
Exploring Environmental Factors and their Impact on Solar Power

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Agenda

- Introduction
- Initial Exploration of Variables
- Correlation to Power
- Extract Significant Variables
- Power Prediction

Introduction

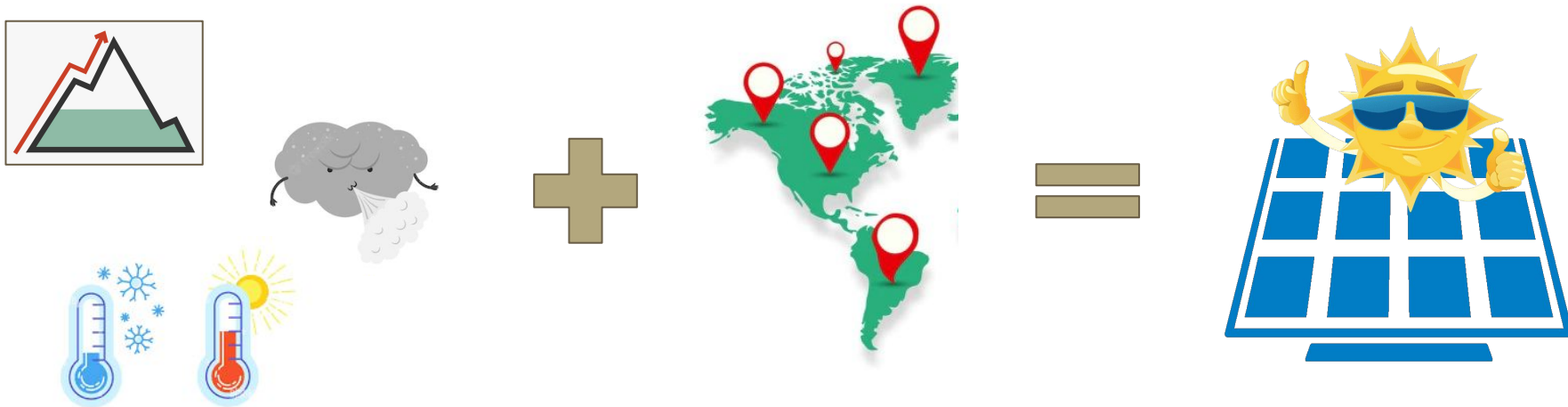
Background

- Solar energy is a rapidly **growing** market.
- Predicting solar power output is **vital**.
 - Maintaining current solar plants
 - Deciding where to build next
- Power is normally modeled **with irradiance** data.
 - Measurements require specific sensors which are **costly in time and money**.
- We want to **predict power** using **only weather** and **location** features.



Objective

- Find the **weather** and **location features** that best **explain** solar **power** output.



Data

- **Power output + environmental factors** from **12 US sites** over **14 months** (2017 to 2018)
- 21K observations
- Features:

Location:

- Latitude
- Longitude

Time:

- Date
- Time sampled
- Season

Weather:

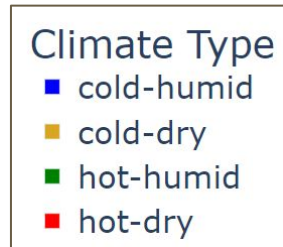
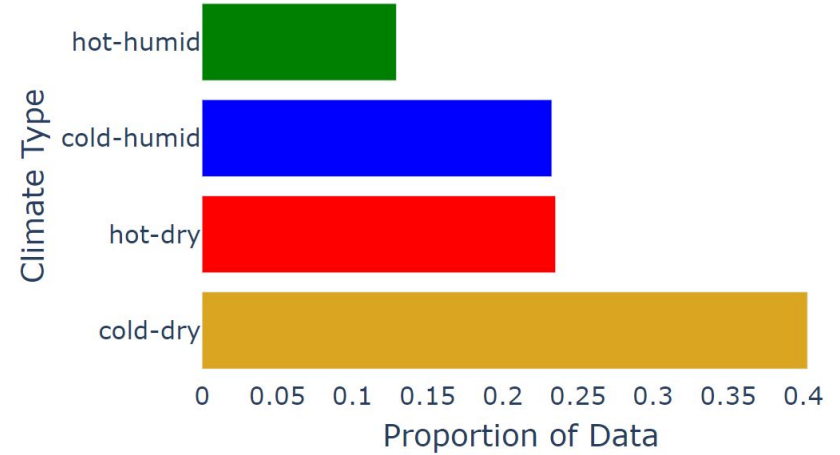
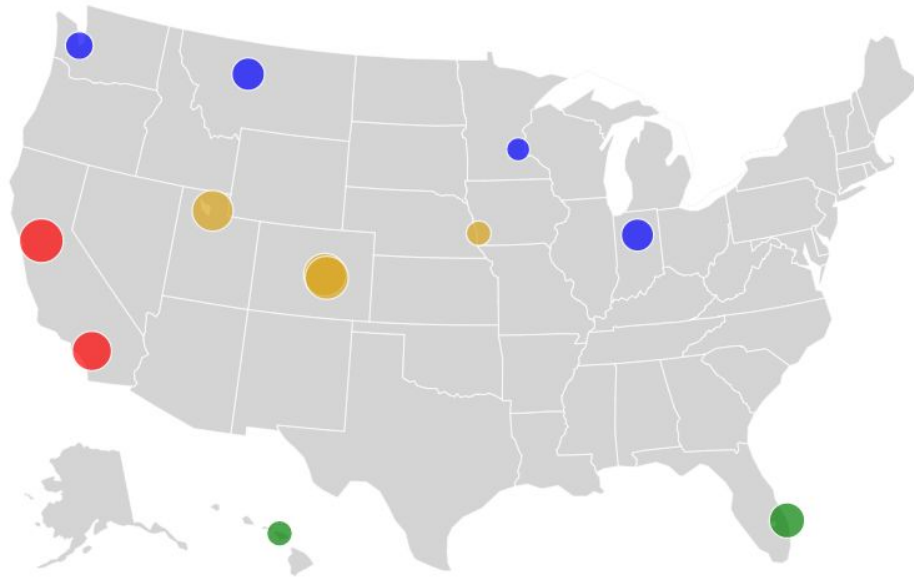
- Altitude
- Humidity
- Ambient temperature
- Wind speed
- Visibility
- Pressure
- Cloud Ceiling

Source:

