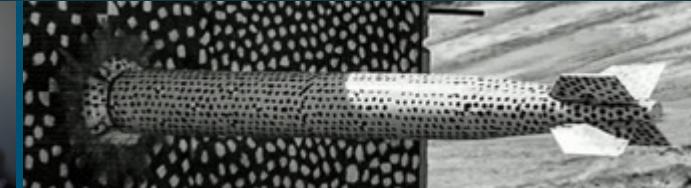
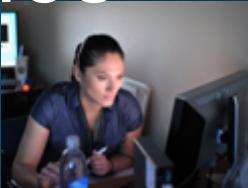




Evaluation of Extreme Weather Impacts on Utility-scale Photovoltaic Plant Performance



SAND2020-11758PE



PRESENTED BY

Nicole D. Jackson

DuraMat Early Career Scientists Seminar Series

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SANDXXXXXX



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PV plants can be exposed to disruptions due to weather events such as hurricanes and hail storms

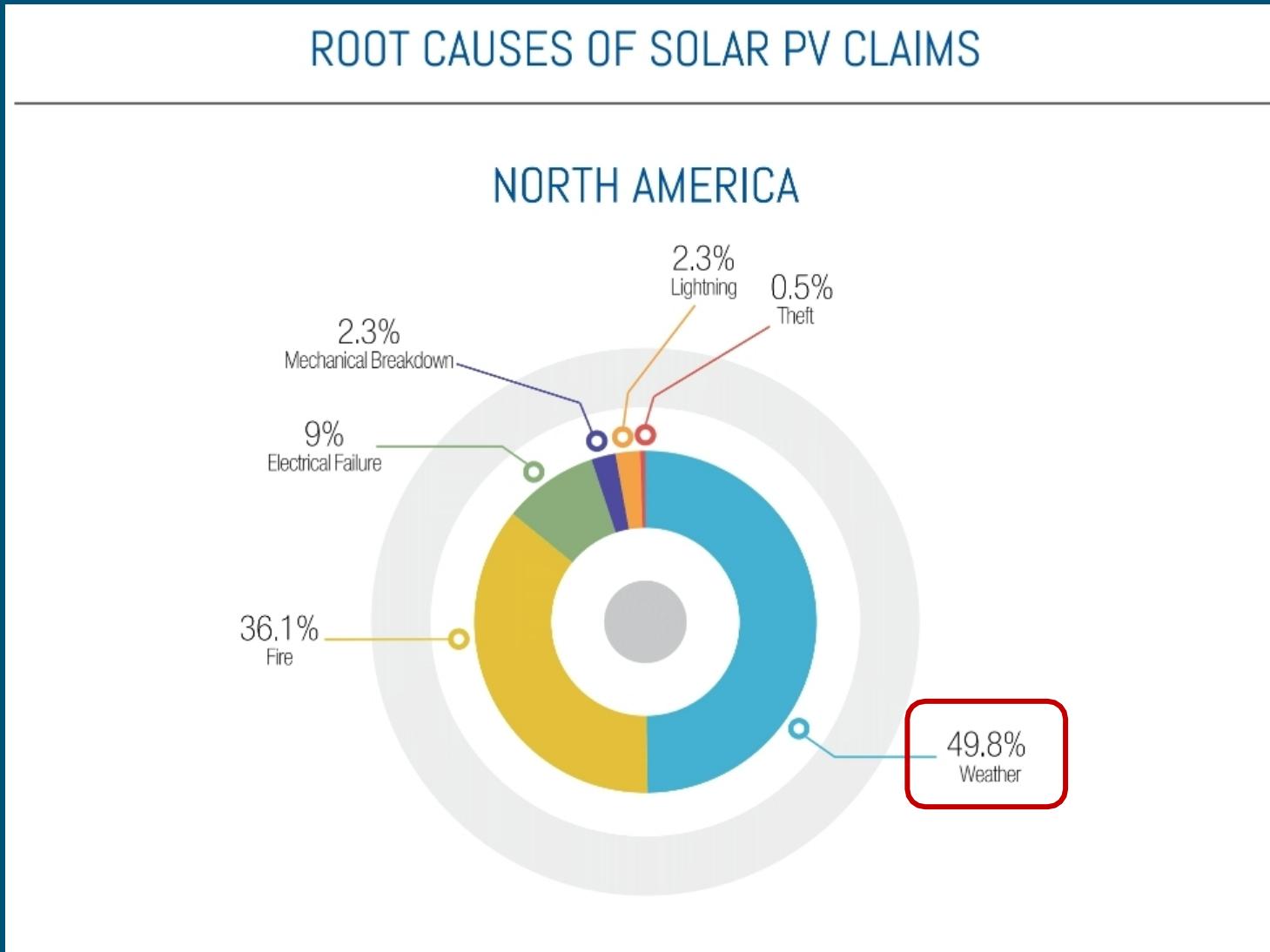


(BMR Energy 2017)



SBS News (2018)

There is an increasing prevalence of weather impacts to PV



(GCube 2016)



- Analysis of site-level performance and weather data to identify trends
- Identification of performance variabilities across sites, climates, and event types

