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Coordinated Ramping Product and Regulation Reserve Procurements in CAISO and MISO using Multi-Scale Probabilistic Solar Power Forecasts (Pro2R)

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Solar Forecasting 2 Annual Review & Workshop
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- Amber Motley, Clyde Loutan, Rebecca Webb, Guillermo Bautista (California ISO)
- Blagoy Borissov, Stephen Rose (Midcontinent ISO)

Pro2R Project Summary

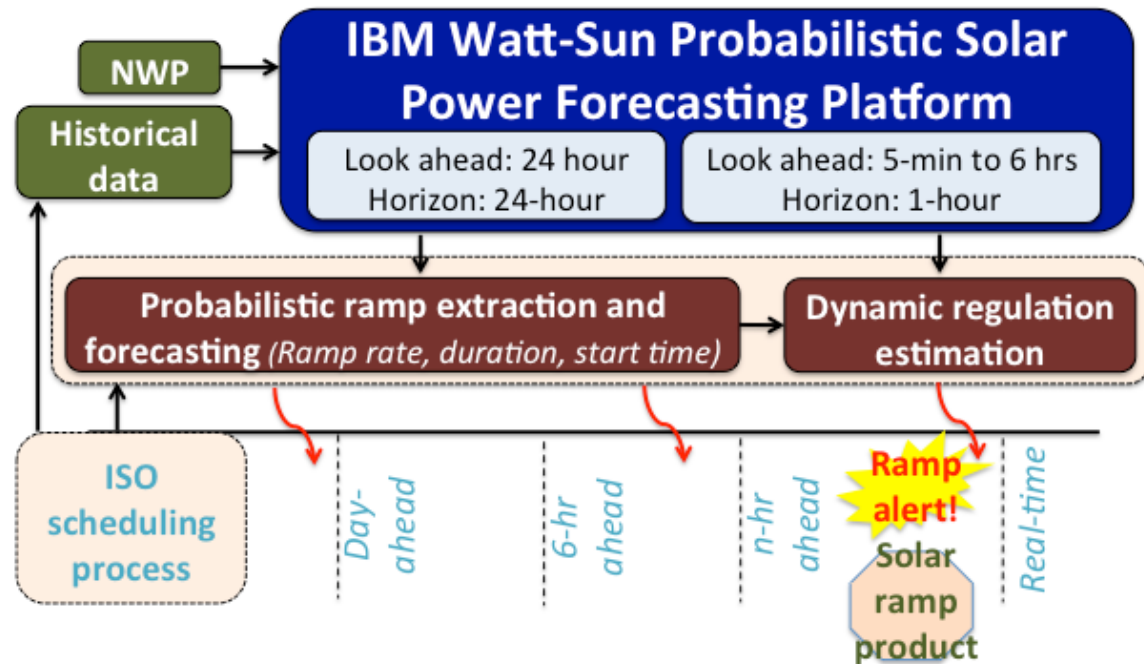
Objective: Integrate probabilistic short- (2-3 hr ahead) and mid-term (day-ahead) solar power forecasts into operations of CAISO & MISO

Approach:

Thrust 1: Advanced big data-driven “probabilistic” solar power forecasting technology using IBM Watt-Sun & PAIRS (Big data information processing and machine learning approaches to blend outputs from multiple models).

Thrust 2: Integrate probabilistic forecasts in ISO operations for *ramp product* & *regulation* requirements

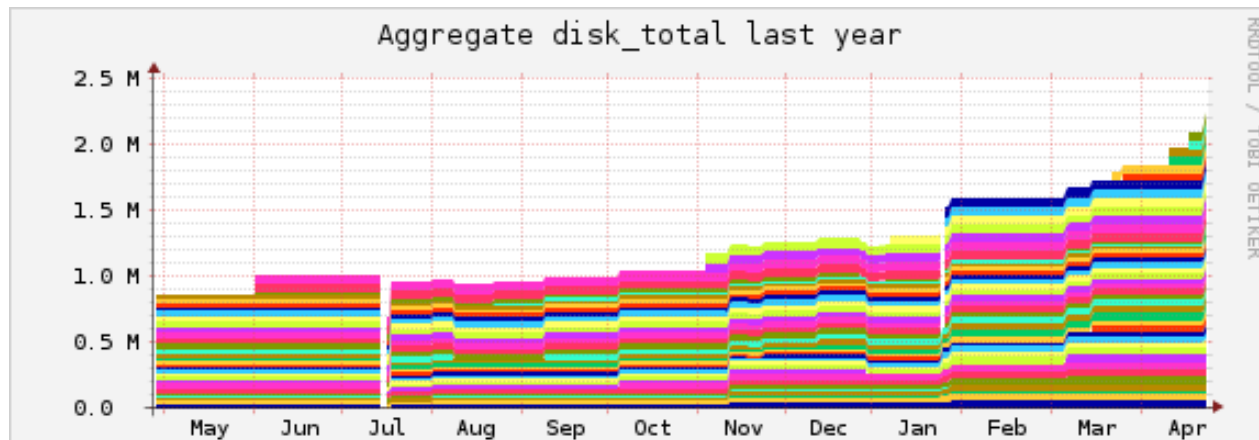
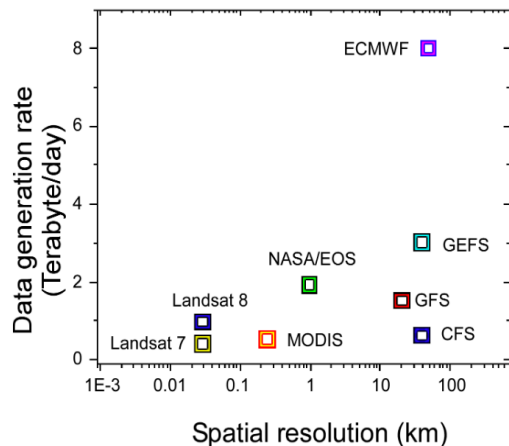
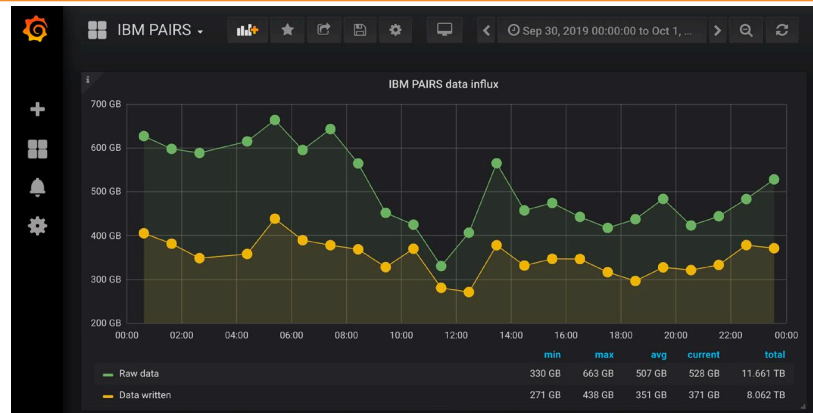
Thrust 3: Provide **situational awareness** via **visualizations** of probabilistic ramp forecasts & alerts



Thrust 1: SFII Built Upon IBM PAIRS Geoscope Platform

- Distributed computational system
- Scales to many hundreds of Petabytes (PB)
- Data processing rate at PB/day.