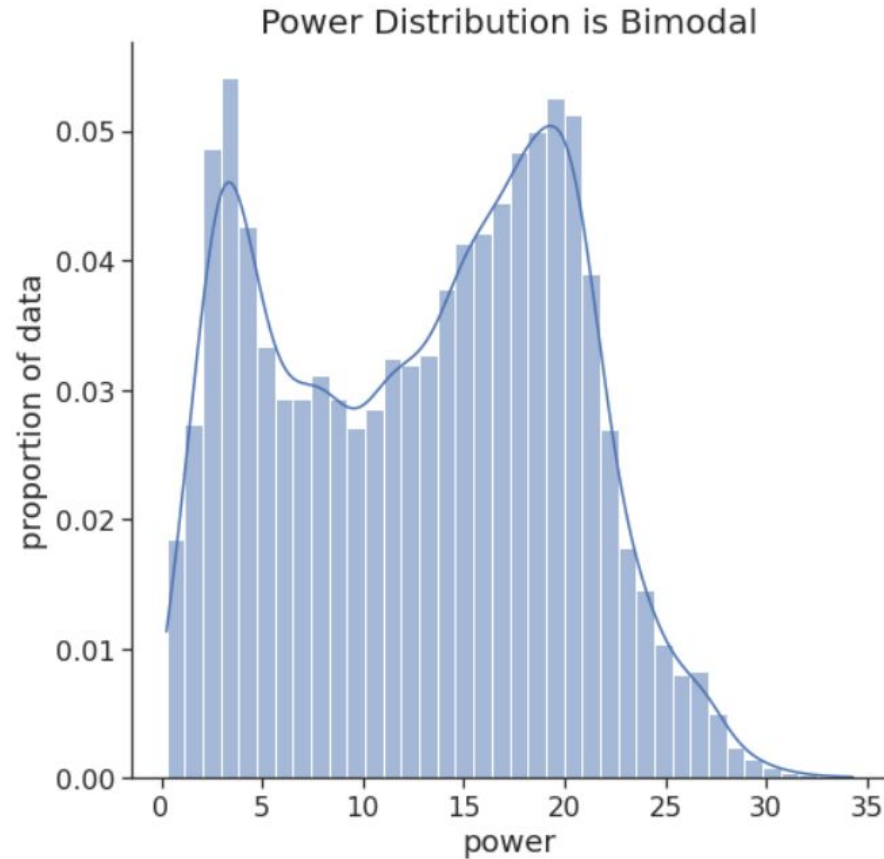
Williams, Jada; Wagner, Torrey (2019), "Northern Hemisphere Horizontal Photovoltaic Power Output Data for 12 Sites", Mendeley Data, V5, doi: 10.17632/hfhwmn8w24.5

Initial Exploration of Variables

Location of Solar Plants

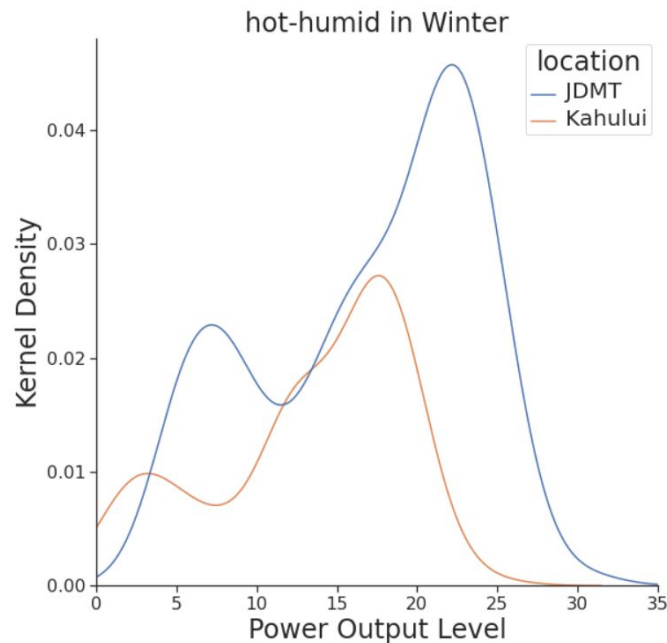
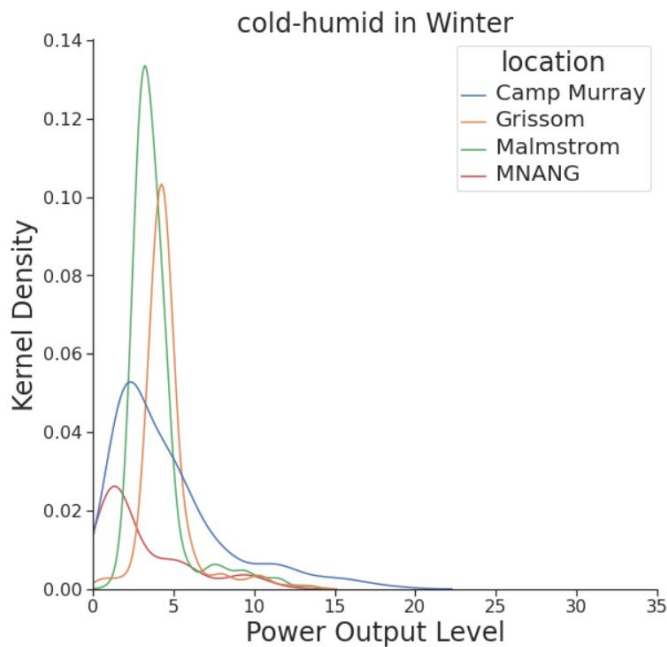


