Providing Uncertainty Information to End Users in the Electric Sector

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VAISALA

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Tutorial on Integration of Uncertainty Forecasts into Power System Operations

Talk Overview

Probability Forecasts

information of the control o

best-estimate
optimal-bid
information
confusing
rule-of-thumb
overwhelming
worst-case-scenario
risk-adjusted



2

Barriers to Adoption

Community effort needed with a 3-pronged strategy

- Policy
- Technology Demonstration
- Education



Excerpt from Vaisala Webinar (Jan 2015)



Our Efforts

Three examples from private sector use cases





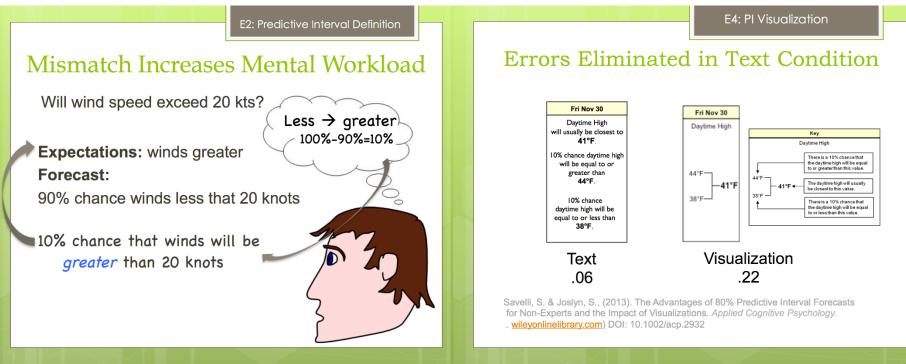


Complex Decision

Process



Human Factors Education



Source: Susan Joslyn, Decision Making Psychology, University of Washington

- Predictive intervals can lead to better decisions and greater trust
- Misinterpretation (reversal errors) quite common
- Cognitive psychology research shows that framing of language and mode of communication are critically important

Example 1: To Power Marketers

Submitting Offers into the Day-Ahead Electricity Market

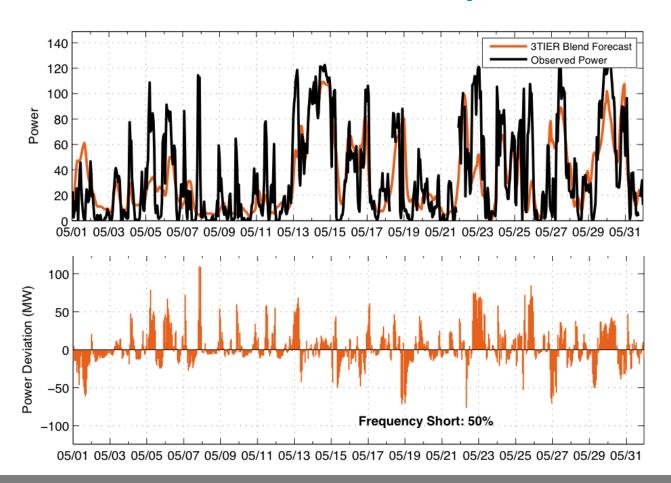


Common Energy Scheduling Practices

- Passing through a deterministic forecast (from ISO or forecast provider)
- "Haircutting" (aka scaling) the deterministic
- Scheduling only what you can cover during high uncertainty periods (e.g., known reserve capacity)

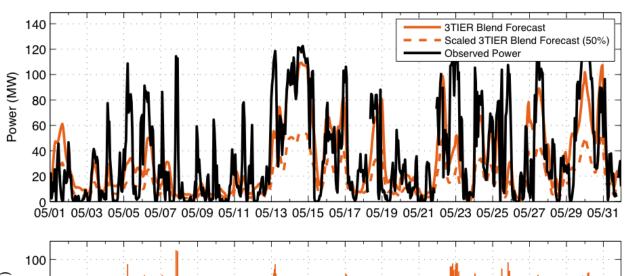
These approaches may help to avoid over-promising generation, but they lose sight of the "upside potential"