Multiple datasets are combined to link performance, weather and O&M records



Industry Data

Site Characteristics

- DC Size
- Climate zone
- Latitude
- Longitude
- Asset clipping limit

Measured data

- Date
- Irradiance
- Energy delivered
- Output power
- Expected energy

O&M Logs

- Failure details
- Event duration

Site Metrics

Yield Loss

- Final system yield
- Reference yield

Performance metrics

- Performance ratio
- Energy performance index

Climate Data

GHCN Weather Stations

- Precipitation
- Snow

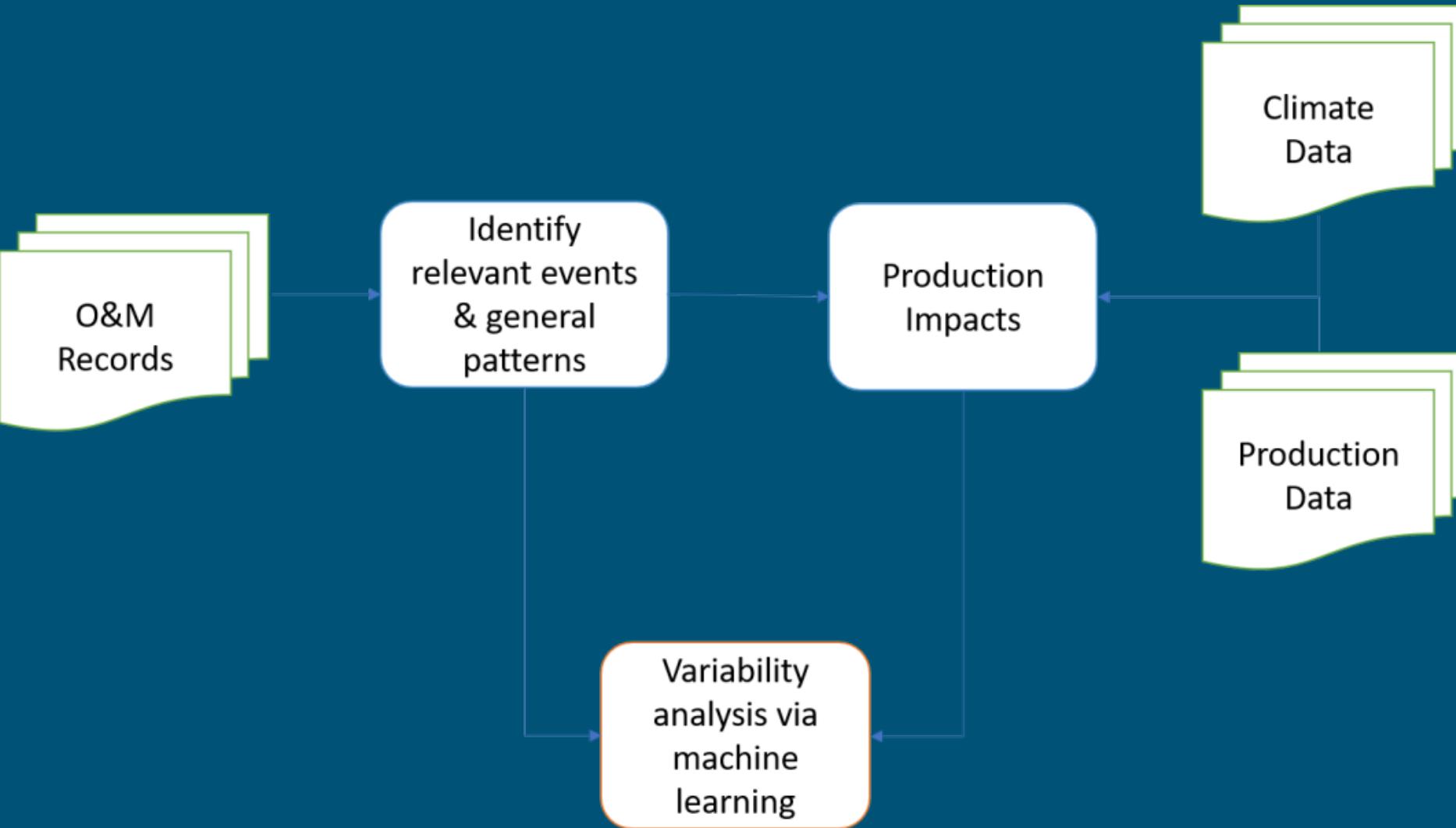
PRISM

- Maximum temperature
- Minimum temperature
- Precipitation

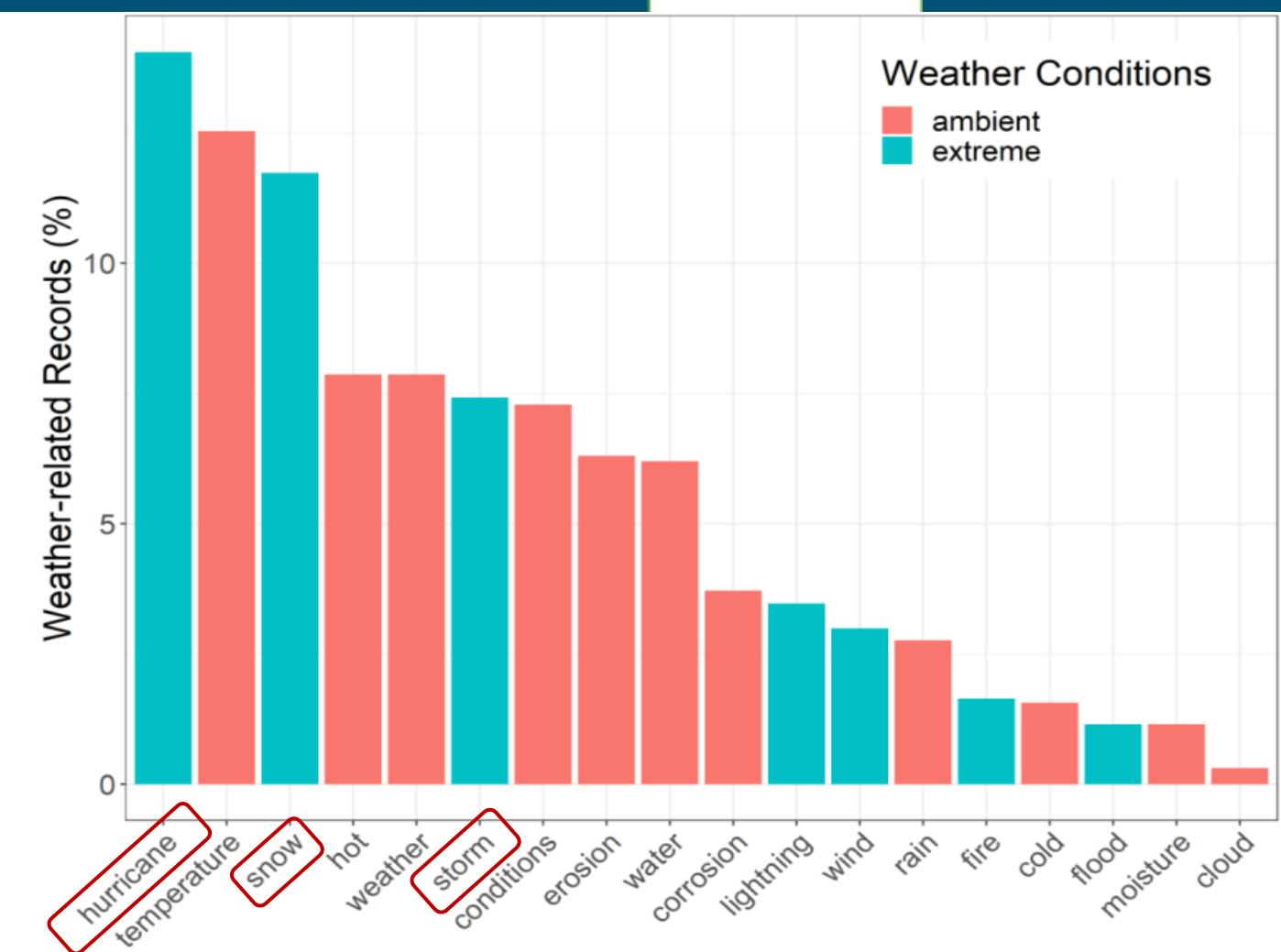
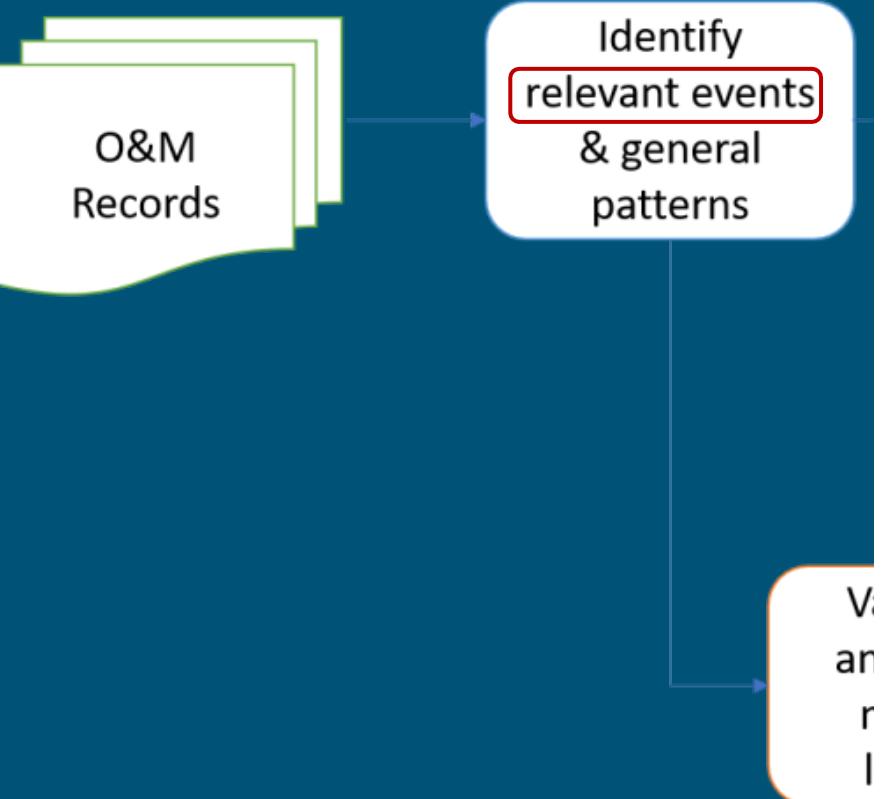
NOAA Storm Events Database

