

Renewable Energy Production Forecast

PROJ0016 - Big Data Project

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Photovoltaic production

Improvements



- Maximum power bounding
- Temperature influence
- Sine scaling
- (timezone fixing)

Resulting Model



$$P_{out} = \min\left(\frac{\eta I \cos(\theta) A}{\sin(alt)}, \text{MAX_POW}\right)$$

where

$$\texttt{MAX_POW} = \min\left(P_{\textit{max}}, P_{\textit{max}} - 0.004(T-25)\right)$$

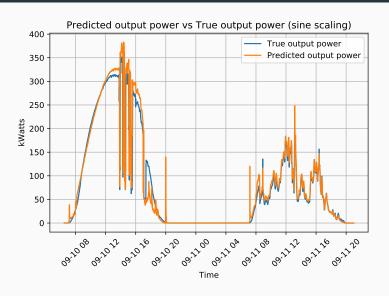
Updated Model Example



Using the thermodynamics laboratory data, the following model is obtained for a period ranging from September $10^{\rm th}$ to September $12^{\rm th}$.

Updated Model Example





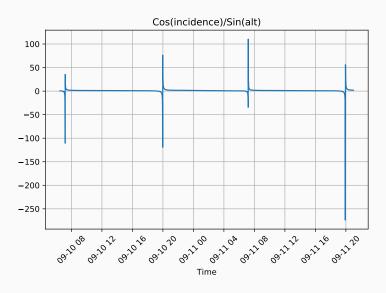
Issue



There seems to be peaks at regular periods of the day. By looking at the ratio between the sine and cosine, we notice the existence of peaks as well.

Issue





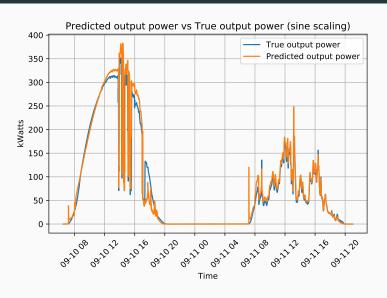
Issue



To fix the issue, we decided to force the output power to be equal to zero whenever both quantities are negative: it means the sun is "behind" the panels and has set.

Final model







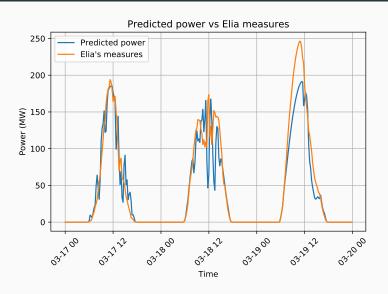
We have access to old (2017-2018) solar statistics data, containing among others the kVA (maximum solar power) of photovoltaic panels installed in the municipalities of Liège.



Using this, as well as some measure of area of photovoltaic panels per kWPeak (\sim 7), we obtained a rough measure of the average area of photovoltaic panels installed in the province of Liège.

Setting the efficiency η to some fixed arbitrary value (0.15), we obtain the following model (for a period of 3 days).







- Use PyStan to fit sensitive parameters to posterior measures made by Elia.
- Potentially use Elia's measure of monitored capacity as starting point

Next objectives



- ullet Deal with uncertainty on parameters $(\eta, \text{ peak area, etc.})$
- Find relevant irradiance forecast data

Photovoltaic panels

enumeration

Test set – Google Maps



In the previous review, we talked about using the Google Maps satellite imagery for detection (not training) because of its high resolution.

Unfortunately, the Google Maps API isn't free. Actually, it is probably reserved to commercial use (cannot remove Google watermark).

Test set – Google Earth Engine



Afterwards, we found the Google Earth Engine API [1]. It combines a multi-petabyte catalog of satellite imagery and geospatial datasets, like Landsat (Nasa) [2] and Sentinel (Esa) [3].

The API is primarily available in JavaScript and secondarily in Pyhton.

The access to this API is restricted, but we obtained the rights.

Test set



Unfortunately, the available satellite imagery had an insufficient resolution for Belgium.

WalOnMap is still our best choice.



Last review, we proposed a few way to investigate in order to find a learning set.

- Asking to a university/repository owner.
- OpenStreetMap solar panel coordinates dataset for England.
- Keep searching for a dataset.



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Last review, we proposed a few way to investigate in order to find a learning set.

- Asking to a university/repository owner. We sent multiple mails, none were answered.
- OpenStreetMap solar panel coordinates dataset for England. Gives only (very) approximate location, not surface nor shape.
- Keep searching for a dataset. We found one comming from Duke University! [4]



This dataset contains the geospatial coordinates and border vertices for over $19\,000$ solar panels across 601 high resolution (5000×5000) images from four cities in California (Fresno, Modesto, Oxnard, Stockton).





