





Leibniz Institute for Tropospheric Research

Estimation of cloud motion vectors: exploring different approaches using a dense network of solar radiation sensors

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Surface solar irradiance can be seen as a spatiotemporal process

An intuitive concept: clouds advect in space through time



Advection characterization is key for

- Variability analysis
- Forecasting applications

But it is **challenging** :

- There are several approaches
- In general, no "ground truth"

Our goal

Build a preliminary setup for <u>analyzing</u> and <u>cross-comparing</u> different cloud advection methods

• Cross-correlation of neighboring ground sensors





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- Optical flow using consecutive satellite images
 - Clear-sky index (kc = GHI/GHIcs) calculated using CAMS-Radiation data
 - TV-L1 algorithm with different regularization parameters





Build a preliminary setup for <u>analyzing</u> and <u>cross-comparing</u> different cloud advection methods

- Cross-correlation of neighboring ground sensors
- Optical flow using consecutive satellite images
- Wind speed from **ERA5** reanalysis
 - At different pressure levels (from 800 to 1000 hPa)





Our testbed location – HOPE-Jülich campaign



Ground irradiance data from the **HOPE-Jülich* campaign**

- 99 sensors, 10x12 km²,
- 15 sec measurements between April and August 2013





HD(CP)2 Observational Prototype Experiment, Macke et al. (2017) DOI: <u>10.5194/acp-17-4887-2017</u> <u>https://www.tropos.de/aktuelles/messkampagnen/blogs-und-berichte/hdcp2-hope-2013</u>

From sensor network to advection



- The correlation is calculated between all pairs of sensors
- A time window of **15 min** is used for the calculation of the correlation
- The correlations are represented as a function of the spatial lags (δx and δy)



From sensor network to advection





Estimating the advection with the ST autocorrelation function can be **very effective**

- \checkmark pure advection
- \checkmark homogeneous flow
- \checkmark no cloud formation/dissipation
- \checkmark clear patterns to track

... but it can also lead to **unsharp ACF** and **undetermination**



The **spatio-temporal structure** of the irradiance field can lead to **undetermination**



-> All pairs of sensors sensors along the cloud edge will behave synchronously 10



Blurry autocorrelation functions are very frequent:

- **no feature** to track (clear sky or overcast)

- **no pure advection** (e.g. presence of different flows)



Conclusion and outlook

• Lessons learned:

- The performances of the spatialautocorrelation method vary with the structure of the irradiance field
- Understanding the performances of ACF is an important prerequisite for our analysis

• Next steps:

- Use of the ST ACF as a reference for the evaluation of ERA5 wind speed and optical flow
- Understand the effect of the spatial scale & weather situation on the performance of the different approaches



Additional slides

Preliminary results





No systematic results:

(a) OF and ERA5 can both perform well(b) ERA5 can be better than OF(c) OF can be better the ERA5(d) Both approach can fail

Questions:

- Does this depend on the **meteorological situation**?
- For **Optical Flow**, how do we **choose the regularisation parameter**?