- Snow
- Storms
- Hurricanes

Integrated assessment of uses inputs from multiple datasets



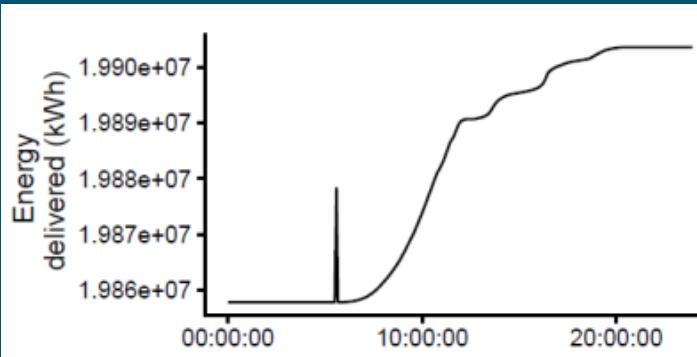
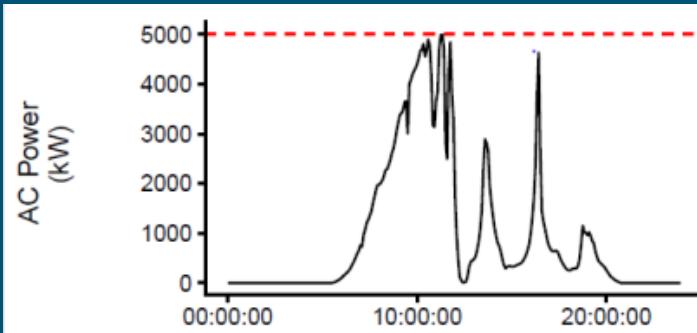
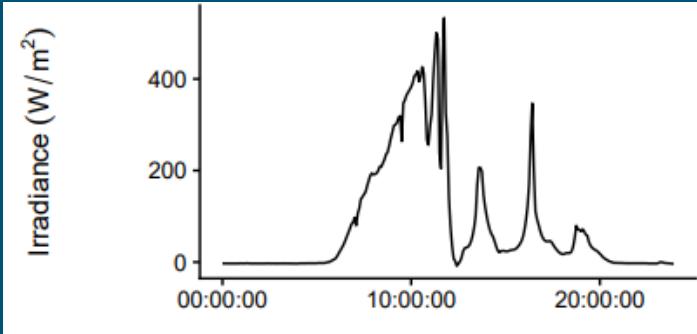
Text analysis of PVROM's O&M records helps identify the most prevalent weather events



Data processing of site-level raw production data



Raw Data



Removed negative values

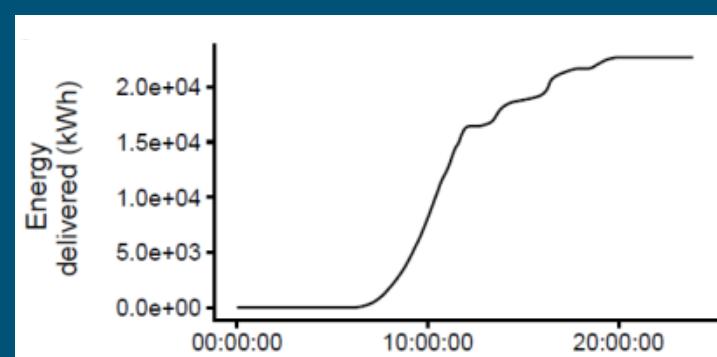
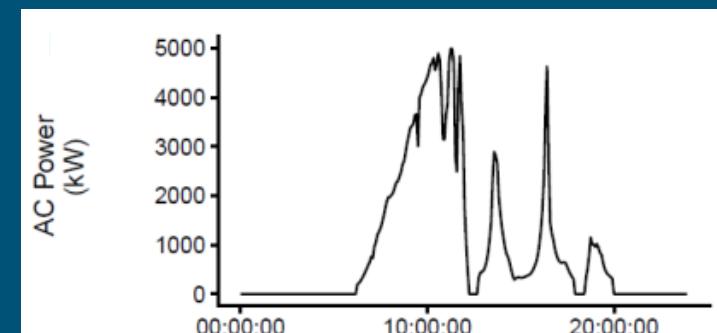
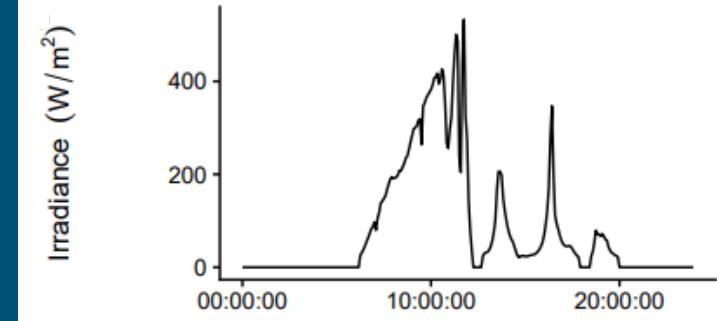


