

The Use of Probabilistic Forecasts in Theory and Practice

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October 22, 2019

A photograph of a wind farm at sunset. Several white wind turbines are visible against a sky with large, orange-tinted clouds. The foreground is a dark, flat field.

The Use of Probabilistic Forecasts in Theory and Practice

Dr. Sue Ellen Haupt

National Center for Atmospheric Research
Research Applications Laboratory

A Team Effort

The Use of Probabilistic Forecasts

Applying Them in Theory and Practice

By Sue Ellen Haupt, Mayte Garcia Casado, Michael Davidson, Jan Dobschinski, Pengwei Du, Matthias Lange, Timothy Miller, Corinna Möhrle, Amber Motley, Rui Pestana, and John Zack

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MUCH OF THE ELECTRIC SYSTEM IS WEATHER dependent; thus, our ability to forecast the weather contributes to its efficient and economical operation. Climatological forecasts of meteorological variables are used for long-term planning, capturing changing frequencies of extreme events, such as cold and hot periods, and identifying suitable locations for deploying new resources. Planning for fuel delivery and maintenance relies on subseasonal to seasonal forecasts. On shorter timescales of days, the weather affects both energy demand

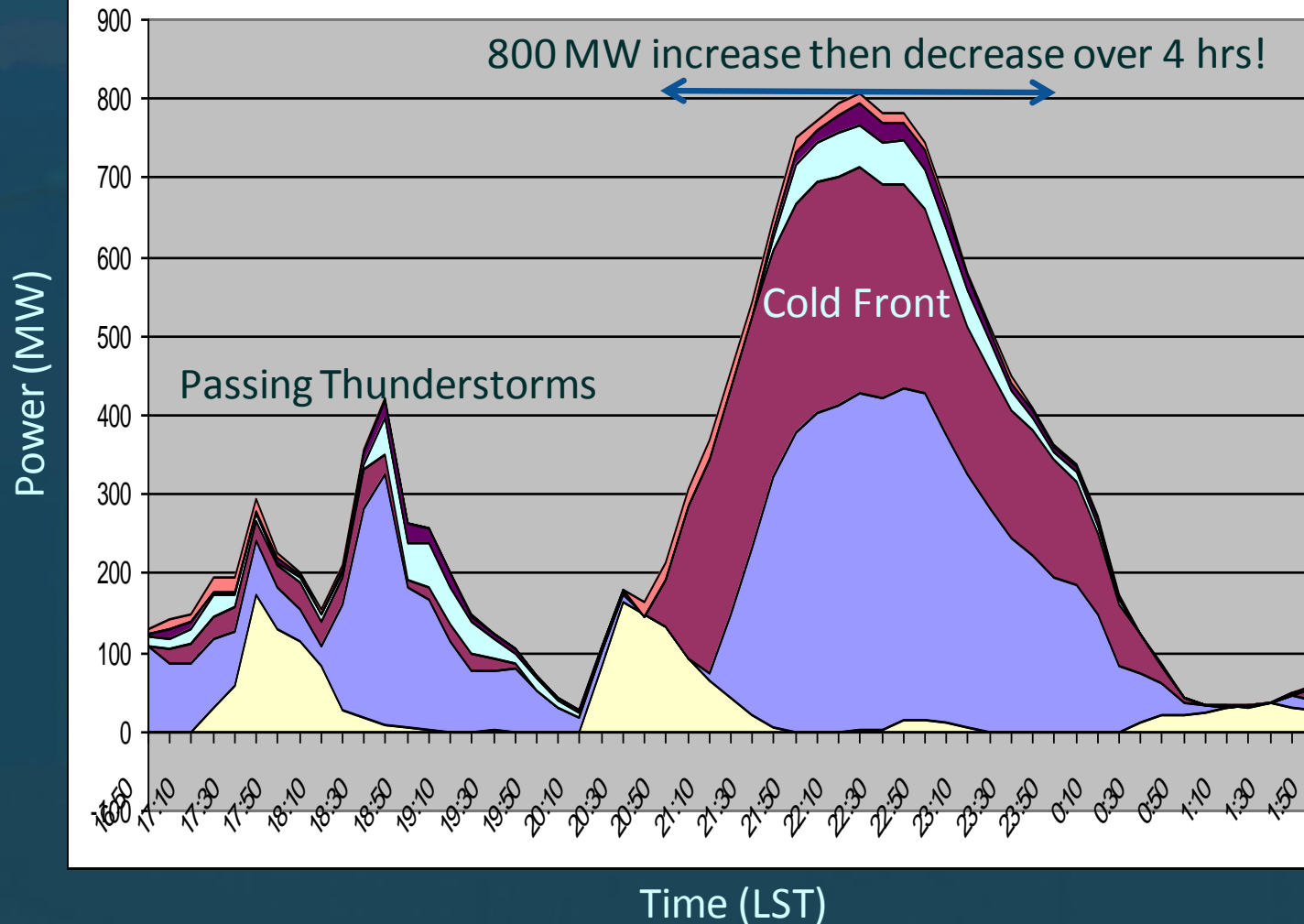
and supply. Electrical load depends critically on weather because so much electricity is used for heating and cooling. As more renewable energy is deployed, it becomes increasingly important to understand how these energy sources vary with atmospheric conditions; thus, predictions are necessary for planning unit commitments. On the scales of minutes to hours, short-term nowcasts aid in the real-time grid integration of these variable energy resources (VERs).

Meteorologists use the dynamical equations of fluid motion to forecast the weather by numerically integrating those equations forward in time in numerical weather prediction (NWP) models. The weather is both variable and



We wish to predict specific events

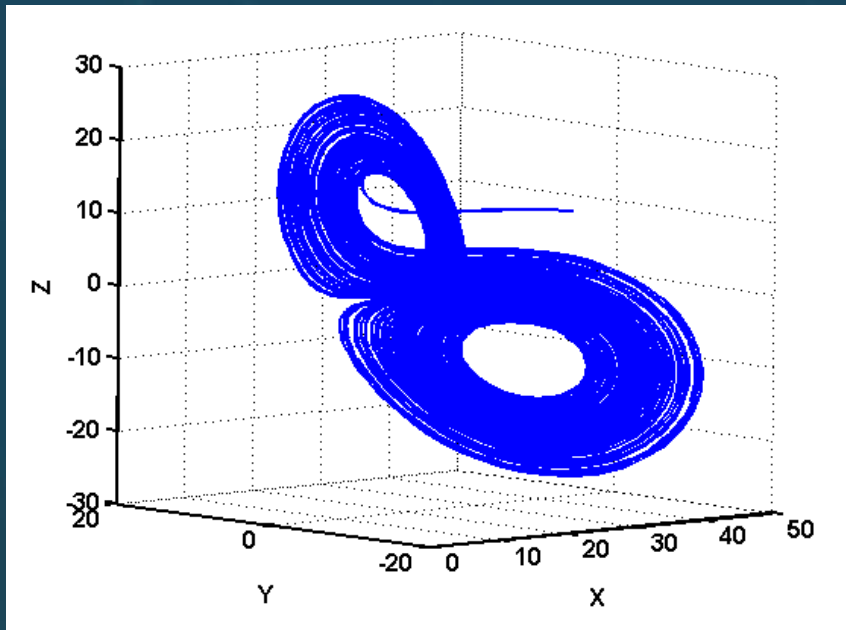
8/03/09 771mw up-ramp from 20:10 - 22:10 followed by a 738mw down-ramp from 22:40 - 00:50



Why is Atmospheric Flow Subject to Uncertainty?



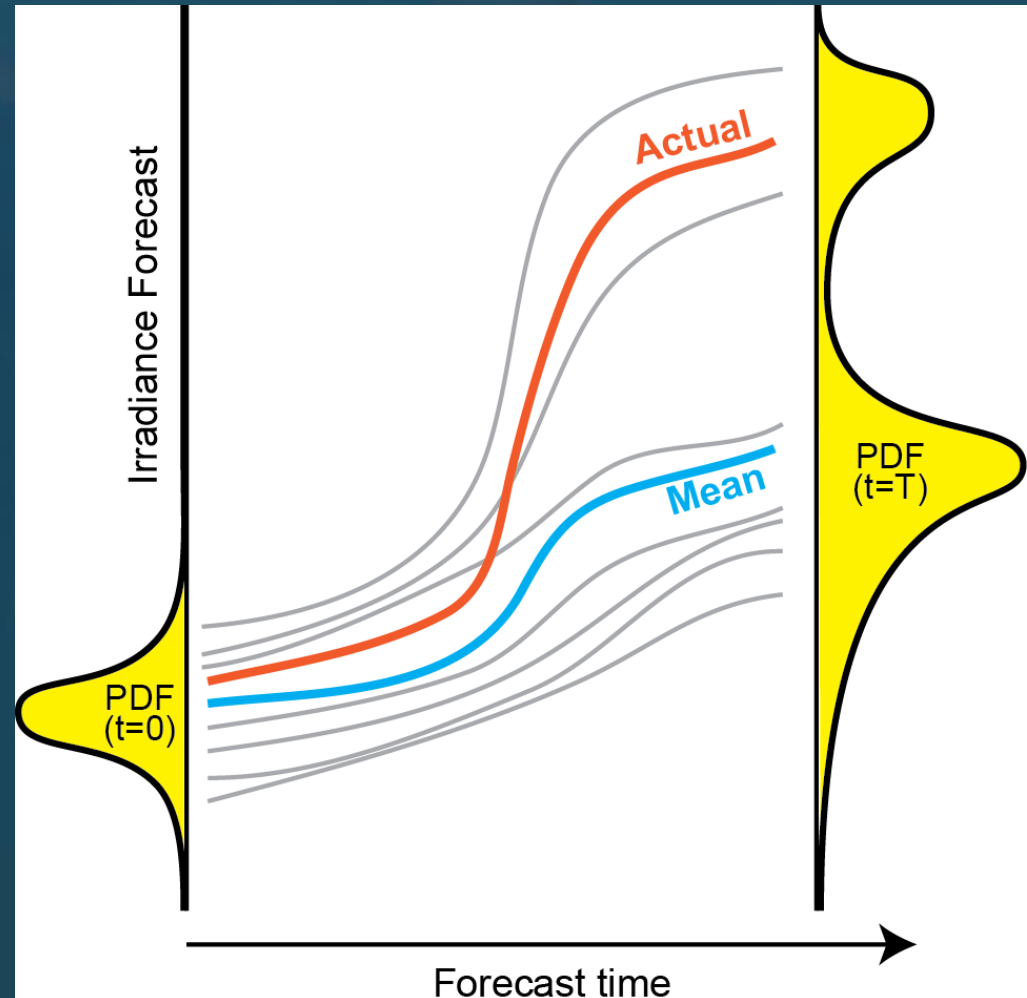
Ed
Lorenz



- Nonlinearity
- Sensitivity to initial conditions
- Chaos → There are limits to predictability
- Think in terms of attractors & manifolds
- Requires probabilistic forecasts

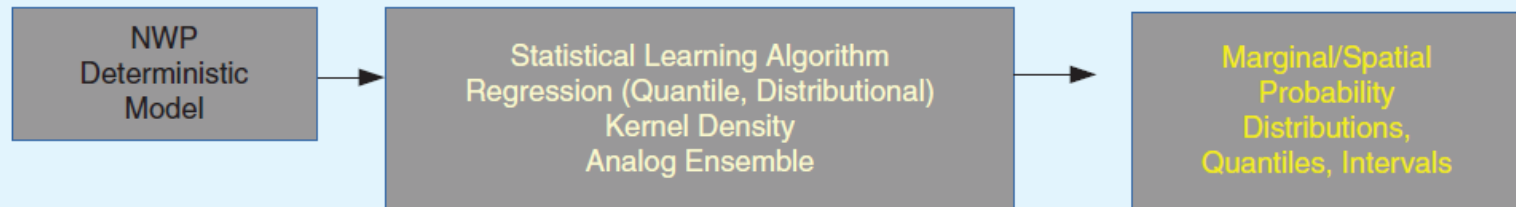
Ensembles & Uncertainty Quantification

- Account for **uncertainties** due to imperfect initial conditions and model formulation
- Produce **more accurate** predictions than any single model realization
- Provide flow-dependent **uncertainty estimates**

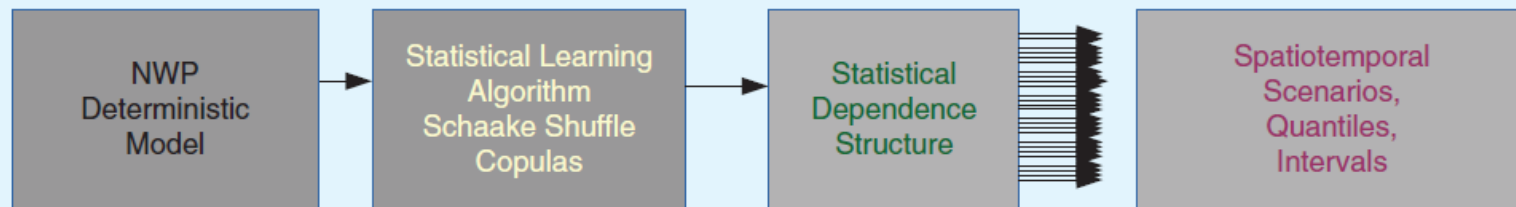


Different Methods of Making Probabilistic Forecasts

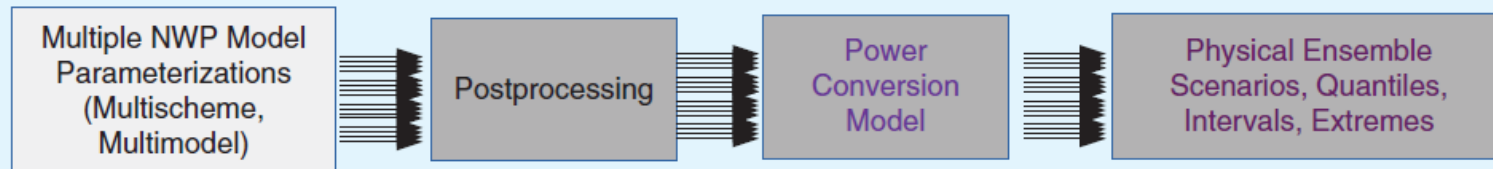
1) Statistical Methods of Probabilistic Forecasts