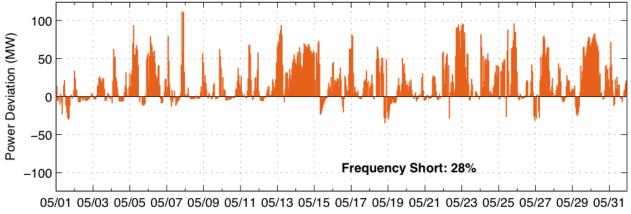
The Problem: Downside Risk Exposure



3TIER Blend minimizes bias and MAE; 50% downside risk

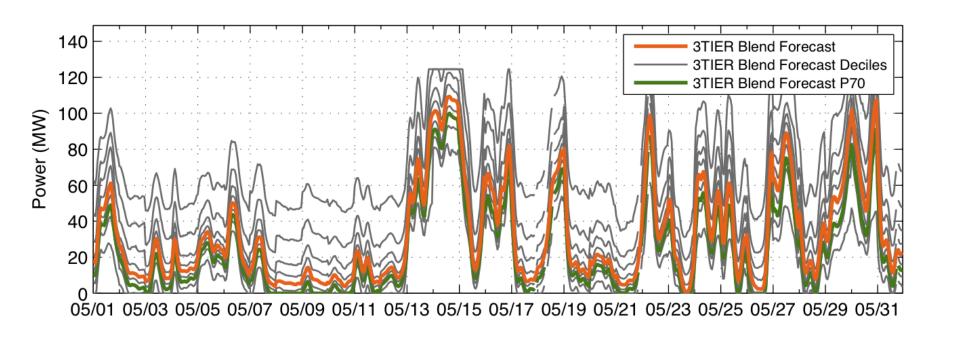
One Strategy: Scaling the Forecast





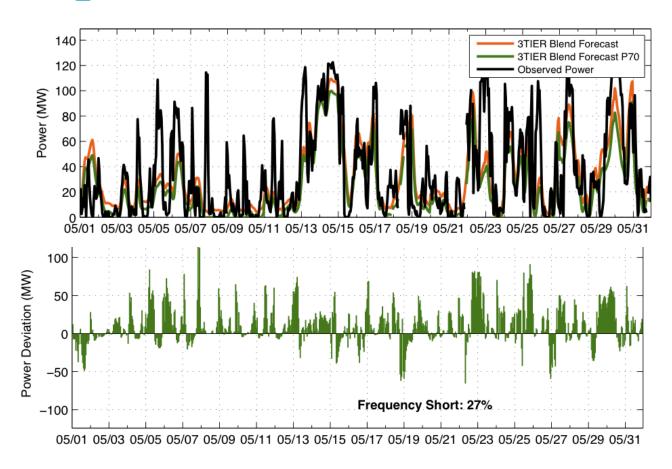
50% scaled 3TIER Blend biased low; 28% downside risk

Another Strategy: Choosing Risk Tolerance



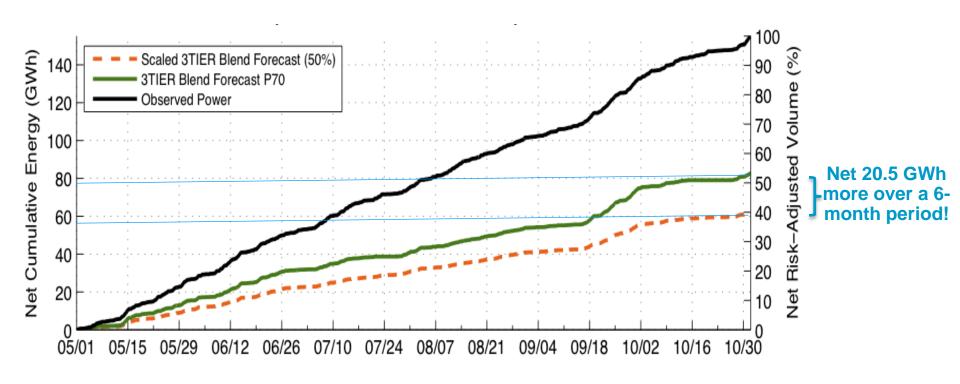
P70 forecast estimates level with 30% downside risk

Checking the Risk Exposure



P70 forecast has 27% downside risk, very close to the scaled forecast risk

Comparing the Two Strategies Over Time



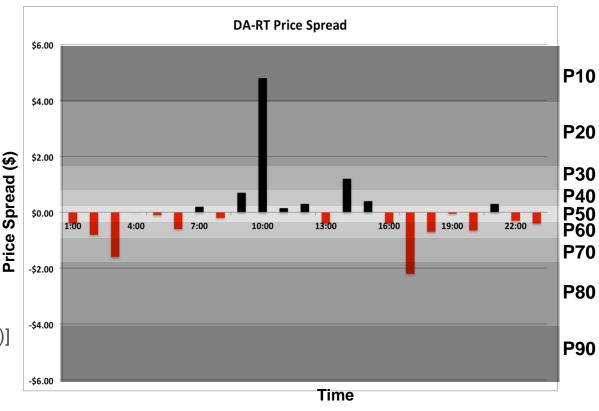
With similar downside risk exposure, the P70 forecast netted 20.5 GWh more energy scheduled into the day-ahead market over 6 months