Entries with negative
irradiance have AC power
= 0



Removed errant spikes in
cumulative energy
delivered to grid

Clean Data



Multiple site metrics are evaluated at the daily time scale



Yield losses

- Final system yield

$$Y_f = \frac{E_{out}}{P_0}$$

- Reference yield

$$Y_r = \frac{H_i}{G_{i,ref}}$$

Performance metrics

- Performance ratio

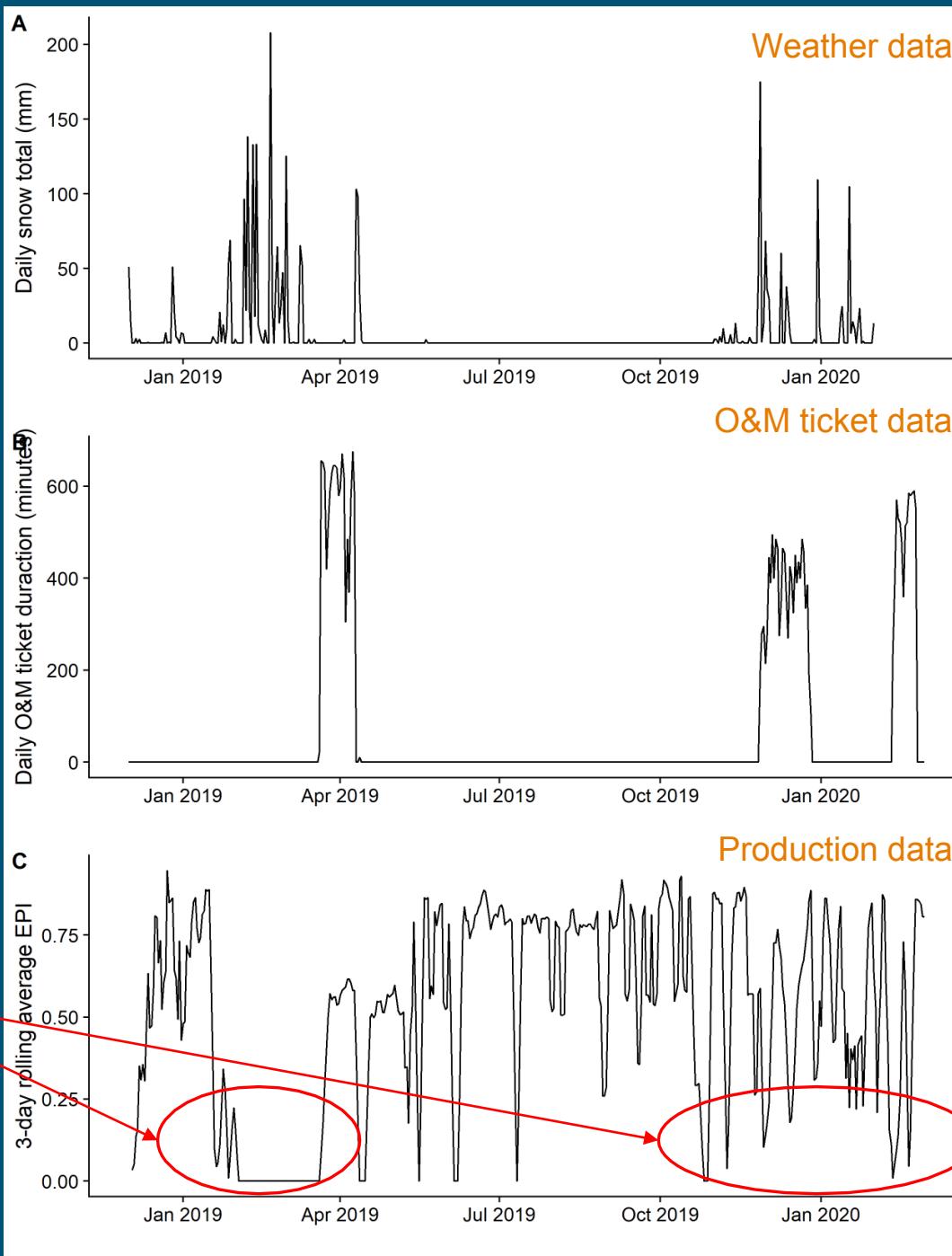
$$PR = \frac{Y_f}{Y_r}$$