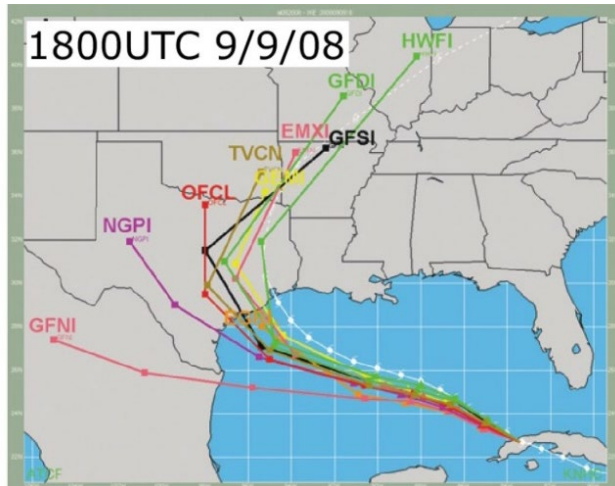
	NAM	HRRR-Subhour	GOES-R
spatial resolution	12 km	3 km	500 m
temporal resolution	1 hr	15 min	5~10 min
daily data volume	9.4 GB	86 GB	203 GB



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Situation-Dependent Model Blending for Solar Forecasting

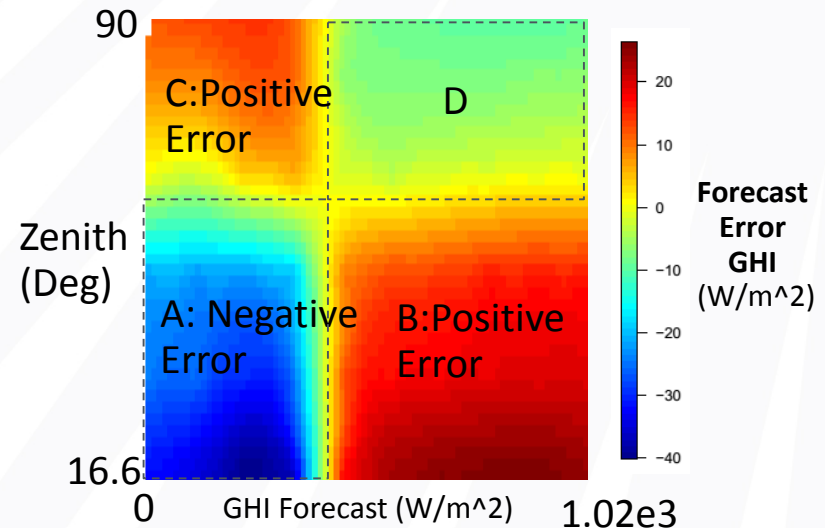
Hurricane Ike path forecasts from
8 different weather models*



Question: What is the error of the models, when, where, under what weather situation?

NAM GHI Forecast Error (Surfrad BND)

– Strongly depends on zenith angle and forecast irradiance. The two parameters create 4 categories of situations:

