



Solar Radiation Prediction

Battery: Solar storage
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Motivation

- Estimate levels of solar radiation
- Why:
 - Solar energy fluctuations
 - Predictability, easier integration to conventional production
- How:
 - Analyze samples from 4 months period (1-9-2016 - 31/12-2016)
 - Machine learning model(s)

The Data

- HI-SEAS weather station, Hawaii
- Collected data/features (32686 rows):
 - Solar radiation [W/m^2]
 - Temperature [F]
 - Atmospheric pressure [Hg]
 - Humidity [%]
 - Wind speed [miles/h]
 - Wind direction [degrees]
 - Time sun rise
 - Time sun set
- Response variable:
 - Solar radiation [W/m^2]

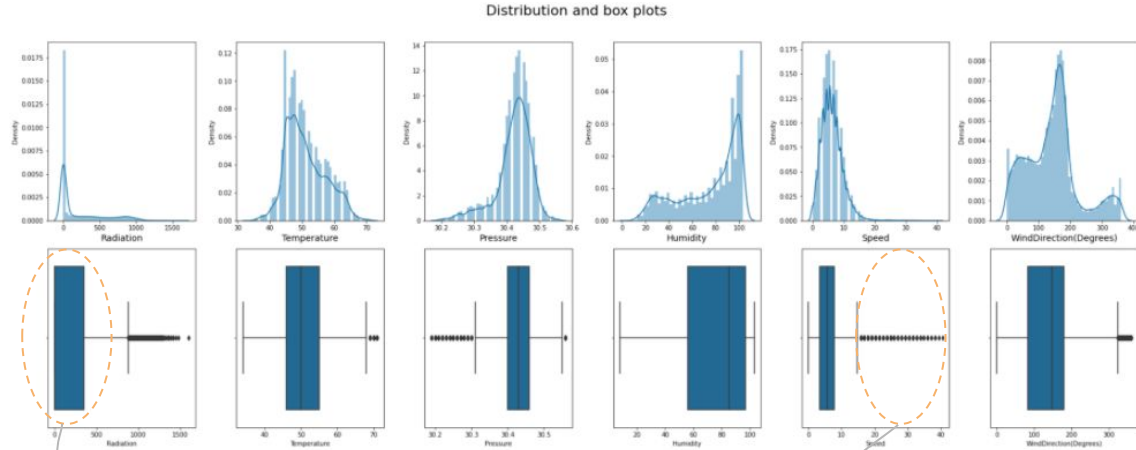
Approach

- Challenge:
 - Time Series
 - Cyclicity
 - Seasonality
- Feature Engineering
 - Deal with Time
- Models
 - Simple Linear Regression
 - Random Forest
 - Gradient Boosting
- Metrics:
 - R-Squared, Mean-Squared Error, CV

Train/Test Split

- Begin Sep. - End Nov.
 - As train set (70%)
- Begin Dec. - End Dec.
 - As test set (30%)
- Training Set, TimeSeriesSplit:
 - 3 Folds to 'simulate' seasonality

Explore - Distribution



Roughly 50% of values
located between 0 and 250
W/m²

Extreme outliers

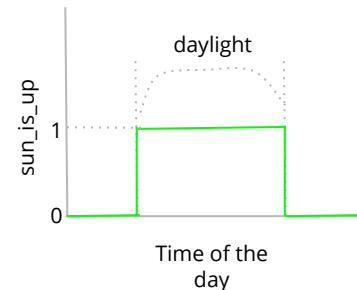
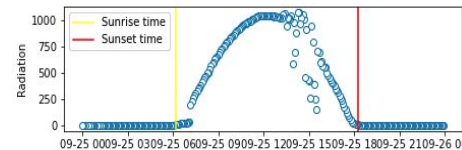
Feature Engineering: Time

2 cases

- Sun_is_up feature:

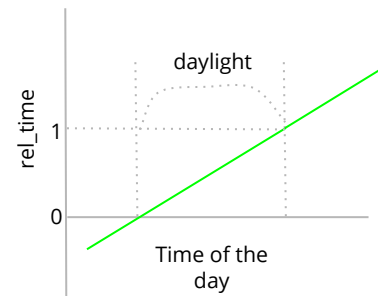
- 0 - when outside of [SunRise - SunSet]
- 1 - when during [SunRise - SunSet]
- Band pass filter, hard cut-off

Typical solar radiation dist.