Power

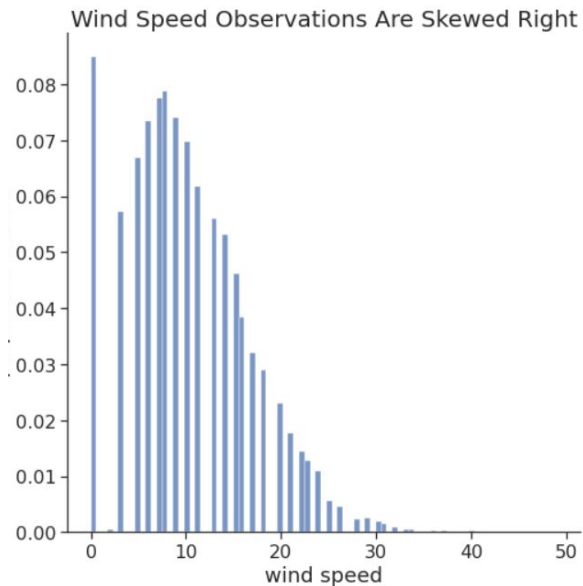
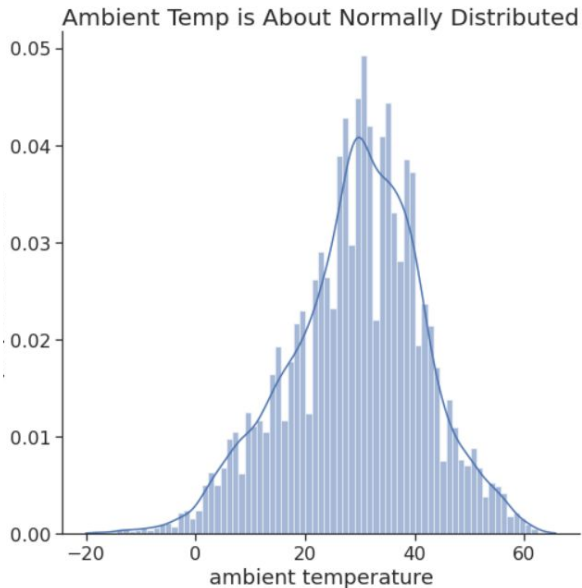
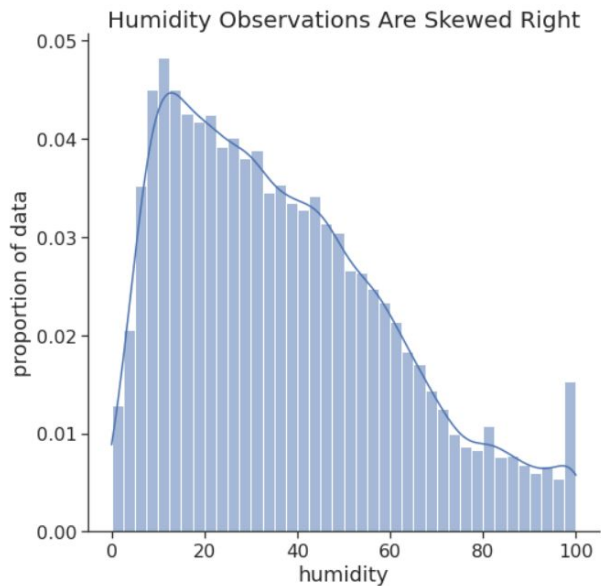


Why Classify by Climate?

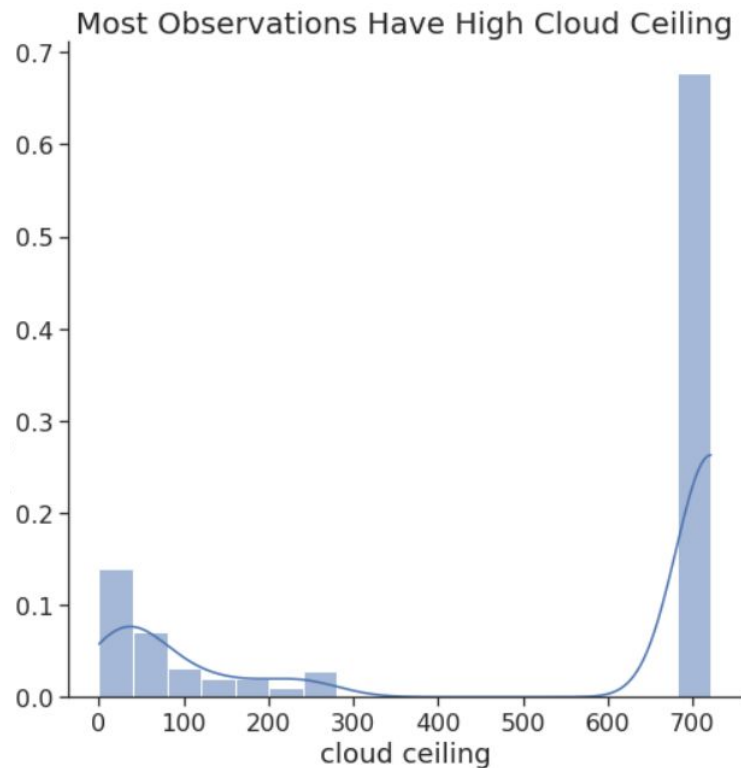
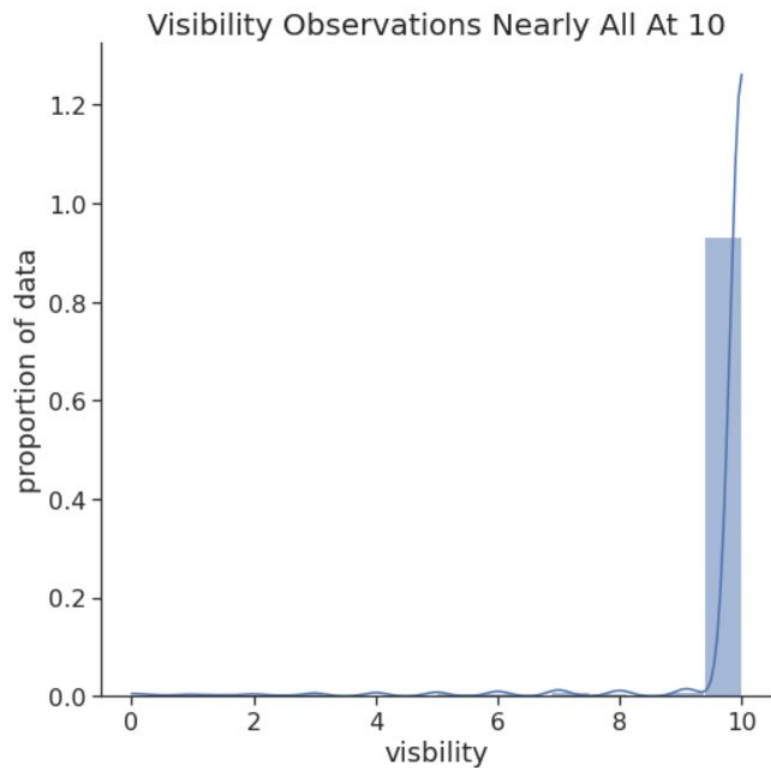
Locations in the Same Climate Have Similar Power Output