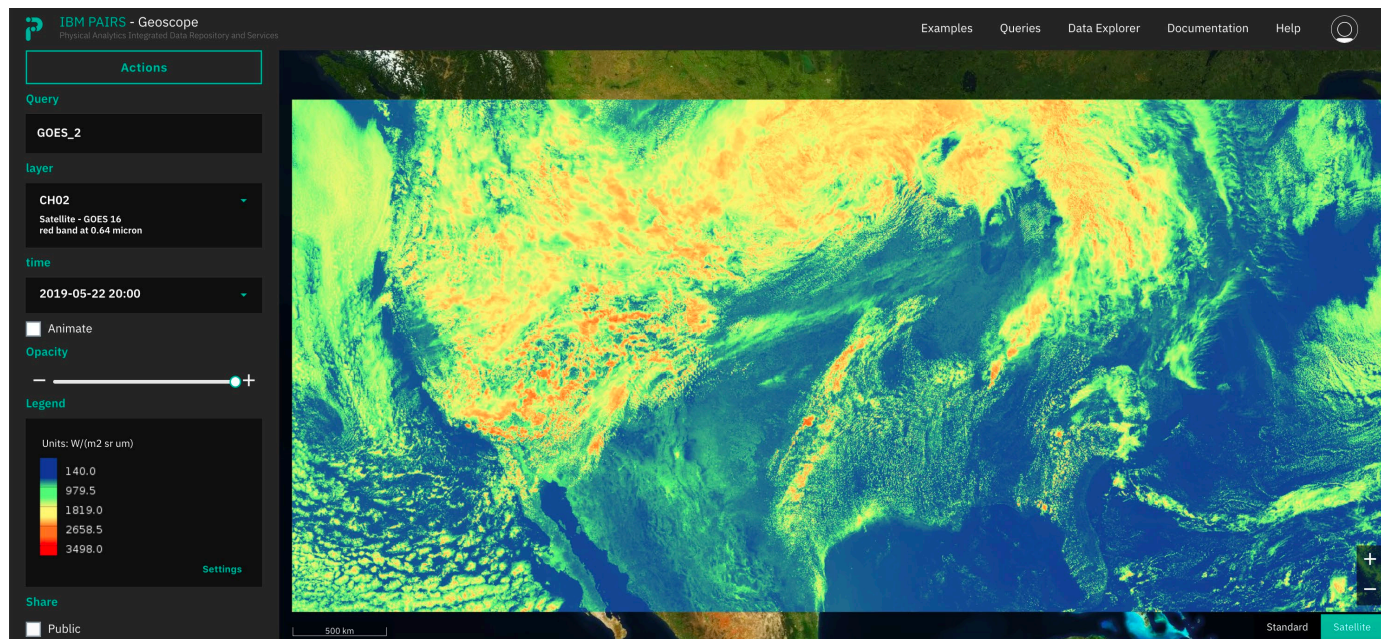


In progress– Expected in Watt-Sun 2.0, 2019-20 Q7

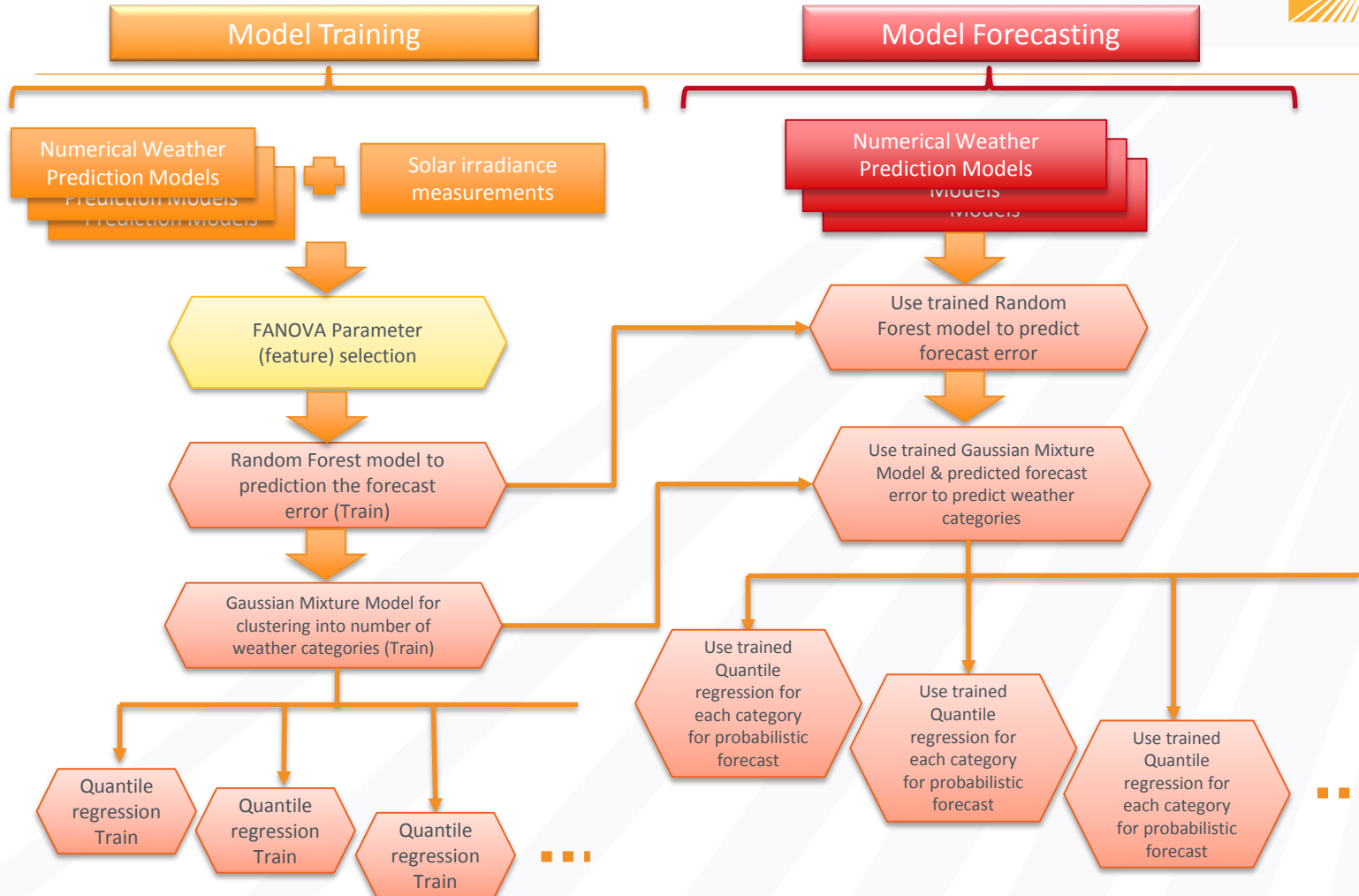
- **Blue band, 0.47 μm :** *monitor dust, haze, smoke and clouds*
- **Red band, 0.6 μm :** *detect fog, estimate solar insolation, depict diurnal aspects of clouds*
- **Near-infrared, 0.86 μm :** *detect daytime clouds, fog, and aerosols; calculate normalized difference vegetation index*
- **Infrared, 10.3 μm :** *correct atmospheric moisture, estimate cloud particle size, characterize surface properties*

Plan:

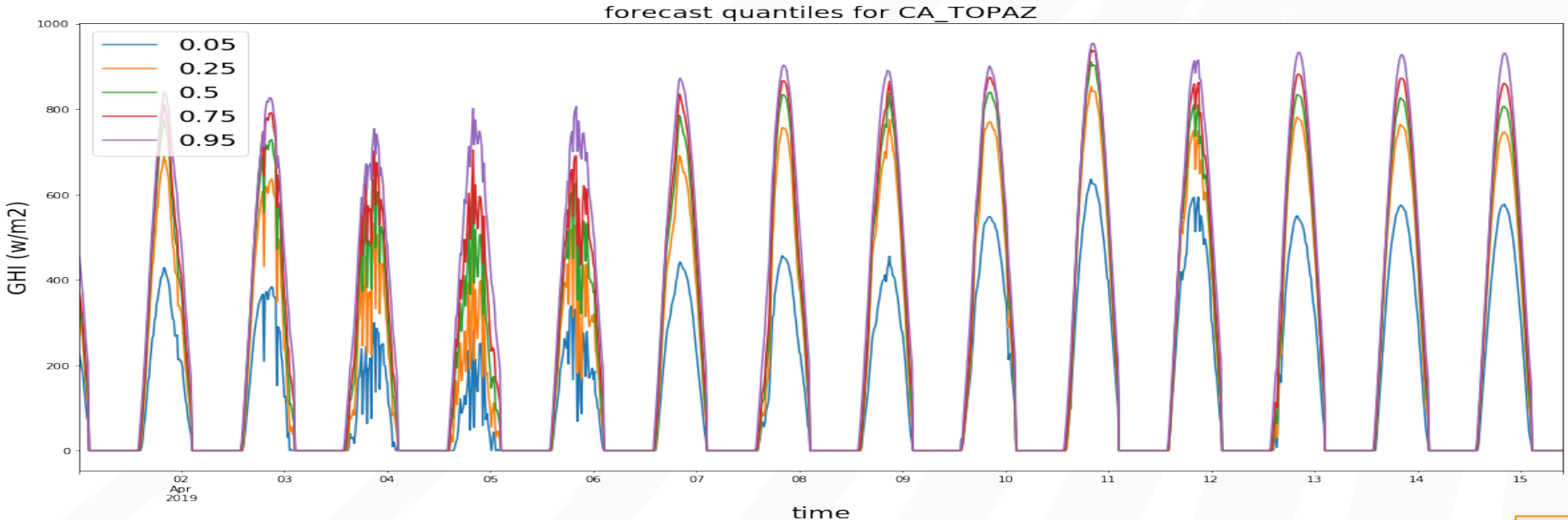
- Translate GOES imagery to Solar Irradiance raster map
- Use Deep Learning combined with Optic Flow to forecast cloud movement, then translate into forecast solar irradiance map.



Overall Flowchart



- Quantiles of solar as function of independent variables
- Example results for 2 hr-ahead forecasts
 - Distributions are asymmetric \rightarrow need *quantile regression techniques*
 - Adjacent days have different distributions; but present CAISO flexiramp requirements are very stable day-to-day because they don't reflect weather forecasts \rightarrow need to *integrate probabilistic forecasts in requirements*