Advanced Offering Strategies

- Accounting for risk that varies with time
 - Energy price expectations
 - Load forecast uncertainty
 - Transmission congestion
- Developing rules or an algorithm customized to detailed risk profile
 - Choose optimal quantile:

$$q = 0.5 + 0.5*tanh[(\pi_{DA} - \pi_{RT})/(r*\pi_{DEV})]$$

r = Pratt coefficient of risk aversion



Additional gains can be made with a customized approach

Example 2: To Balancing Authorities

Scheduling Firm Capacity Hour-Ahead and Protecting Against Down Ramp Events

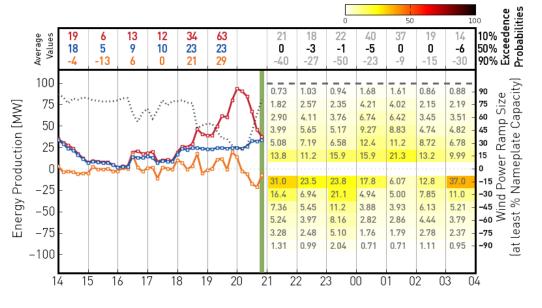


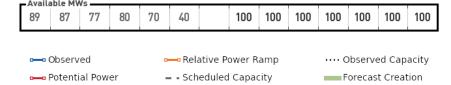
Taking a Risk-Based Approach

- Current operating practice is to <u>protect against ramp events at all times</u>, even when the risk is very low.
- Although current ramp forecast guidance has low to moderate skill and large uncertainty, it is not worthless. There is value.
- Value can be extracted by considering the uncertainty of each forecast situation directly, thereby minimizing missed ramp events over the long term, while maximizing the scheduling, transmission, and use of VG when it is produced.
- Guidance on ramp events is needed in probabilistic form to facilitate the user to take action only when the estimated risk is above the user's tolerance for it.

Ramp Event-Based Forecasts

Hour Ahead Ramp Exceedance Forecast





- Is the risk of a significant down ramp event next hour larger than your tolerance for it?
 - 31% for -15 MW or more
 - 16% for -30 MW or more
 - 5% for -60 MW or more
 - 1% for -90 MW or more
- Answer depends on the user or situation:
 - cost of missed event
 - tolerance for false alarms

Comparing Risk Tolerances

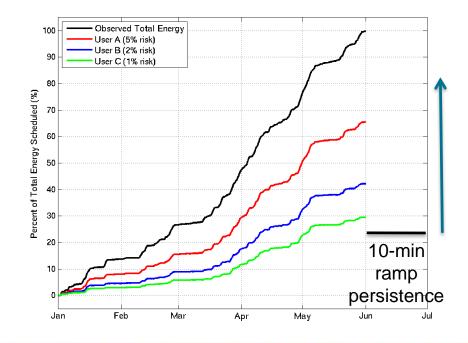
Examples of User-Specificity: Which kind of user are you?

- » Consider hypothetical forecast users A, B, and C.
- » Each aims to schedule maximal firm wind energy capacity 1 hour ahead and must protect against missed down ramp events.
- » They can tolerate any number of false alarms.
 - > User A accepts a 5% risk of being caught short and can afford to schedule 66% of the total realized energy.
 - > User B is more risk averse, accepting a 2% chance of being short, and schedules 42%.
 - > User C is the most risk averse, accepting only a 1% risk, and is able to schedule 30%.

Risk Adjusted Volume (MWh)

Forecast Value

(\$\$\$)