- Energy performance index

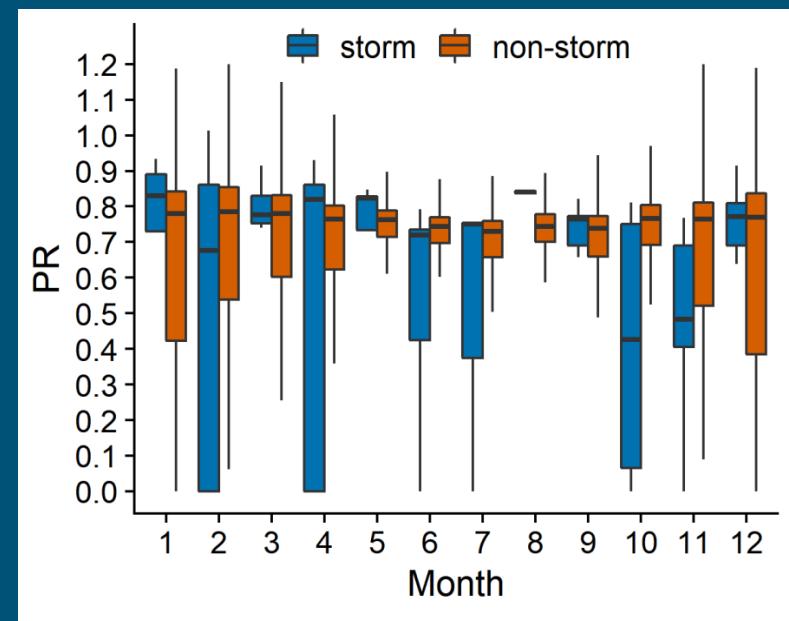
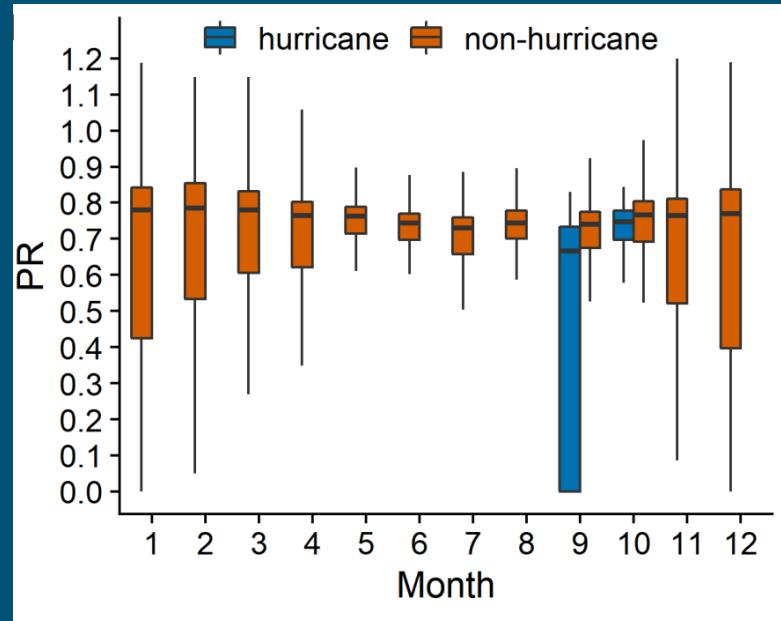
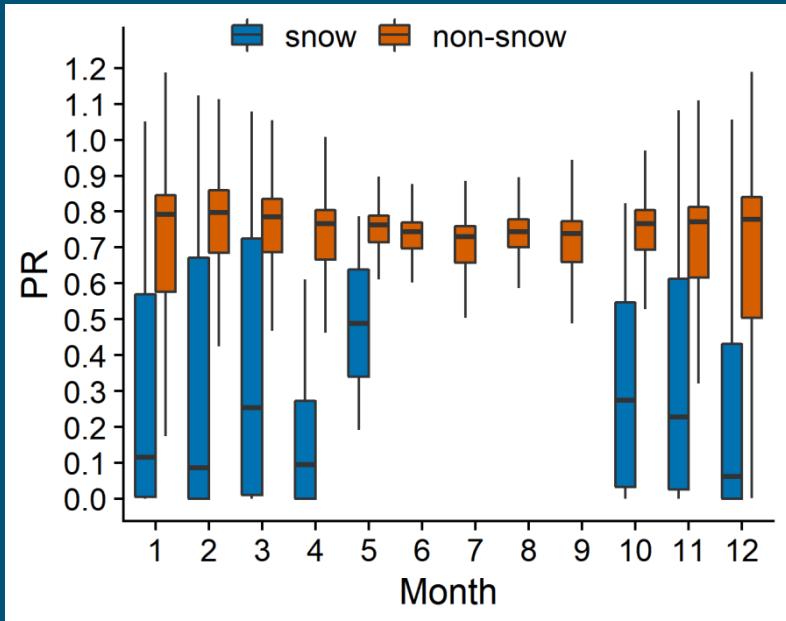
$$EPI = \frac{\text{Measured energy}}{\text{Expected energy}}$$

Data fusion of snow, production, and O&M data show nearness of phenomena

“low”
performance



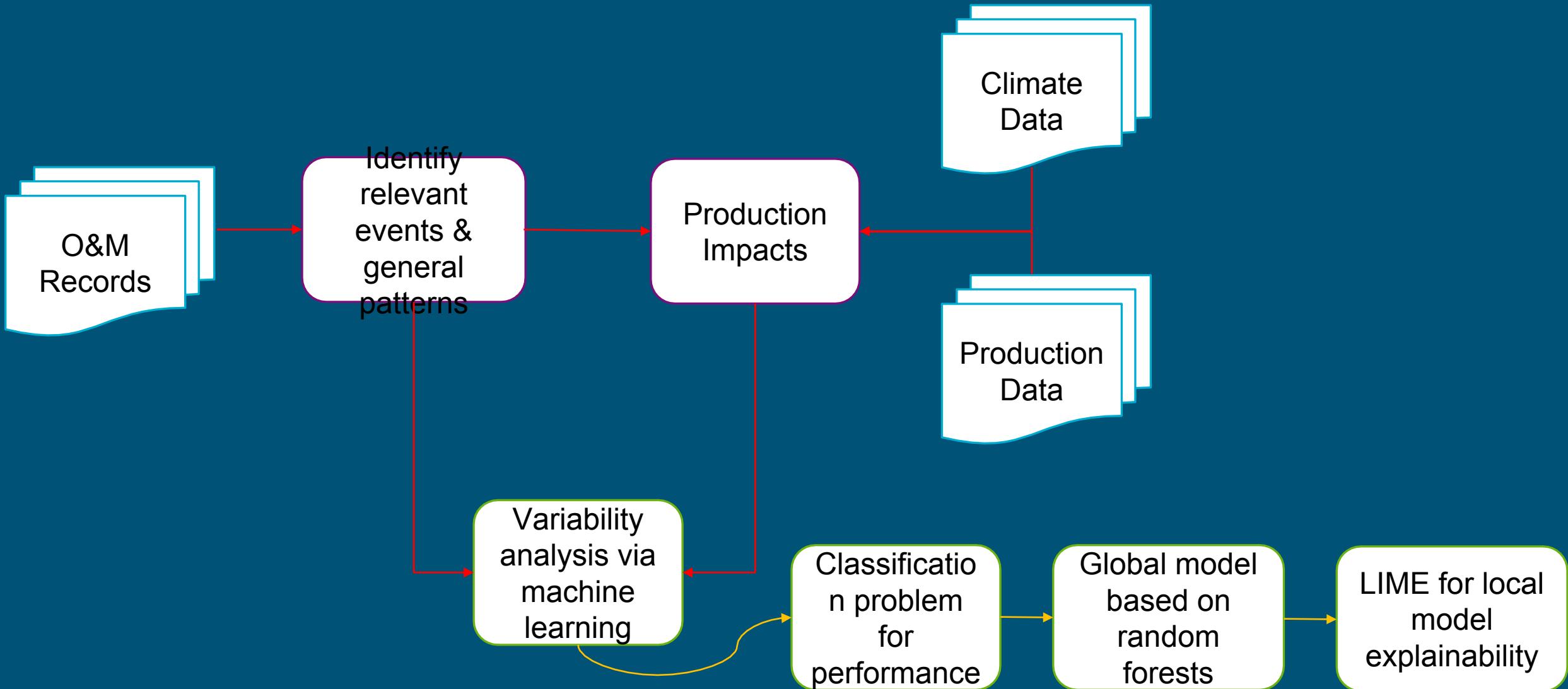
Event and non-event days show variable response across months and weather types



