

# SOLAR ENERGY GENERATION PREDICTION

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- Data Collection
- Initial Findings
- Modeling
- Conclusions/ Recommendations
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- Streamlit App



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# PURPOSE OF INVESTIGATION

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# BACKGROUND

Due to the variable nature of renewable energies, as well as the uncertainty and variability inherent in any electrical architecture, predicting the power generation from each site is very challenging.

Consequently, the integration of solar energy generation sites with larger electrical grids can be complicated. One way of easing this burden is accurately reporting and predicting power generation from these variable generation sites.

# PROBLEM STATEMENT

Using the solar generation output of a specific solar farm site, along with the corresponding weather data, can a model predict the solar generation output of that specific site within 10% of max output?

Furthermore, can the same model architecture optimized for a specific site be used to predict the generation of another site or even an entire campus?



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# Data Collection

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# DATA COLLECTION

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Two years of high-granularity solar energy generation data from 42 sites across five campuses at La Trobe University, Victoria, Australia.

## Includes:

- Solar Generation Data reported at 15 minute intervals
- Weather data from Australian Bureau of Meteorology (BOM) using longitude and latitude of each campus
  - Includes Apparent Temp, Air Temp, Dew Point Temp, Relative Humidity, Wind Speed, Wind Direction



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# INITIAL FINDINGS

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# INITIAL FINDINGS

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**01**

## **CAMPUS/SITE VARIABILITY**

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Campuses ranging  
from 1 to 27 Solar  
Generation Sites

**02**

## **DAILY VARIABILITY**

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Weather and electrical  
architecture  
variability

**03**

## **SEASONALITY**

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Seasonality present  
on daily and seasonal  
scales

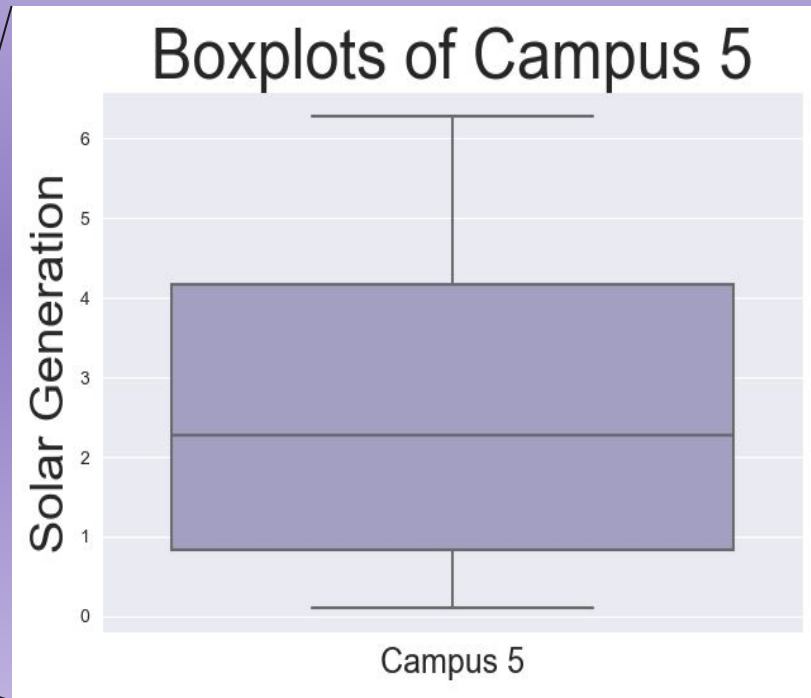
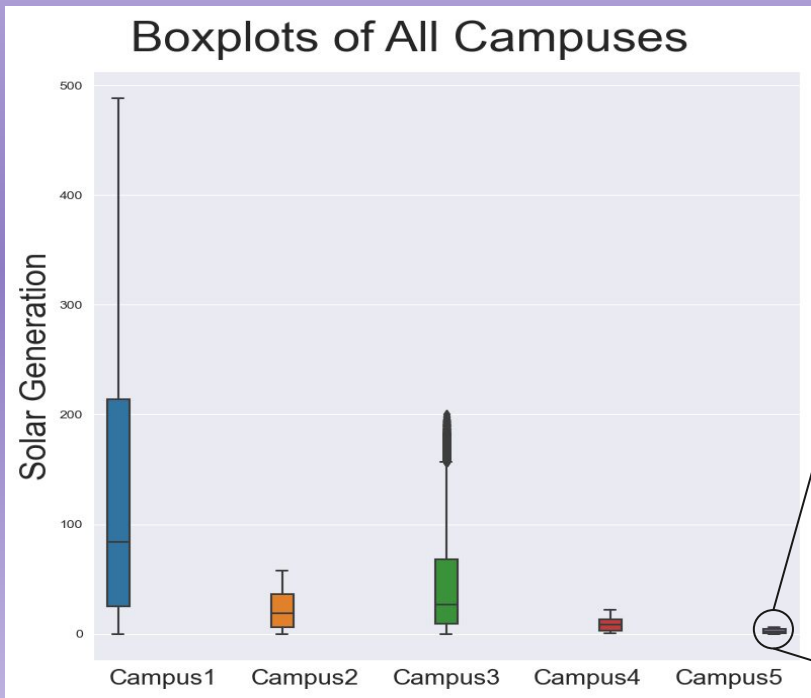
**04**