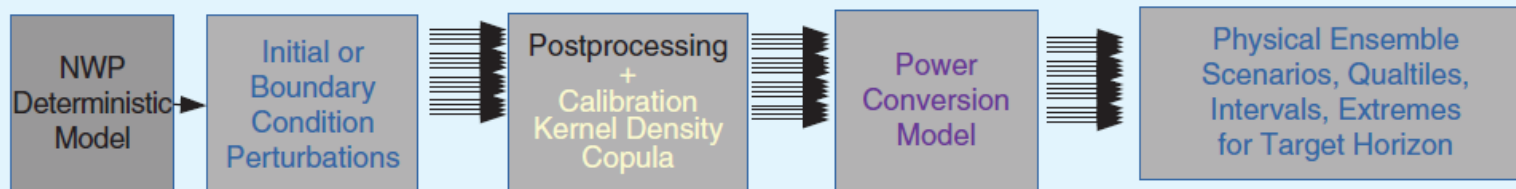
2) Statistically Based Scenarios



3) Physically Based Ensemble Forecasts



4) Perturbation-Based Ensemble Forecasts



Ensemble vs. Statistic (one case)

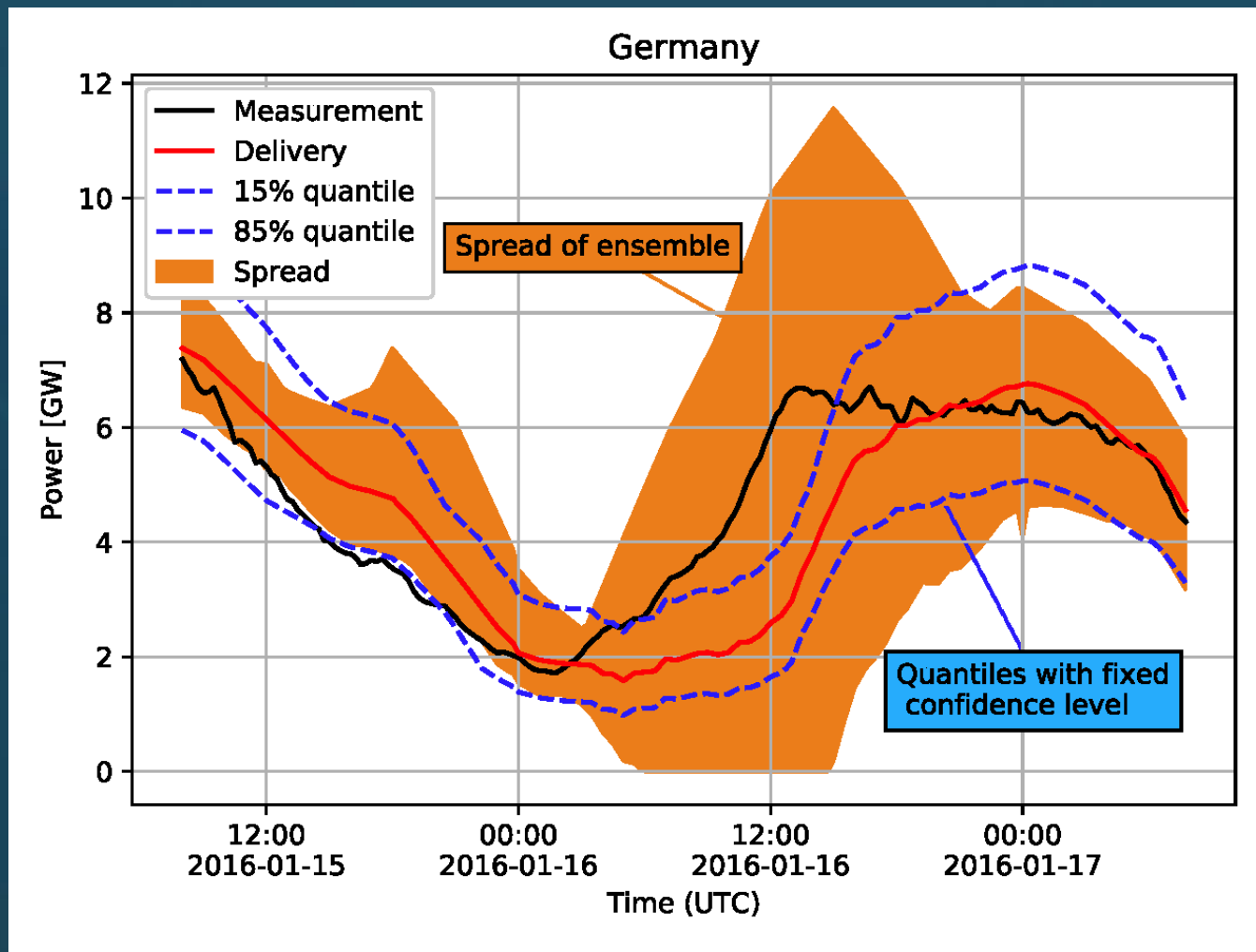
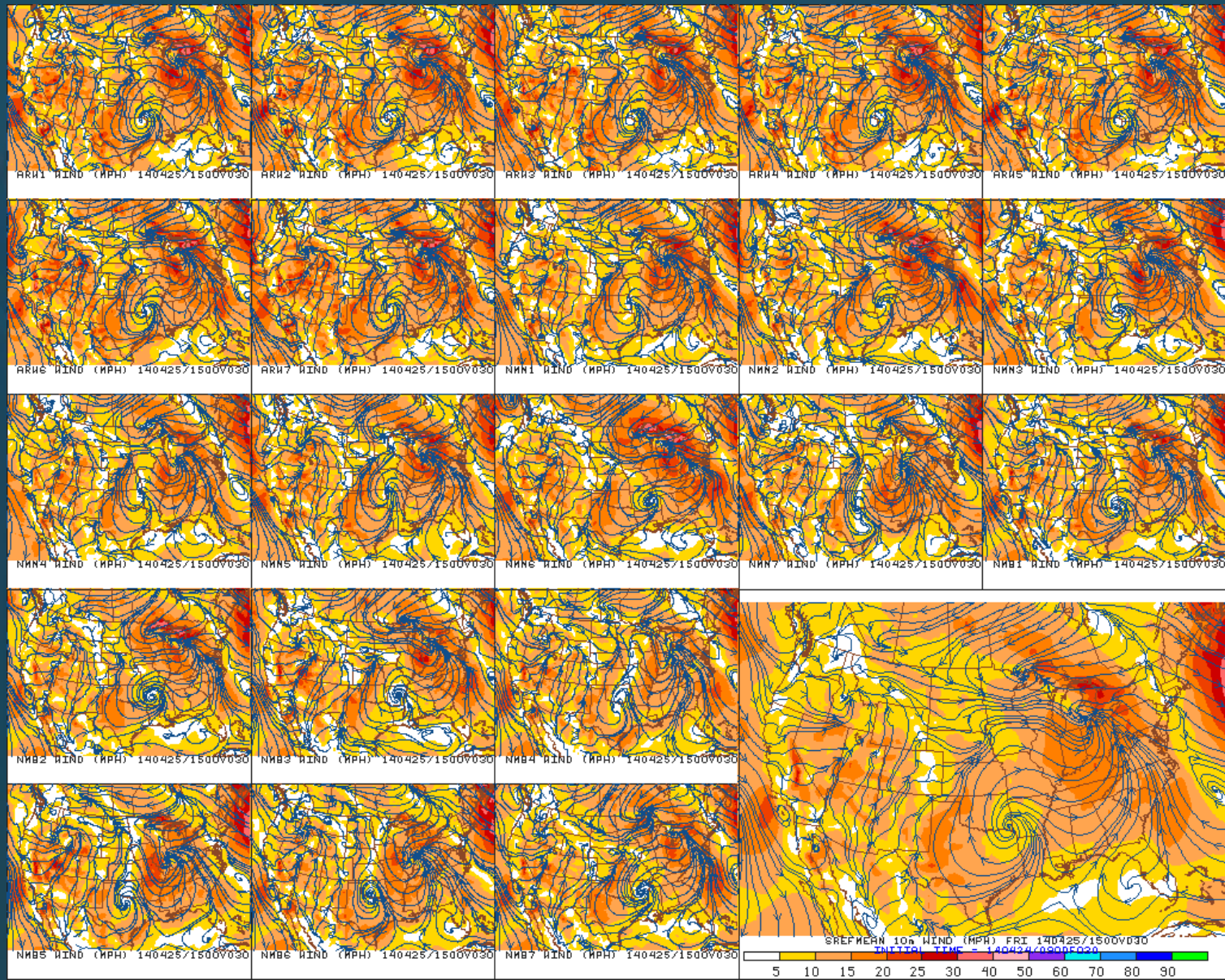


figure 2. A comparison of uncertainty information between a statistical approach based on a single NWP (blue dashed lines) and spread (deviation) of multiple NWP forecasts (orange shading). (Source: Energy & meteo systems.)

Short Range Ensemble Forecast System

30 hr forecast

10 m
Wind
500 mb
Height



How do we determine a Good Match?

- Since probabilistic, need to evaluate based on large number of forecasts

The Brier Score

- Mean square error of a probability forecast

$$BS = \frac{1}{n} \sum_{i=1}^n (f_i - x_i)^2$$

where n is the number of forecasts

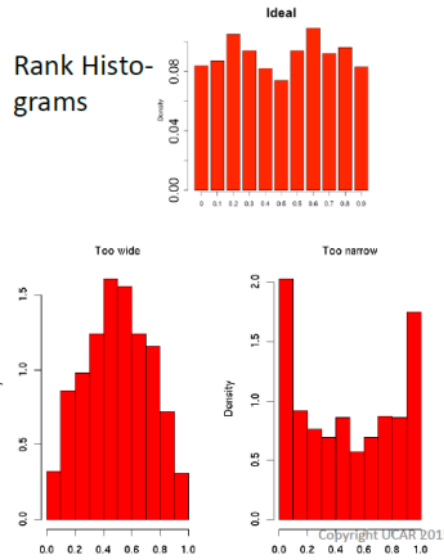
f_i is the forecast prob on occasion i

x_i is the observation (0 or 1) on occasion i

- Weights larger errors more than smaller ones



Rank Histograms



$$BS = \frac{1}{n} \sum_{i=1}^n N_i (f_i - \bar{x})^2 - \frac{1}{n} \sum_{i=1}^n N_i (\bar{x}_i - \bar{x})^2 + \bar{x}(1 - \bar{x})$$

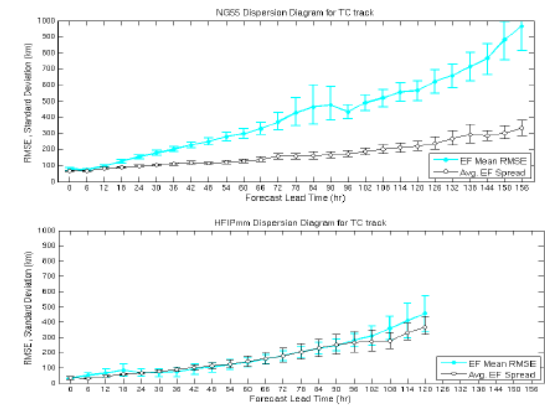
Reliability
Resolution
Uncertainty

Frequency matches actual

Distinguish different events

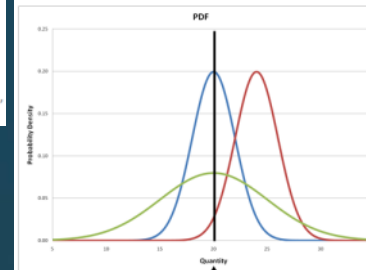
Variability of Observations

Spread-skill

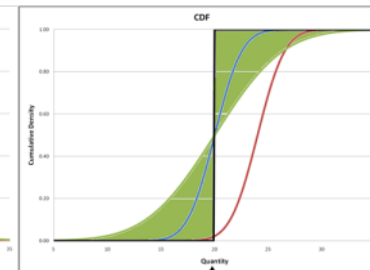


Continuous Ranked Probability Score

Forecast PDFs and Observation



Forecast and Observed CDFs

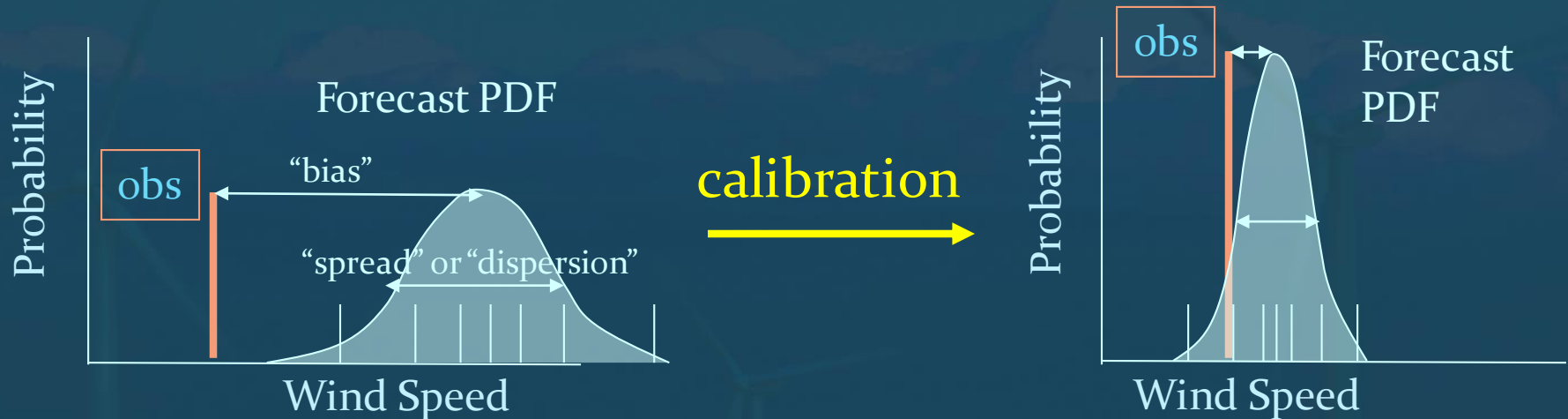


$$CRPS = \int_{-\infty}^{\infty} [F_{fcst}(y) - F_{obs}(y)]^2 dy$$

$F_{fcst}(y)$ CDF of forecast
 $F_{obs}(y)$ CDF of observation

$F_{obs}(y) = 0$ for $y < \text{observed value}$
 $F_{obs}(y) = 1$ for $y \geq \text{observed value}$

How do we improve the Match?