Mean PR by event status

	Snow	Hurricane	Storm
Event	0.296	0.594	0.627
Non-event	0.650	0.669	0.634

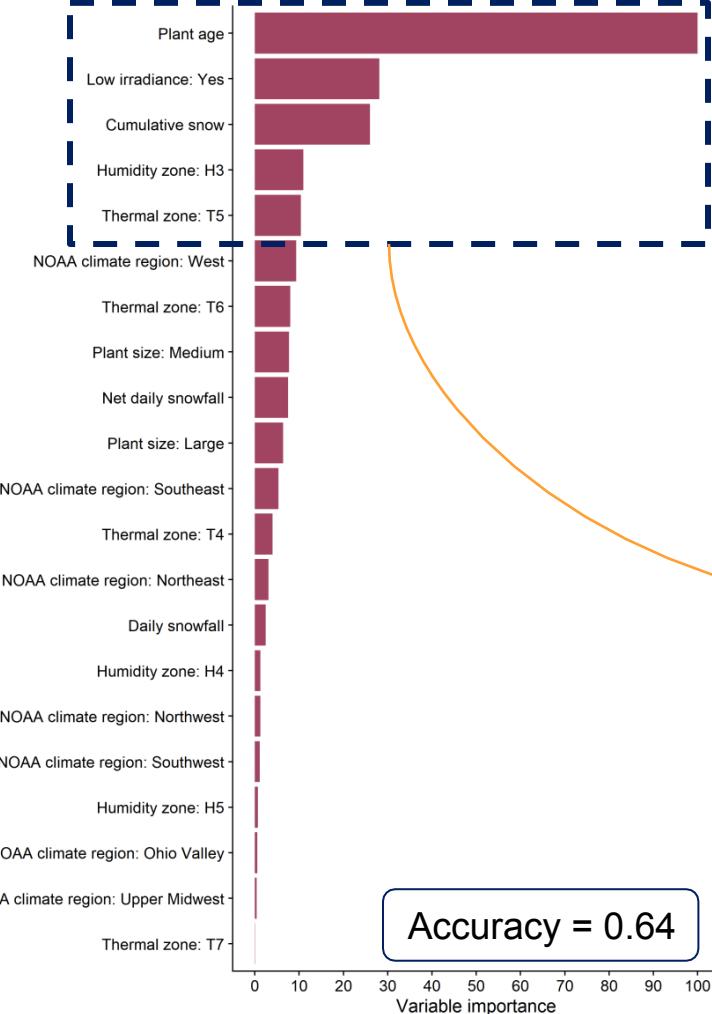
Machine learning used to gain insights into drivers of performance during weather events



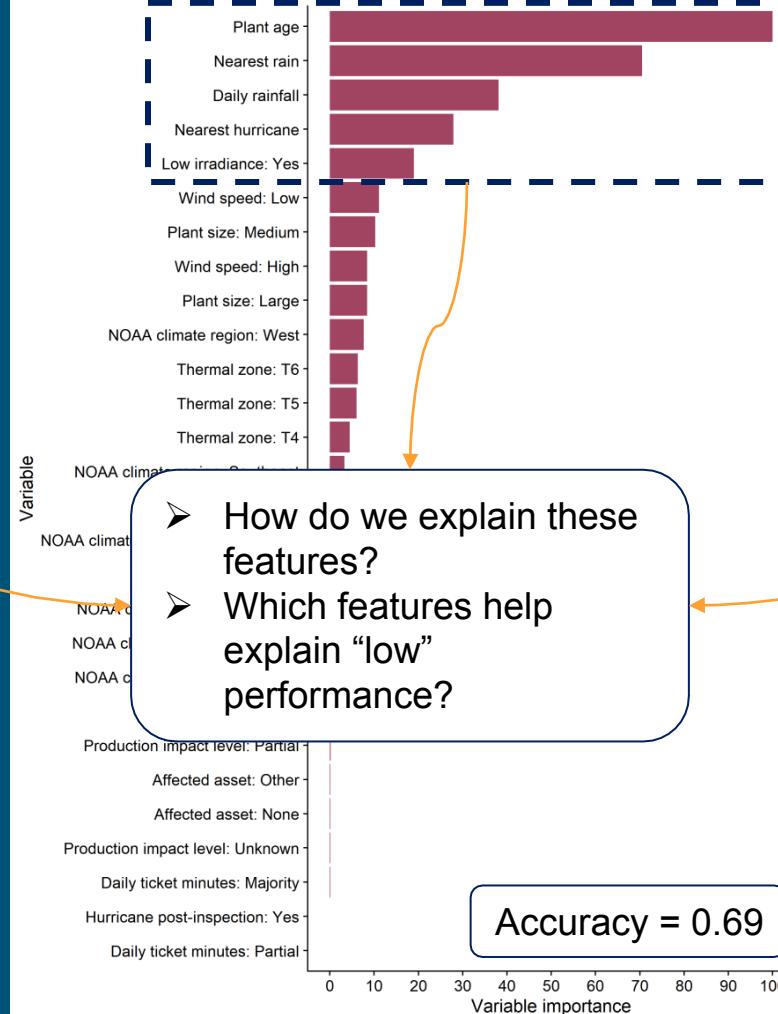
Random forest machine learning implementation gives insight to feature importance and accuracy for global models



