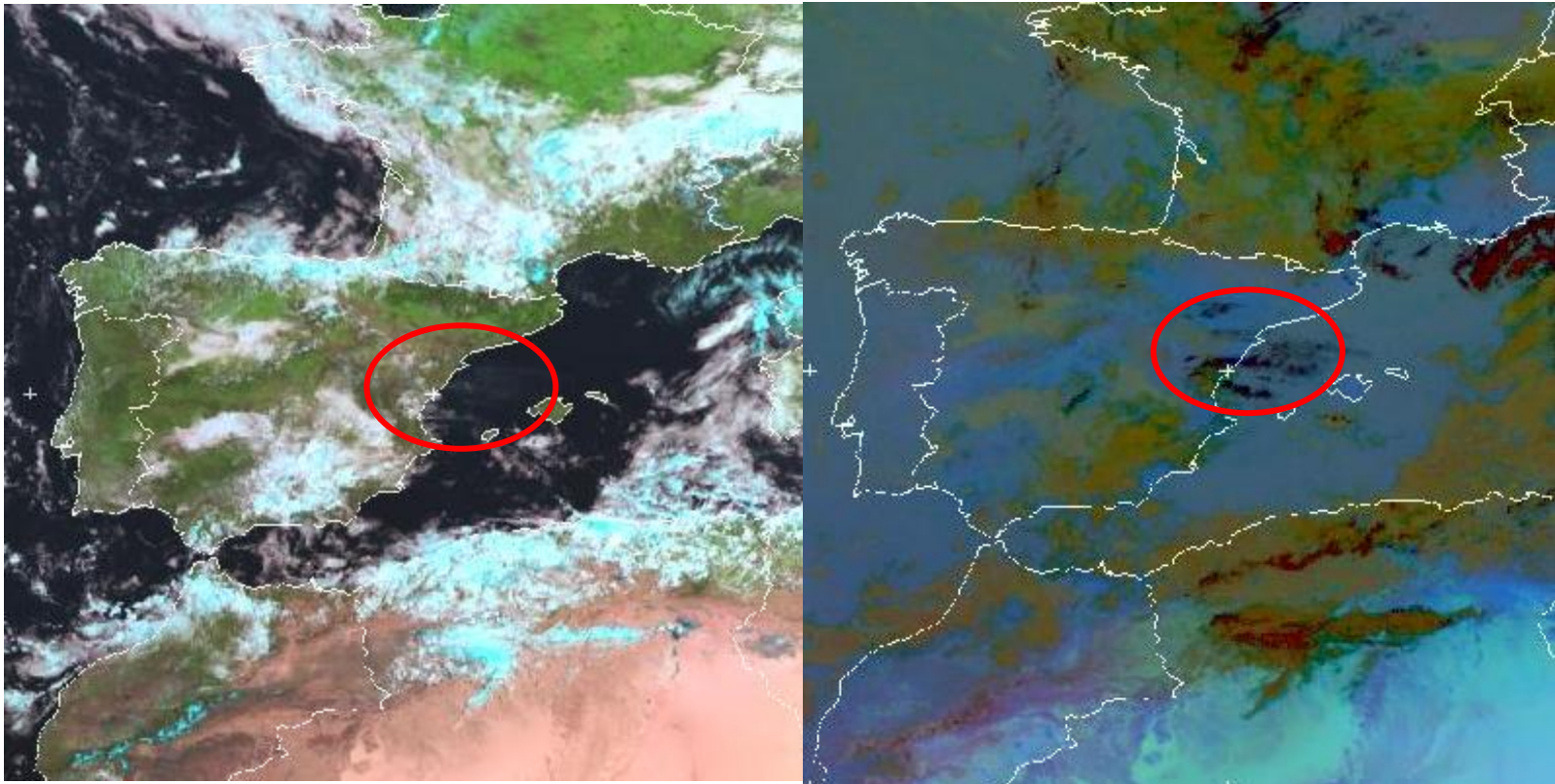
Looks ,known‘ to the operator

We just see iced tops of clouds.  
We can't distinguish  
thick ice clouds (e.g. 10% DNI)  
from thin ice clouds (e.g. 50% DNI)

