

Electricity Markets and Forecasts, how they interact

ESIG Tutorial – Forecasting and Markets Workshop

Aidan Tuohy and Erik Ela EPRI Grid Operations and Planning

Denver, June 4 2019



Objectives of Tutorial

- Provide general background for those less familiar on two important and increasingly interrelated topics
 - Forecasting of renewables
 - Electricity market operations
- Provide more detailed discussion on several advanced R&D topics and latest learnings from EPRI and others
- Discussion on the topics so that we can all learn from each other!



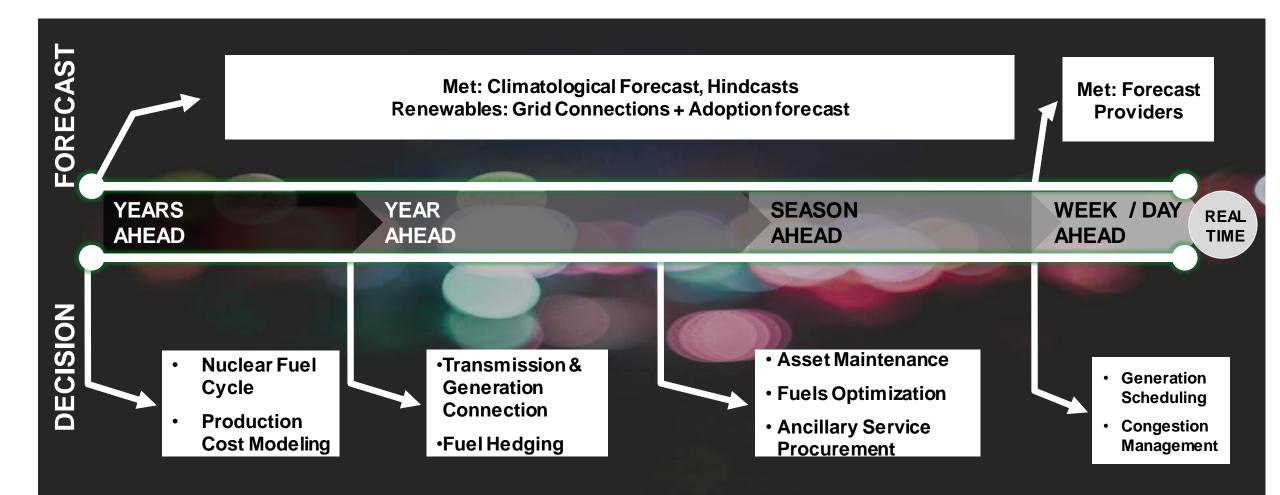


Agenda

TIME	TITLE	SPEAKER		
8:00-8:30	Session Overview	Aidan Tuohy and		
		Erik Ela, EPRI		
8:30-9:15	Wind and Solar Forecasting Utilization 101	Aidan		
9:15-10:00	Energy and Ancillary Services Markets 101	Erik		
10:00-10:30	Break			
10:30-11:00	Advanced Forecasts 201: How advanced forecasts could	Aidan		
	be used in operation			
11:00-11:30	Electricity Markets 201: Capacity Markets, FTRs,	Erik		
	emerging trends			
11:30-12:00	Forecasting and Market Design Integration: Advanced	Erik/Aidan and		
	Research	Audience		



Bulk System Forecast Uses

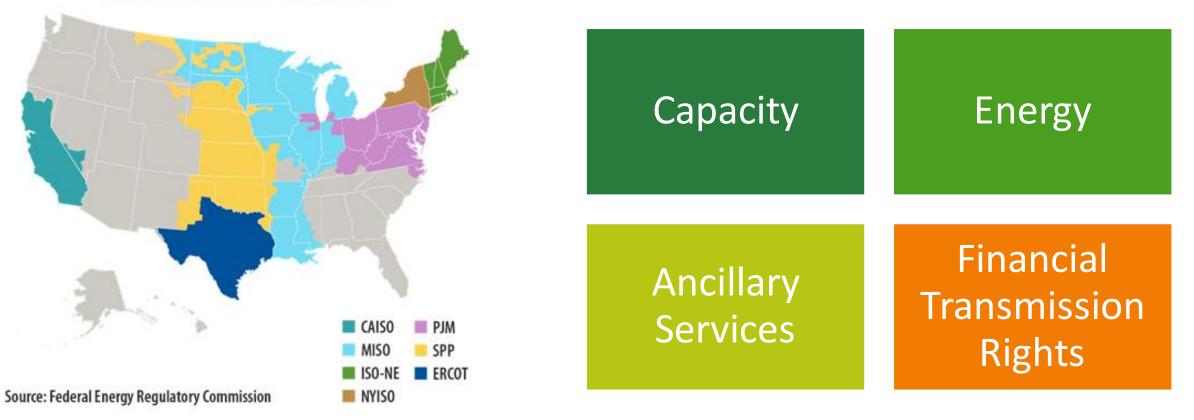




North American Electricity Market Design

Wholesale Electric Power Markets

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Together...Shaping the Future of Electricity