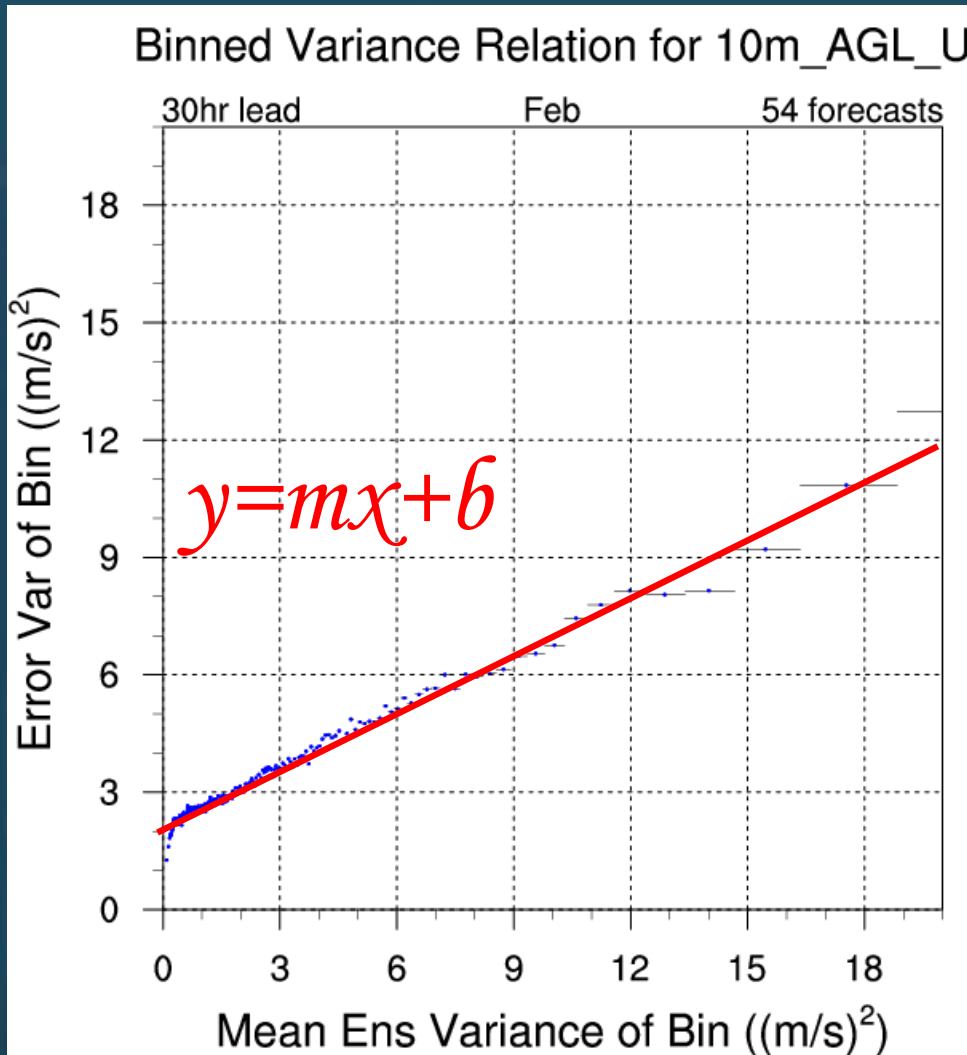


- Centered (the right answer)
- Sharp (narrow range)
- Reliable (quantile predicted matches quantile observed averaged over time)

Example Calibration

Example Calibration Techniques

- Linear Variance Calibration
- Ensemble Kalman Filter
- Quantile Regression
- Bayesian Model Averaging
- Kernel Density Methods
- Analogue Method
- Many others, including logistic regression, nonhomogeneous Gaussian regression, EMOS,



Linear Variance Calibration

Kolczynski et al. (2009, MWR)).

Using Probabilistic Forecasts

Topics in Paper

- Used as Input for Grid Security Calculations (German TSOs)
- Used to Fine-Tune Unit Commitment (SPP) – Tim Miller
- Capturing Extreme Conditions (ERCOT) – Pengwei Du
- Forecasting for Extreme Events, such as High-speed Shutdown Risk (Ireland)
- TSO Use (Red Electrica – Spain)

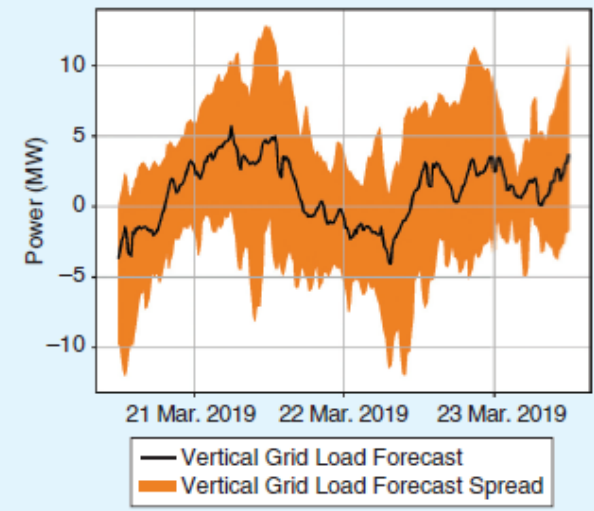


figure 3. A forecast of vertical grid load (solid black line) for one substation between TSO and DSO grids with upper and lower bands indicating uncertainty. Negative values refer to transport to the lower-voltage grid; positive values represent transport to the higher-voltage grid. <AU: Please

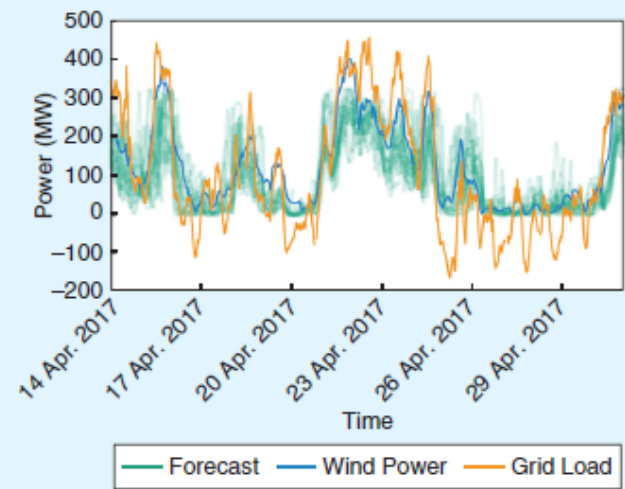
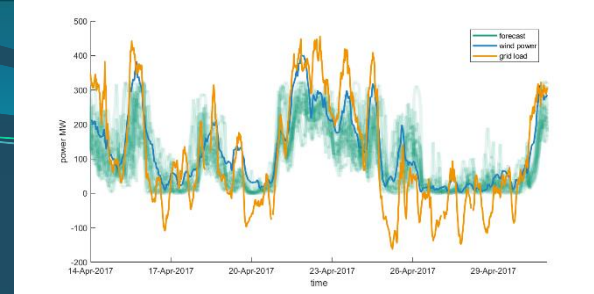


figure 4. An example of a load flow measurement at a transformer between extra-high and high voltage levels. The blue and green curves show, respectively, deterministic and scenario-based forecasts of the wind-power share at this transformer.

Summary



- The atmosphere is inherently Chaotic
- Ensemble prediction embraces and quantifies the uncertainty, producing
 - Better mean forecasts
 - Estimates of uncertainty
- The ensemble should be calibrated
- Research is showing
 - Better ways of creating ensembles
 - Better ways of blending ensemble information via postprocessing

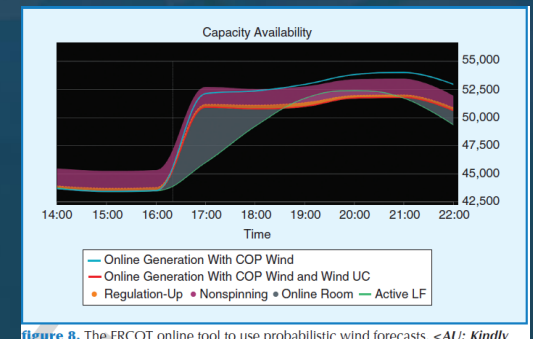


figure 8. The ERCOT online tool to use probabilistic wind forecasts. <AU: Kindly

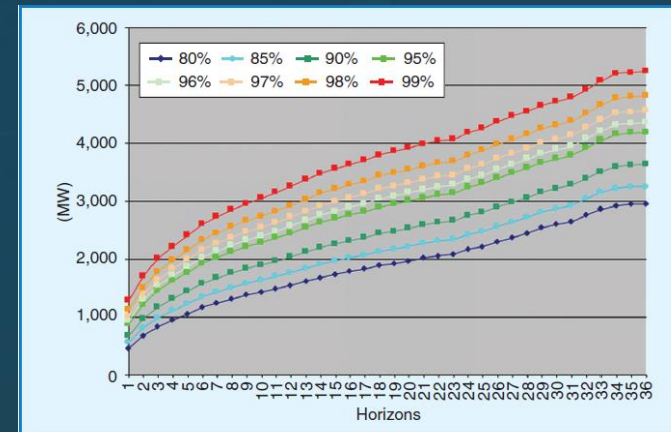


figure 12. A graph from the REE system of probabilistic reserves at different time horizons and different confidence intervals. <AU: Please note that the raw image file

- Such probabilistic forecasts can enhance decision-making



MAINTAINING AWARENESS OF UNCERTAINTY

OCTOBER 2019 ESIG WEBINAR

*Helping our members work together to
keep the lights on... today and in the
future.*