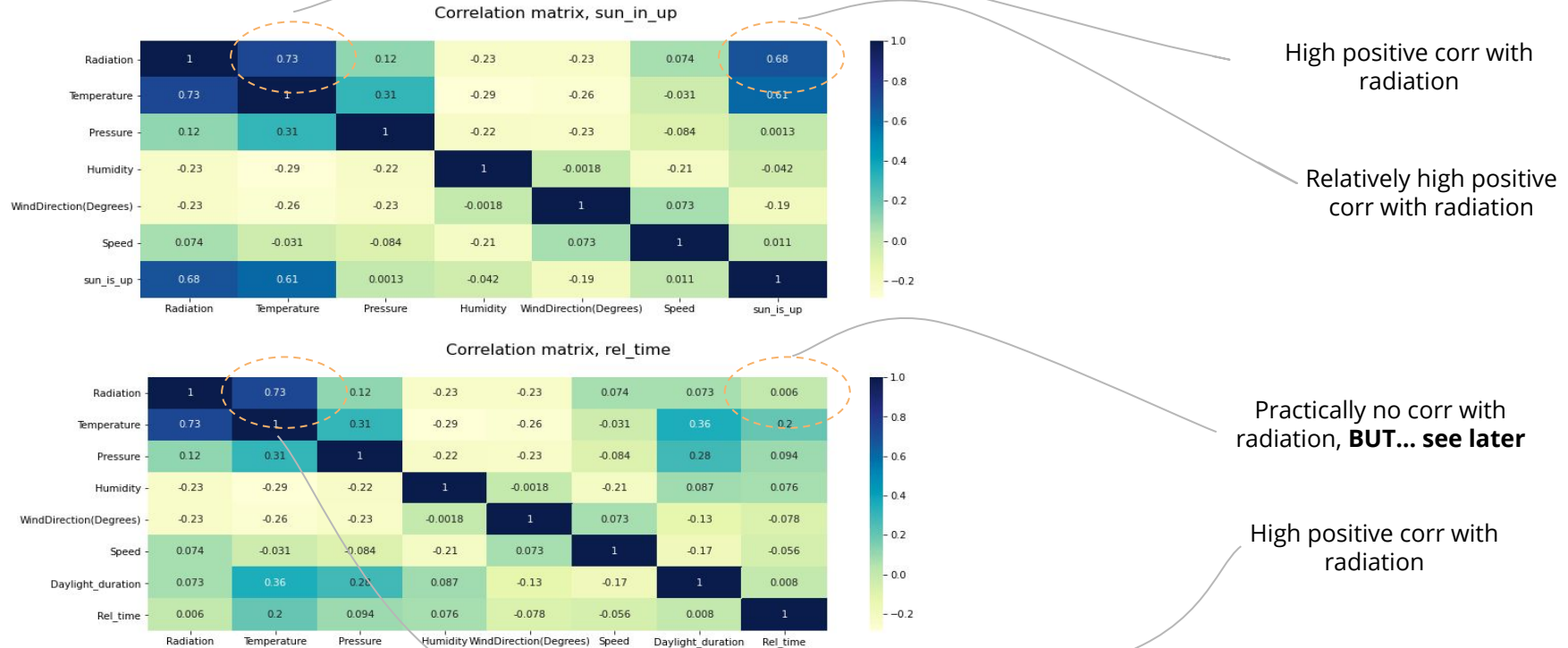


- Rel_time feature:

- $(\text{Current_time} - \text{SunRise_time}) / \text{Daylight_duration}$
 - $\text{Daylight_duration} = \text{SunRise_time} - \text{SunSet_time}$
 - Rel_time < 0, if before sunrise
 - Rel_time = 0 at exact sunrise
 - Rel_time (0,1) between sunrise and sunset, **linear**
 - Rel_time = 1 at exact sunset
 - Rel_time > 1 if after sunset



Explore - Correlation



Models

- Multivariate Linear Regression
- Random Forest
 - 3 folds - Cross Validation (sklearn.timeseriessplit)
 - Randomized best hyper-parameters search
- Gradient Boosting
 - 3 folds - Cross Validation (sklearn.timeseriessplit)
 - Randomized best hyper-parameters search