Can't see dust over land

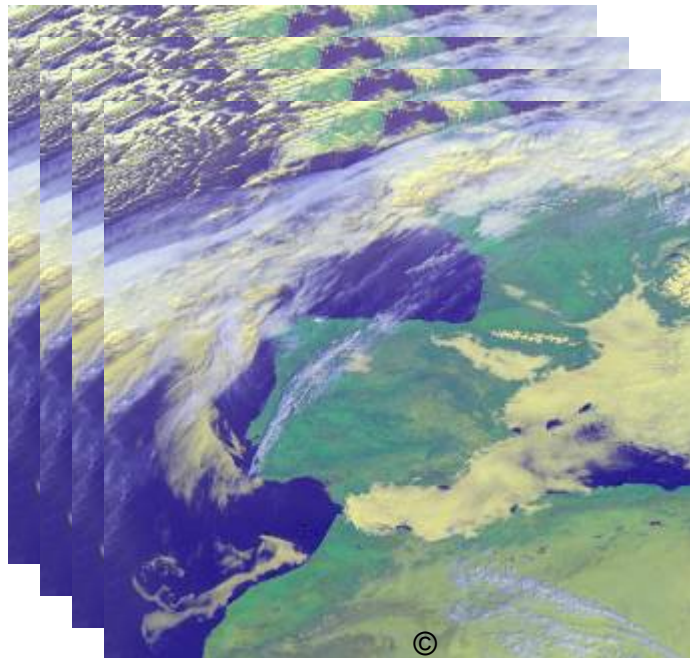


**Dust and cirrus – special colors**  
**Dust pink, thin cirrus black, available at night !!!**



3th May 2011, 0900 UTC

# Automatic nowcasting – which DNI to expect?



APOLLO cloud  
retrieval for MSG

Cloud mask

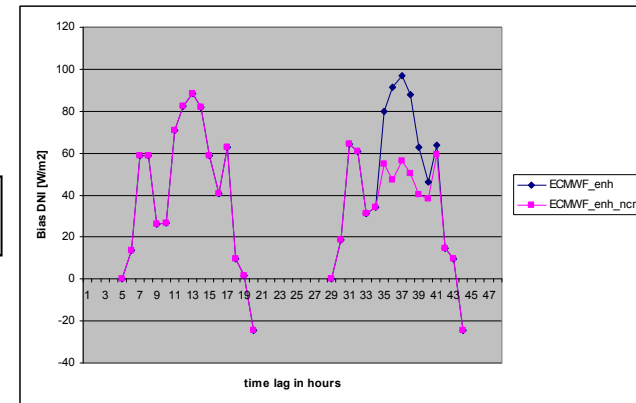
- water clouds
- thin cirrus clouds
- 15 min update

DNI forecast



Receptor model  
cloud mask  
forecast

Day and night !!