## **CORRELATION**

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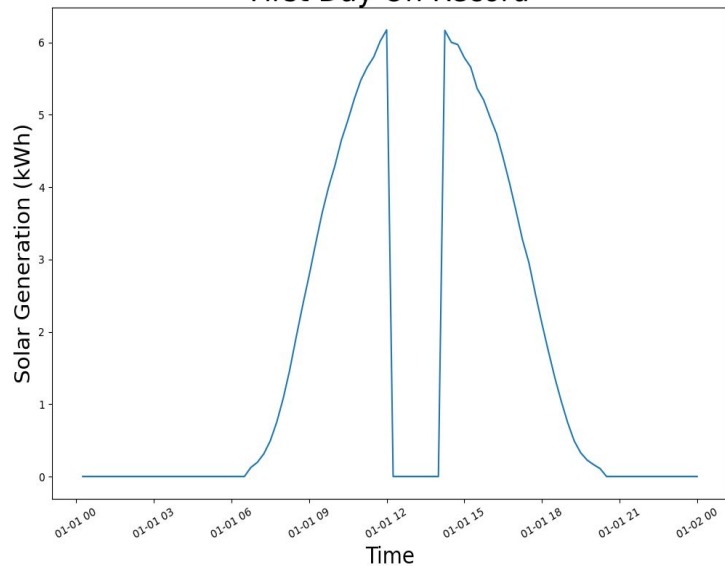
Auto- and partial  
autocorrelation  
reflects seasonality  
and variability