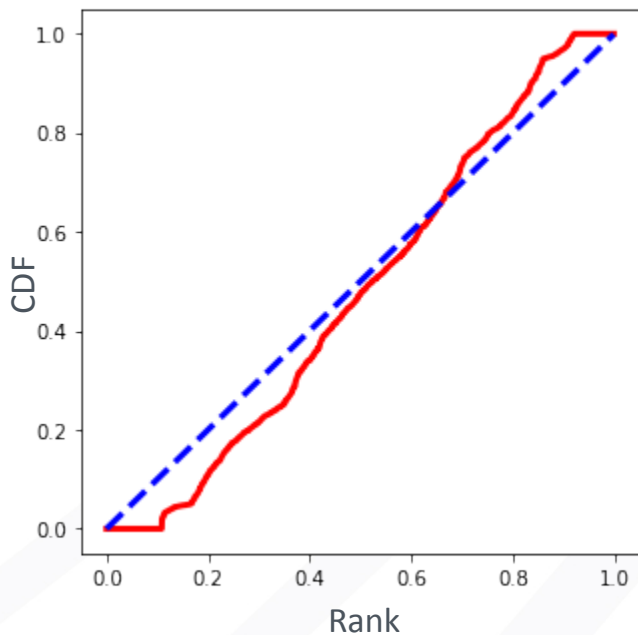


P-P plot-based Error Metric

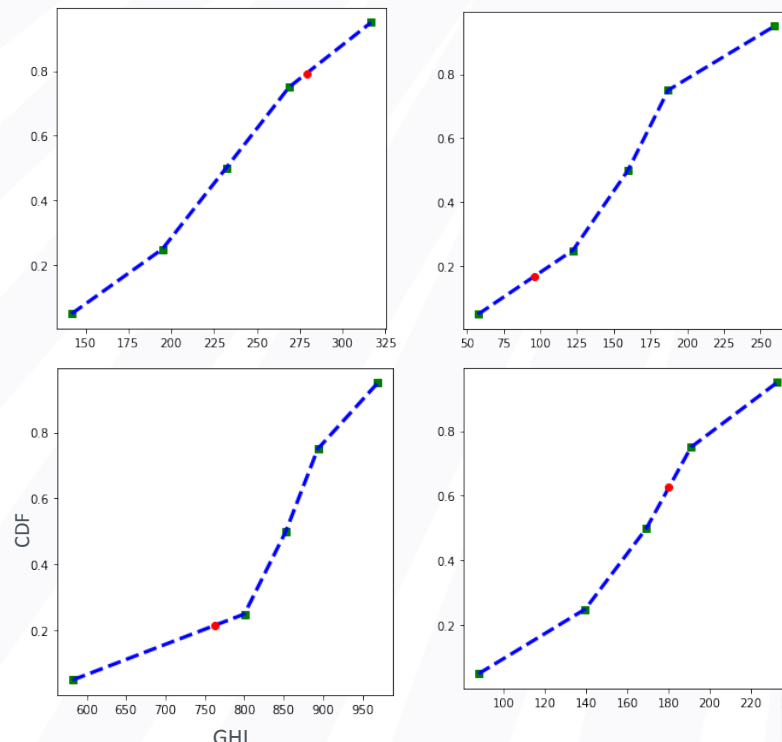
- A P-P plot (probability–probability plot) assesses how **closely two data sets agree**
- By plotting the two cumulative distribution functions against each other

P-P plot (2 hr forecast)

Mean absolute difference: 0.0543



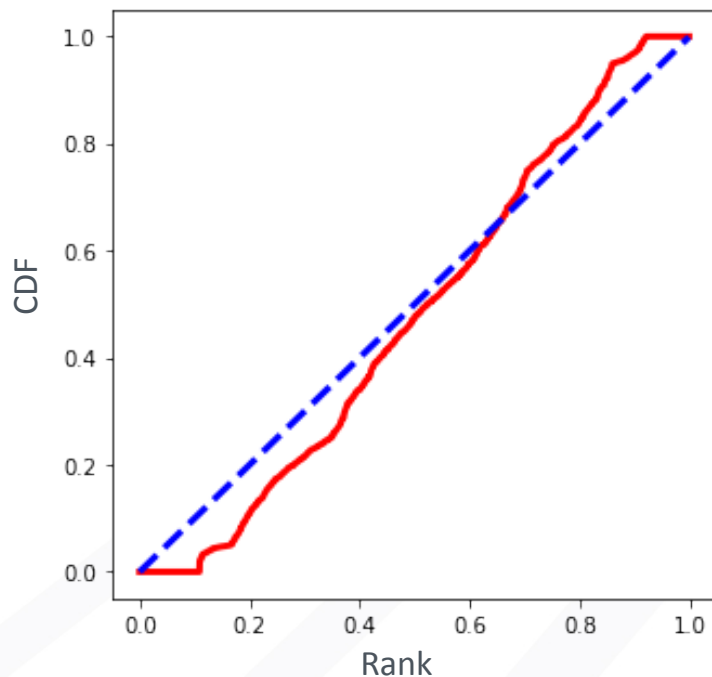
Example Empirical CDF Curves (2 hr forecast)



P-P plot 2 Hr Ahead Error Metric: Comparison with Bias Corrected HRRR

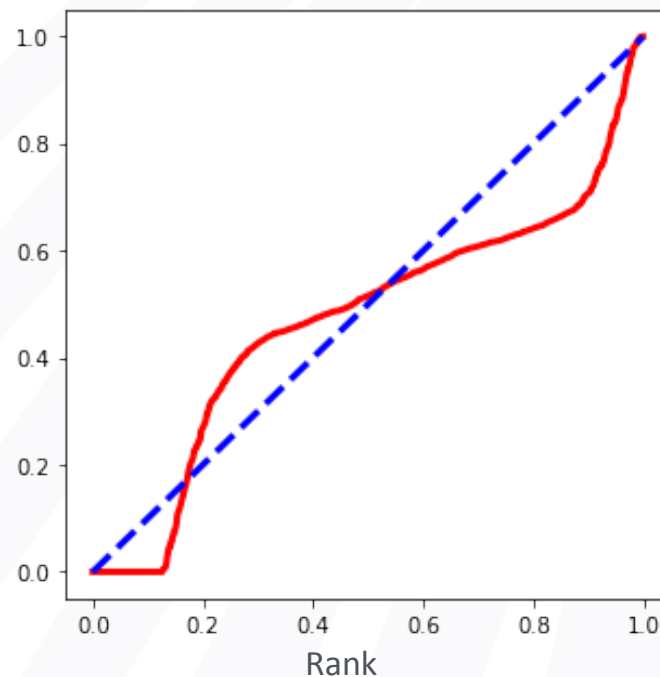
Empirical CDF

Mean absolute difference: 0.054



Bias Corrected HRRR (normal distribution)