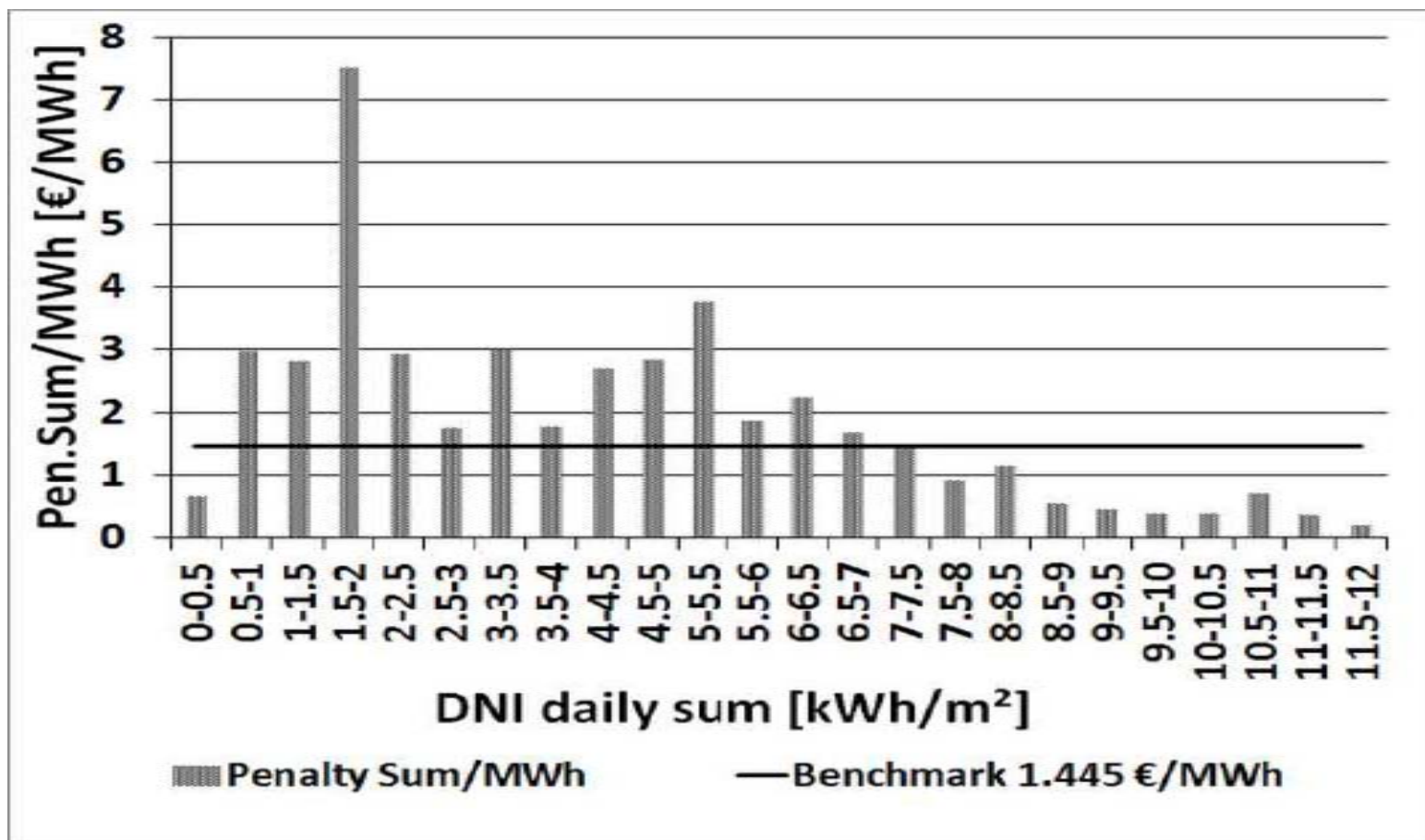




Some comments on the economic value  
of forecasts....