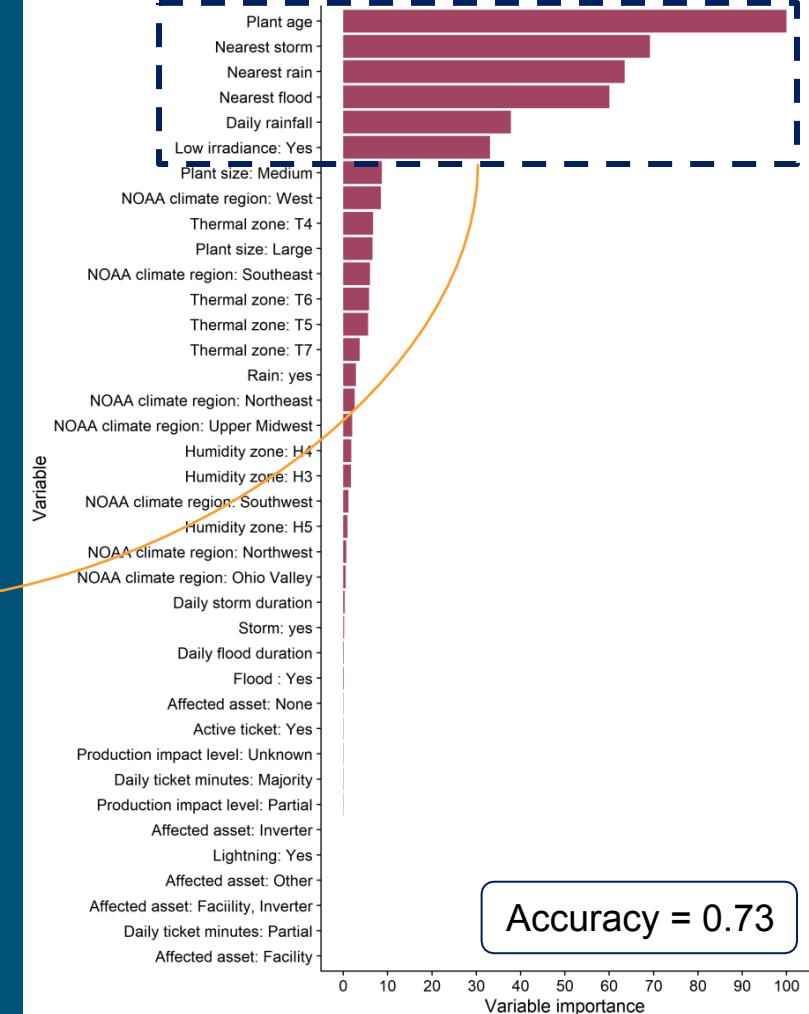
Snow



Hurricane



Storm



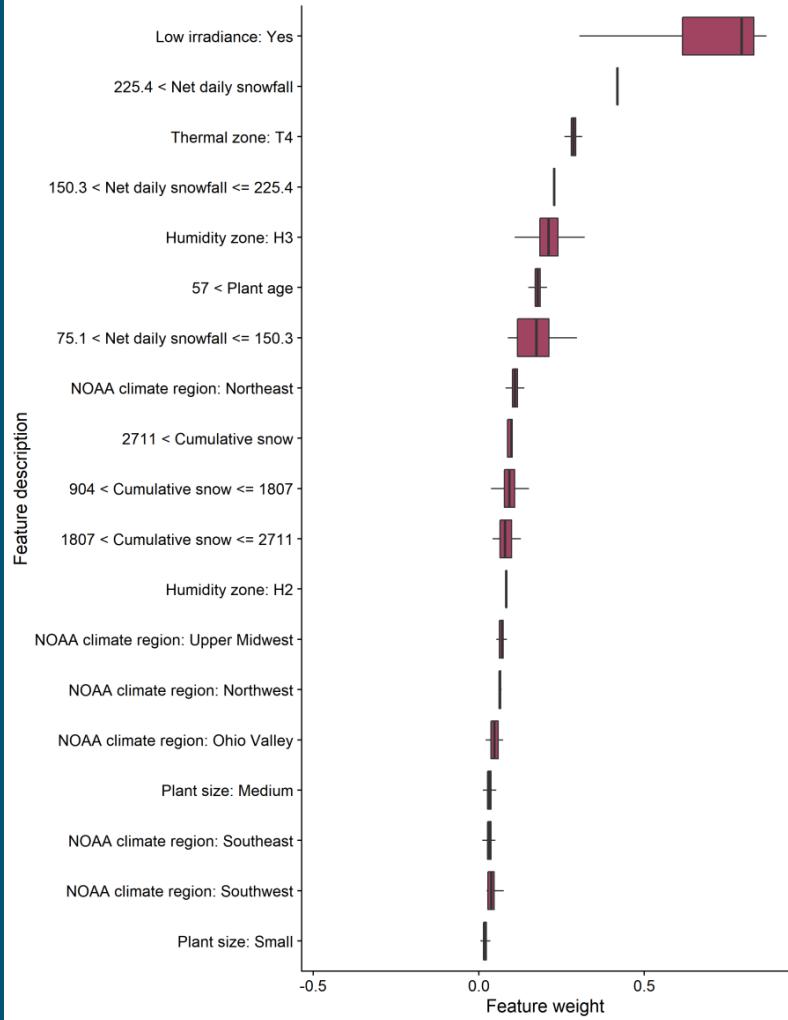
- How do we explain these features?
- Which features help explain “low” performance?

Local Interpretable Model-Agnostic Explanations (LIME)

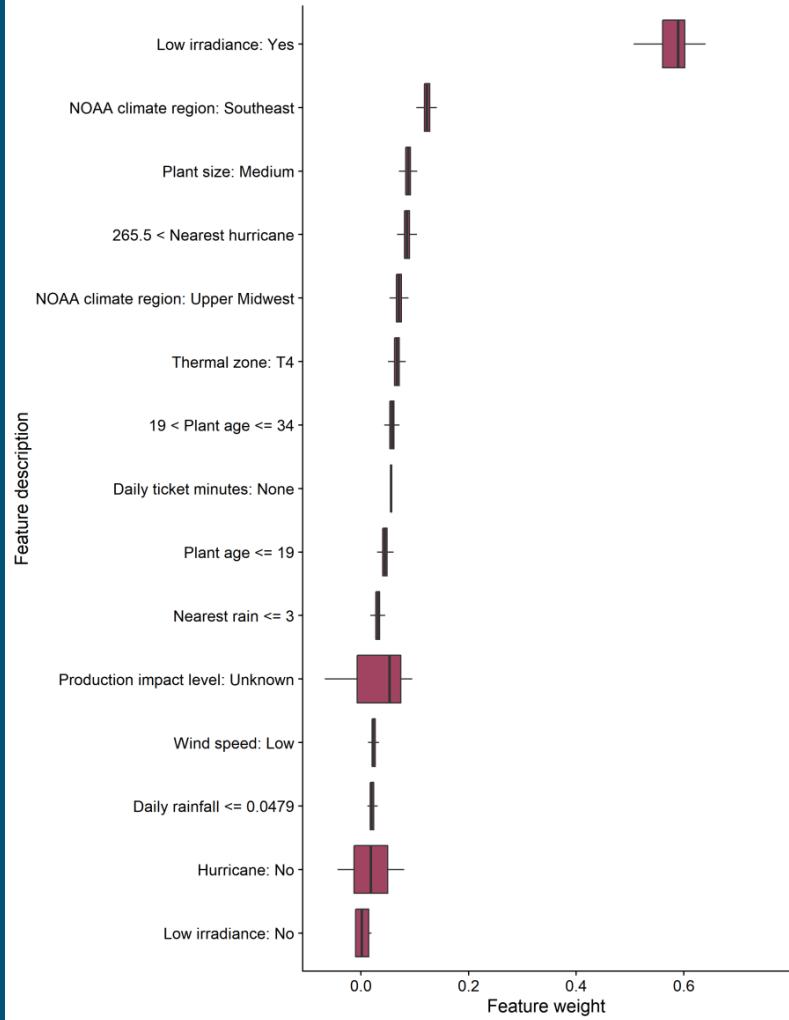
used to identify drivers of low performance



