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Using Probabilistic Solar Power Forecasts

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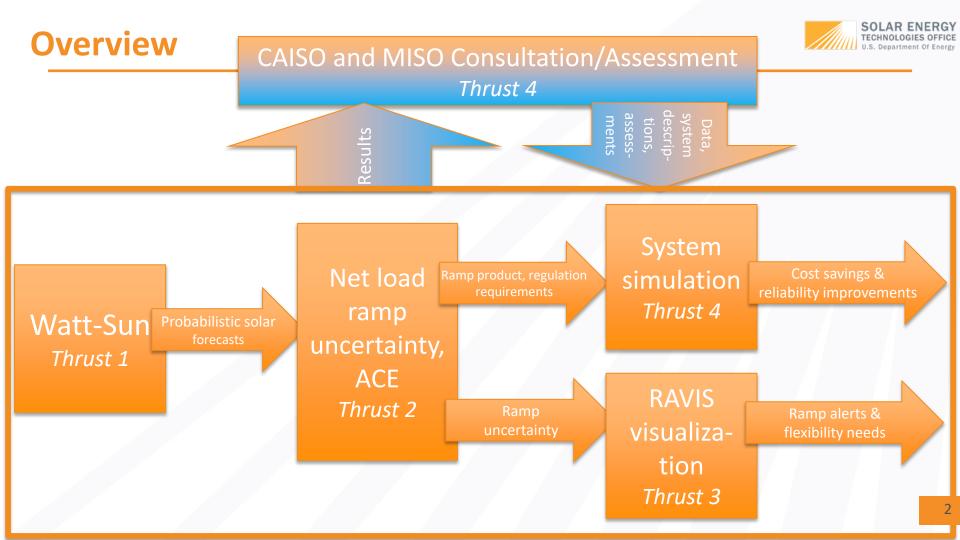
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SETO Workshop on Solar Forecasting May 5–6, 2021

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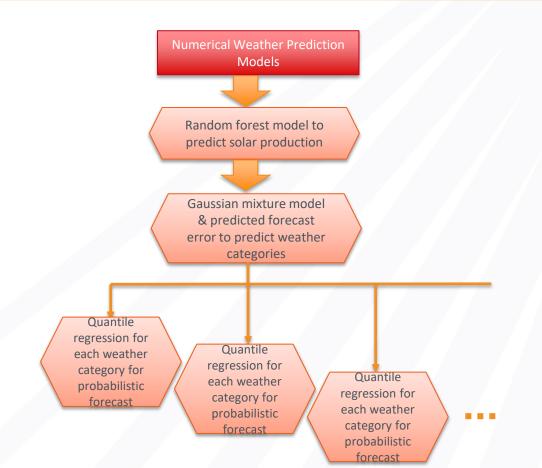




- 1. Comparison of Watt-Sun to base method: P-P index
- 2. Using solar forecasting prediction intervals to predict ramp and regulation: the Pareto method
- 3. Simulation of cost-reliability effects of solar-conditioned ramp requirements on the California Independent System Operator (CAISO) system
- 4. Visualization: Resource Forecast and Ramp Visualization for Situational Awareness (RAVIS).

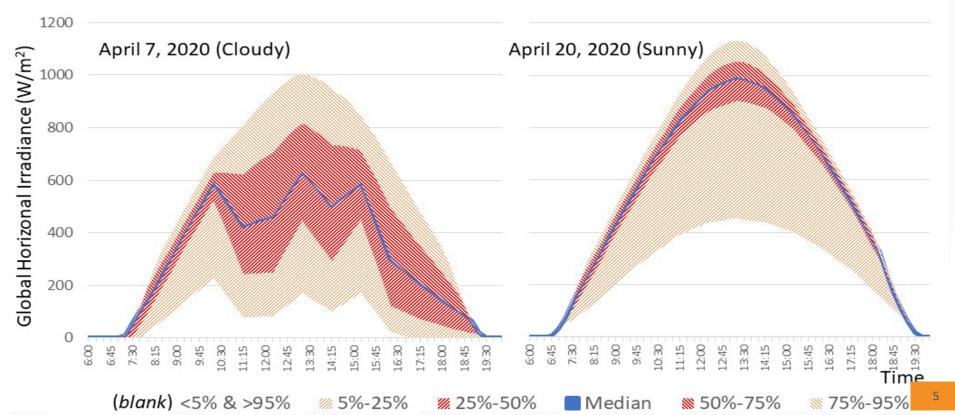
Probabilistic Watt-Sun Flowchart (IBM)