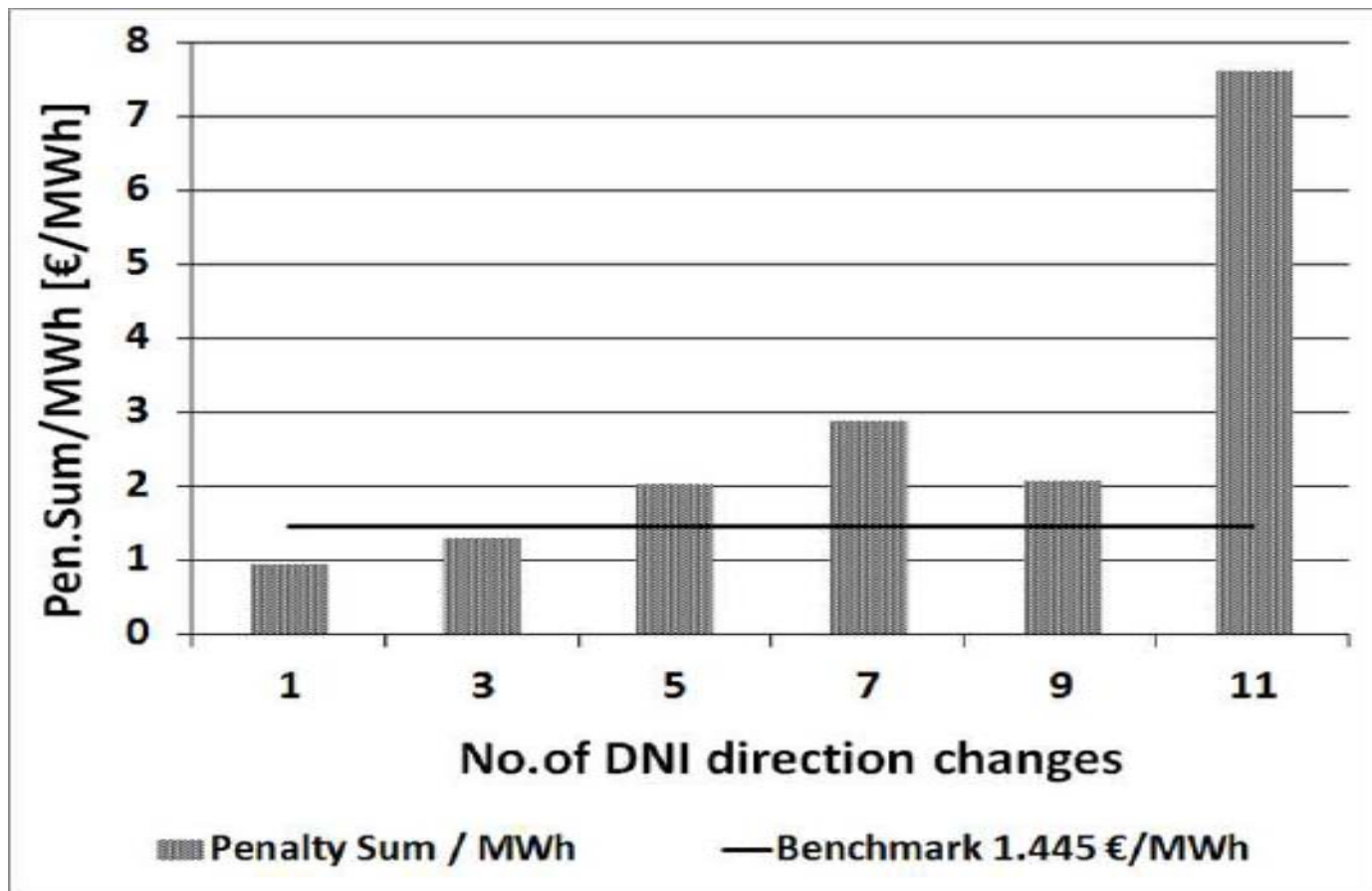


Kraas et al., SolarPaces 2010;  
submitted to Solar Energy

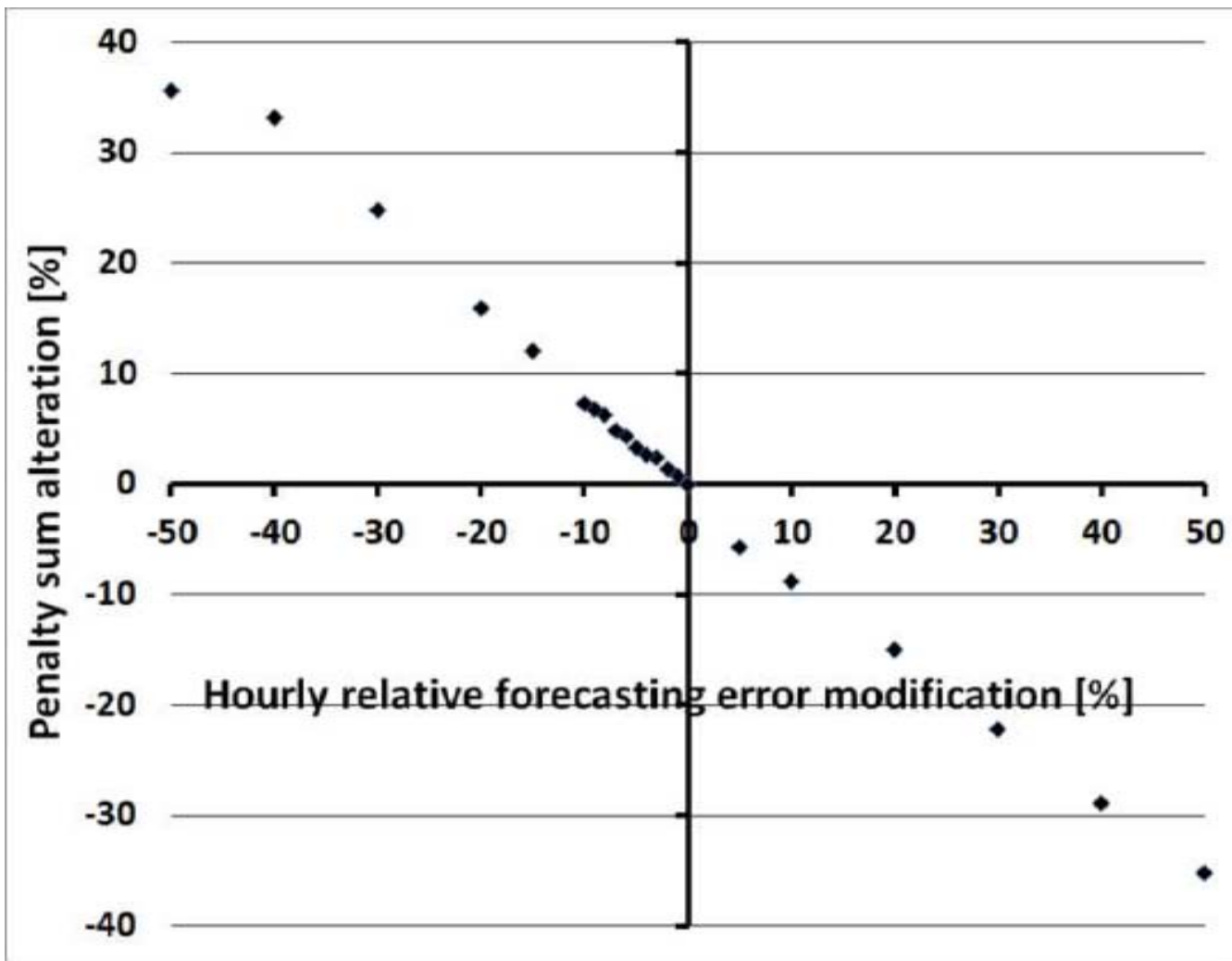
Based on Meteologica MOS  
July 2007 to December 2009  
Benchmark = average penalty