Figure 1: Modesto city, picture 1, zoomed 10 times



The solar panel geospatial coordinates are very precise and provided as polygon vertices in either a csv or a json.

```
"polygons": [
  "polvgon id": 1.
 "centroid latitude": 36.926310139710594,
 "centroid longitude": -119.84055537864529,
 "centroid latitude pixels": 107.6184581.
 "centroid longitude pixels": 3286.151487,
 "city": "Fresno".
 "area pixels": 136.192872,
 "area meters": 1513.254134.
 "image name": "11ska460890".
 "nw corner of image latitude": 36.92633611,
 "nw corner of image longitude": -119.8516222,
 "se corner of image latitude": 36.91323333.
 "se_corner_of_image_longitude": -119.8343,
 "datum": "NAD83".
 "projection zone": "11".
 "resolution": 0.3.
 "jaccard index": 0.914019659.
 "polygon vertices lat lon": [
   -119.84030285285368.
   36.92625126752494
   -119.84068167074253,
   36.92636019465177
```

Figure 2: SolarArrayPolygons.json 20



These coordinates are not useful as-is for training.

We have written (using mainly OpenCV [5]) a code learning_set.py to

- 1. transform the polygons into black and white classification images.
- 2. slice the images into smaller (256 \times 256) images.
- 3. produce (X, Y) image pairs.



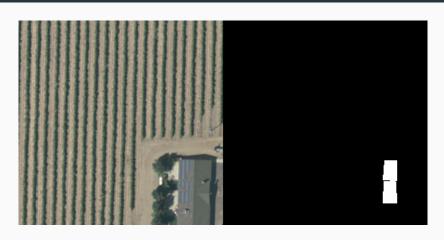


Figure 3: Example of (X, Y) image pair.

Learning set – Data augmentation



We plan to rotate and translate these (X, Y) image pairs in order to increase the quantity of available training data.

Next objectives



Now that we have suitable learning and testing sets, we are ready to build our first detection model(s).

Wind production

Data



New data has been retrieved:

- Wind speed measures from February 2012 to now (one single location in Wallonia for now)
- Physical model output using the wind speed measures and the wind turbines data collected for last review, from 2012 to now
- Elia's wind power measures from February 2012 to now

Models



For now, only four types of models have been investigated.

- A Gaussian Process based on a Rational Quadratic Kernel predicting the power from the wind speed (at a unique location)
- A Gaussian Process based on a Rational Quadratic Kernel predicting residuals between the measures of Elia and the physical model
- A Random Forest predicting the power from the wind speed, physical model, and timestamp.
- Extra-Trees predicting the power from the wind speed, physical model and timestamp.



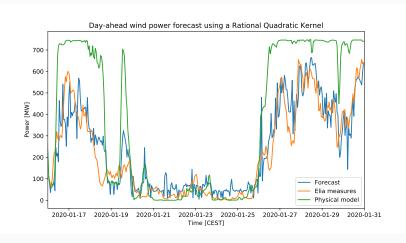


Figure 4: Gaussian Process with Rational Quadratic Kernel - 1 month of training data



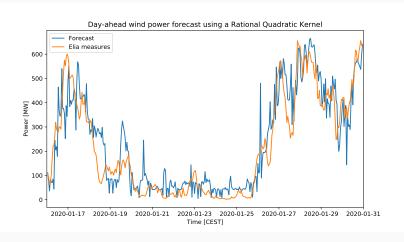


Figure 5: Gaussian Process with Rational Quadratic Kernel - 1 month of training data



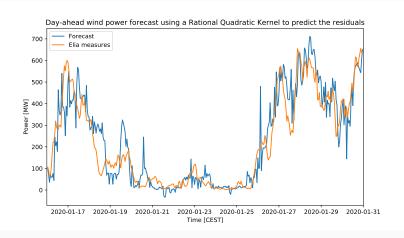


Figure 6: Gaussian Process with Rational Quadratic Kernel - 1 month of training data



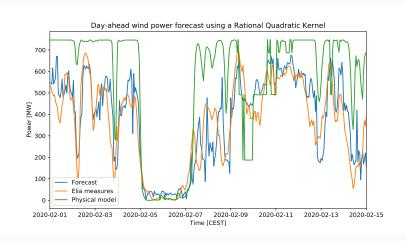


Figure 7: Gaussian Process with Rational Quadratic Kernel - 1 month of training data