`nr_estimators = [100, 300, 500, 800]`

Average to control overfit

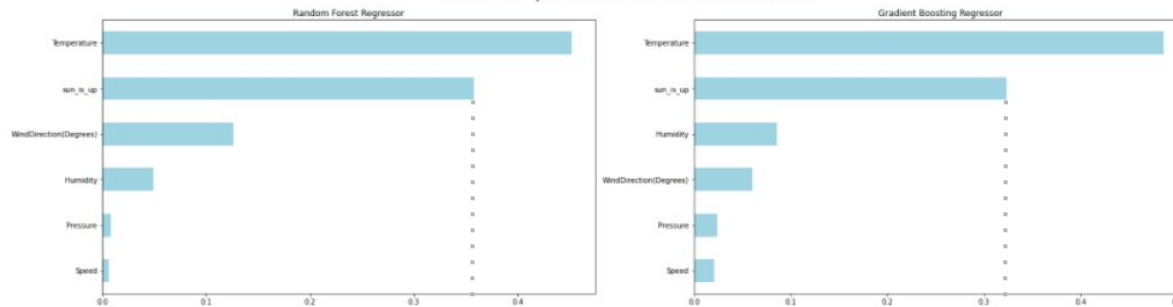
Boosting stages, control overfit

Feature Importance

Gini/Information Gain

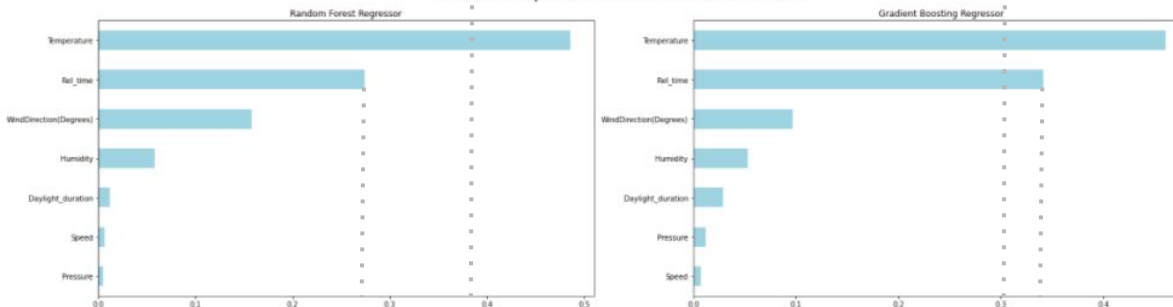
sun_is_up feature

Feature importances in the two ML models



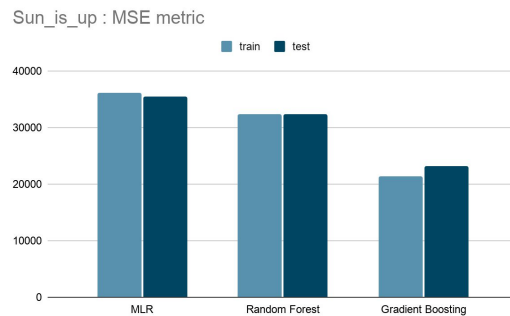
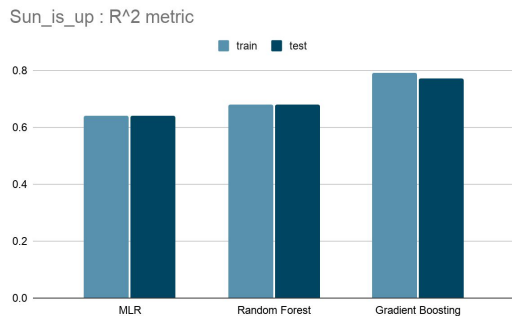
rel_time feature

Feature importances in the two ML models



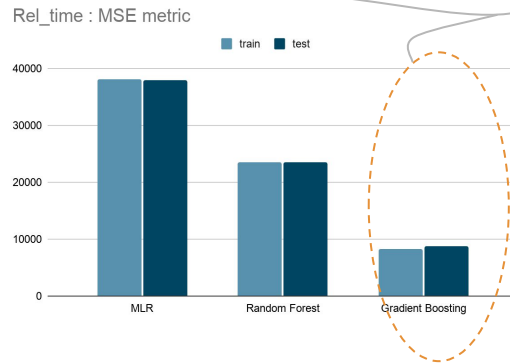
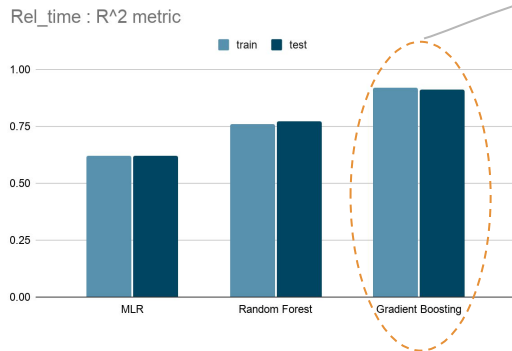
Results - Metrics

sun_is_up feature



Best combination of model/feature

rel_time feature



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  "feature_case": "rel_time",  
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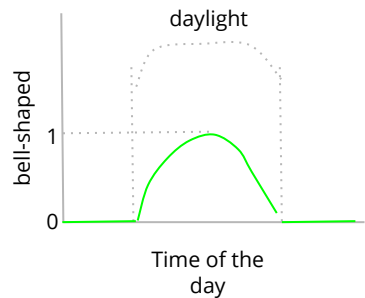
Future work

- Feature Engineering

- Model time within daylight period as bell-shaped
- Exclude samples outside daylight period
 - Since solar radiation is practically zero during night hours

- Models

- Try more models
 - **XGBoost...**
 - **SVR**



Questions