SouthwestPowerPool



SPPorg

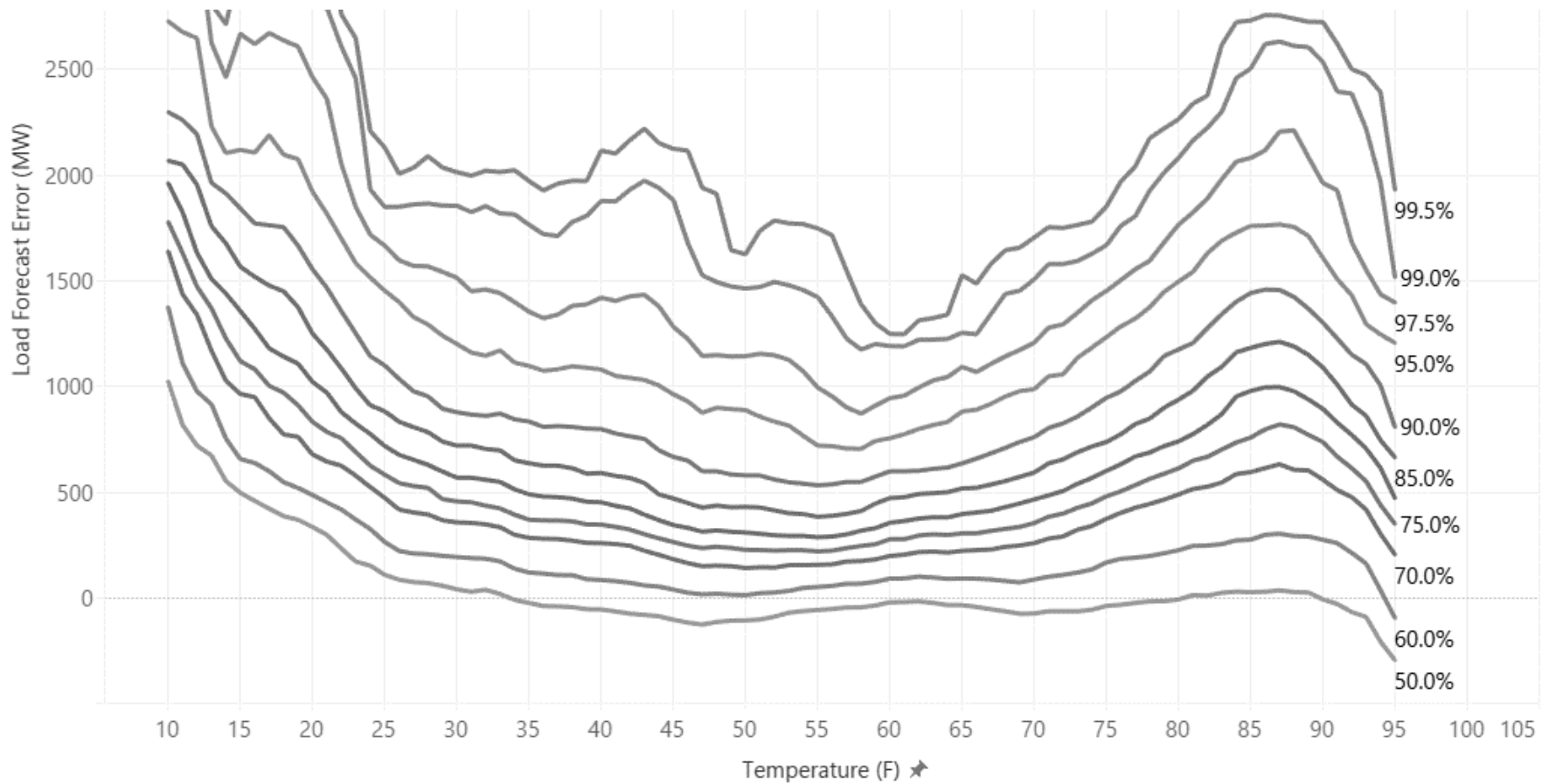


southwest-power-pool

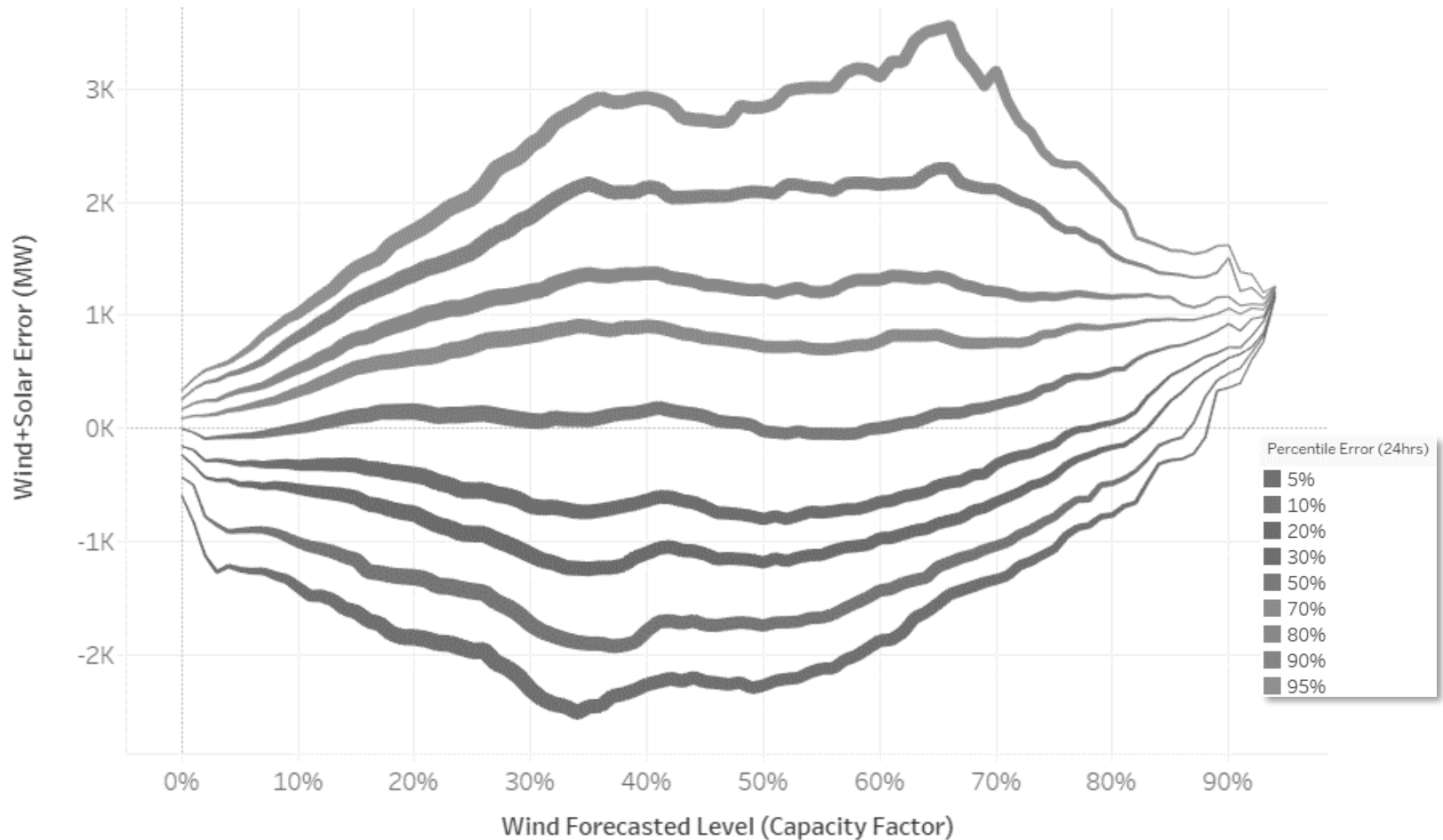
SPP'S FOCUS

- Ensure adequate capacity in real-time to manage longer-term uncertainties
 - Load forecasts
 - Variable resource forecasts
 - Resource capacity plans

LOAD FORECAST ERROR



WIND & SOLAR FORECAST ERROR



EXAMPLE HIGH RISK SCENARIO REPORT

- 1, 4, & 8 Hour Horizons (assessing potential errors and available resources over each horizon)
- Flags potential capacity deficiencies over the next 7 days triggering further investigation
- $\text{URT Capacity Margin} = \text{Available Capacity} + \text{RampableHeadroom} - \text{Wind Error} - \text{Load Error} - \text{Resource Error}$

EXAMPLE HIGH RISK SCENARIO REPORT

Capacity Margin (Initial Assessment)

Un-utilized capacity in each horizon

Uncertainty Capacity Margin:

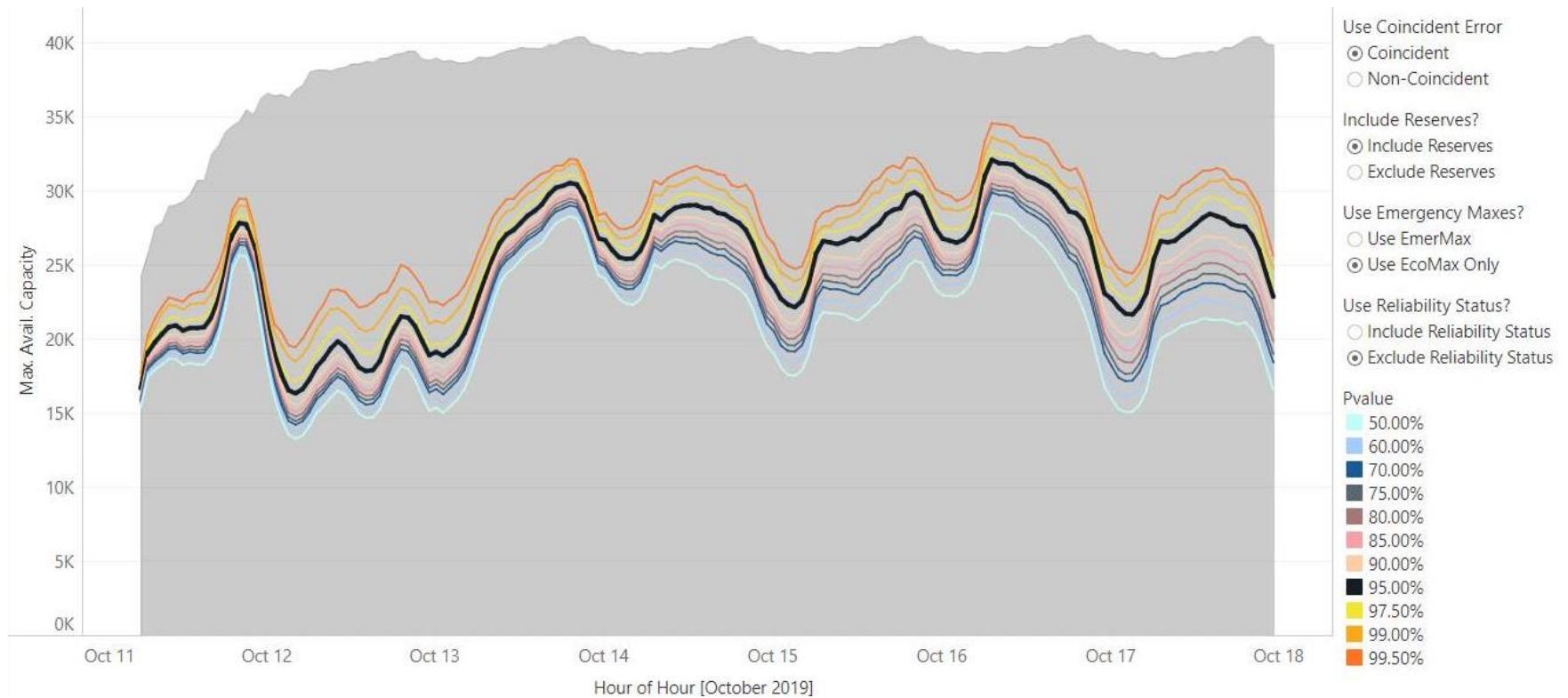
| Value | Thu, Nov-08-2018 | Fri, Nov-09-2018 | Sat, Nov-10-2018 | Sun, Nov-11-2018 | Mon, Nov-12-2018 | Tue, Nov-13-2018 | Wed, Nov-14-2018 | Alert Threshold |
|-----------------|------------------|------------------|------------------|------------------|------------------|------------------|------------------|-----------------|
| URT Cap. Margin | 2197 | 722 | 2564 | 3327 | 3091 | 103 | 1368 | 2000 |
| Rampable HR | 1945 | 0 | 214 | 586 | 959 | 0 | 70 | ???? |
| Capacity | 2864 | 3908 | 5548 | 5460 | 4849 | 3117 | 4390 | ???? |
| Wind Error | 976 | 1373 | 1386 | 1084 | 1081 | 1202 | 1280 | ???? |
| Load Error | 1016 | 1035 | 1035 | 1016 | 1016 | 1035 | 1035 | ???? |
| Resource Error | 619 | 777 | 777 | 619 | 619 | 777 | 777 | ???? |
| Market Interval | 11/08/2018 18:00 | 11/09/2018 06:00 | 11/10/2018 07:00 | 11/11/2018 06:00 | 11/12/2018 18:00 | 11/13/2018 06:00 | 11/14/2018 06:00 | |

Combined Net Load Error

UNCERTAINTY CAPACITY MARGIN

| Day of Hour | Hour of Hour | Pvalue | | | | | | | | | | |
|---------------------|-----------------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| | | 50.00% | 60.00% | 70.00% | 75.00% | 80.00% | 85.00% | 90.00% | 95.00% | 97.50% | 99.00% | 99.50% |
| October 11, 2019 | 6 | 8,856 | 8,659 | 8,456 | 8,338 | 8,216 | 8,063 | 7,872 | 7,567 | 7,248 | 6,899 | 6,554 |
| | 7 | 8,344 | 8,075 | 7,834 | 7,701 | 7,565 | 7,375 | 7,127 | 6,741 | 6,347 | 5,917 | 5,557 |
| | 8 | 9,625 | 9,342 | 9,064 | 8,916 | 8,748 | 8,538 | 8,272 | 7,775 | 7,314 | 6,744 | 6,183 |
| | 9 | 9,628 | 9,331 | 9,014 | 8,861 | 8,664 | 8,452 | 8,113 | 7,555 | 7,029 | 6,157 | 5,734 |
| | 10 | 10,337 | 9,990 | 9,681 | 9,494 | 9,294 | 9,060 | 8,735 | 8,176 | 7,576 | 6,691 | 6,184 |
| | 11 | 10,425 | 10,042 | 9,710 | 9,527 | 9,312 | 9,076 | 8,733 | 8,155 | 7,564 | 6,837 | 6,409 |
| | 12 | 11,042 | 10,618 | 10,277 | 10,086 | 9,872 | 9,617 | 9,265 | 8,720 | 8,146 | 7,318 | 6,784 |
| | 13 | 11,376 | 10,978 | 10,621 | 10,430 | 10,216 | 9,941 | 9,573 | 9,004 | 8,406 | 7,410 | 6,779 |
| | 14 | 12,425 | 12,022 | 11,656 | 11,465 | 11,233 | 10,943 | 10,540 | 9,956 | 9,403 | 8,305 | 7,553 |
| | 15 | 12,453 | 12,032 | 11,649 | 11,441 | 11,203 | 10,954 | 10,549 | 9,943 | 9,315 | 8,344 | 7,514 |
| | 16 | 13,598 | 13,158 | 12,754 | 12,538 | 12,285 | 12,018 | 11,617 | 11,005 | 10,401 | 9,404 | 8,519 |
| | 17 | 13,021 | 12,563 | 12,152 | 11,905 | 11,646 | 11,368 | 10,982 | 10,412 | 9,923 | 8,972 | 8,193 |
| | 18 | 12,011 | 11,565 | 11,142 | 10,892 | 10,654 | 10,389 | 9,970 | 9,461 | 8,960 | 8,413 | 7,671 |
| | 19 | 9,741 | 9,366 | 8,992 | 8,769 | 8,527 | 8,290 | 7,969 | 7,392 | 6,928 | 6,259 | 5,730 |
| | 20 | 9,042 | 8,686 | 8,318 | 8,114 | 7,897 | 7,664 | 7,323 | 6,846 | 6,341 | 5,671 | 5,195 |
| | 21 | 9,922 | 9,543 | 9,177 | 8,977 | 8,757 | 8,515 | 8,224 | 7,757 | 7,222 | 6,496 | 6,060 |
| | 22 | 11,133 | 10,732 | 10,318 | 10,139 | 9,894 | 9,625 | 9,335 | 8,843 | 8,364 | 7,655 | 7,362 |
| | 23 | 15,075 | 14,602 | 14,137 | 13,925 | 13,671 | 13,370 | 13,046 | 12,459 | 11,967 | 11,249 | 10,896 |