Used Quantile Regression to Deploy Probabilistic Forecast Models

- Quantiles of solar as a function of independent variables
- Example results for 2-hour-ahead forecasts





Evaluation of Watt-Sun



Assessment: Relative improvement of the PP-Plot metric

Temporal:

- Train: Sept. 1, 2018–Feb. 29, 2020
- Test: Mar. 1, 2020–June 1, 2020

Spatial:

24 stations

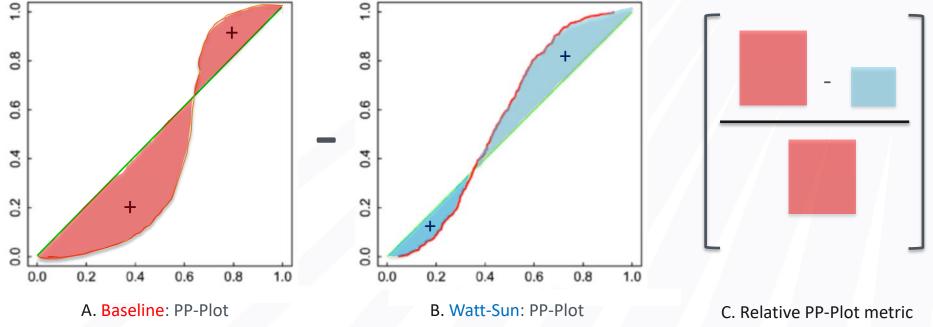


Map: 24 reference stations

Calibration Quality of Watt-Sun vs. Persistence



Assessment: Relative improvement of the PP-Plot metric

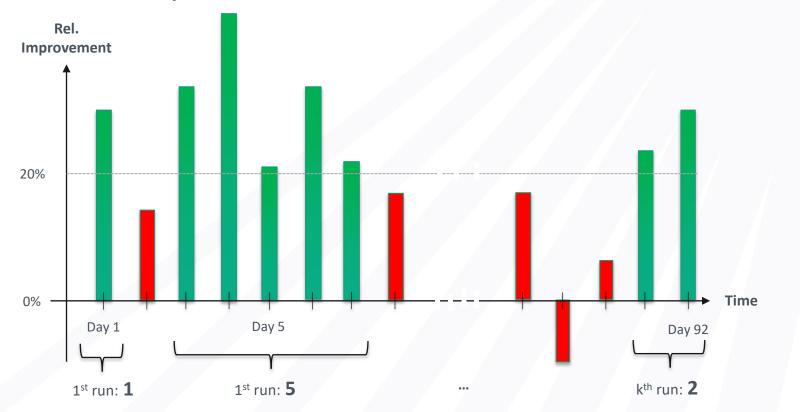


improvement of Watt-Sun

Daily Values of Metric

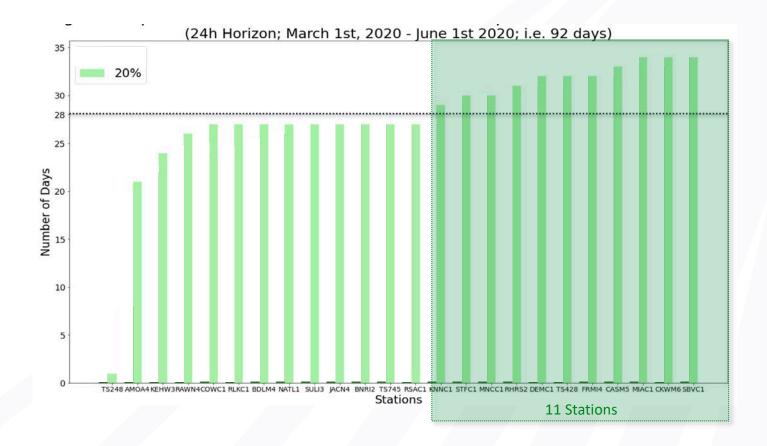


Assessment: Relative improvement of the PP-Plot metric



Longest Run per Station with Metric Improvement >20%





9





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Flexible Ramping Product: CAISO, MISO, SPP



- CAISO: Need to forecast three components in real-time:
 - Expected 5-minute ramp forecast
 - Uncertainty in up direction (97.5th percentile)
 - Uncertainty in down direction (2.5th percentile).
- Uncertainty distributions are presently unconditional.
 - CAISO is revising to condition requirements on wind, load, solar forecasts (<u>www.caiso.com/StakeholderProcesses/ Flexible-ramping-</u> <u>product-refinements</u>).

Quantile Analysis: Continuous Classifier Shows Potential