Deutsches Zentrum  
für Luft- und Raumfahrt e.V.  
in der Helmholtz-Gemeinschaft



Benchmark = average penalty





# Conclusions


- User requirements and meteorological basics
- Numerical weather prediction – verification measures have been explained – NWP can be used, potential for improvements is also large
- First results on spatial correlation of forecast errors
- Global maps – need of hourly aerosol forecasts, need of aerosol forecasts at all – regionally different
- Ground measurements are difficult to use in the multi-hour forecast horizon
- Satellites help to characterize any location in advance with respect to typical cloud behaviour
- Satellite-based nowcasting provides decision support to the plant operator – both visual and quantitatively
- Value of forecasts quantified for the market participation case
- Value of forecast improvements is a nearly linear effect – also if taking a non-linear market into account – not obvious



## References

- **Beyer, H.G., Martinez, J.P., Suri, M.** et al., 2009. MESOR Report on Benchmarking of Radiation Products. Deliverable 1.1.3, Management and Exploitation of Solar Resource Knowledge (MESOR), European Commission 6<sup>th</sup> framework programme, Contract Number 038665.
- **Breitkreuz, H., Schroedter-Homscheidt, M., Holzer-Popp, T., Dech, S.**, 2009. Short Range Direct and Diffuse Irradiance Forecasts for Solar Energy Applications Based on Aerosol Chemical Transport and Numerical Weather Modeling. Journal of Applied Meteorology and Climatology, 48 (9), pp. 1766-1779. DOI: 10.1175/2009JAMC2090.1.
- **Beyer HG, Polo Martinez J, Suri M, Torres JL, Lorenz E, Müller SC, Hoyer-Klick C and Ineichen P.** D 1.1.3 “*Report on Benchmarking of Radiation Products*”. Report under contract no. 038665 of MESoR, 2009. Available for download at <http://www.mesor.net/deliverables.html>, (January 12, 2012).