Humidity, Temperature, Wind Speed

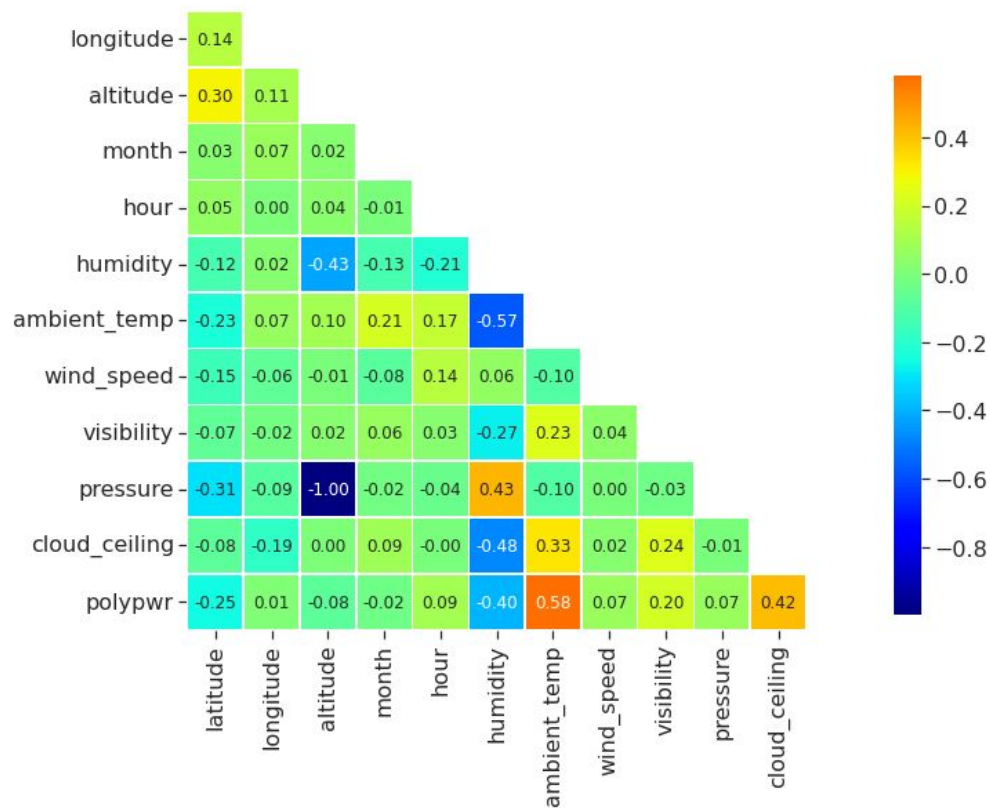


Visibility and Cloud Ceiling

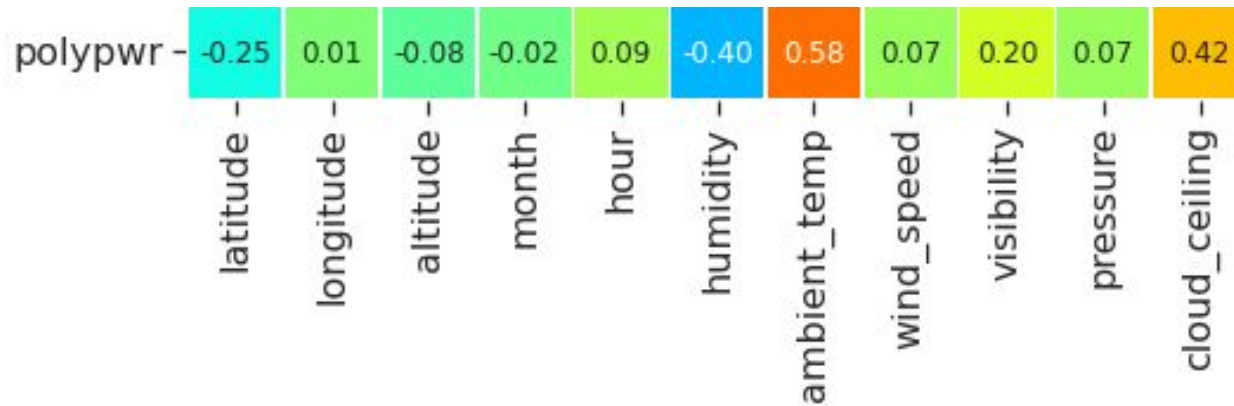


Correlation to Power

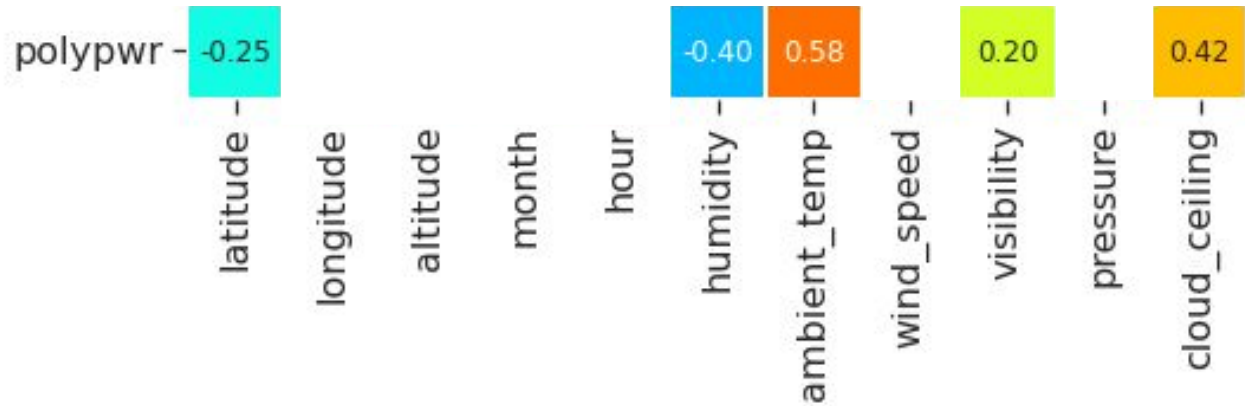
Correlation for All Sites



Correlation Between Variables



Correlation Between Variables

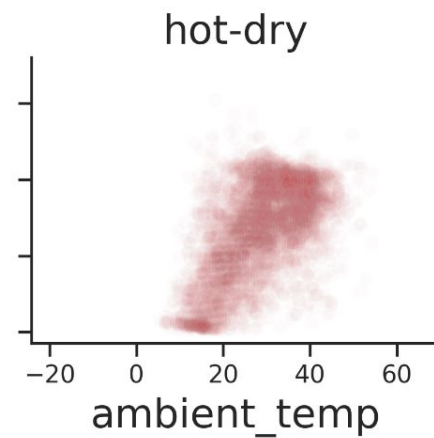
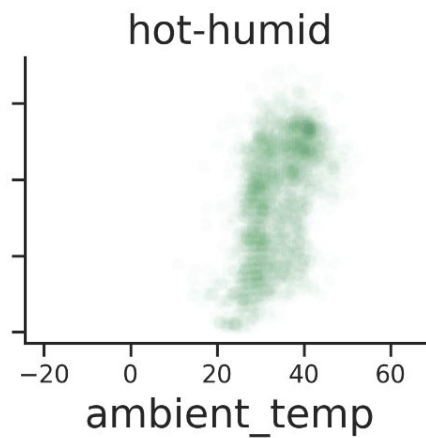