# CAMPUS / SITE VARIABILITY

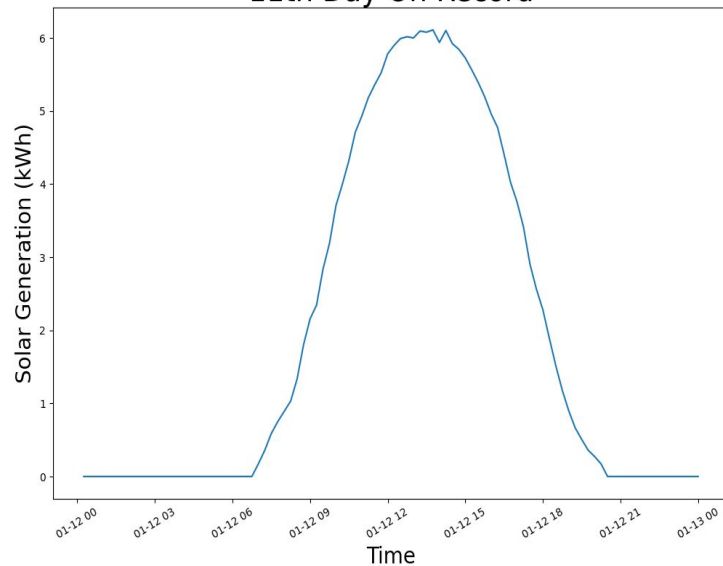


# DAILY VARIABILITY

First Day On Record

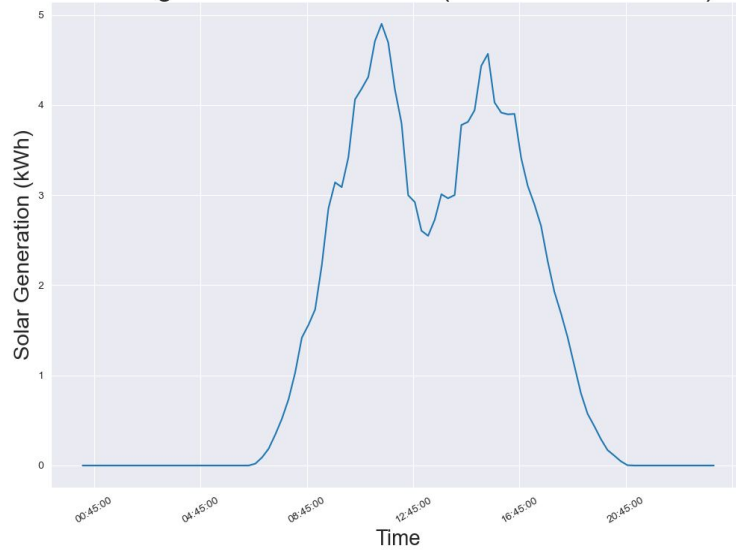


11th Day On Record

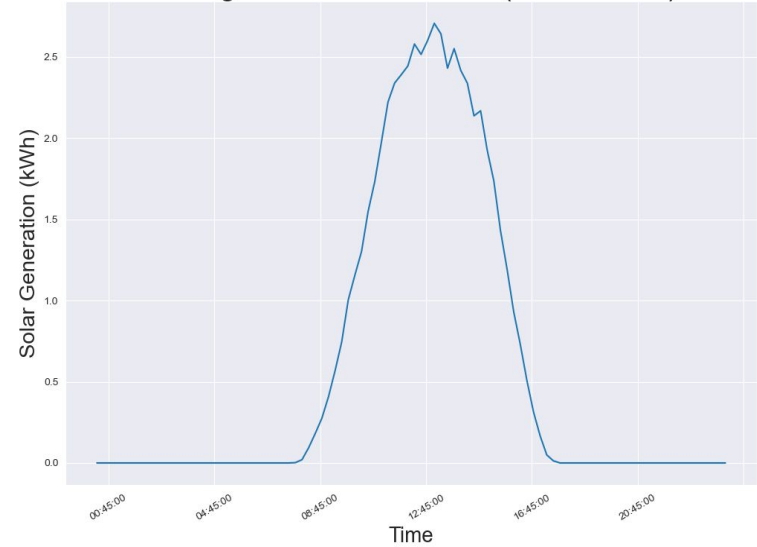


# SEASONALITY

Averaged Solar Generation (Summer 2020/2021)

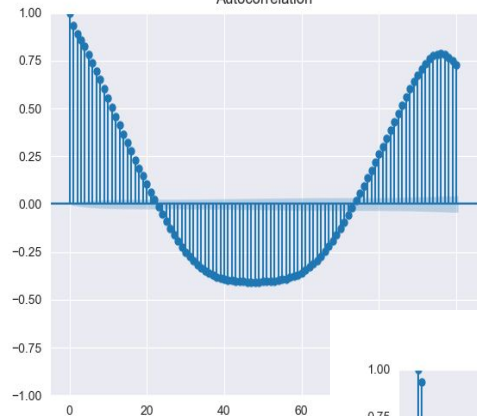


Averaged Solar Generation (Winter 2021)