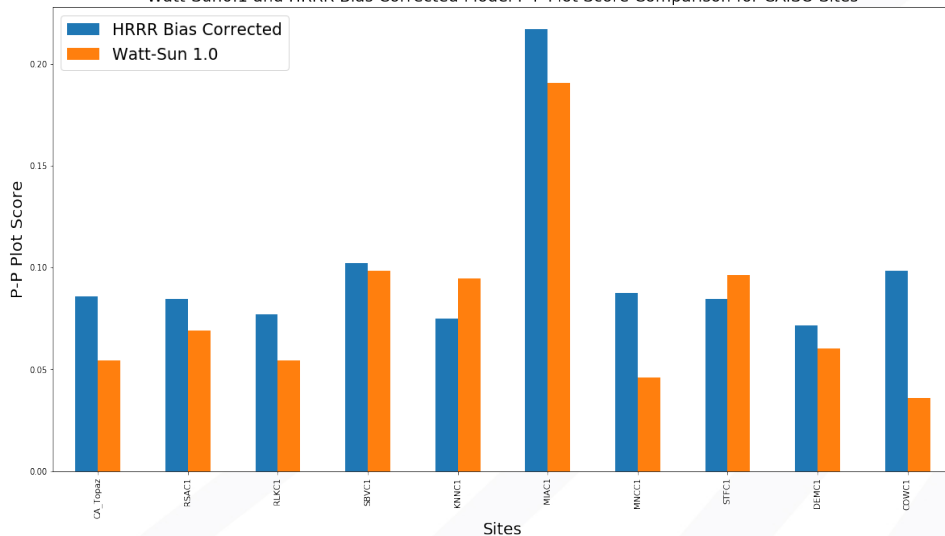
Mean absolute difference: 0.086



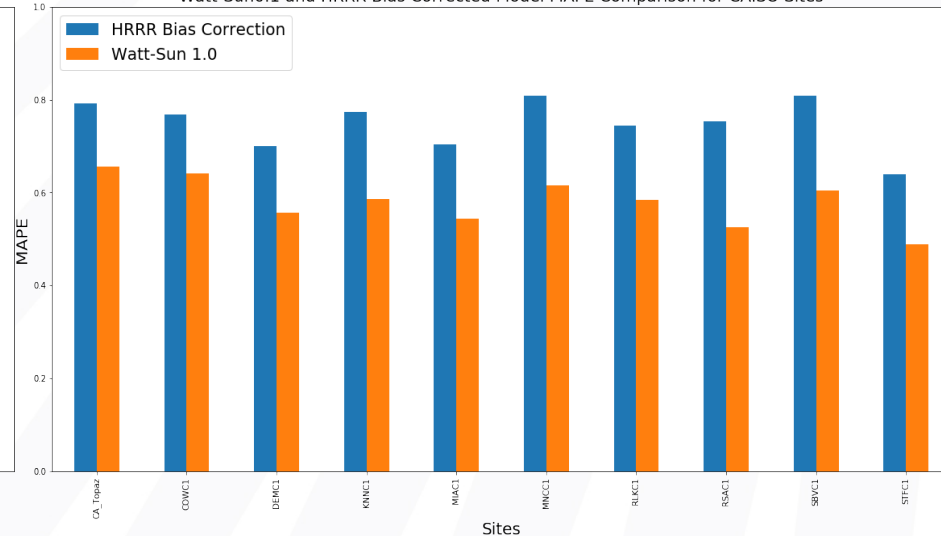
Thrust 1: Forecast Error Report -- CAISO

- P-P Plot score and MAPE comparison (normalized by max GHI; daylight hours only)
- Watt-Sun 1.0 outperforms HRRR Bias Corrected in all sites in terms of MAPE, in most sites in terms of P-P Plot score

Watt-Sun0.1 and HRRR Bias Corrected Model P-P Plot Score Comparison for CAISO Sites



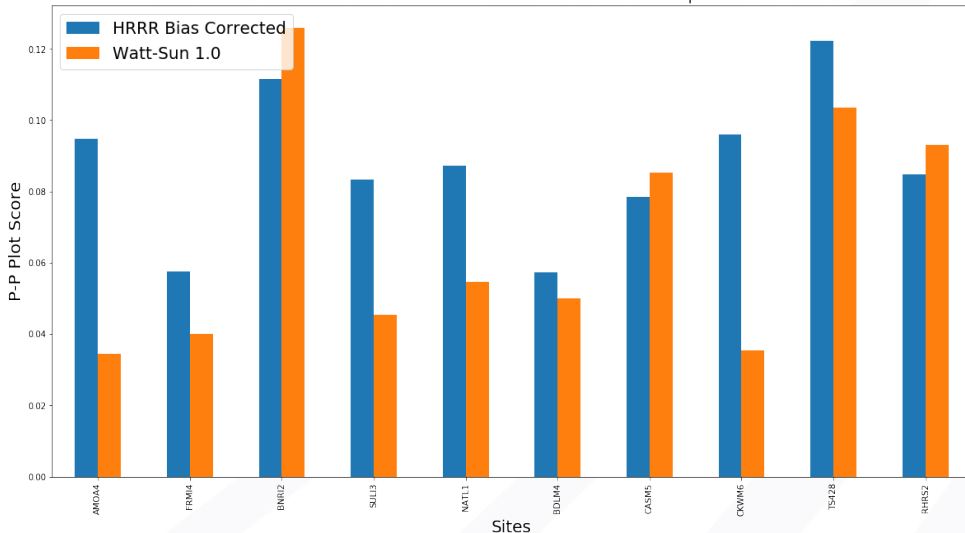
Watt-Sun0.1 and HRRR Bias Corrected Model MAPE Comparison for CAISO Sites



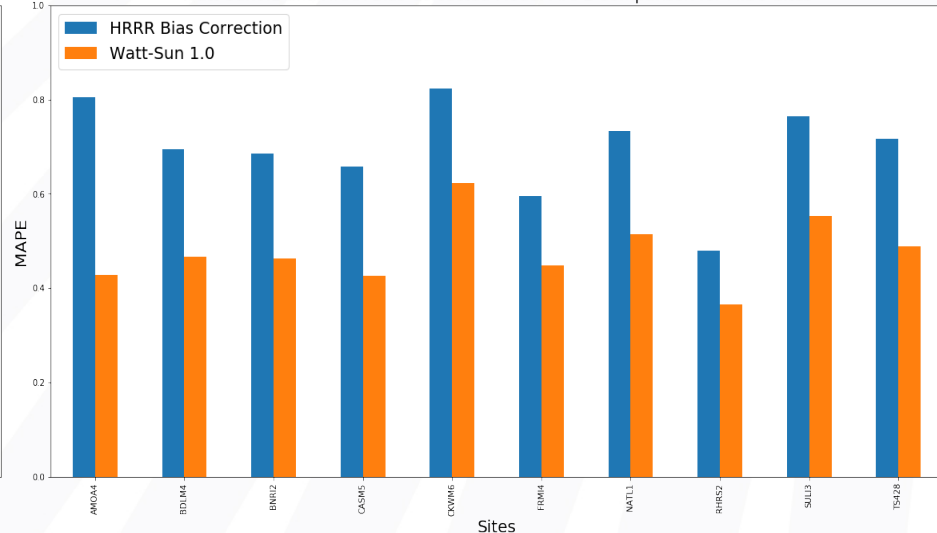
Thrust 1: Forecast Error Report -- MISO

- P-P Plot score and MAPE comparison (normalized by max GHI, daylight hours only)
- Watt-Sun 1.0 outperforms HRRR Bias Corrected in all sites in terms of MAPE, in most sites in terms of P-P Plot score

Watt-Sun0.1 and HRRR Bias Corrected Model P-P Plot Score Comparison for MISO Sites