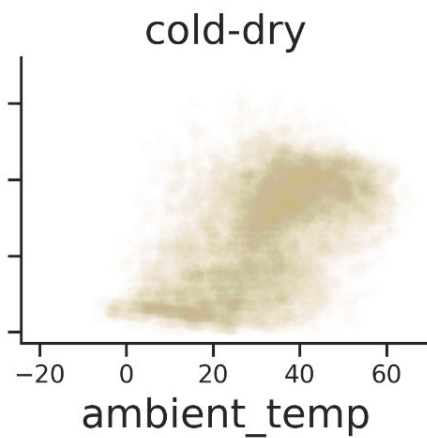
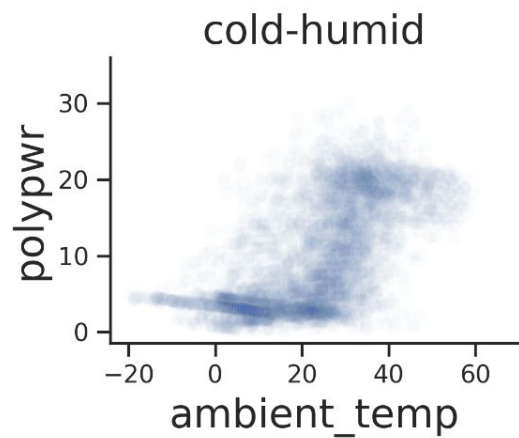
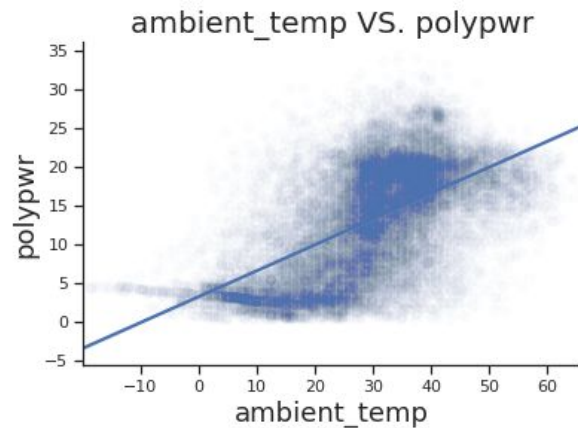


Extracting Variables that Explain Power

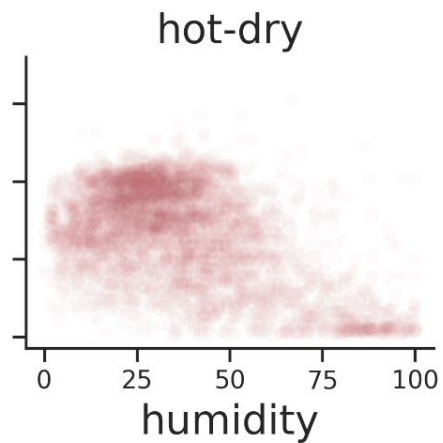
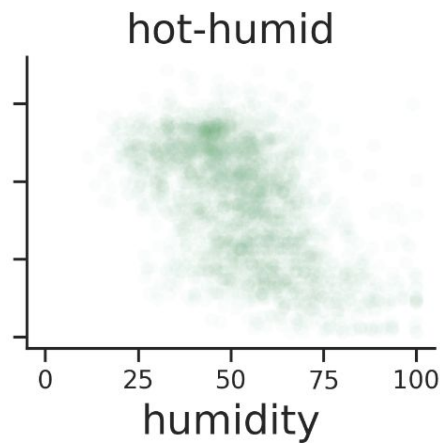
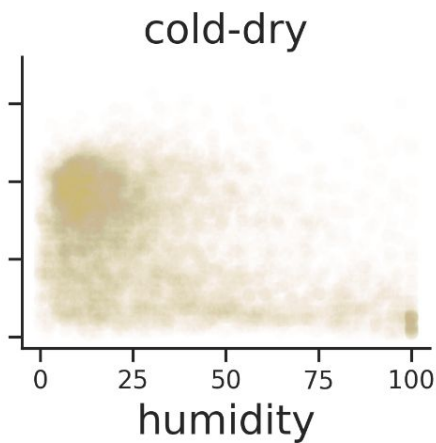
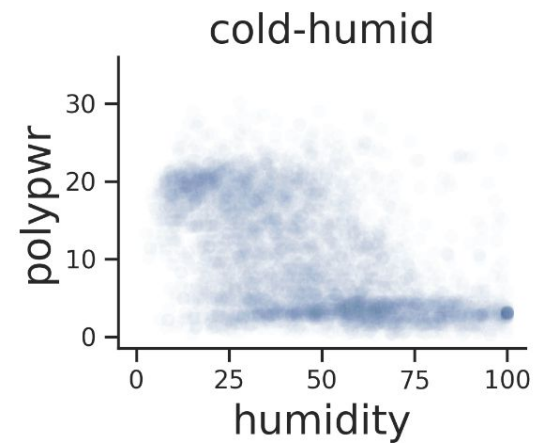
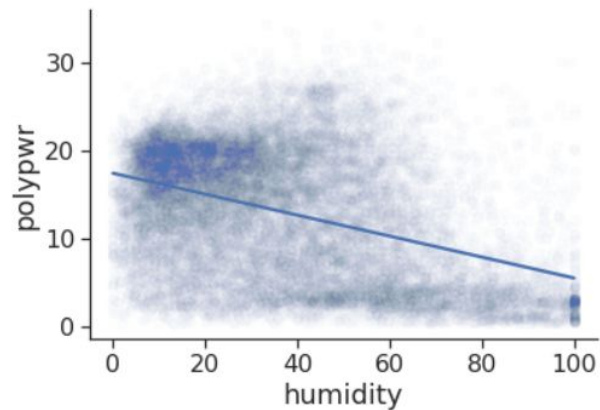
Variables that Explain Power