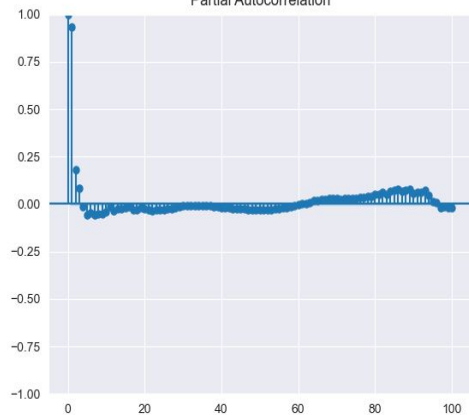
# CORRELATION

Autocorrelation

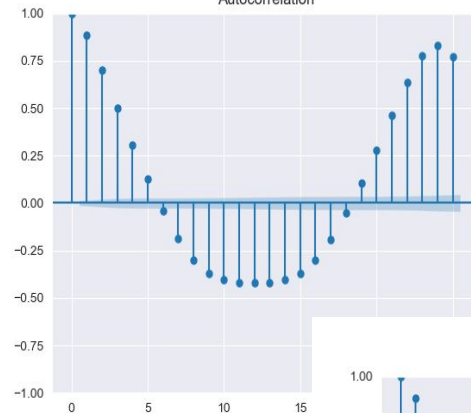


15 - Min  
Interval

Partial Autocorrelation

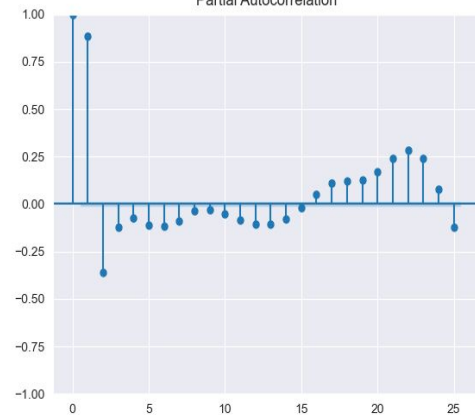


Autocorrelation

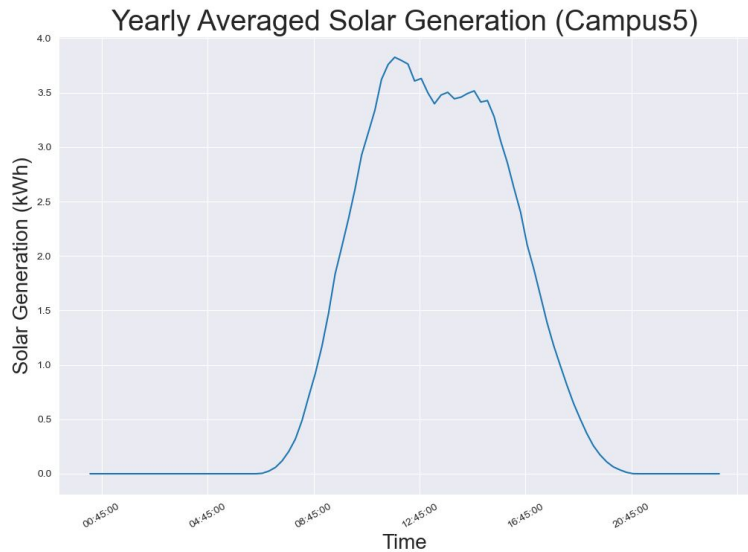


Resampled  
Hourly

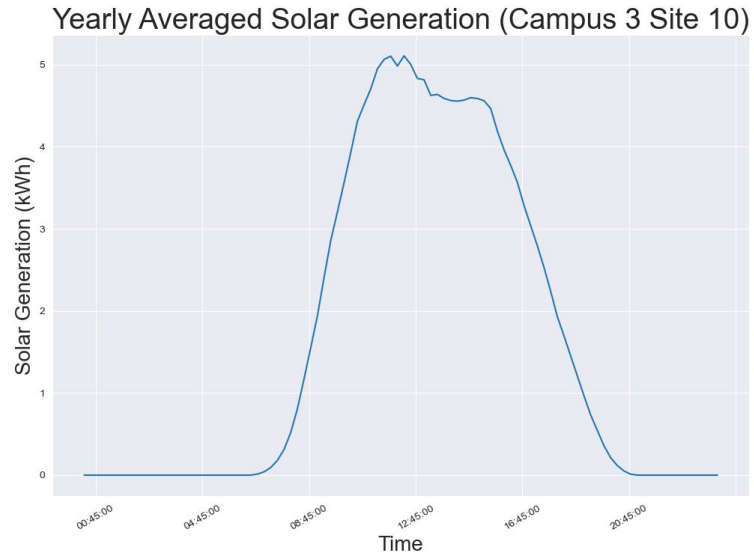
Partial Autocorrelation



# CAMPUS 5



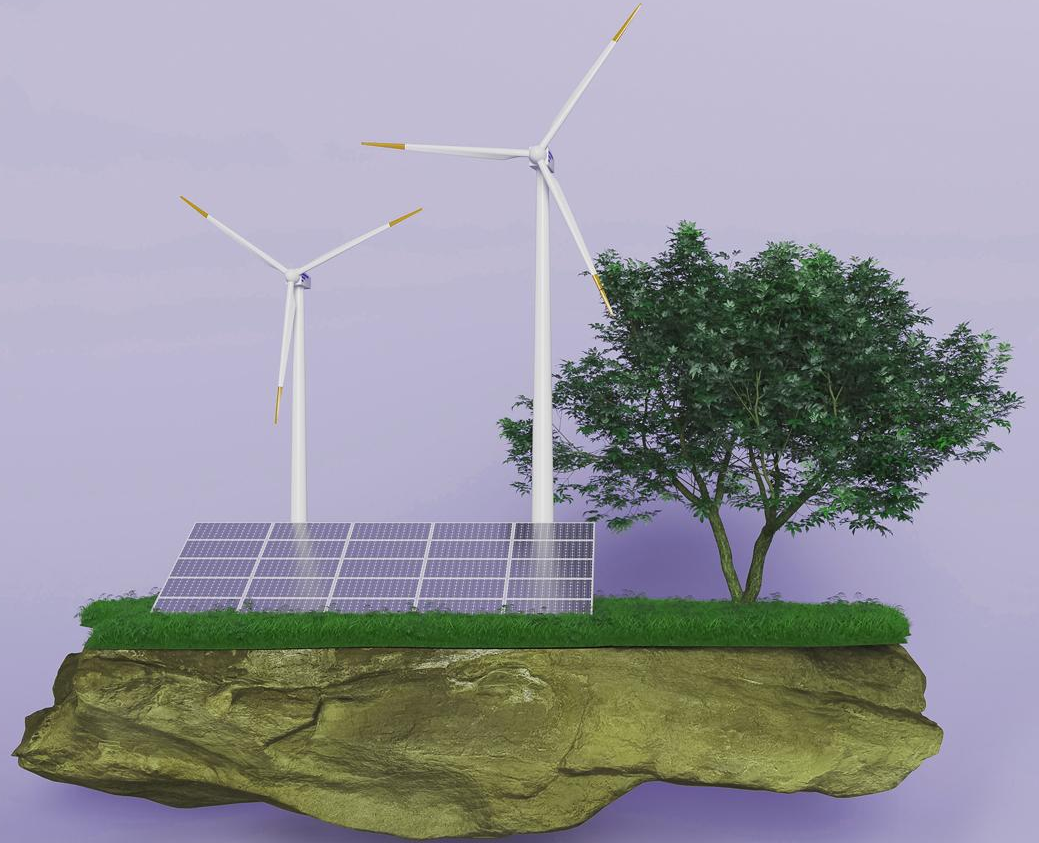
# CAMPUS 3 - SITE 10



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# MODELING

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# MODELING