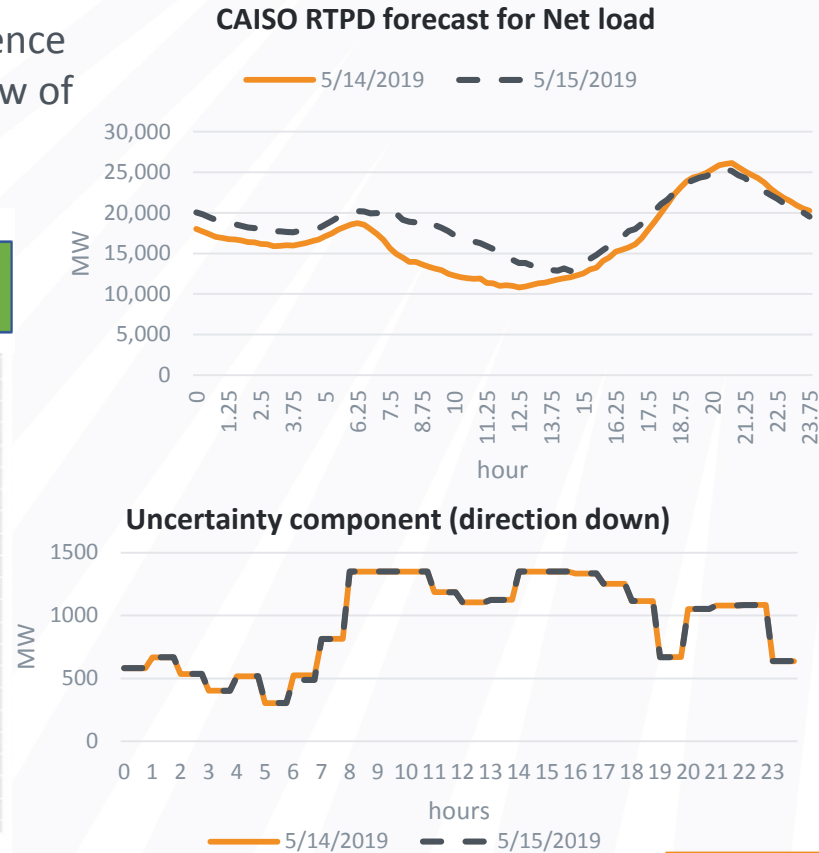
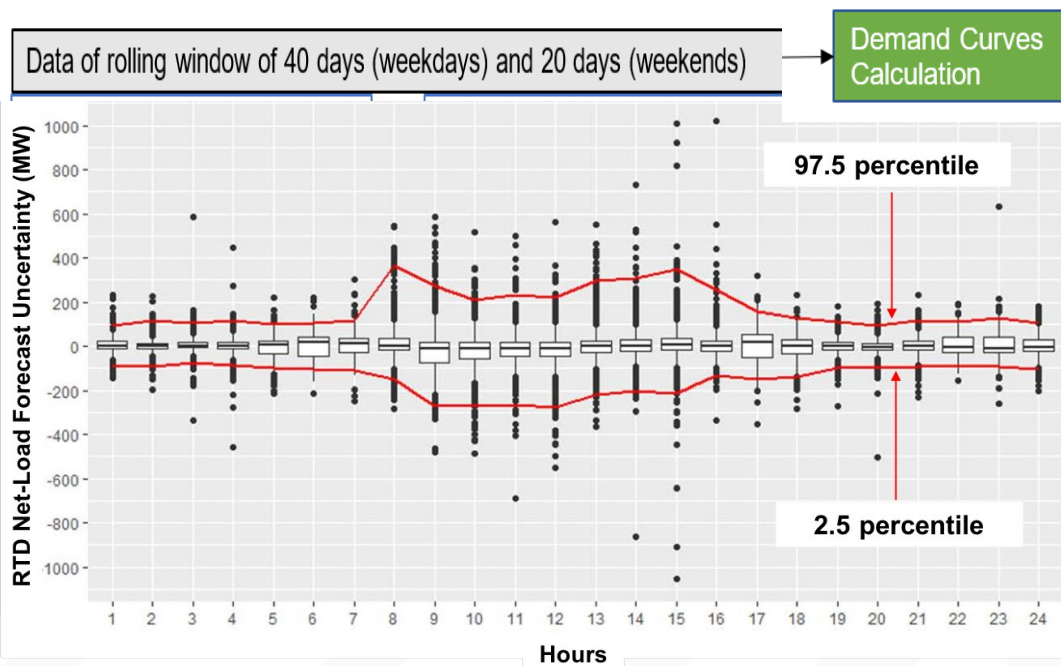


Watt-Sun0.1 and HRRR Bias Corrected Model MAPE Comparison for MISO Sites

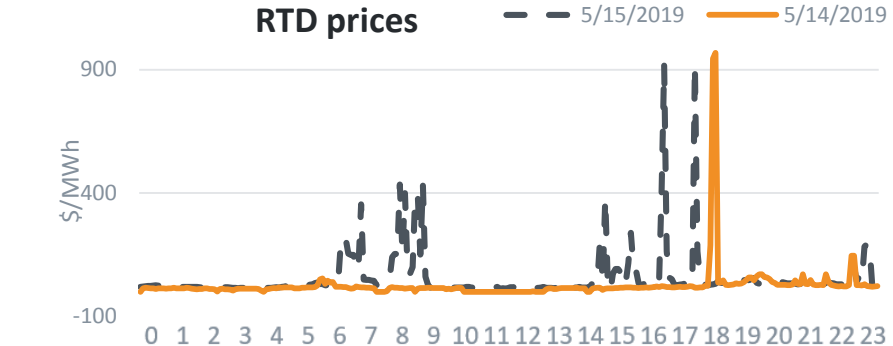
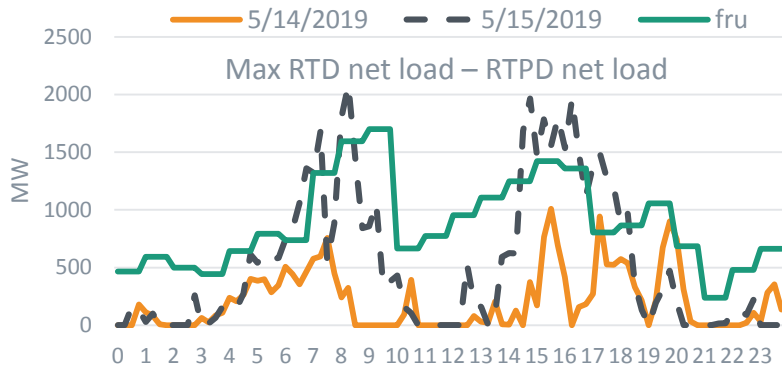


Thrust 2: Integration of probabilistic forecasts into ISO operations: Flexible Ramping Product

Baseline approach CAISO bases FRP requirements on confidence intervals for net load uncertainty by analyzing a rolling window of last 20-40 days for the same hour

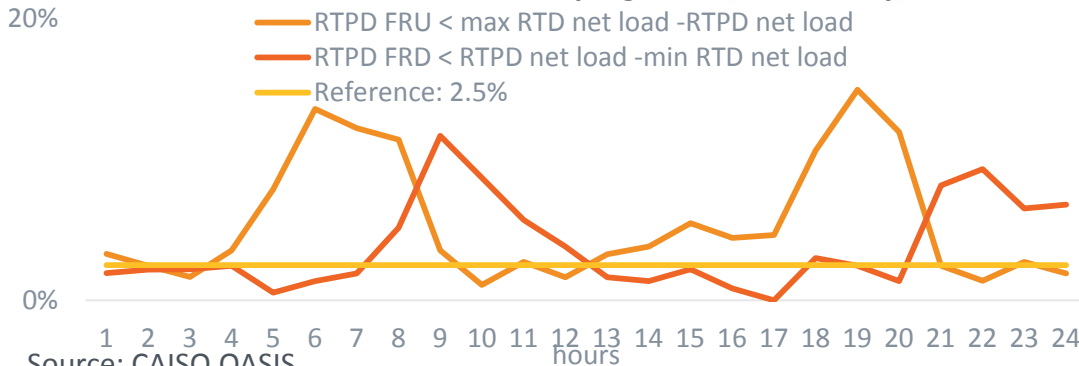


Baseline method for FRU uncertainty component potentially leads to under-procurement & price spikes



*Two days had different net load errors between RTPD & RTD
→ different RTD prices*