CAPACITY VS. UNCERTAINTY





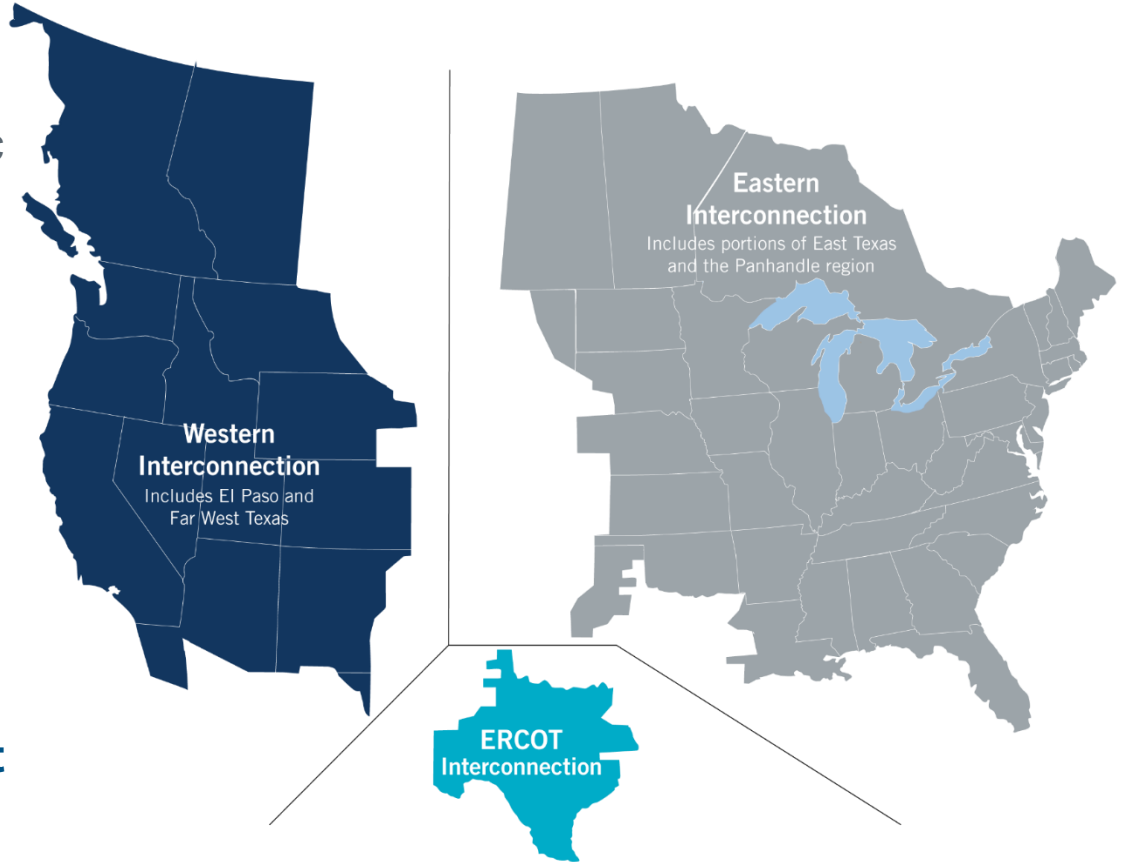
***INTEGRATE
PROBABILISTIC
INFORMATION INTO GRID
OPERATIONS***

Pengwei Du
ERCOT
10/22/2019

What is ERCOT?

The Texas Legislature restructured the Texas electric market in 1999 and assigned ERCOT four primary responsibilities:

- **System Reliability**
- **Competitive Wholesale Market**
- **Open Access to Transmission**
- **Competitive Retail Market**



ERCOT is a nonprofit organization and regulated by the Public Utility Commission of Texas, with oversight by the Texas Legislature.

ERCOT is not a market participant and does not own generation or transmission/distribution wires.

Current Records

Peak Demand Record: 74,666 megawatts (MW)*

- August 12, 2019, 4-5 p.m.

Weekend Peak Demand Record: 71,915 MW*

- Sunday, August 11, 2019, 5-6 p.m.

Winter Peak Demand Record: 65,915 MW

- Jan. 17, 2018, 7-8 a.m.

Wind Generation Records (instantaneous)

- Output: 19,672 MW
 - Jan. 21, 2019, 7:19 p.m.
- Penetration (load served): 56.16%
 - January 19, 2019, 3:10 a.m.
 - Total MW Served by Wind = 17,406 MW

Recent Monthly Peak Demand Records

2019

- March: 60,756 MW (March 5, 7-8 a.m.)
- August: 74,666 MW (Aug. 12, 4-5 p.m.)*
- September: 68,817 MW (Sept. 6, 4-5 p.m.)*

2018

- January: 65,915 MW (Jan. 17, 7-8 a.m.)
- May: 67,265 MW (May 29, 4-5 p.m.)
- June: 69,123 MW (June 27, 4-5 p.m.)
- July: 73,473 MW (July 19, 4-5 p.m.)
- November: 56,317 MW (Nov. 14, 7-8 a.m.)

2017

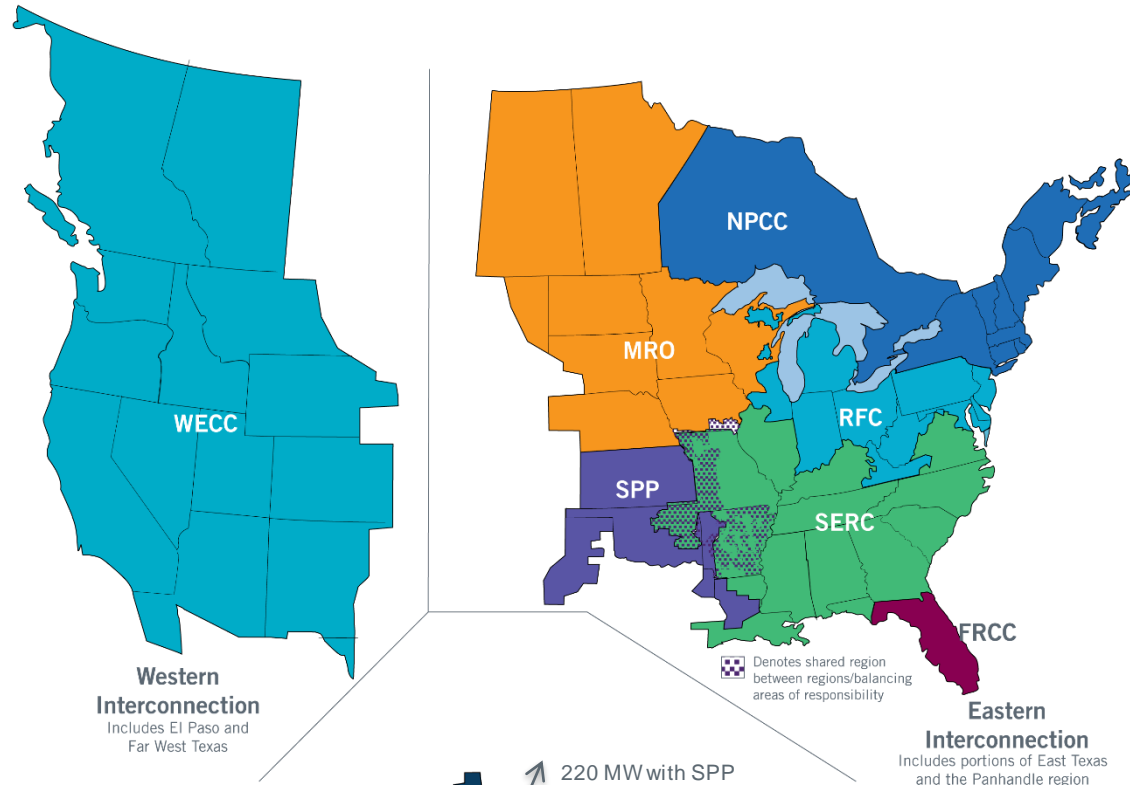
- April: 53,486 MW (April 28, 4-5 p.m.)
- October: 62,333 MW (Oct. 9, 4-5 p.m.)

**New records are preliminary, subject to change in final settlement*

The ERCOT Region

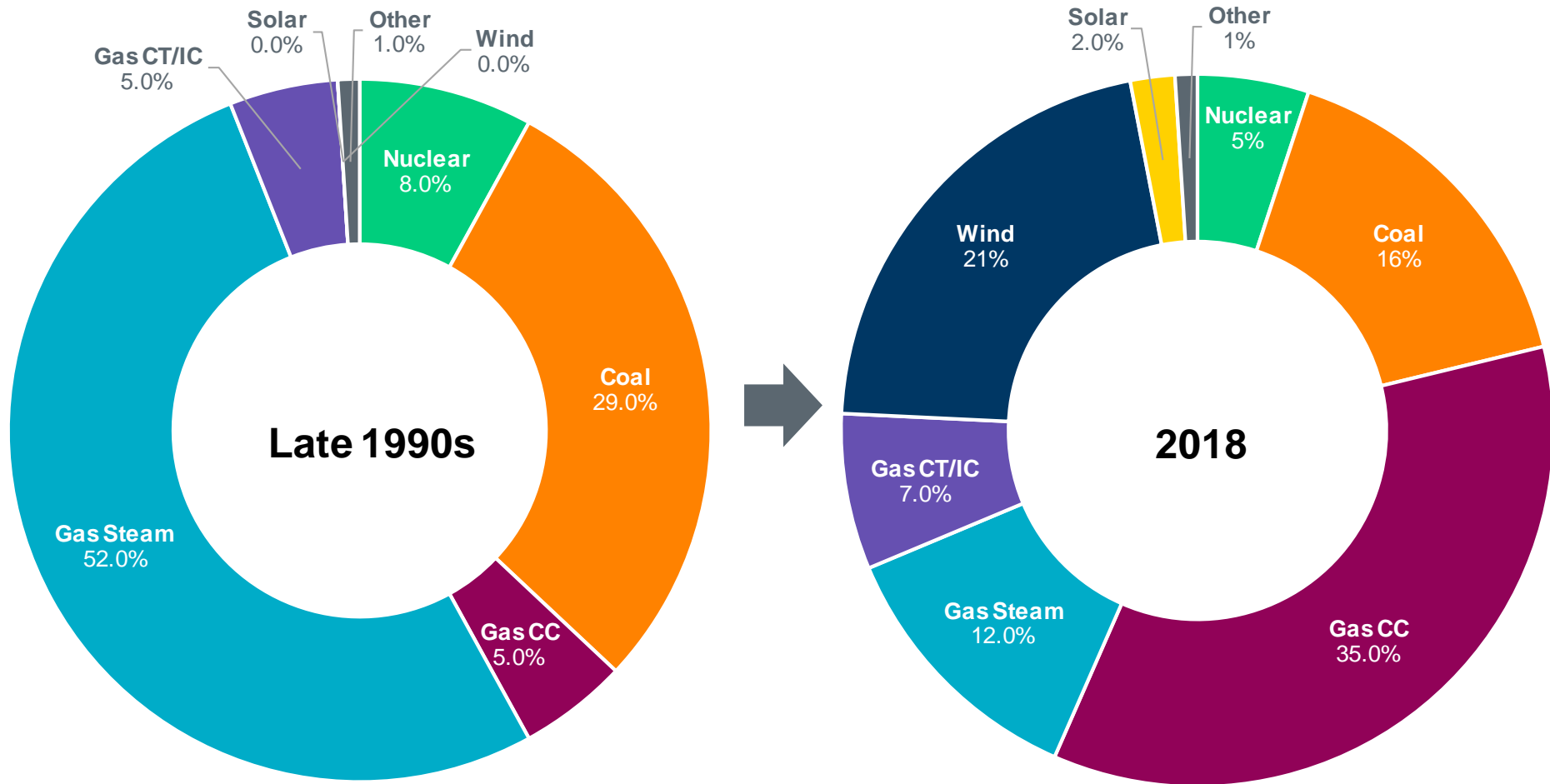
The interconnected electrical system serving most of Texas, with limited external connections

- 90% of Texas electric load; 75% of Texas land
- More than 46,500 miles of transmission lines
- 600+ generation units



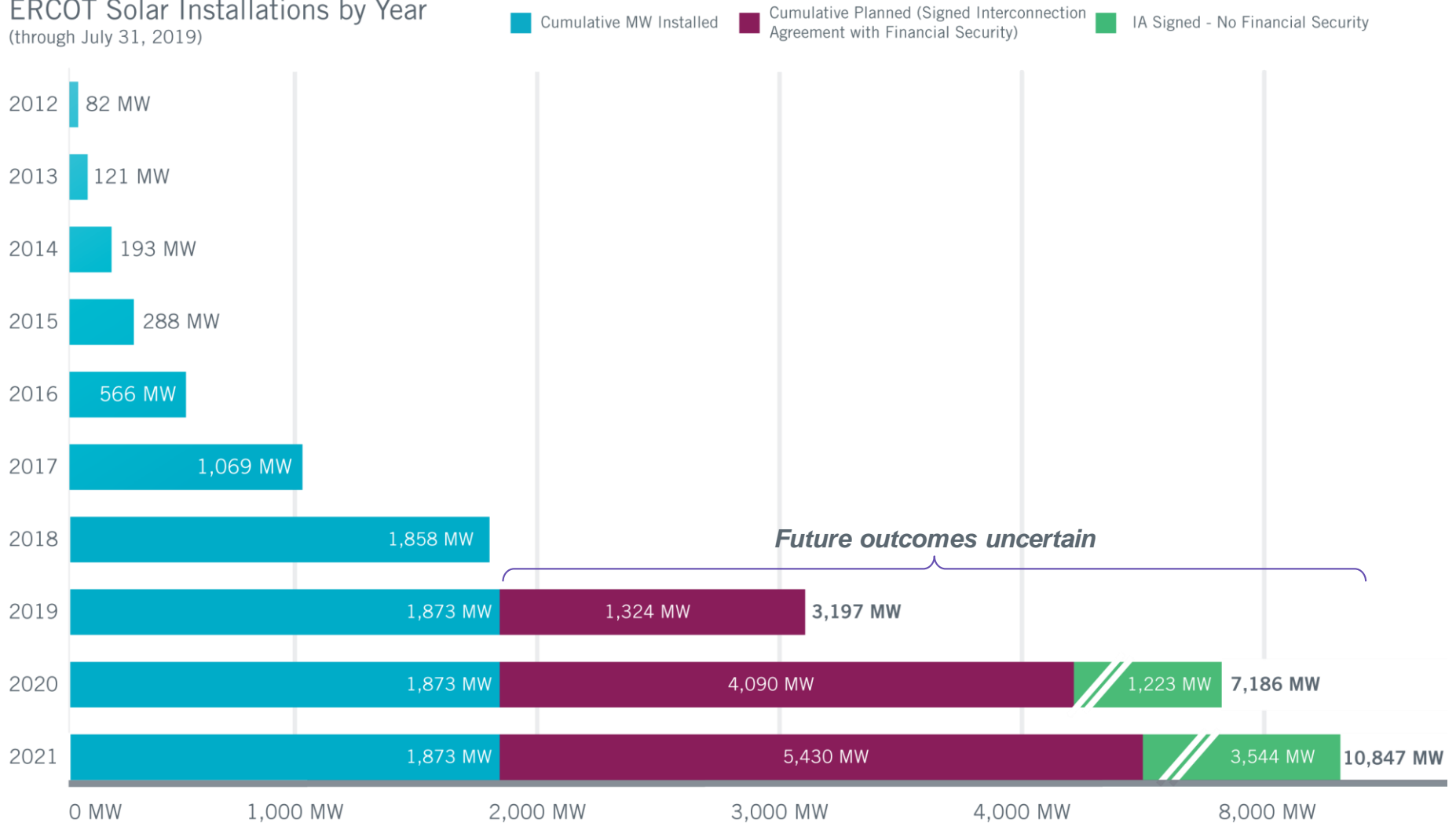
ERCOT connections to other grids are limited to ~1250 MW of direct current (DC) ties, which allow control over flow of electricity