6% Increase

\$65K / month *

*for a 100MW wind facility with a 30% capacity factor and \$50/MWh energy price

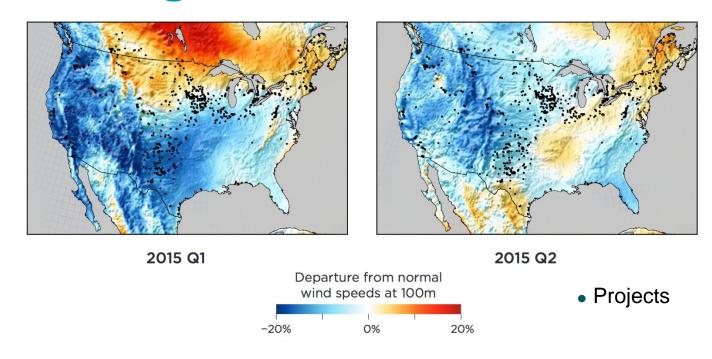
Example 3: To Owner/Operators

Hedging Seasonal Anomalies from Budgeted Renewable Generation



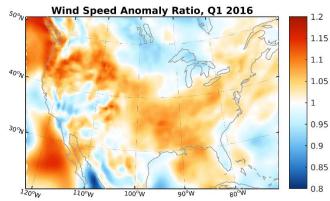
Recent Large Anomalies in Wind

2015

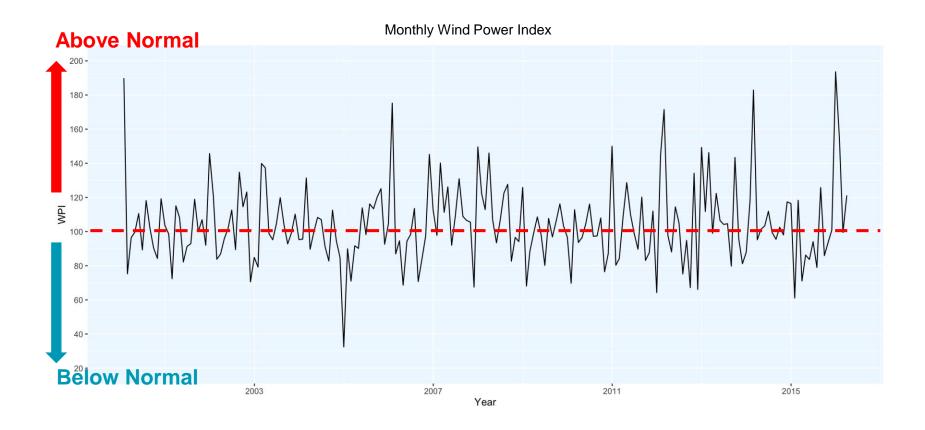


2016



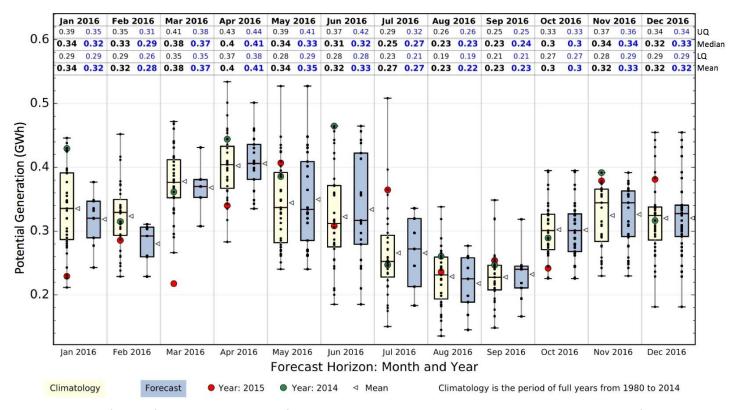


Monthly Anomalies At One Project



Some projects can deviate wildly from normal on a monthly basis.

Probabilistic Budget Setting Guidance



An example figure from our seasonal forecasting product that includes an operational reforecast to set climatology or "normal" for each month and the expected variation around normal based on the historic record (yellow bars) and the future climate state (blue bars).

Boxplots used to convey summary statistics and possible extremes.

Take Home Messages

- Probabilistic forecasts widely available, not widely used
- Requires community efforts to promote adoption
 - Policy
 - Market Rules & Transparency
 - Regulatory Mandates & Independent System Operator Practices
 - Technology Demonstrations
 - Dynamic Reserve Allocation & Stochastic Unit Commitment
 - Education
 - Conferences & Workshops
 - Industry Journals & Magazines
 - Webinars & Interactive Customer Engagements
- Usage may prove advantageous to bottom line (3 examples)
 - Power Marketers: Day-Ahead Market Offers
 - Balancing Authorities: Hour-Ahead Firm Capacity
 - Owner/Operators: Monthly Budget Setting Guidance

Thank You...

Further Reading:

- Scheduling More Wind: Exploiting Probabilistic Forecasts
 - Grimit and Lerner, 2016: 6.8, 7ENERGY, 96th AMS Annual Meeting, New Orleans
 - Grimit et al. 2015: Article, North American Clean Energy, Jan/Feb 2015, pp. 16-17
 - http://www.nacleanenergy.com/magazine/janfeb_2015/index.html#p=16
- Making Energy Balancing Decisions Based on Very Uncertain Wind Power Ramp Forecasts
 - Grimit et al., 2011: Poster 753, 2ENERGY, 91st AMS Annual Meeting, Seattle
- Benchmarking the Accuracy of Seasonal Forecasting for Renewable Energy Resource Anomalies
 - Grimit et al., 2017: J7.6, 8ENERGY/15AI, 97th AMS Annual Meeting, Seattle