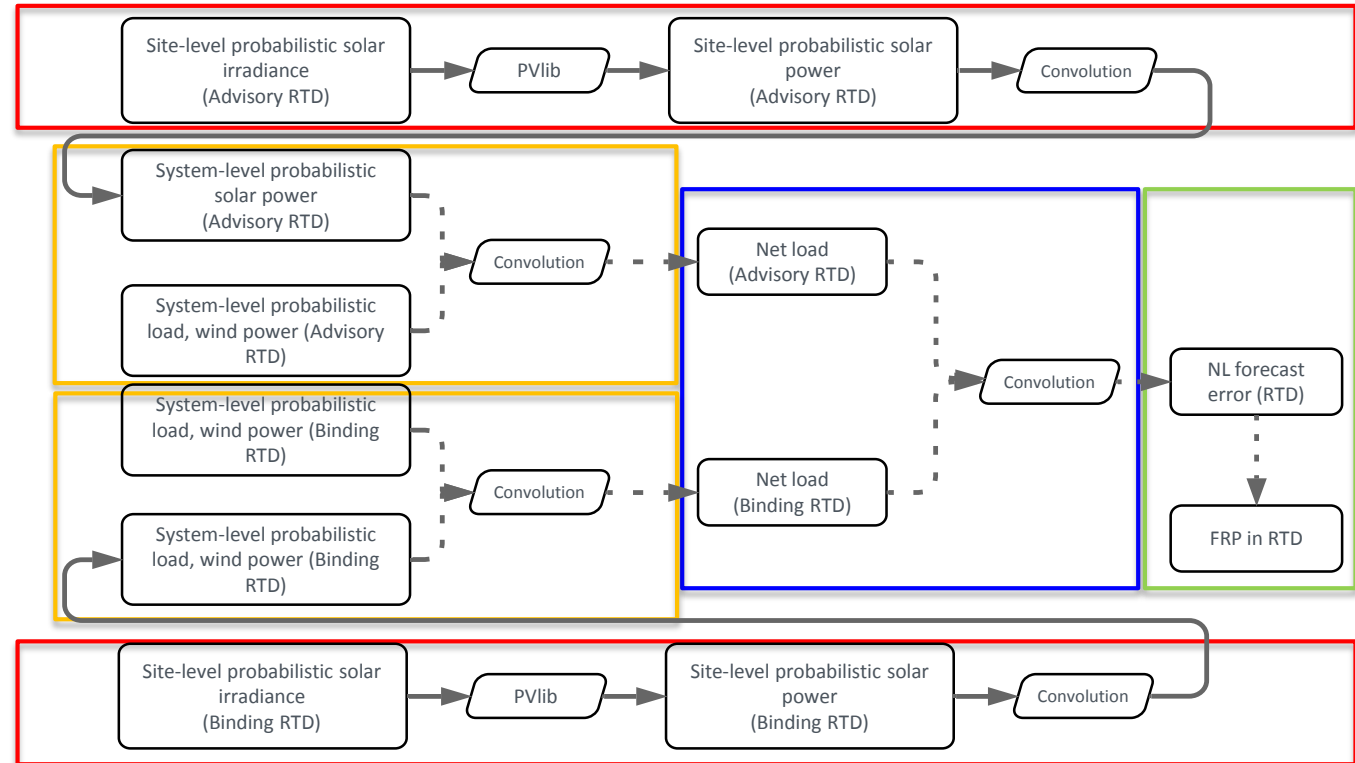
Percent of 15-min intervals in March-May 2019 that underestimated flexible ramping need (CAISO only)



Estimates of FRP uncertainty component do not necessarily correspond to the target confidence interval

Probabilistic net load forecast-based ramping product

- I. Aggregation
- II. Net load (NL)
- III. NL forecast error
- IV. FRP Requirements



Value of probabilistic forecast-based FRP procurement

F: Frequency

Compare baseline to probabilistic requirements

F_a

Under procurement (Risky)

Over procurement (Conservative)

Probabilistic FRU \approx Baseline FRU

Probabilistic FRU $>$ Baseline FRU

Probabilistic FRU $<$ Baseline FRU

Compare requirements

F_b

F_c

F_d

F_e

Actual RTD flexibility need $>$ Baseline FRU

Actual RTD flex need \leq Baseline FRU

Actual RTD flex need $>$ Probabilistic FRU

Actual RTD flex need \leq Probabilistic FRU

Actual need (realization)



Performance of prob. forecasts

Production cost

\approx

Decrease

Increase

Increase

Decrease

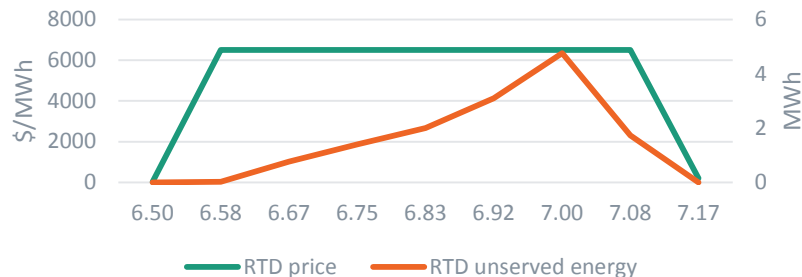
Power balance violations (scarcity events)

\approx

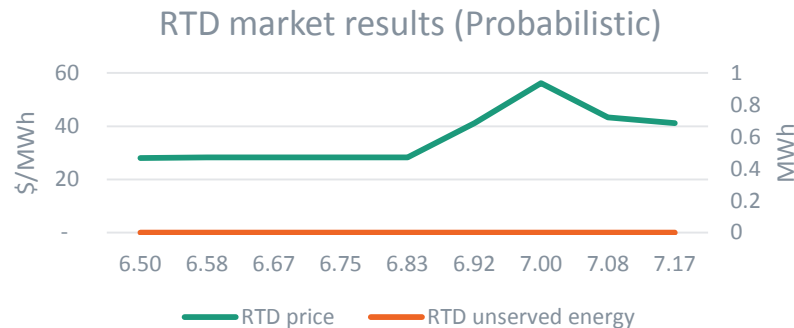
Market simulations: Probabilistic vs. Baseline

Test system: Modified 118 bus IEEE Reliability Test System, mimicking CAISO gen mix (1/10th CAISO system)

Baseline *Power balance violations during 7 RTD market intervals:*
RTD market results (Baseline)