- Climate Type
- Temperature
- Humidity
- Cloud Ceiling
- Visibility

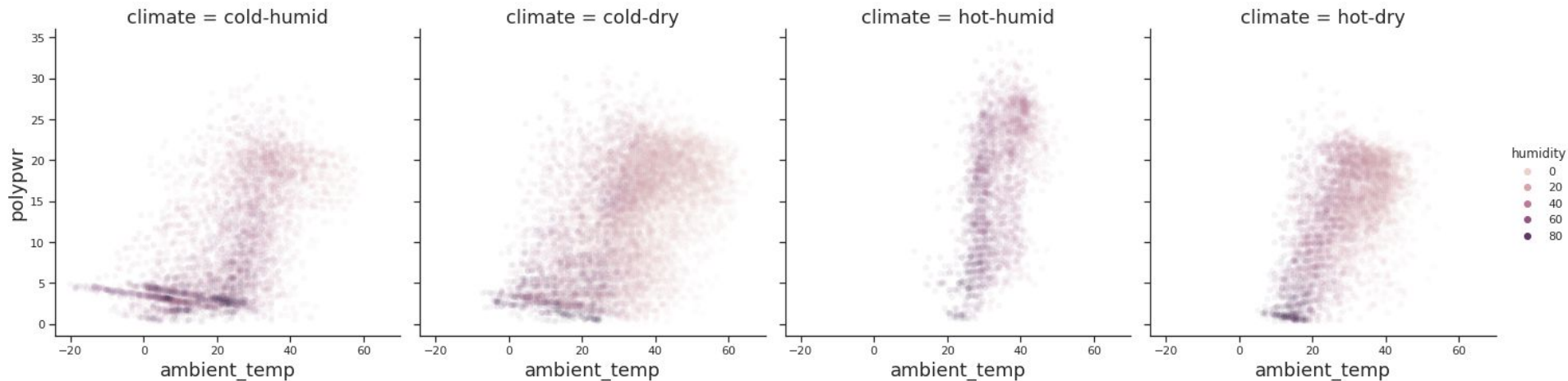
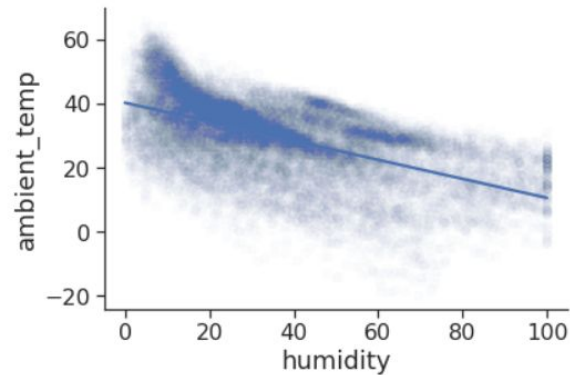
Temperature VS. Power



Humidity VS. Power

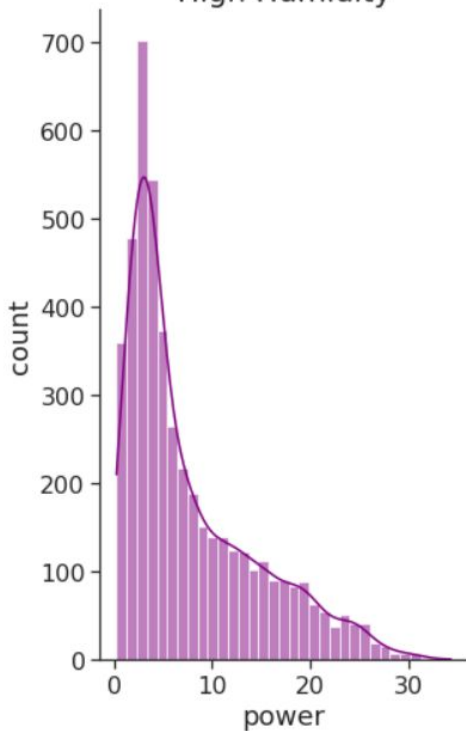


Why Consider both Temperature and Humidity?

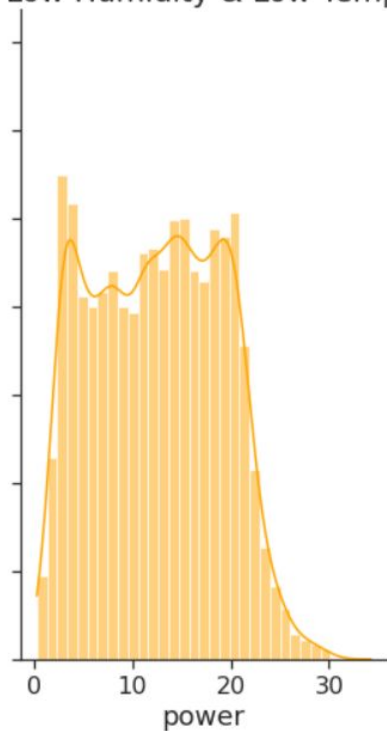


Why is Power Bimodal?