to Adjust Net Load Ramp "Up Uncertainty" (JHU)

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Upward error quantile regression results:

500



500

400

Low Solar Uncertainty

600

Forecasted GHI width W/M²

700

High Solar Uncertainty

bottom, respectively: 90th, 75th, 50th, 25th estimated percentiles) Up-ramp requirement Up-ramp requirement !-- function of solar unc Error MW Up-ramp requirement Jpward

400

Upward Error MW

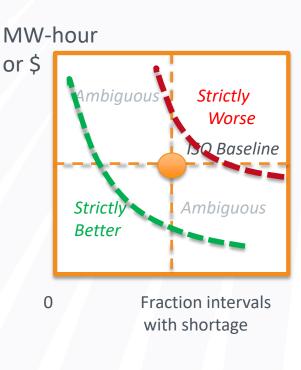
700

600

Out-of-Sample Pareto Analysis of Requirements

Methods:

- Compare performance of ramp capacity (MW) requirement method relative to independent system operator (ISO) baseline method using multiple criteria
- Criteria:
 - 1. Reliability: Fraction of intervals in which ramp capacity (MW) need exceeds requirement ("shortage")
 - 2. Cost: Total MW-hour/\$ cost of requirement.
- Assessment procedure:
 - Simulate rolling estimation method (out-of-sample test)
 - Baseline: Histograms of N previous days' realizations of MW need in that interval
 - Alternative: Statistical or machine learning-based estimate of MW need
 - All methods: Rescale amount or vary target reliability
 - Trade-off: More requirements → less shortage but more cost

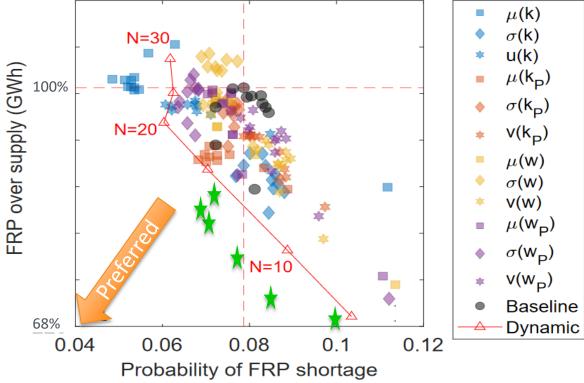






kNN/PCA-Based Method for Flexible Ramp Requirements (UTD)