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01

**BASELINE**

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MAE: 1.141  
RMSE: 2.447

02

**ARIMA**

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MAE: 1.524  
RMSE: 1.934

03

**LAGGED LINEAR**

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MAE: 0.260  
RMSE: 0.629

04

**UNIVARIATE**

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MAE: 0.1871  
RMSE: 0.5765

05

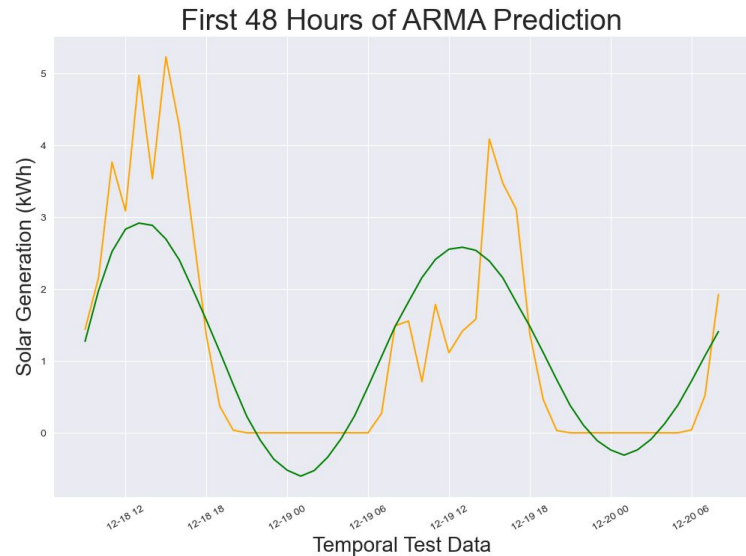
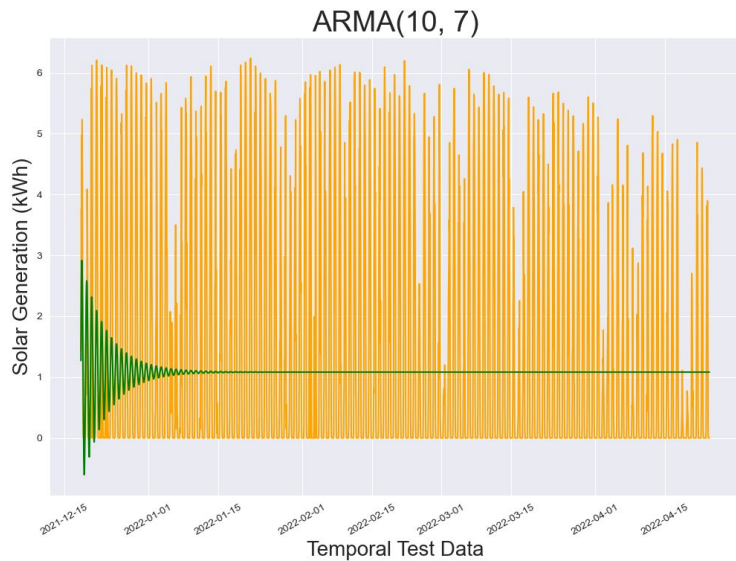
**MULTIVARIATE**