Using Wind and Solar Forecasts in Power System Operations and Planning

Aidan Tuohy Principal Project Manager

ESIG Tutorial June 4, 2019



 Image: Market with the second seco



Agenda

Evolution of forecast use in operations and planning

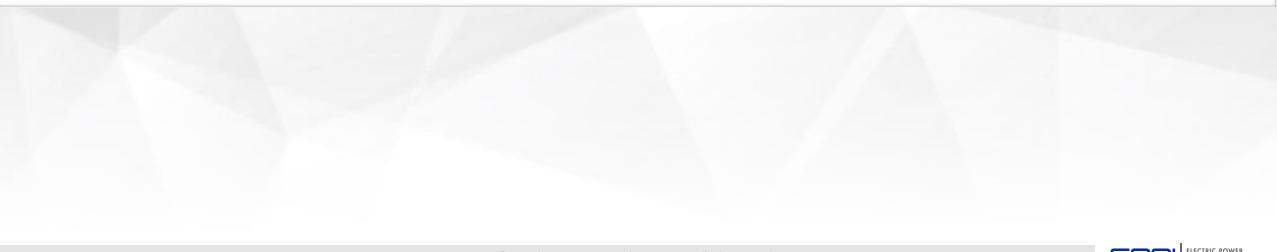
Evaluation of forecasting

Forecasting in integration studies





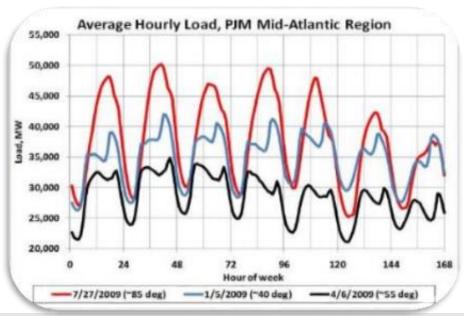
Where are forecasts used in operations today?





We Already Have a Weather Dependent System

- Load: temperature, humidity, wind
- Distribution: wind, snow and ice
- Transmission: temperature, fire, ice, wind
- Generation: temperature and extreme temperature events







Status of Renewable Forecasting in US Markets

All ISOs/RTOs now use centralized wind power forecasting

- Many other non-ISO areas also use wind power forecasting
- Most areas have solar forecasting in use or coming online

Used for multiple applications

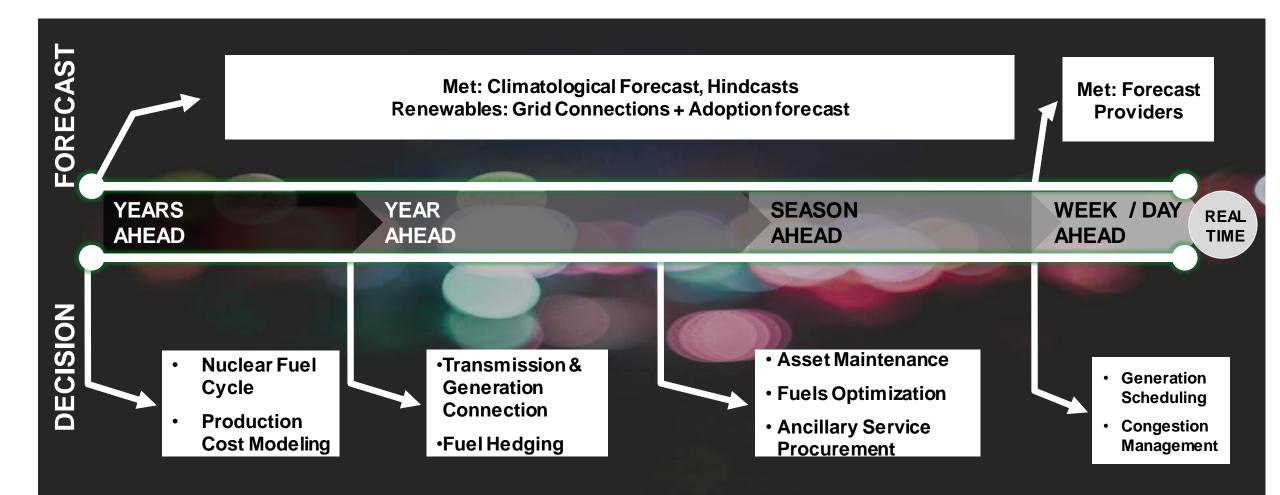
- Day(s) ahead and short term (5-60 mins) forecasts used
- Used in clearing markets and operational planning
- Different time horizons, frequency of provision and granularity provided to different users based on their needs