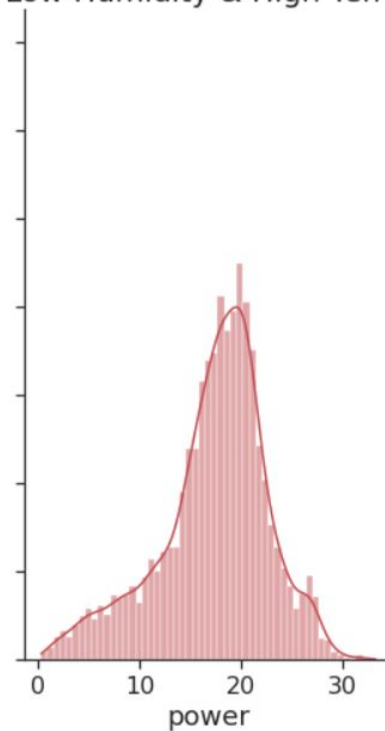
High Humidity



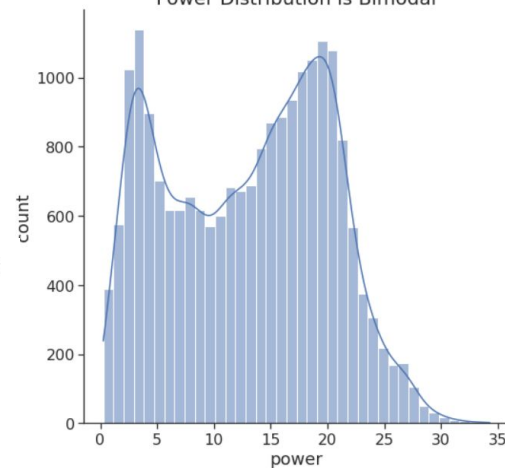
Low Humidity & Low Temp



Low Humidity & High Temp



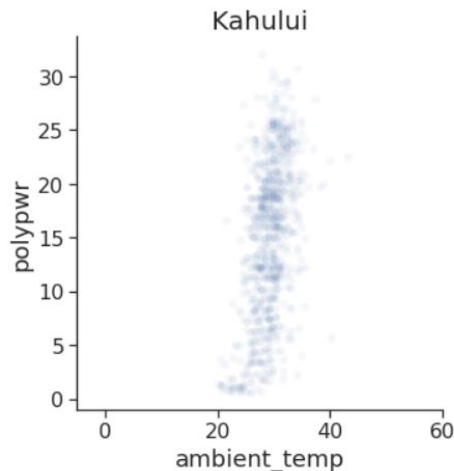
Power Distribution is Bimodal



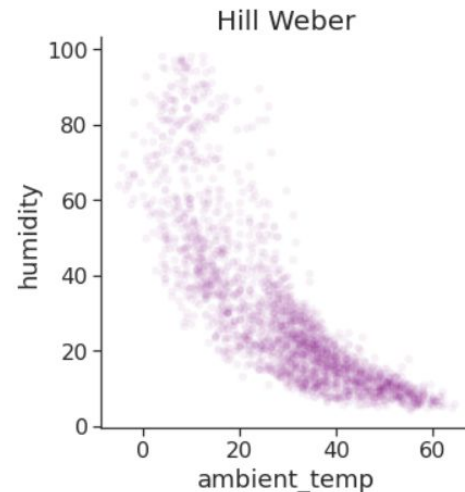
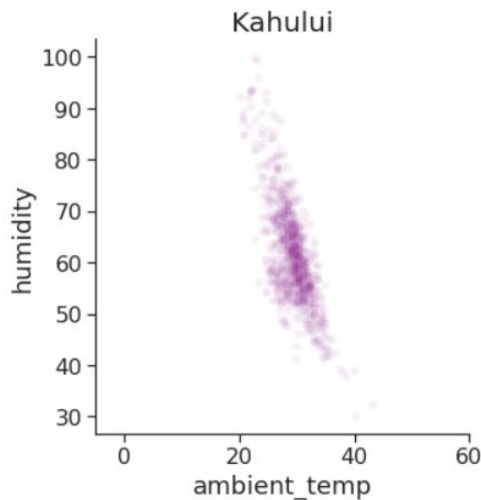
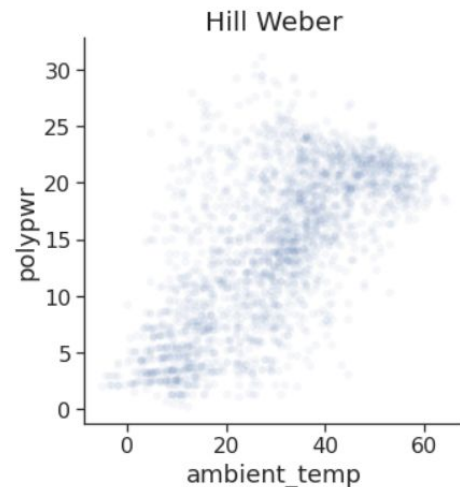
Case Study

Kahului 's
temperature remains
rather **constant**.

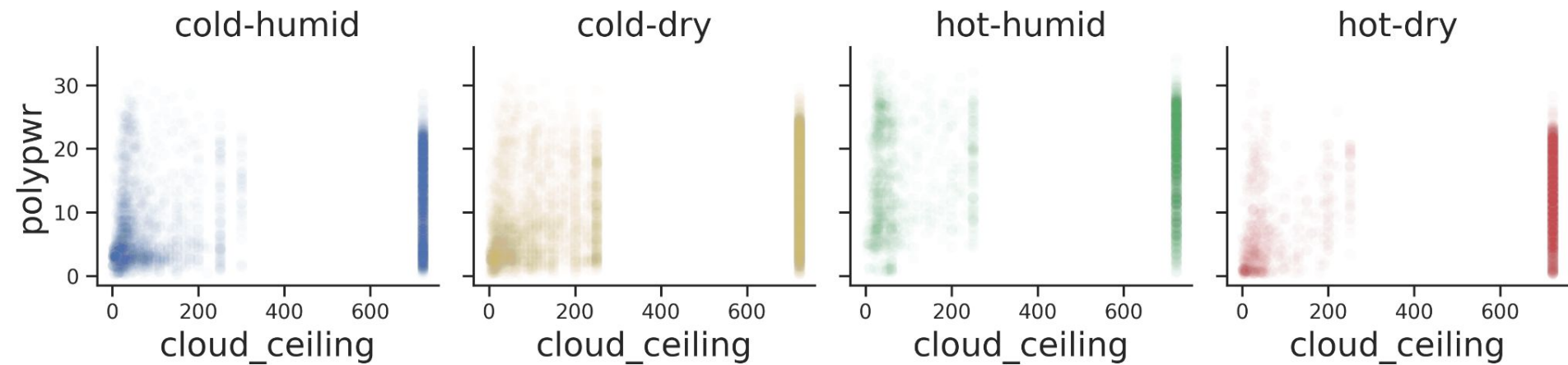
We can conclude that
for Kahului,
temperate alone is
not a good indicator
of **power**.