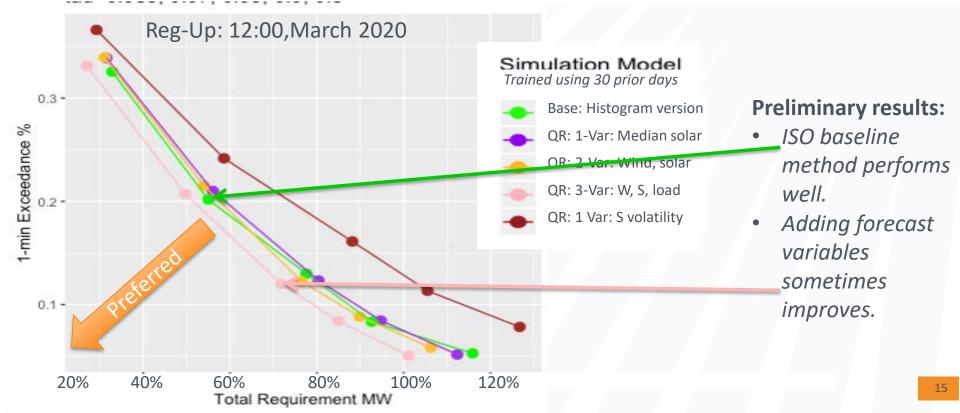
- Shown: kNN-based Flexible Ramping Product (FRP) requirements: Reliability-oversupply trade-offs Feb. 2020.
 - 1-D classifiers from solar site 2 using various predictors.
- Multisite/PCA classifiers perform even better.



Out-of-Sample Pareto Analysis of Weather-Aware *Regulation* Requirements Using Solar Forecasts (JHU)



Quantile regression-based regulation requirements: Tradeoffs between <u>reliability</u> (fraction of 1-min average adjusted ACE > requirement) and <u>MW supply</u>. (Unconditional and 1- and 3-variable rolling regressions)







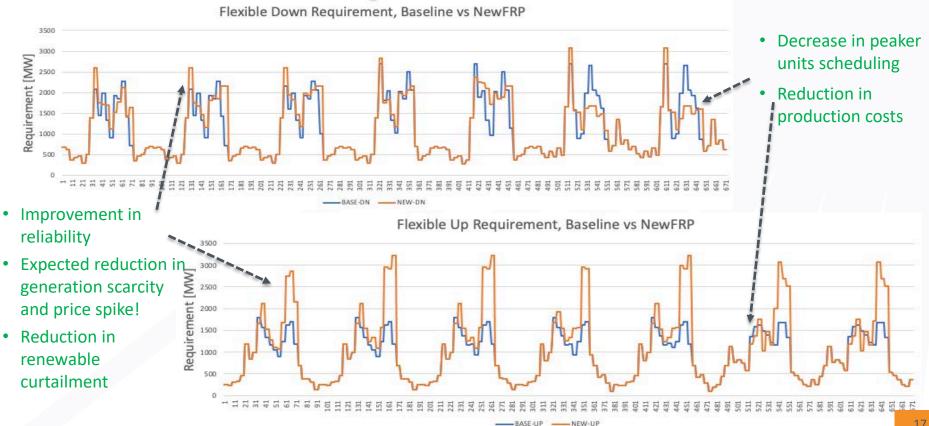
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Other relevant materials for this simulation task:

- Presentation 1: ESIG Spring Technical Workshop, 2019
- Presentation 2: Solar Forecasting 2 Annual Review and Workshop, 2019
- IEEE GreenTech 2020 conference paper: <u>https://www.nrel.gov/docs/fy20osti/75544.pdf</u>

Benefits Assessment Using Simulations (NREL)

FRP Analysis March 9–15: Baseline vs. New Flexible Ramp Requirements



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Experimental Setup for Economic Analysis from Simulation

- Flexible Energy Scheduling Tool for Integrating Variable generation (FESTIV) tool modified with CAISO operating rules
- Three scenarios considered:
 - **1.** Baseline—used historical FRP requirements obtained from the Open-Access Same-Time Information System
 - 2. New FRP—used new machine learning-based FRP requirements calculated by the team
 - **3.** *Perfect*—perfect knowledge (i.e., forecasts) representing operations without netload uncertainty.
 - Will be used to obtain uncertainty-induced costs under scenarios 1 and 2 by subtracting the total costs from the "*perfect*" run

Note: Another metric used in economic analysis will be *"total FRP procurement costs"* (total of "FRP procured" times the "FRP market clearing prices/duals" in various intervals). Only intervals with binding FRP constraints will have nonzero FRP marginals.