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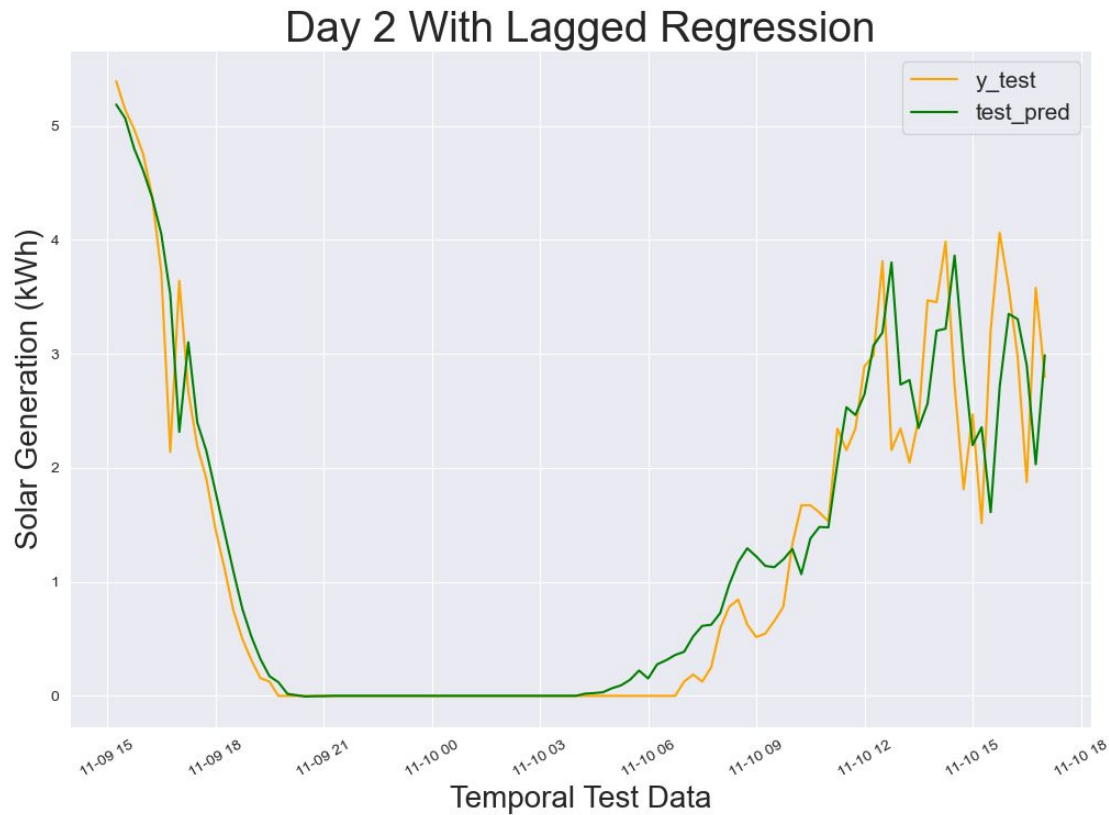
MAE: 0.4346  
RMSE: 0.8450



# ARIMA (10, 0, 7)



# LAGGED LINEAR



# BEST MODEL APPLIED TO ALL (MIN/MAX SCALED)

<u>Campus/Site</u>	<u>MAE Testing</u>	<u>RMSE Testing</u>	<u>Max Production</u>
Campus 5	0.02992	0.09386	6.281
Campus 1	0.03372	0.08035	488.088
Campus 2	0.03548	0.09301	57.363
Campus 3	0.03617	0.09281	200.215
Campus 4	0.03916	0.12786	21.938
Site 6	0.028186	0.07417	28.734
Site 8	0.03348	0.0897	26.969
Site 10	0.03832	0.11368	7.75
Site 12	0.05323	0.15264	17.594



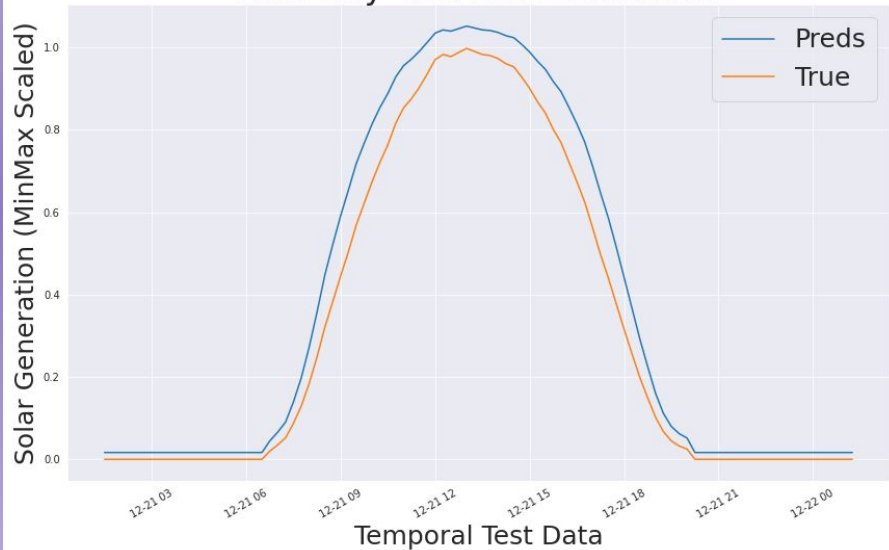
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# POST MODEL ANALYSIS