Changing Resource Capacity Mix



Utility Scale Solar Generation Capacity – July 2019

ERCOT Solar Installations by Year
(through July 31, 2019)



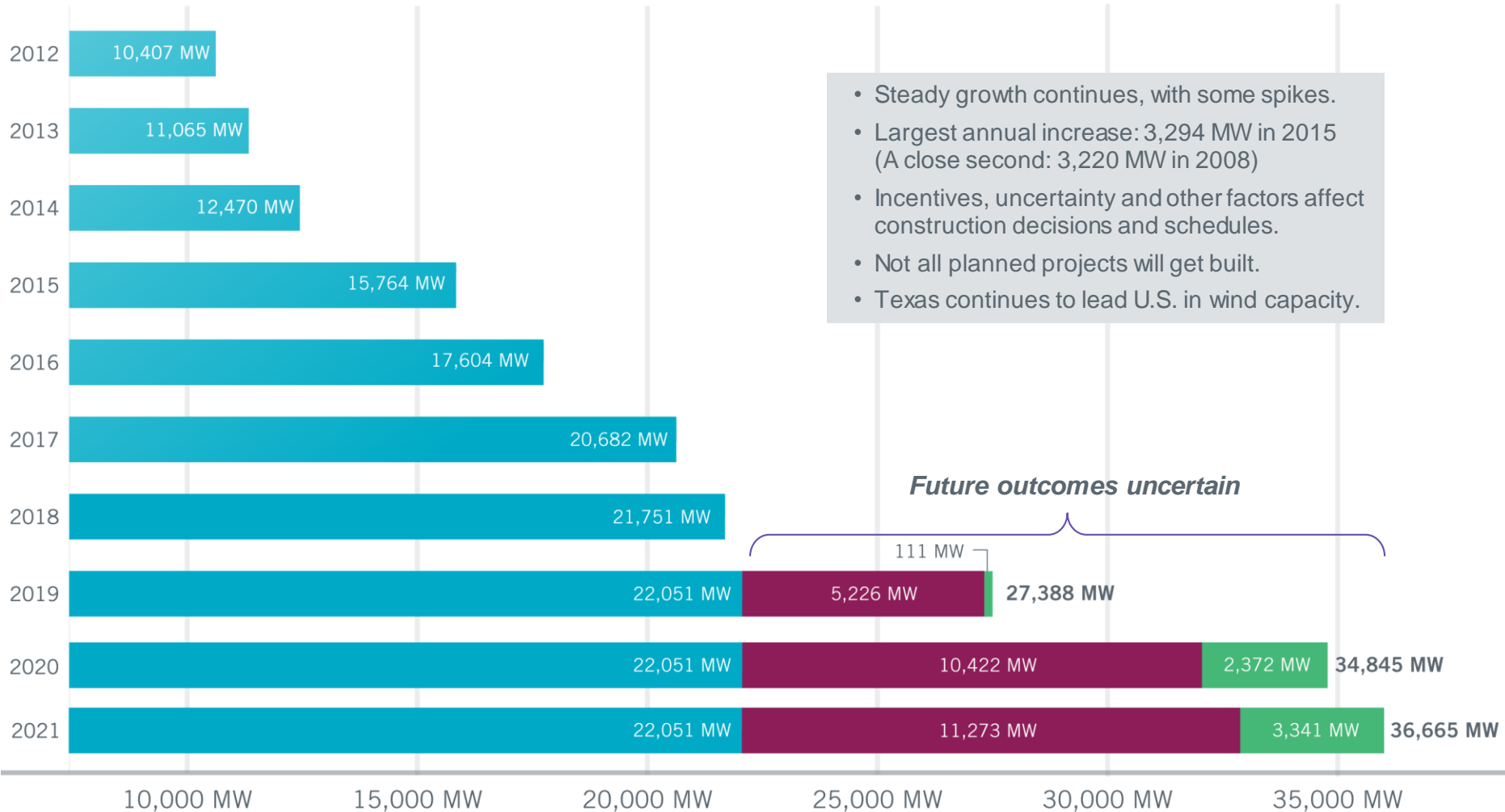
Wind Generation Capacity – July 2019

ERCOT Wind Installations by Year
(through July 31, 2019)

Cumulative MW Installed

Cumulative Planned (Signed Interconnection Agreement with Financial Security)

IA Signed - No Financial Security



Wind Forecasting at ERCOT

WIND GENERATION RESOURCE (WGR)

- Asset Registration Information
- Outage/De-rate Schedule
- Real Time Telemetry



WIND FORECAST SERVICE

- Two Forecast Service Providers each provide:
- Hourly Forecast for each WGR for each hour of next week, updated hourly
- Intra-hour Forecast for next 2-hr, at 5-min resolution, updated every 15-min
- Extreme Event Forecasts and Probabilistic Forecasts, used for risk assessment

energy & meteo
systems



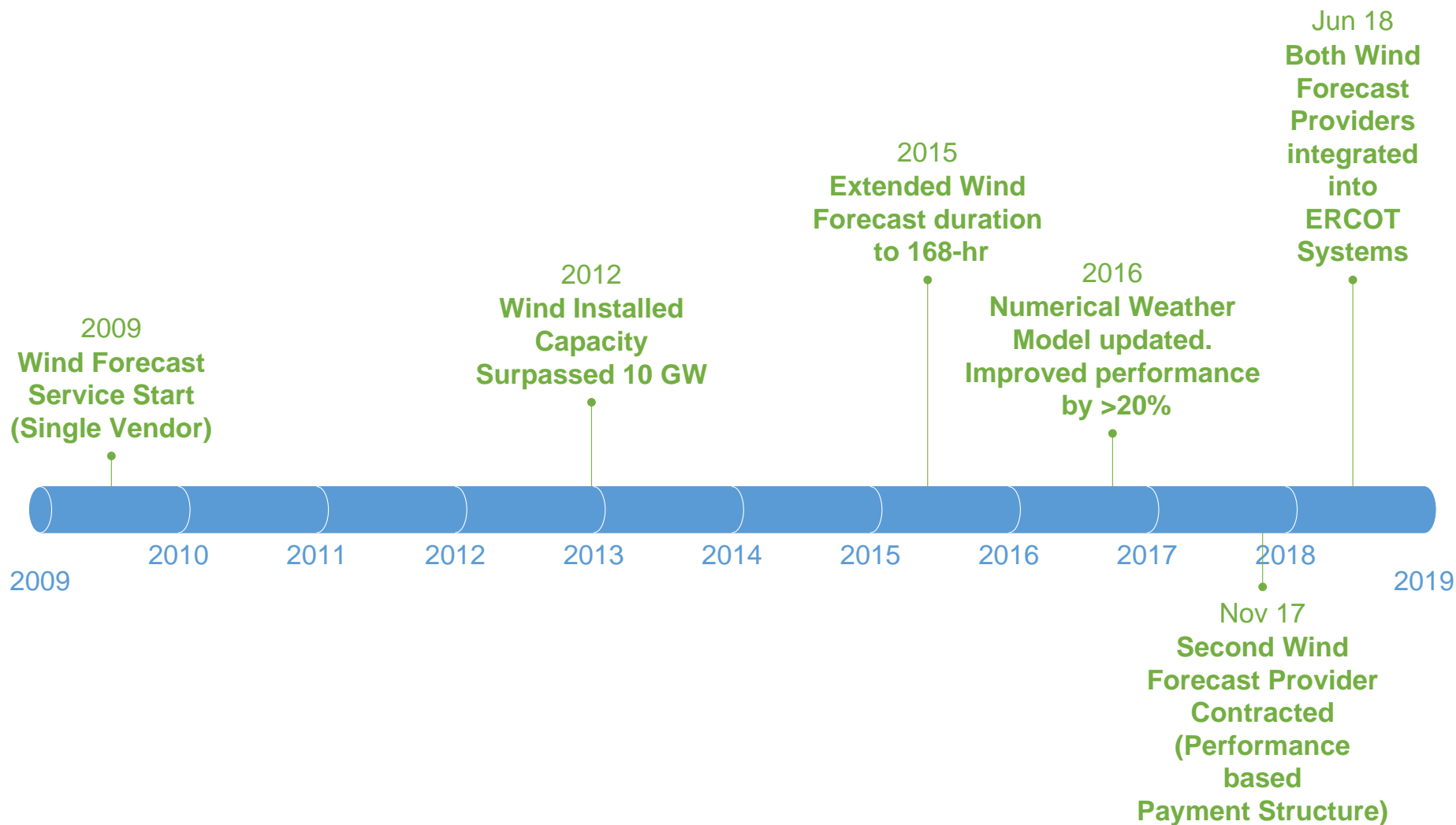
ERCOT SYSTEMS

- ERCOT Control Room can select “active forecast” for every hour and every 5-min.
- **Hourly Forecast** Used in all Look Ahead Studies (ex. Outage Coordination, Reliability Unit Commitment, Next Day Study)
- **Intra-hour Forecast** To be used in Real Time Dispatch
- Extreme Event Forecasts and Probabilistic Forecasts, used for risk assessment



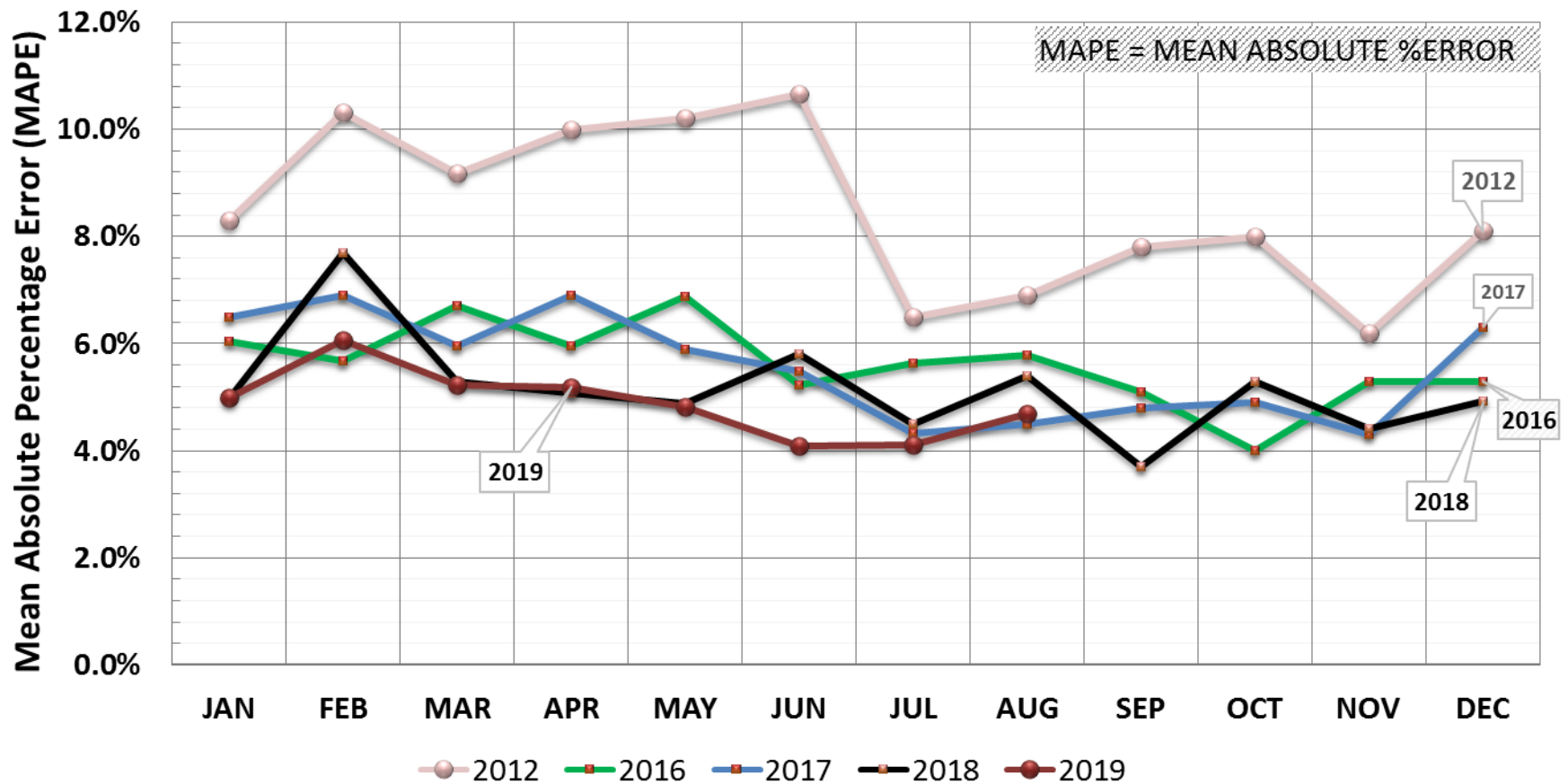
* WGR Operators can reduce the Current Operating Plan (COP) High Sustained Limit (HSL) to reflect operating condition changes.