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•FESTIV market clearing process modified with CAISO operating rules

IEEE 118 simulated for 3 weeks in March (9–29, 2020):

	Baseline (Scenario 1)	New FRP (Scenario 2)	Perfect NO uncertainty
Total Production Cost [\$M]	23.45	23.05 (1.7% savings)	23.00
Total Uncertainty Cost [\$M]	0.45	0.05 (~90% savings)	

Uncertainty-induced costs reduced by a daily average of 44%

- Savings in production cost from lower FRP (reduction in peaker units)
- ■Higher FRP reduces generation scarcity events and real-time price spikes
 ■More flexible generation, with lower min. gen → reduced curtailment.

Simulation Results – Large 1,820-Node WECC/CAISO System

~1,820 buses: 1,782 buses for CAISO and 38 for other WECC regions and trading hubs

Copperplate analysis for 3 weeks in March (9–29, 2020):

Total Production Cost (\$M)	Baseline (Scenario 1)	New FRP (Scenario 2)	Savings in uncertainty costs (comparing with "perfect" run)	
Week 1 (March 9-15)	70.72	70.56 (0.3% savings)	1.13%	
Week 2 (March 16-22)	49.35	49.29 (~0.09% savings)	6.3%	
Week 3 (March 23-29)	44.95	44.94 (0.01% savings)	0.27%	
All weeks (March 9-29)	165.02	164.75 (0.16% savings)	~1.7% (~\$270К)	

\$0.27M savings in uncertainty-induced costs in 21 days ~ extrapolate to \$4.7M/y*

*Note: Good annual estimations of cost savings will require longer-term simulations under myriads of scenarios

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Simulation Results – Large 1,820-Node WECC/CAISO System

Due to the computational burden of simulations with full network, few different days in March 2020 were simulated.

With full network analysis: \$0.43M savings in 8 days simulated

Total Production Cost (\$M)	Baseline (Scenario 1)		Ne	w FRP (Scenario 2)	Savings in uncertainty costs (comparing with "perfect" run)	
5 days: March 16-20	66.9			66.5 (0.5% savings)	2.34% (~\$332K)	
1 day: March 23	11.94		11.91 (~0.23% savings)		10.6% (~\$28.3K)	
3 days: March 23-25	39.74		39.65 (0.24% savings)		~1.7% (~\$97К)	
Total FRP procurement Cost (\$K)		Baseline (Scenario 1)		New FRP (Scenario 2)		
5 days: March 16-20 7		76.14	67.53 (11.3% s		savings ~ 8.6K)	
3 days: March 23-25		60.38		28.9 (52.15% savings ~ 31.5K)		

High solar penetration and conventional generation flexibility limited scenarios may further increase the cost benefits of using probabilistic forecasts based FRP procurements

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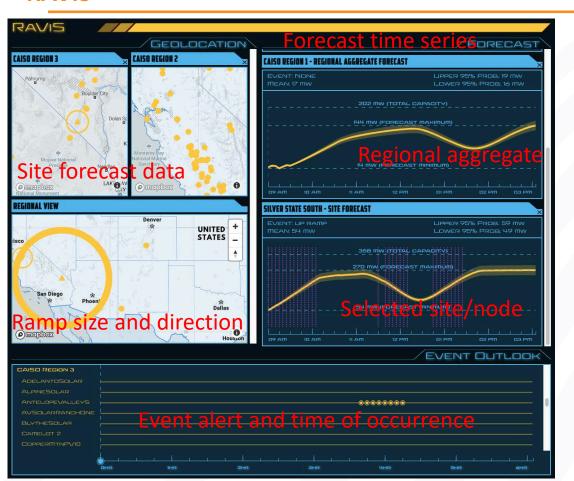


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Open-source link to the source code: <u>https://github.com/ravis-nrel/ravis</u> Public-facing demo with past forecasts (2017 solar eclipse): <u>https://maps.nrel.gov/ravis</u>

Resource Forecasts and Ramp Visualization for Situational Awareness RAVIS





RAVIS: A modular dashboard for viewing:

- Forecast timeline
- Spatially relevant forecasts and events
- Details of specific events as desired, including market data.