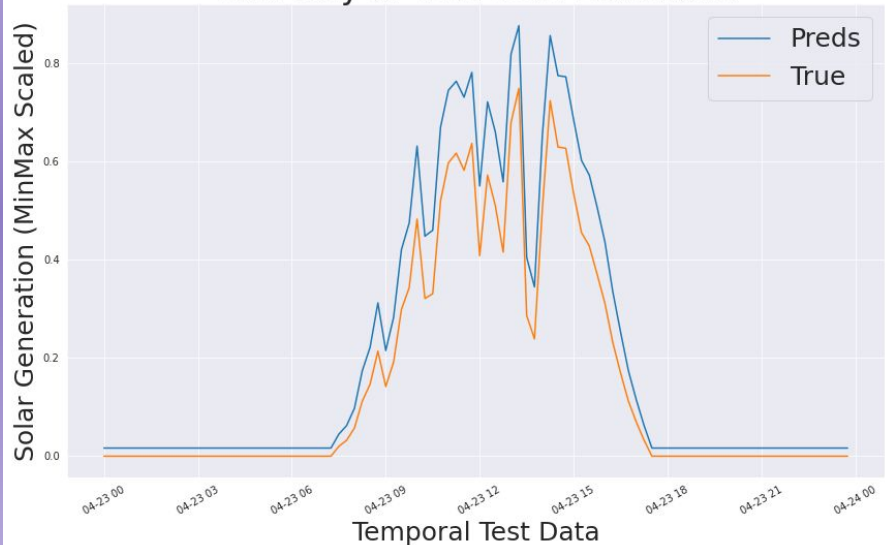
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# POST MODEL ANALYSIS