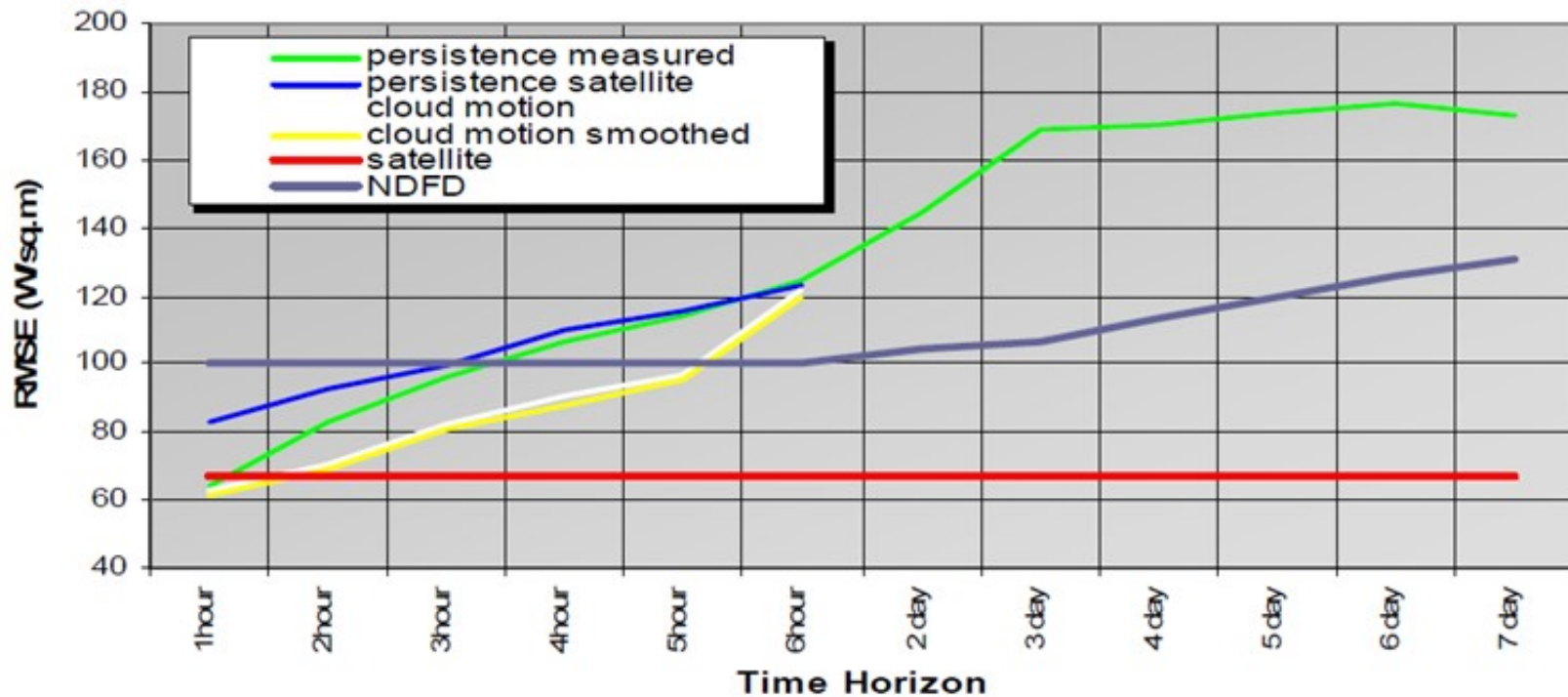


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- **Lara-Fanego, V., Ruiz-Arias, J.A., Pozo-Vázquez, D., Santos-Alamillos, F.J., Tovar-Pescador, J.**, Evaluation of the WRF model solar irradiance forecasts in Andalusia (southern Spain). Solar Energy, 2012
  - **Mathiesen P, Kleissl J.** *“Evaluation of numerical weather prediction for intra-day solar forecasting in the continental united states”*. Solar Energy. 2011;85(5):967-77.
  - **Perez R, S Kivalov, J Schlemmer, K Hemker Jr., D Renne, TE Hoff,** *“Validation of short and medium term operational solar radiation forecasts in the US”*, Solar Energy, 2010
  - **Remund, J.**, Photovoltaic and solar forecasting – state of the art, IEA PVPS report, section 3.4, 2012 (in edit stage at the moment)
  - **Wittmann, M., Breitkreuz, H., Schroedter-Homscheidt, M., Eck, M.**, Case Studies on the Use of Solar Irradiance Forecast for Optimized Operation Strategies of Solar Thermal Power Plants, IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing, 1, 1, 2008



Additional slides – dealing with accuracies





Source Perez et al, 2010:  
RMSE for GLOBAL IRRADIANCE forecasts  
US SURFRad stations  
Similar results for DNI are not available so far