Design: RAVIS's technology suite is assembled to *provide optimum visualization facility.*

- Takes advantage of web application technologies and tooling
- These technologies *enable deployment in any environment*, using any operating system.

Can Also Show Net Load Time-Series Forecasts and Available Generation Headroom/Flexibility





Flexible architecture to visualize more data:

- Net-load forecasts, or each component
- Available generation flexibility
- Network nodes
- Transmission
- Prices

Sorted ramp events: "most active" (e.g., most events in the next 5 hours)



Summary of Features Visualized in RAVIS

1) Site-specific probabilistic solar power forecasts:

Hovering reveals forecast data and other metadata; ramp size and direction are shown by circle and arrow.

- 2) Time-series forecast data: Clicking a site/region shows individual site-specific probabilistic forecast time series.
- 3) Event alerts (e.g., ramp alerts), at site as well as regional level

4) User can configure visualization parameters:

- a) Specific sites or user-defined aggregate regions
- b) Ramp definition customizable by end user.

5) Data from market clearing engine integrated with forecasts viewer:

- a) On map: Network topology, nodal prices, transmission congestion
- b) On time series: Available generation flexibility, plotted against net load forecasts.

Open Source, Publicly Accessible/Extensions

- 1) Open-source link to source code: https://github.com/ravis-nrel/ravis
- 2) Public-facing demo (2017 solar eclipse data): https://maps.nrel.gov/ravis

3) Documentation published @ https://www.nrel.gov/docs/fy21osti/79746.pdf. Edwards, Paul, Haiku Sky, and Venkat Krishnan. 2021. RAVIS: Resource Forecast and Ramp Visualization for Situational Awareness—An Introduction to the Open-Source Tool and Use Cases. Golden, CO: National Renewable Energy Laboratory. NREL/TP-5D00-79746.

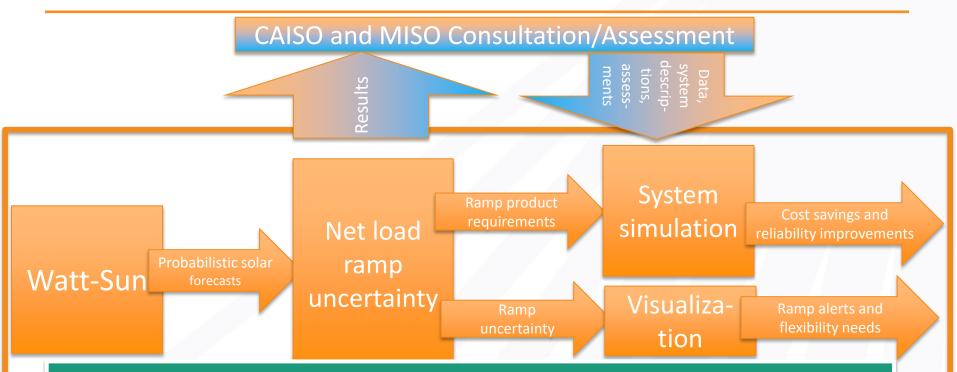
4) EnergyViz conference publication @ https://www.nrel.gov/docs/fy21osti/79786.pdf

Future extensions:

- Transmission and distribution grids with grid-edge visibility
- Ongoing project: Sensor data and cyber threat integration
 - (U.S. Department of Energy Office of Cybersecurity, Energy Security, and Emergency Response (CESER) funded project Situational Awareness and Grid Analytics (SAGA) project at NREL.)