Timelines of Wind Forecast



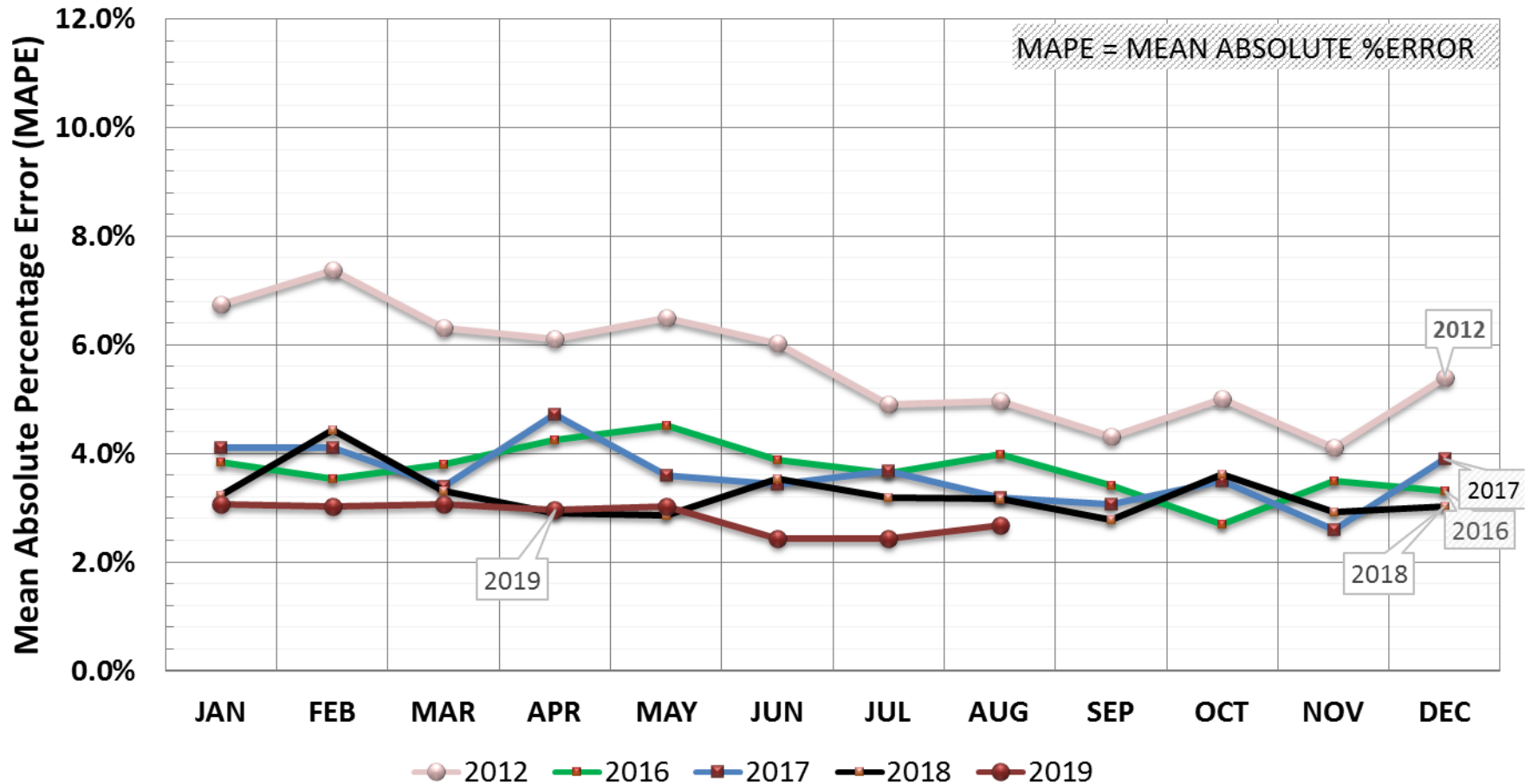
Wind Forecast Errors (Day-Ahead)

DAY AHEAD WIND FORECAST PERFORMANCE



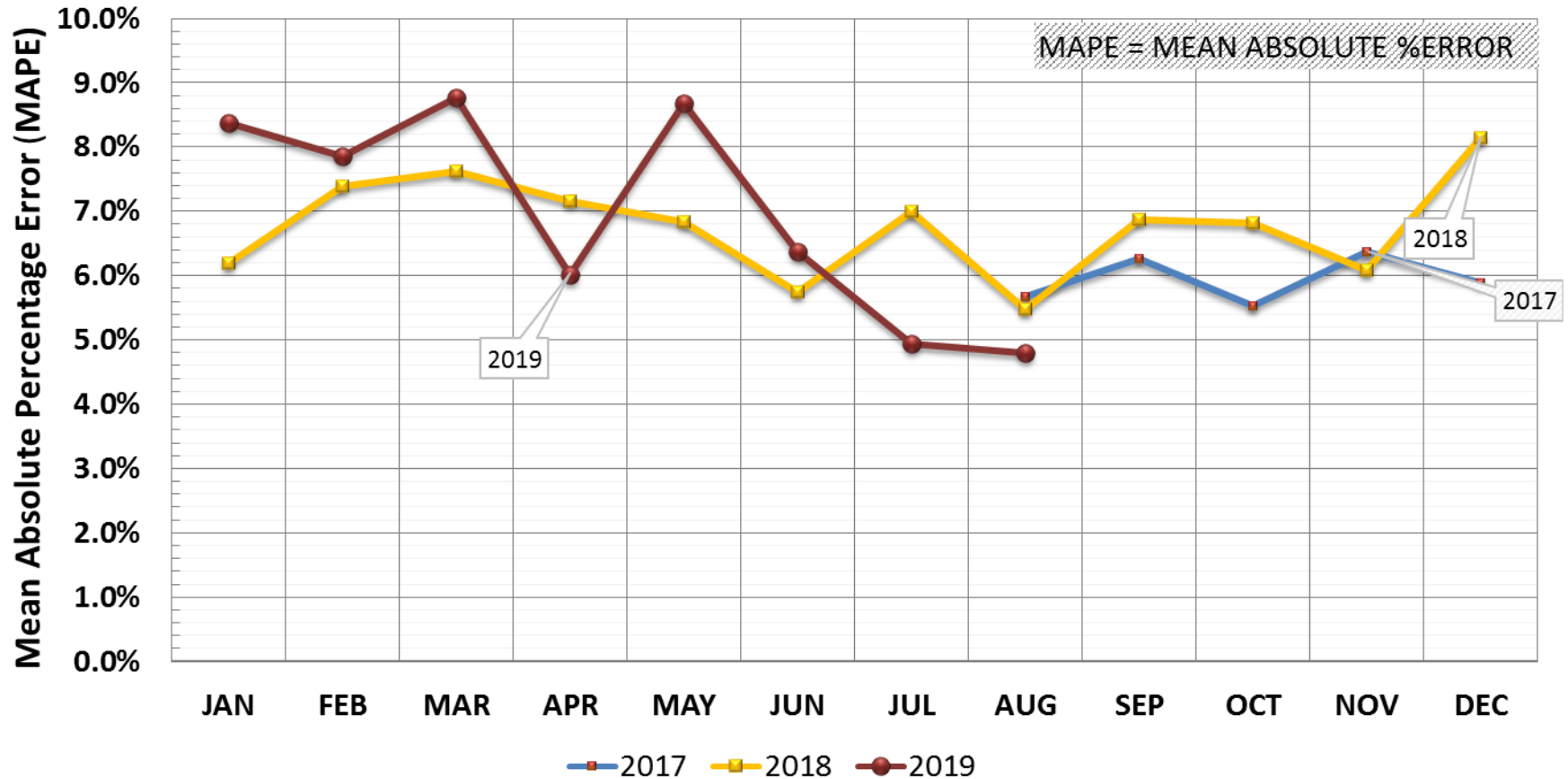
Wind Forecast Errors (Hour-Ahead)

HOUR AHEAD WIND FORECAST PERFORMANCE



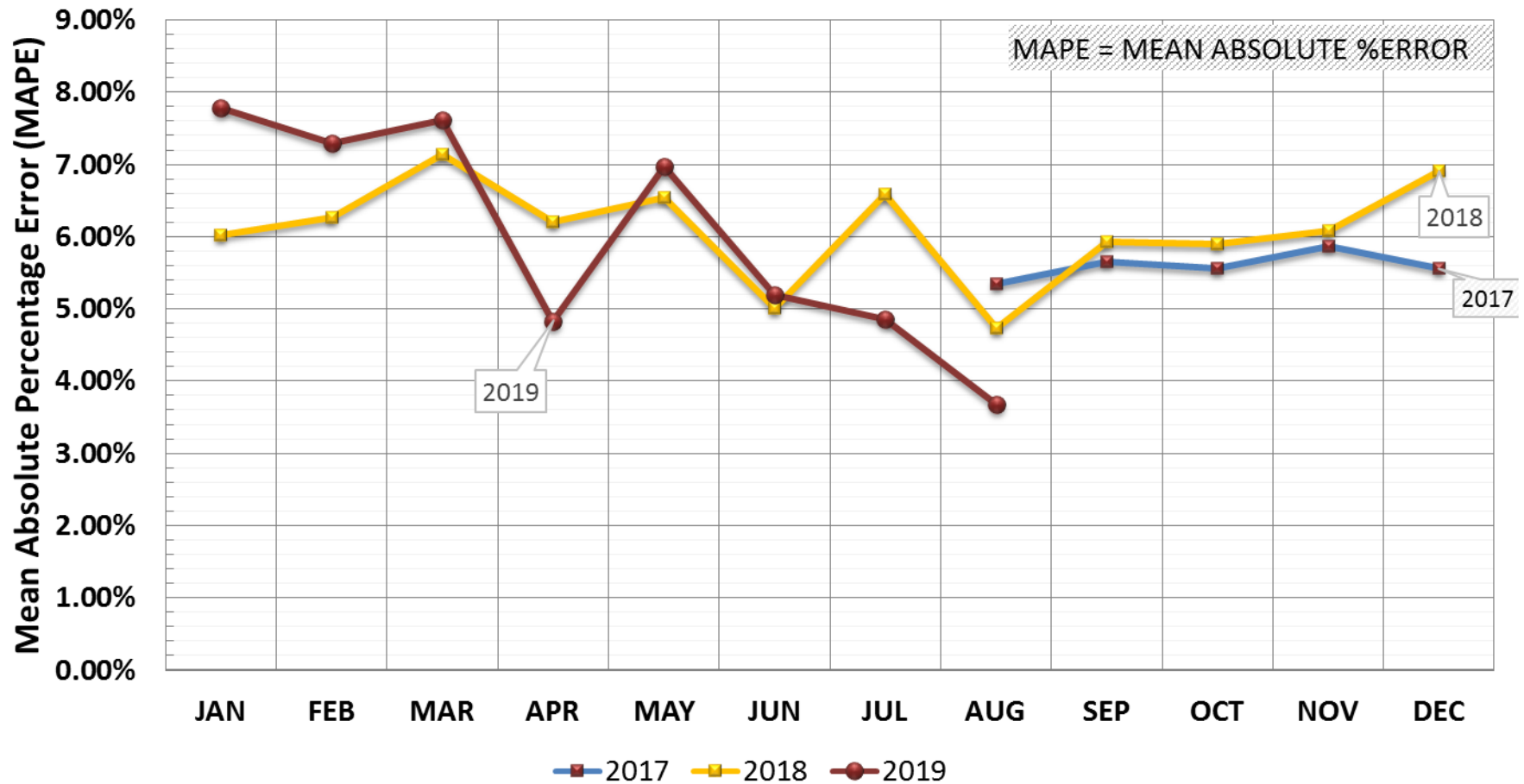
PVGR Forecast Errors (Day-Ahead)

DAY AHEAD PVGR FORECAST PERFORMANCE



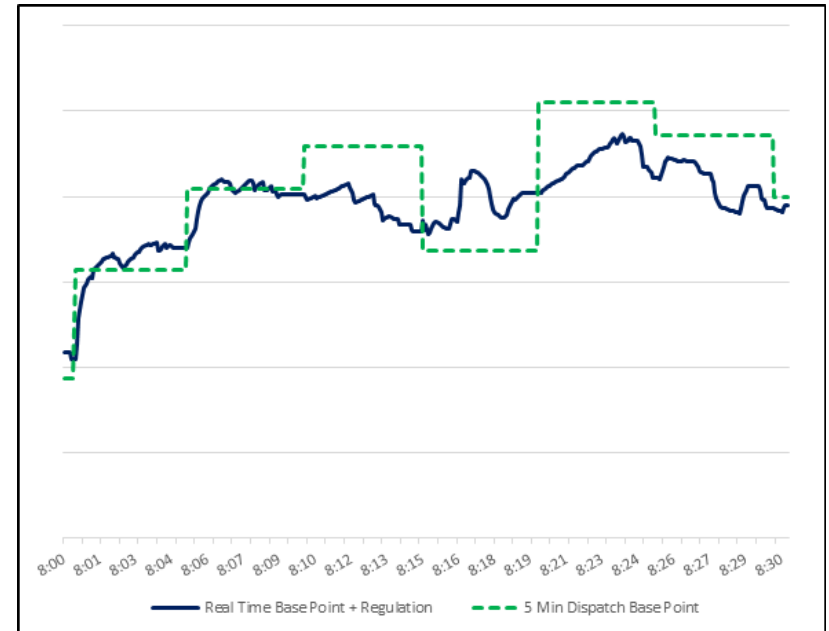
PVGR Forecast Errors (Hour-Ahead)

Hour Ahead PVGR Forecast Performance



Ancillary Services – Regulation Service

- SCED sends dispatch instruction every 5 minutes.
- LFC sends Regulation signals to generation resources every 4 second service to move up or down to maintain frequency at nominal.



Ancillary Services – Non-Spinning Reserve Service (NSRS)

- Non-spinning Reserve Service
 - 30 minute product that can be provided by unloaded capacity, offline Generators, and Load Resources
 - Wind power and load forecast error is one of the inputs used for calculating the requirement for this service
 - The net forecast error uncertainty that NSRS will be used to mitigate will depend on the risk of net load ramp.
 - During higher net load ramp risk hours, we procure NSRS to cover up to 95th percentile of net forecast uncertainty.
 - During low net load ramp risk hours, we may only procure NSRS to cover 70th percentile of net forecast uncertainty.