## Example on forecast metrics

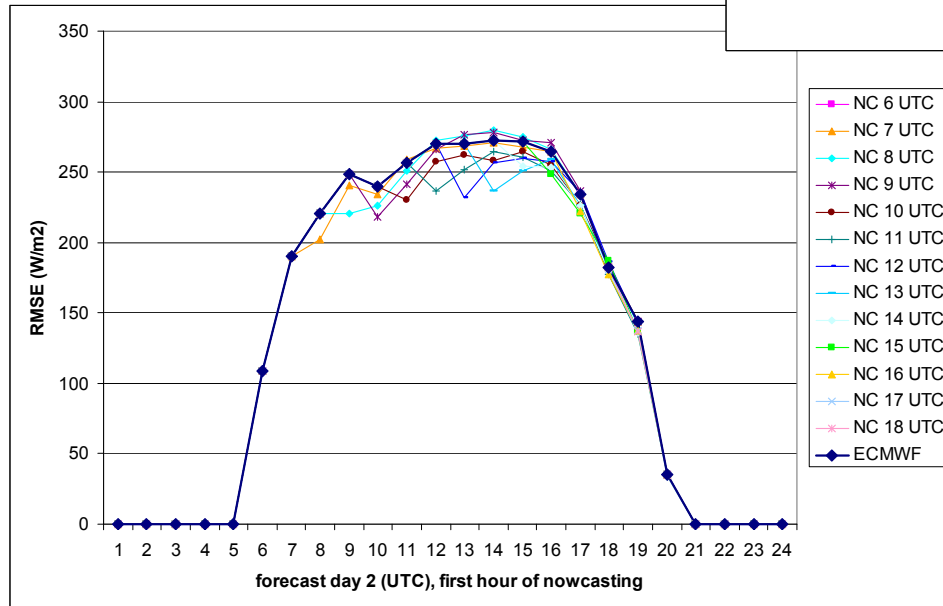
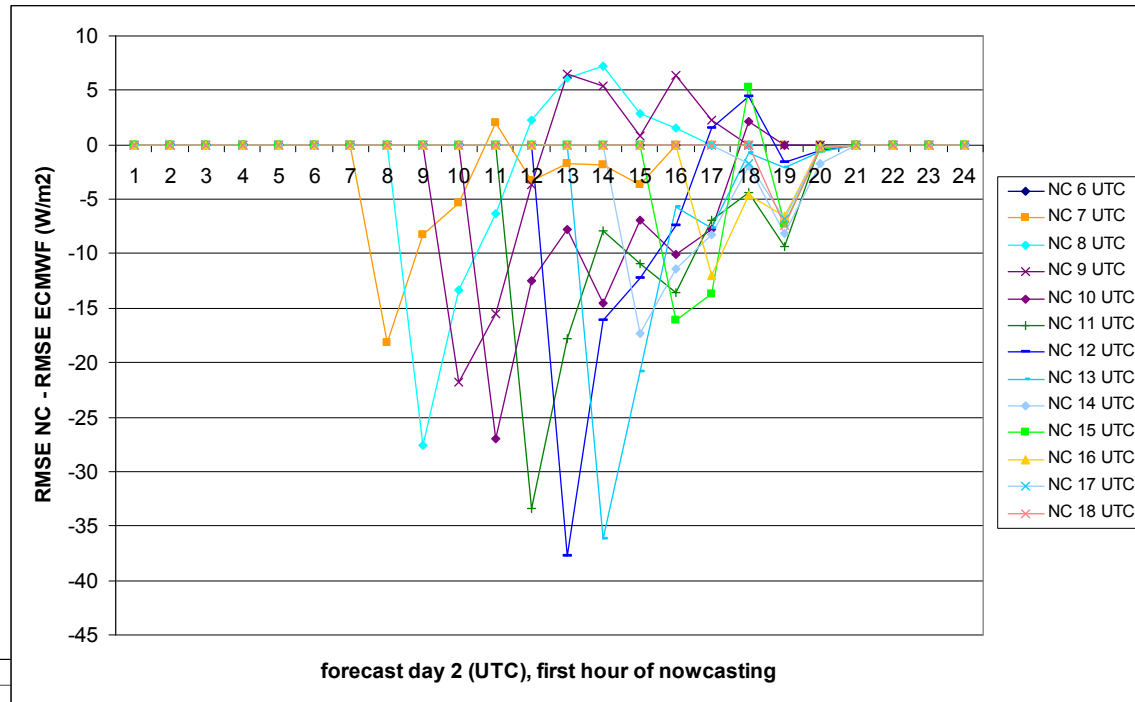
- ECMWF global irradiance forecasts
  - 0.25 degree spatial resolution
  - Day 2 forecast to deal with day ahead market
  - 1 hourly resolution generated from 3 hourly ECMWF output for day ahead/intra day markets
  - GHI2DNI conversion scheme for concentrating solar power
- BSRN, PSA and Andasol-3 surface radiation ground measurements
- MSG SOLEMI irradiance retrievals
- **MESOR QC/Benchmarking standards** (see references)