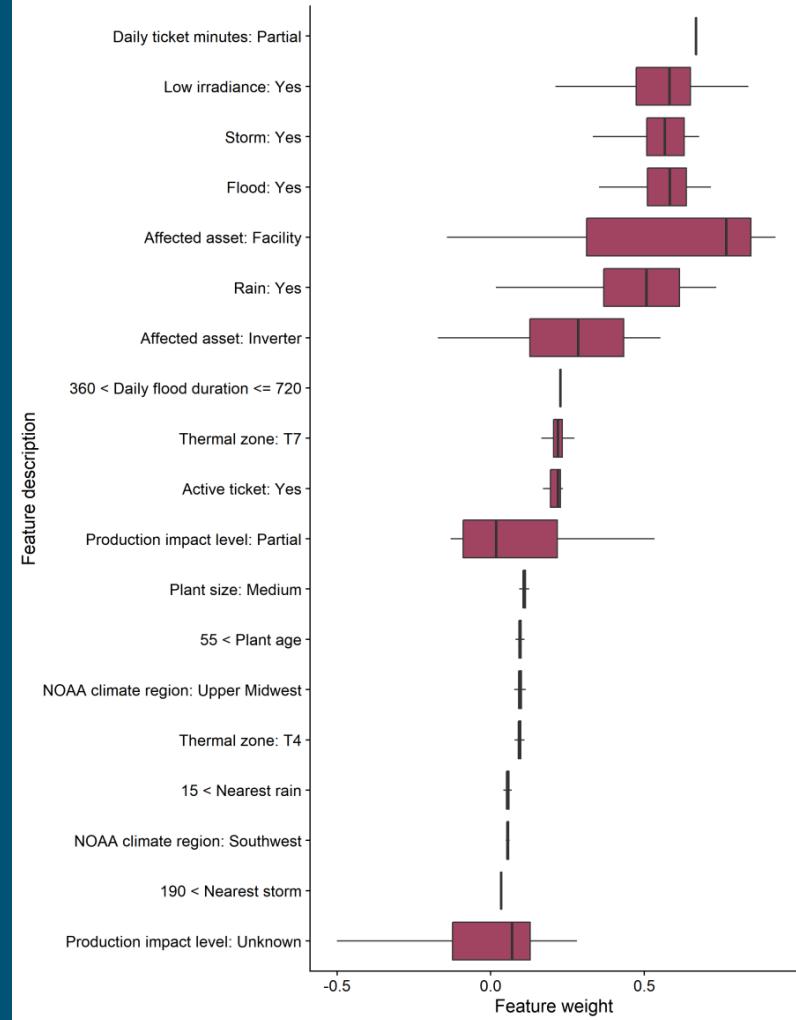
Snow



Hurricane



Storm

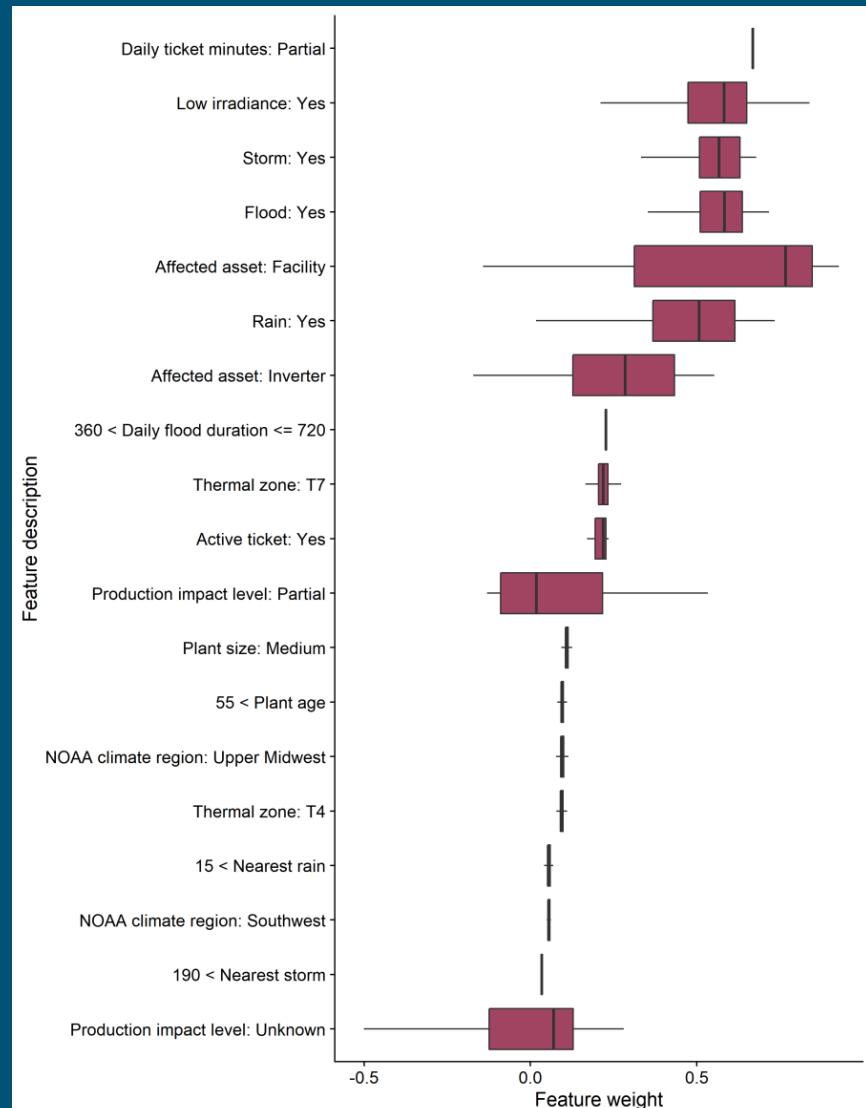


In conclusion, integrated assessment of O&M, production, and climate data shows differentiated responses to weather events

- Snow events have largest performance reduction
- Unique global models developed for each event
- LIME results for storms include O&M-related features

- Future work directions
- Comparison of “clear sky” to weather impacted days
- Incorporation of temperature effects
- Predictive modeling development for “day ahead”

scenario development



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Thank you for your time!

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