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# **Use of Several Sources of Spatio-temporal Information** to Improve Short-term Photovoltaic Power Forecasting

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## Context



- > High penetration rate of renewables characterized by a high variability:
  - **Jeopardize** the safe operating of the **grid system**
  - Financial penalties over production mismatches in a marketoriented context
- > A growing diversity of sources that:
  - Represent various phenomena at different time & spatial scales
  - Can be **combined** to better seize weather dynamics [1]

# **Objectives**

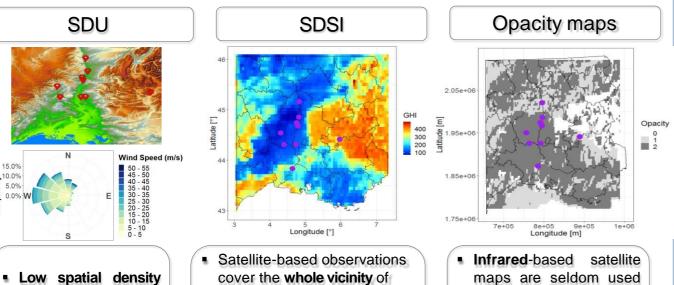


- > Develop an efficient and reliable PV production forecasting model for short-term horizons (15' - 6h):
  - Combine several spatio-temporal sources of data
    - → 9 spatially distributed PV units (SDU)
    - → Satellite-derived surface irradiance (SDSI)
    - → Infrared-based satellite observations (Opacity maps)
  - Select relevant information w.r.t. the forecasting horizon
    - Reduction of the computational burden

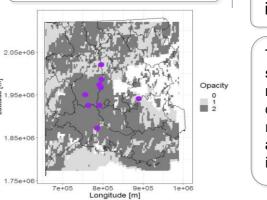
## Methodology



## Multi-inputs approach



- cover the whole vicinity of the site, but at a low
- resolution (e.g. 5 km x 5 km) Dimensionality issue (a radius of 100 km leads to 900 features)
- SDSI is derived from visible spectrum channels



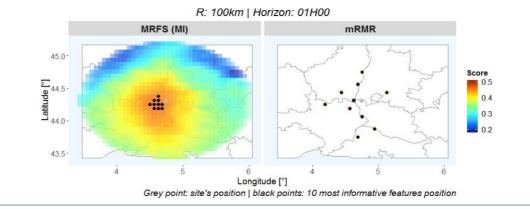
- maps are seldom used in the field of irradiance forecasting [3,4] and are missing from the PV production forecasting domain
- Opacity maps provide **nighttime** observations

### **Feature selection**

To cope with the **dimensional burden** induced by satellite-based datasets, it is common practice to train forecasting models on a set of well chosen pixels

Typically, a maximum relevance feature selection scheme (MRFS) is used. The mutual information (MI) criterion is chosen due to its capacity to assess nonlinear relationships. This scheme tends to select aggregated pixels that carry redundant information.

To address this issue, we propose the minimum redundancy and maximum relevance (mRMR) selection framework that selects spatially distributed features while assuring low redundancy among the selected information.



## Results



- A Random Forest (RF) model is considered for its good performances
- Inputs are clear-sky normalised (this approach mitigates the deterministic component of the signal associated with sun's path and highlights the stochastic component due to cloud motion)
- Training set: 2015 / Terting set: 2016

$$nRMSE^{x}(h) = \sqrt{\frac{1}{N}\sigma_{t=1}^{N} \left(\frac{P^{x}_{t+h} - P_{t+}^{x}}{P_{c}^{x} h}\right)^{2}}$$

of plants compared

distribution

plants is

dominant

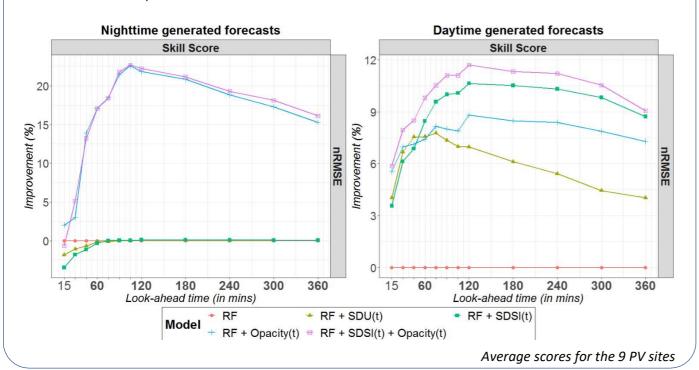
to other studies [2]

Spatial

of the

against winds

$$SS_{nRMSE^{x}}(h) = \frac{nRMSE_{Ref}^{x}(h) - nRMSE_{Mod}^{x}(h)}{nRMSE_{Ref}^{x}(h)}$$



# **Conclusions & Perspectives**



### Conclusions

- The mRMR feature selection scheme enables the selection of sparse features, which **reduces the redundancy** among the retained information
- The integration of observations from SDU exhibits accuracy improvements up to 7,5% compared to a purely temporal approach
- For this specific case study, information contained in satellite-derived datasets is richer despite a coarser spatial resolution: we observe improvements of the nRMSE score up to 11% w.r.t. the temporal-based approach. This finding is supposed to result from the **poor spatial distribution** of the power plant network.
- Opacity data offer a clear advantage over SDSI observations for nighttimegenerated forecasts
- The ST nature of opacity maps leads to more accurate daytime generated forecasts for the first leading times, but their informative potential is slightly lower than that of **SDSI** maps for horizons greater than 1 hour ahead
- Opacity maps appear as a variable able to compete with SDSI, but they cannot replace them. Quite the contrary, these sources of information are complementary and, when used together, they improve forecasting accuracy

## **Perspectives**

- Extend the present investigations by assessing the influence of these sources of information over probabilistic forecasting
- Include observations from all-sky imagers to improve the accuracy for very short
- The proposed contributions developed in this poster are detailed in a forthcoming journal paper

# Bibliography



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# About the author



- M.Sc. from Grenoble INP (2016)
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