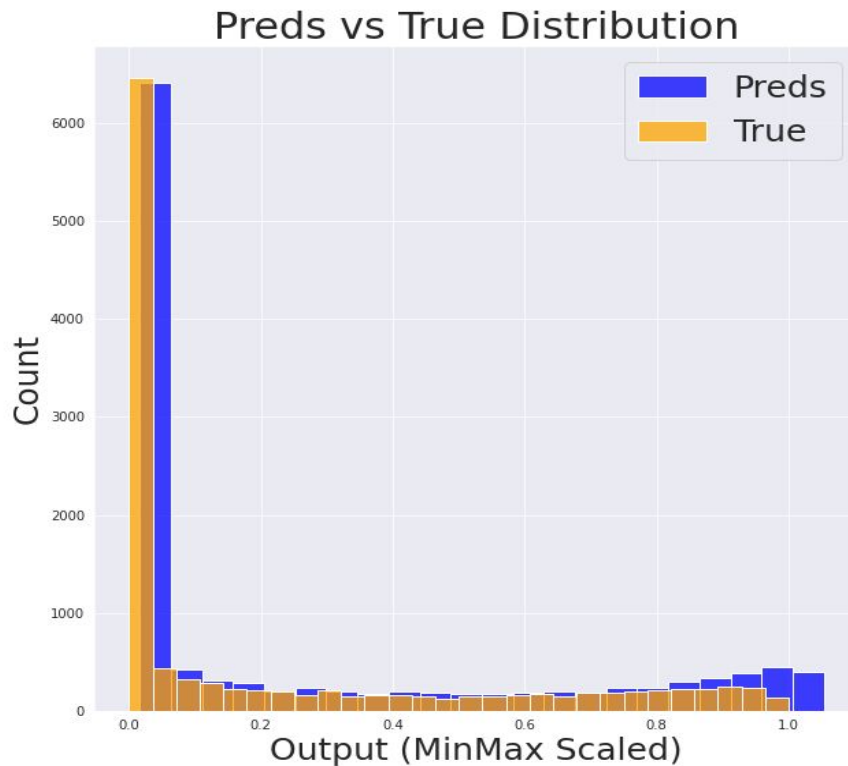
First Day Test Set Predictions



Last Day of Test Set Predictions

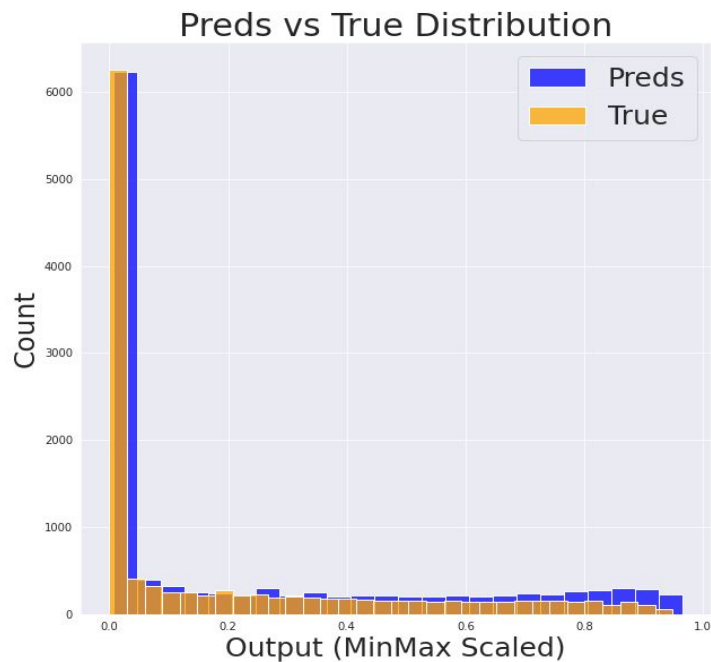


# POST MODEL ANALYSIS

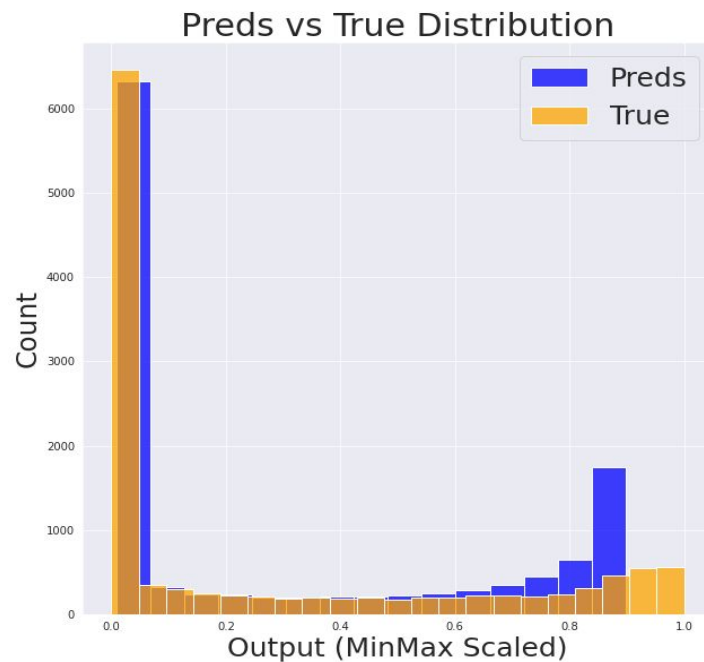


# POST MODEL ANALYSIS

## Campus I



## Site 10



# POST MODEL ANALYSIS

<u>Campus/Site</u>	<u>Preds MAE</u>	<u>Preds RMSE</u>
Campus 5	0.0571	0.0757
Campus 5 Subtracted	0.0406	0.0601
Campus 1	0.0458	0.0651
Campus 1 Subtracted	0.0383	0.0601
Site 10	0.0468	0.0692
Site 10 Subtracted	0.0399	0.0653





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# CONCLUSIONS/ RECOMMENDATIONS

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# CONCLUSIONS

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Based on the wide variety of analyses and modeling conducted, the best model could predict solar generation within 10% of max output.

Moreover, the same model architecture can be used to predict the generation of other sites and even entire campuses?



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# **FUTURE RESEARCH**

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# FUTURE RESEARCH

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**01**

## **SUNLIGHT**

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Accurate Sunrise  
and Sunset Feature

**02**

## **CLOUD COVER**

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Weather Data  
Specific to Cloud  
Cover

**03**

## **ALL SITES/CAMPUSES**

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Employ different  
models to each  
site/campus and  
levels

**04**

## **ARCHITECTURE VARIABILITY**

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Maintenance  
schedules or  
electrical  
malfunction reports  
feature

# FUNCTIONING APP



You can replace the image on the screen with your own work.  
Just right-click on it and select “Replace image”

# THANKS

Do you have any questions?



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