# Verification of ECMWF-based direct irradiance forecasts, PSA, 2005

Day 2, PSA, 2005 GHI_g>0	2 day Persistence Daily sum	ECMWF/DLR Daily sum	Persistence hourly	ECMWF/DLR hourly
Obs. DNI mean [Wh/m <sup>2</sup> ] or [W/m <sup>2</sup> ]	<b>6278</b>	<b>6278</b>	<b>485</b>	<b>485</b>
Bias [Wh/m <sup>2</sup> ] or [W/m <sup>2</sup> ]	<b>24</b>	<b>556</b>	<b>0.4</b>	<b>42</b>
RMSD [Wh/m <sup>2</sup> ] or [W/m <sup>2</sup> ]	<b>3380</b>	<b>2194</b>	<b>344</b>	<b>241</b>
Rel. RMSD [%]	<b>54</b>	<b>35</b>	<b>71</b>	<b>50</b>
Corr.	<b>0.33</b>	<b>0.71</b>	<b>0.52</b>	<b>0.76</b>



# Quantification sat-based nowcast impact vs. latest deterministic ECMWF forecast (case of Southern Spain)

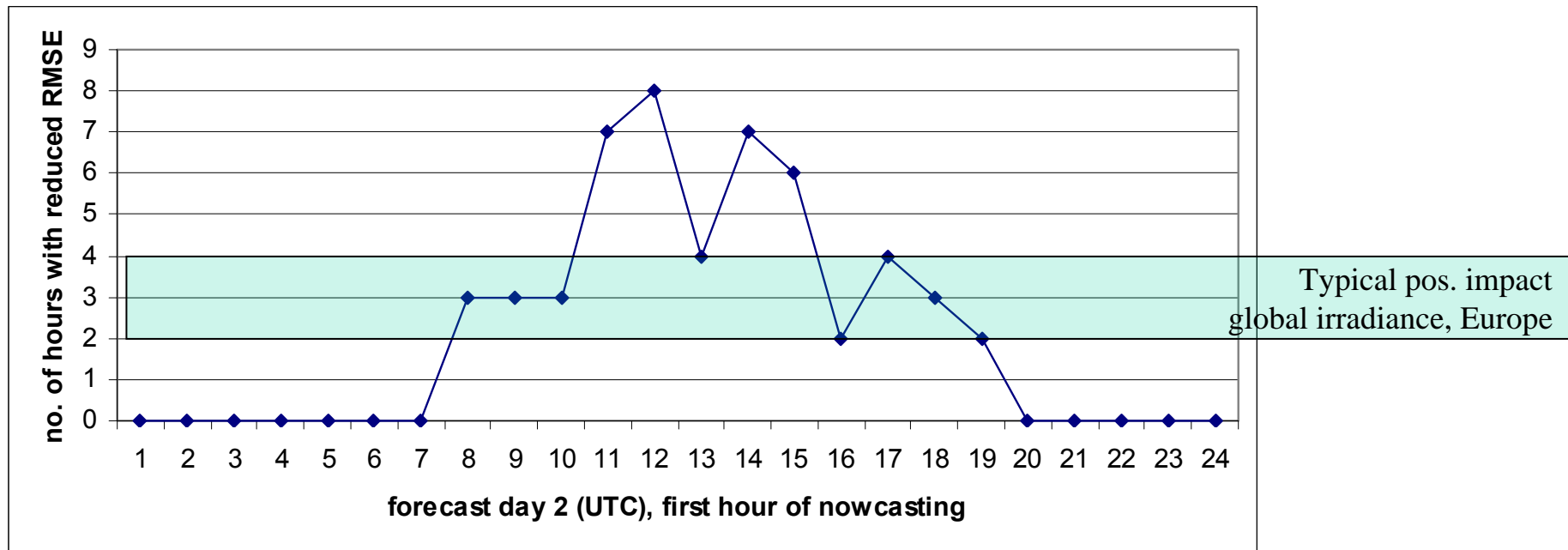


2007, all days  
Nowcasting only if discrepancy  
between ground measurement  
and forecast is observed

Mainly bias reduction !!!

# Duration of positive impact of satellite nowcasting

## Case of Southern Spain



Typical values in Europe  
for direct irradiance  
-> we don't know ....