Conclusion

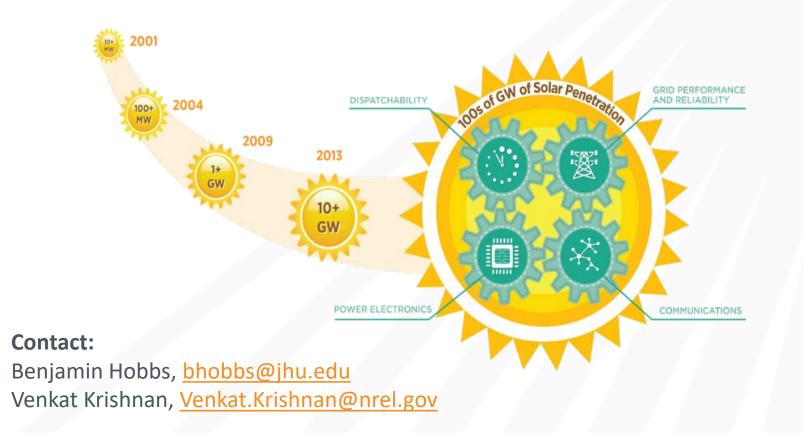
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- We conclude that probabilistic forecasts are a highly promising way to condition ancillary service requirements on up-to-date weather forecasts.
- *Future:* Convolve wind, retail load, and solar forecasts for fuller picture.

Questions?

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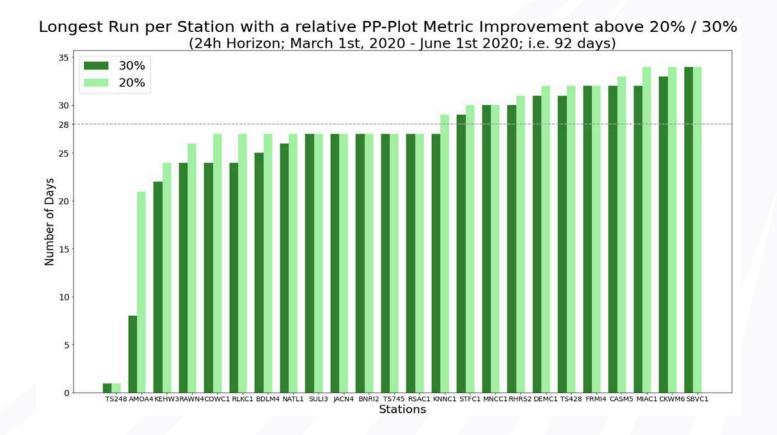




Backup/Extra Slides

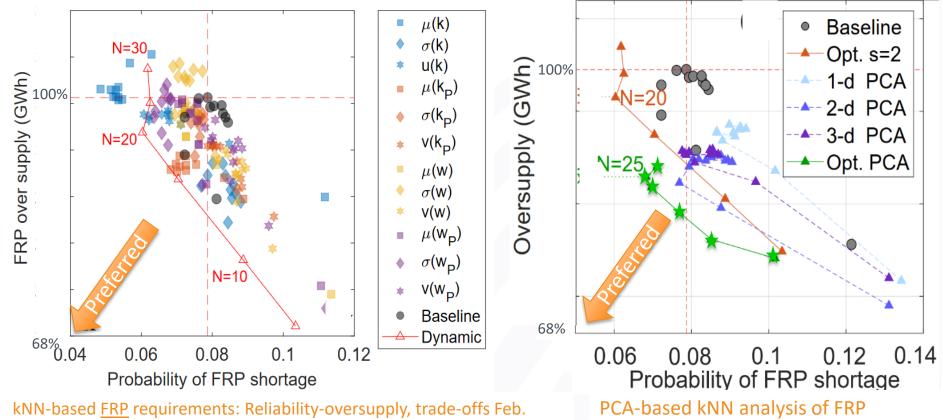
Continuous Improvements of Watt-Sun





PCA/kNN-Based Method for FRP Estimation (UTD)





2020. (1-D classifiers from solar site 2 using various predictors)

requirements from multiple sites