Results - Gaussian Process



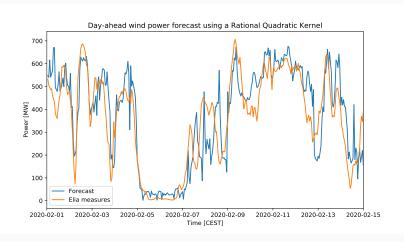


Figure 8: Gaussian Process with Rational Quadratic Kernel - 1 month of training data

Results - Gaussian Process



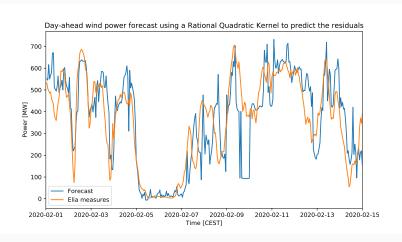


Figure 9: Gaussian Process with Rational Quadratic Kernel - 1 month of training data

Results - Gaussian Process - Issues



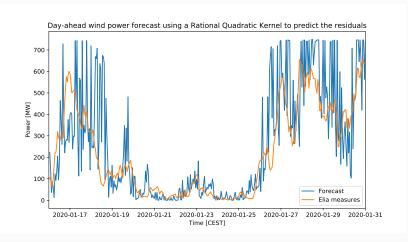


Figure 10: Gaussian Process with Rational Quadratic Kernel - 3 months of training data

Results - Gaussian Process - Issues



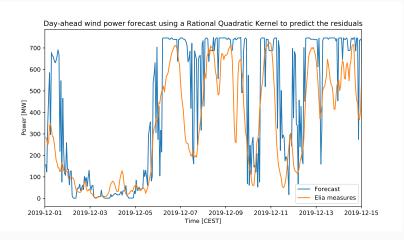


Figure 11: Gaussian Process with Rational Quadratic Kernel - 1 month of training data

Result - Random Forest



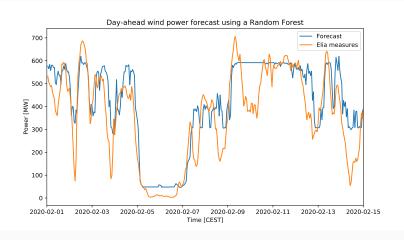


Figure 12: Random Forest - 1 month of training data

Result - Random Forest



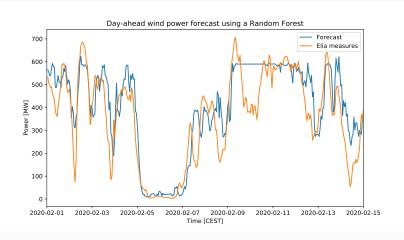


Figure 13: Random Forest - 1 year of training data

Result - Random Forest



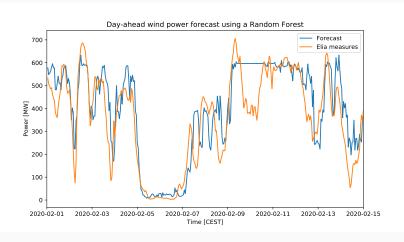


Figure 14: Random Forest - 8 years of training data

Result - Extras Trees



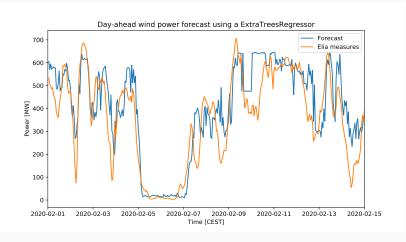


Figure 15: Extra Tree - 1 month of training data

Result - Extras Trees



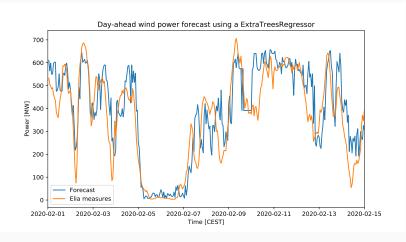


Figure 16: Extra Tree - 1 year of training data

Result - Extras Trees



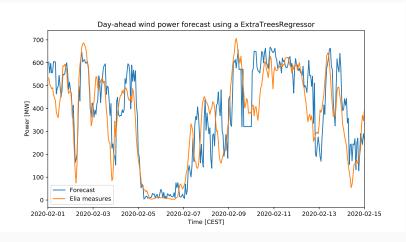


Figure 17: Extra Tree - 8 years of training data

Questions ?

References

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OpenCV. URL: https://opencv.org/.