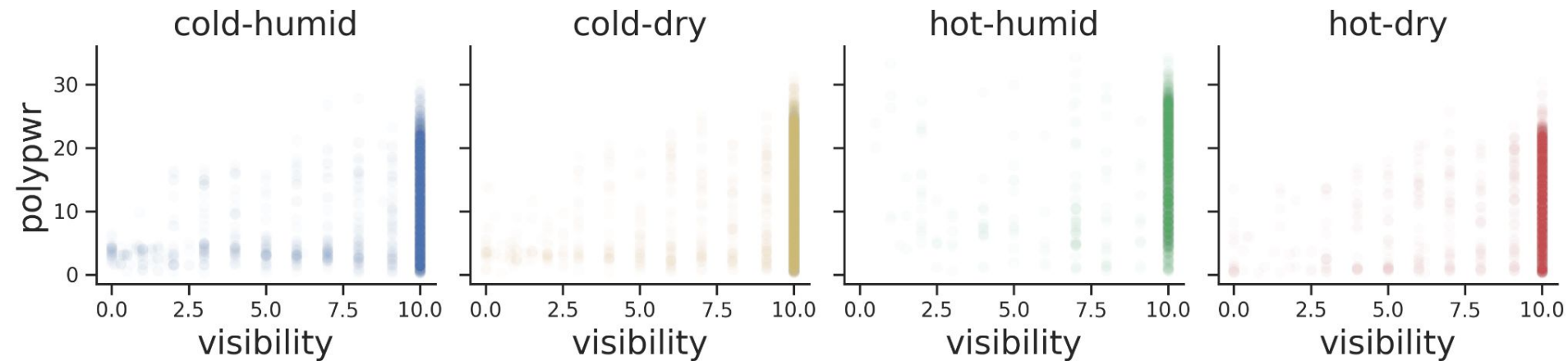
vs.



Cloud Ceiling VS. Power



Visibility VS. Power



Power Prediction

Model Performance for different climates

<i>Optimal Model</i>	<i>Climate Type</i>	<i>Training Set Size</i>	<i>RMSE</i>	<i>MAE</i>	<i>R²</i>
GLM	hot-dry	4455	3.024	2.223	0.745
GLM	cold-dry	7630	4.316	3.211	0.599
HistGradBoosting Regression	hot-humid	2448	5.683	4.529	0.442
GradientBoosting Regression	cold-humid	4407	3.782	2.548	0.700

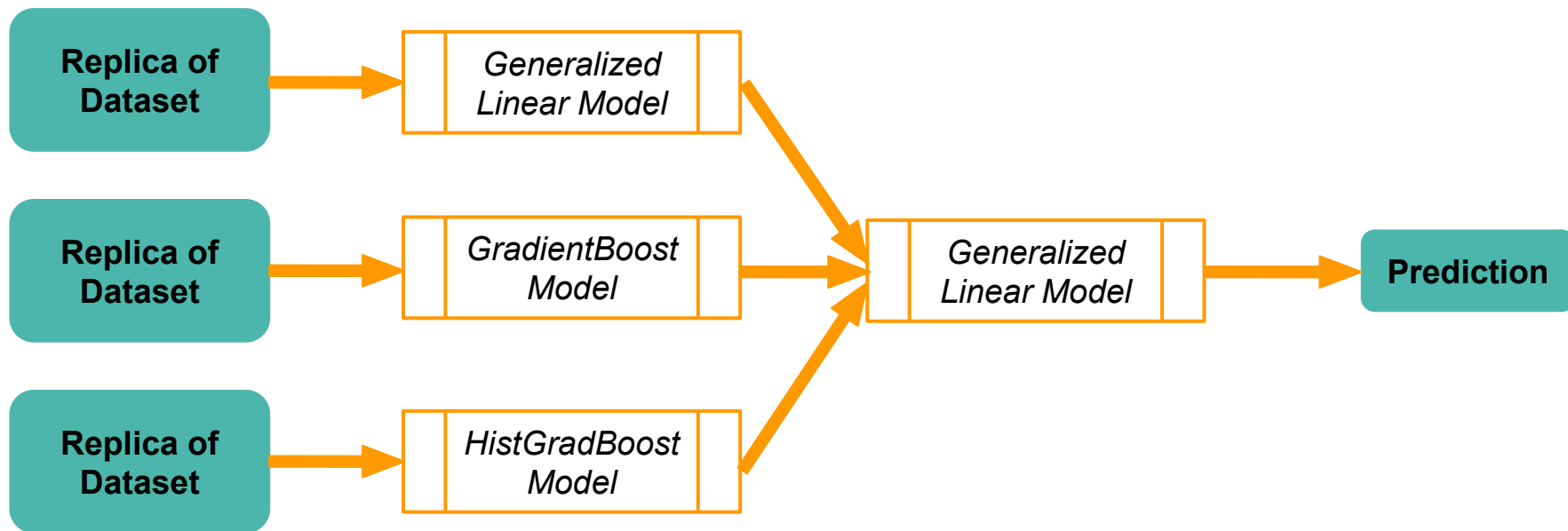
In total we have four climate types, we train the model for each climate type. The optimal model for each climate is listed above.

* GLM: Generalized Linear Model

HistGradBoosting: Histogram Gradient Boosting Regression Model

“Optimal model” means achieving lowest average of RMSE and MAE

Stacking Model



Trade-offs of Stacking Model

<i>Model</i>	<i>Climate</i>	<i>Training Set Size</i>	<i>RMSE</i>	<i>MAE</i>	<i>R²</i>	<i>Training Time (in seconds)</i>
GLM	hot-dry	4455	3.024	2.223	0.745	0.4363
StackingReg	hot-dry	4455	3.029	2.188	0.743	8.8630
GLM	cold-dry	7630	4.316	3.211	0.599	0.5601
StackingReg	cold-dry	7630	4.249	3.142	0.611	15.6120
GradientBoosting	cold-humid	4407	3.782	2.548	0.700	1.3830
StackingReg	cold-humid	4407	3.768	2.521	0.711	9.3810
HistGradBoosting	hot-humid	2448	5.683	4.529	0.442	0.3631
StackingReg	hot-humid	2448	5.591	4.460	0.460535	6.2752

Conclusion

- In the absence of irradiance data, **environmental data** can be used to make **accurate predictions for solar power output**.
- **Which variables** best explain solar power output?
 - Climate type
 - Temperature
 - Humidity
 - Cloud ceiling
- Why is **power bimodal**?
 - High humidity → low power
 - Low humidity + high temperature → high power

Thank you!