Integrating and Managing Renewables

- With increasing integration of renewables, large variations in renewable generation are possible to occur more frequently.
- It is essential to monitor and identify durations when large variations in renewable generation are expected and take actions that preposition the rest of ERCOT's generation fleet in order to maintain reliable grid operations.
 - Adding a new desk in the Control Room to provide increased situational awareness to renewables [2016]
 - Created displays to identify incorrect information telemetered for renewables and mitigate these.
 - Developed capability to monitor inertia and sufficiency of procured Responsive Reserve requirements in Real Time and Day Ahead.
 - Built tools for assessing adequacy of available generation capacity and Non Spin reserves for serving variability expected in net load near Real Time.

Capacity Availability Tool

Capacity Availability Tool (CAT)

Show Actuals: ☒ Use Available Gen: ☒ Use STWPF: ☒ Use LF UC: ☒

Wind: 1h - 3h Percentile: 4h - 6h Percentile: Load: 1h - 3h Percentile: 4h - 6h Percentile: [REFRESH](#)

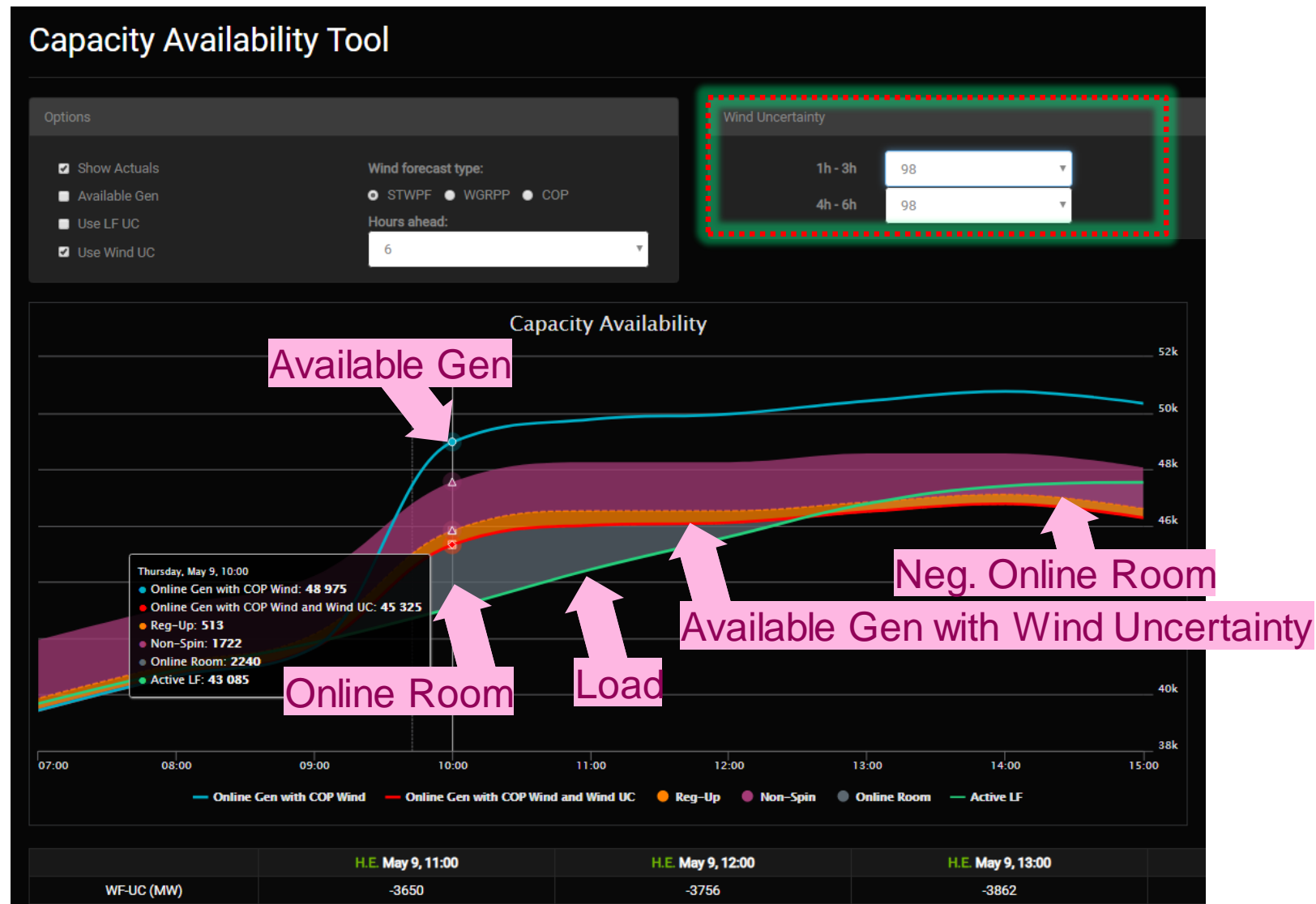
Display 1 Display 2 Display 3

Chart

IOHR: 50 Items

| ID | Time | DC Tie Sched | Reg Up | Non Spin | Active LF | Active LF with Actual | Load UC (%) | Load UC (MW) | Active LF with LF UC | COP Wind | Wind Forecast | Actual Wind | Wind UC (%) | Wind UC (MW) |
|-----|----------------------|--------------|--------|----------|-----------|-----------------------|-------------|--------------|----------------------|----------|---------------|-------------|-------------|--------------|
| H-2 | 5/9/2019 8:00:00 AM | -264 | 439 | 2065 | 39489 | 39681.31 | 0 | 0 | 39681.31 | 8544.399 | 8270.698 | 8137.77 | 0 | 0.00 |
| H-1 | 5/9/2019 9:00:00 AM | -214 | 440.1 | 2065 | 40131 | 40865.68 | 0 | 0 | 40865.68 | 8426.8 | 8319.798 | 8749.06 | 0 | 0.00 |
| H00 | 5/9/2019 10:00:00 AM | -164 | 479 | 2065 | 41516 | 41702.14 | 0 | 0 | 41702.14 | 8751.199 | 8662.103 | 8621.32 | 0 | 0.00 |
| H01 | 5/9/2019 11:00:00 AM | 11 | 513 | 1722 | 43085 | 43085 | 0.0226 | 973.721 | 44058.72 | 8651.302 | 8782.401 | | -0.1378544 | -3.566.78 |
| H02 | 5/9/2019 12:00:00 PM | 111 | 515 | 1722 | 44445 | 44445 | 0.0269 | 1195.571 | 45640.57 | 8814.997 | 8950.598 | | -0.1262395 | -3.266.26 |
| H03 | 5/9/2019 1:00:00 PM | 84 | 431 | 1722 | 45607 | 45607 | 0.0312 | 1422.938 | 47029.94 | 8867.4 | 9004.102 | | -0.1146247 | -2.965.75 |
| H04 | 5/9/2019 2:00:00 PM | 86 | 338.9 | 1722 | 46782 | 46782 | 0.03386667 | 1584.35 | 48366.35 | 8992.9 | 9130.299 | | -0.1276528 | -3.302.83 |
| H05 | 5/9/2019 3:00:00 PM | 89 | 341.9 | 1453 | 47410 | 47410 | 0.03653333 | 1732.045 | 49142.05 | 9115.804 | 9252.304 | | -0.1406809 | -3.639.91 |
| H06 | 5/9/2019 4:00:00 PM | 63 | 331 | 1453 | 47542 | 47542 | 0.0392 | 1863.646 | 49405.64 | 9124.101 | 9261.802 | | -0.153709 | -3.976.99 |

Capacity Availability Tool



Percentile

98th

Most
Conservative

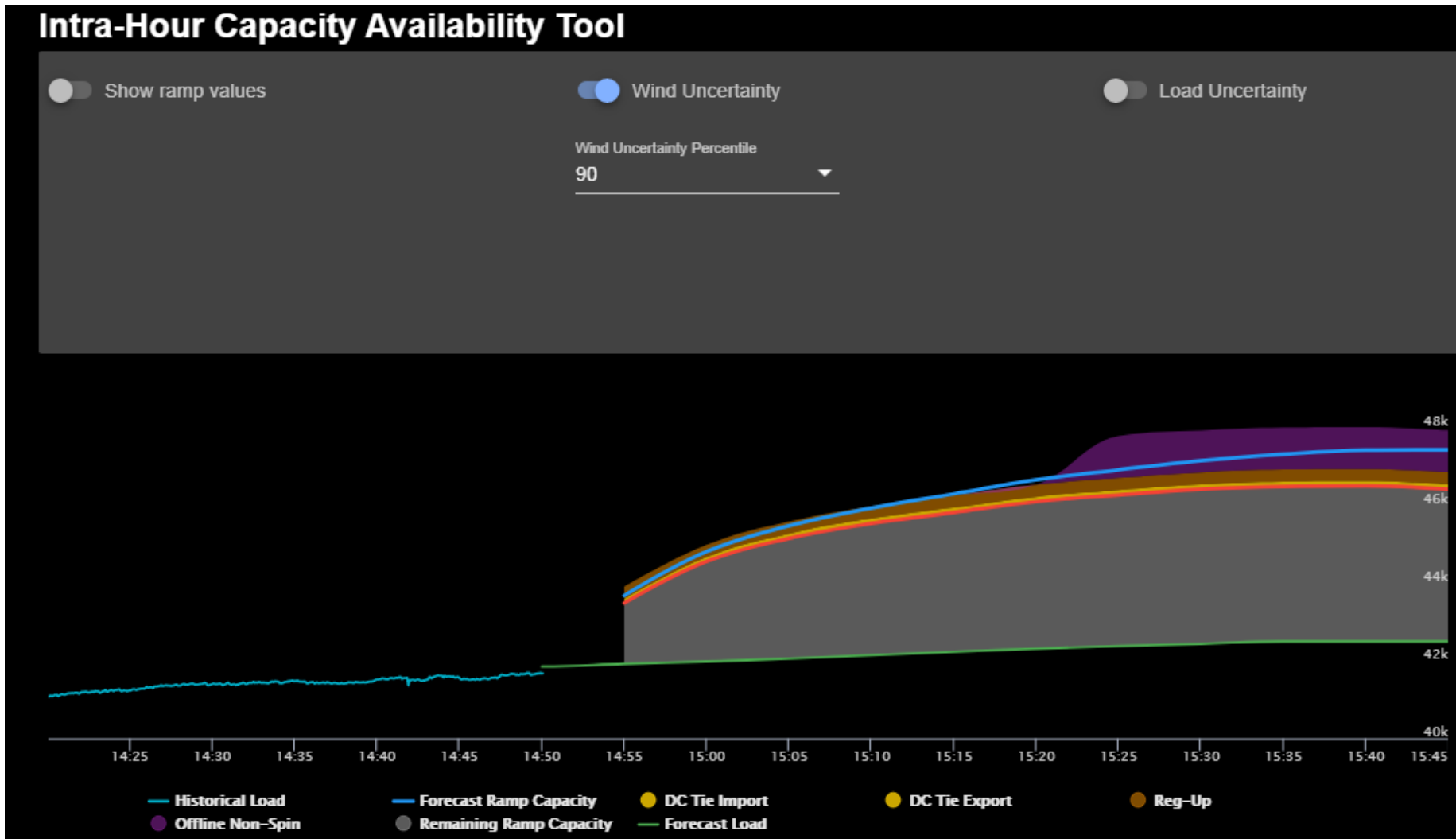
95th

Moderately
Conservative

90th

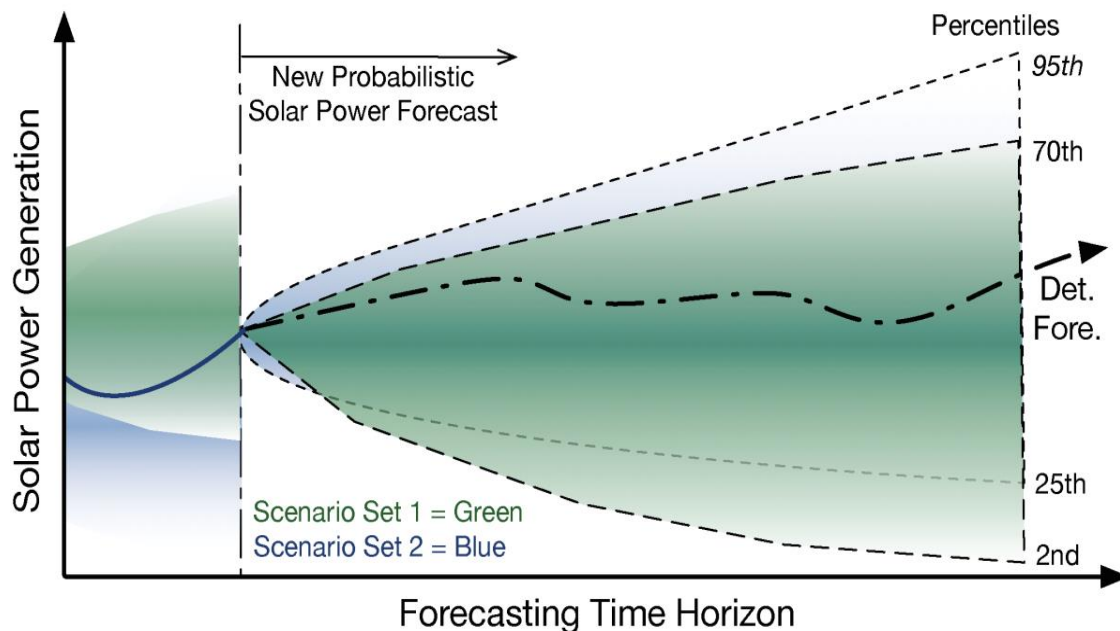
Mildly
Conservative

Operational Tool to Monitor Ramp for Next 1-hour



Solar Uncertainty Management and Mitigation for Exceptional Reliability in Grid Operations (SUMMER-GO)

- SUMMER-GO will bring probabilistic solar forecasts into ERCOT's real-time operation environment through automated reserve and dispatch tools that increase economic efficiency and improve system reliability.
- ERCOT is partnering with NREL, UT Dallas and Maxar for this 3-year project (2018-2020) funded by the Department of Energy





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