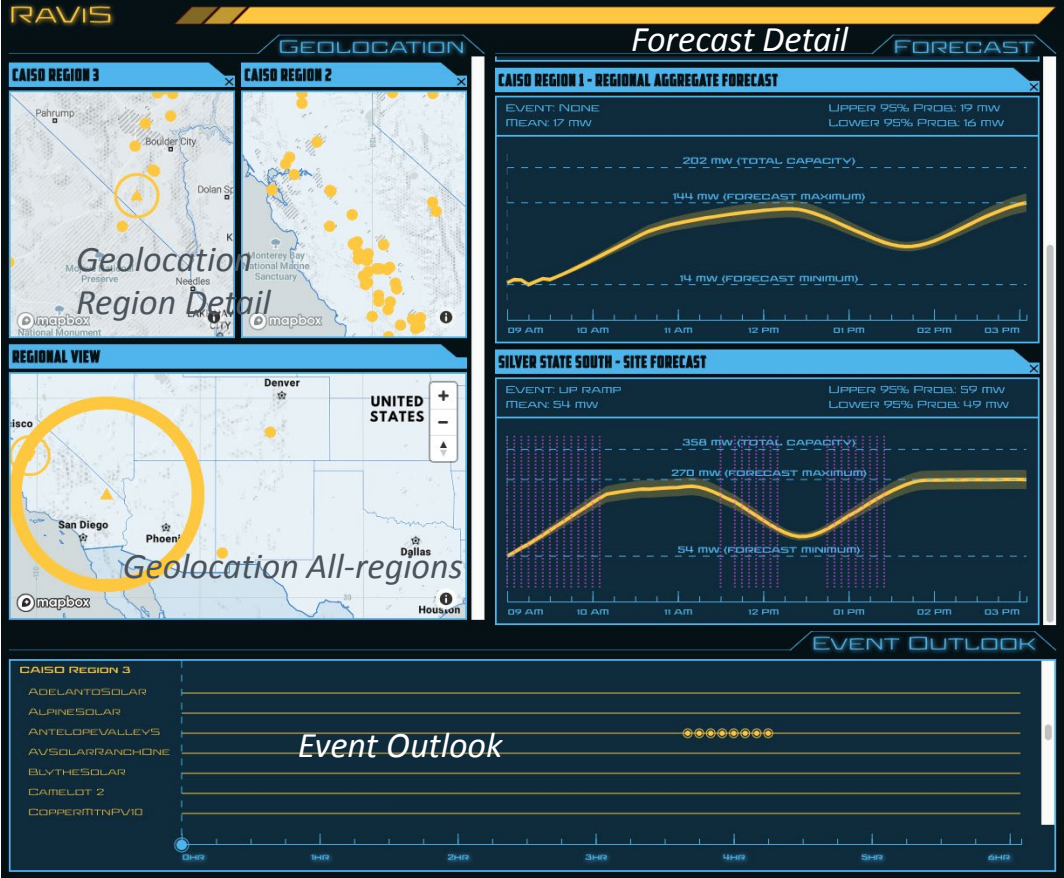
Probabilistic *No power balance violations in RTD market:*
RTD market results (Probabilistic)



- *When baseline is riskier* → we expect reliability benefits when high ramps realized
- *When baseline is conservative* → we expect production cost reductions when ramps not realized



Thrust 3: RaViS - Visualization tool



The Ramp Visualization for Situational Awareness (RaViS) tool provides situational awareness and visualization capabilities using probabilistic solar and net-load time series.

Features:

- Integrates IBM forecast data
- Refresh rate of 60 seconds
- User interface: Single page web application and open source
- Shows site specific metadata via hover
- Highly flexible, easily configurable

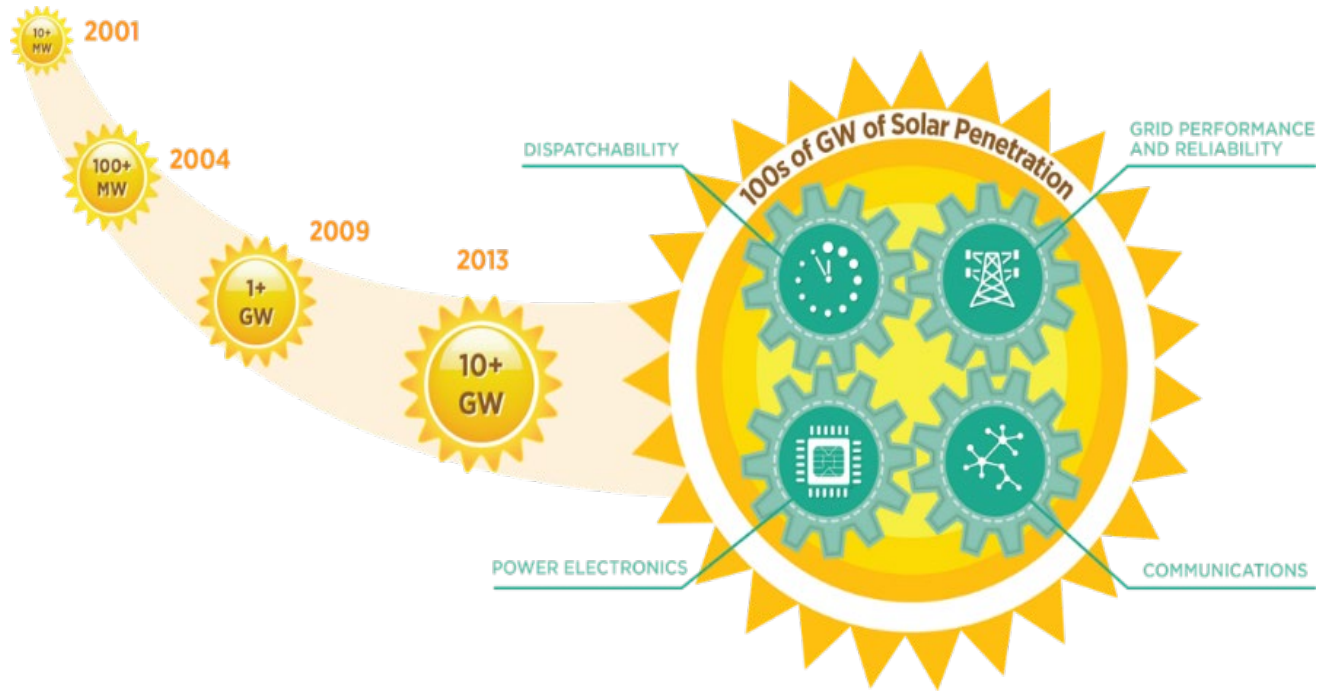
Future work:

- Net-load ramps
- Adaptable to other kinds of events: outage/trip, cyber threats

- **Thrust 1:** Probabilistic Watt-Sun Versions 2.0 & 3.0: integrates GEOS-R, multi-expert machine learning, & blended models
- **Thrust 2:**
 - Statistical/ML modeling of relationship of probabilistic solar forecasts to net load uncertainty on regulation & FRP timescales
 - Simulation-based testing on ISO-scale systems of improved requirements: cost & reliability
 - Interaction with CAISO & MISO on data, method value, simulations, & implementation pathways
- **Thrust 3:** RaVIS development, including integrating latest Watt-Sun methods, net load data, market information, and ISO feedback

- Identified the major methodological issues in integrating probabilistic forecasts into system products
 - BP2 will address those issues and possible improvements in forecasting and industry practice (e.g., product definition, timeline)
 - Expected outcome: a blueprint for research in forecasting and industry practice
- Project results expected to highly influence industry practice
 - ISO staff confirmed that they are evaluating potential improvements to their existing requirements approach
 - Integrating probabilistic forecasts is the most promising way to address needs for requirements to reflect up-to-date weather forecasts

Questions?



EXTRA SLIDES

Thrust 2.1: Requirements for ramping product

- RTPD
 - Requirements are **increased occasionally** due to greater uncertainties of NL forecast errors
 - 25%-58% of hours see >10% reductions in FRU and FRD requirements

	Percentage of time when reduction is over 10%	
	FRU	FRD
May 27, 2019	38%	25%
May 28, 2019	42%	50%
May 29, 2019	46%	50%
May 30, 2019	58%	38%
May 31, 2019	42%	46%