Ramp forecasts are also used by some operators

- Some may prefer to just use one forecast
- Some are using ensembles provided by vendor, or multiple vendors
- DER forecasting is becoming more challenging and important
- Cost allocation of forecast provision varies by region

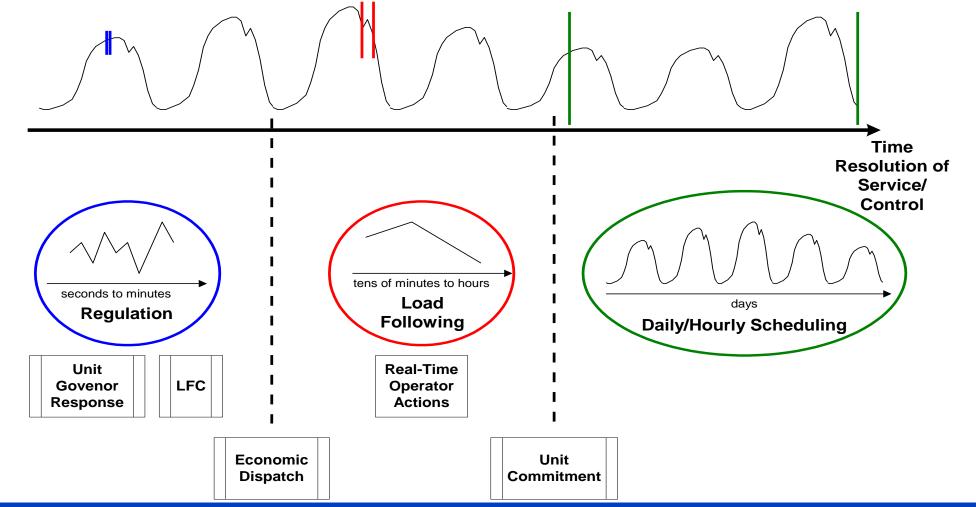


Bulk System Forecast Uses





Timeframes of Interest for Forecasting



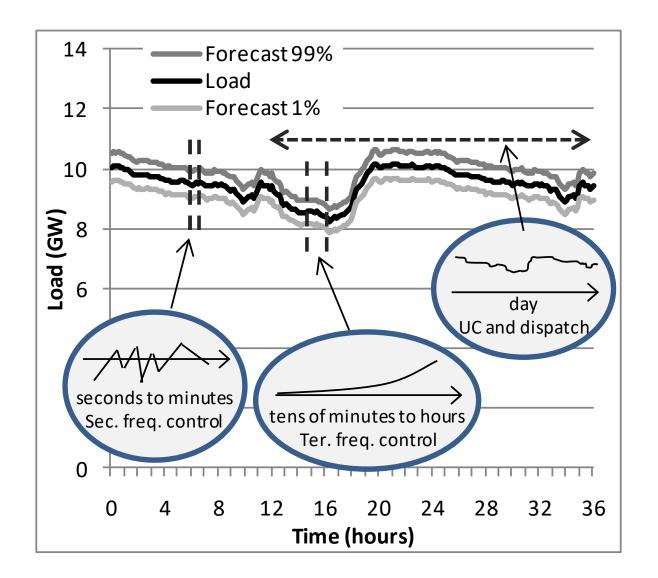
We use these timeframes for operations, but also in planning studies to examine impacts of variability and uncertainty on resource needs



Unit Commitment and Dispatch without Wind/Solar

- During operation demand drifts from the prediction and power plants or power lines may fail
- Reserves keep the system in balance
- UC plans the operation for the next day
- ED dispatches the resources available
- Reserves cover the periods in between

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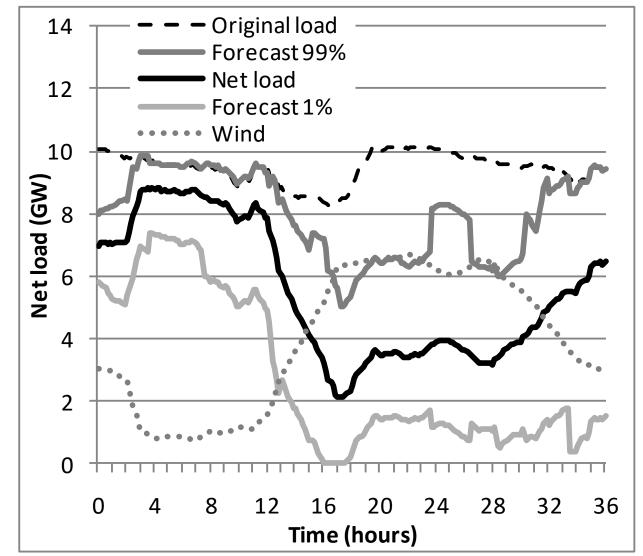


UC with Wind/Solar

- Expected residual load (load minus wind power production)
- Variability and uncertainty increases
- More flexibility required

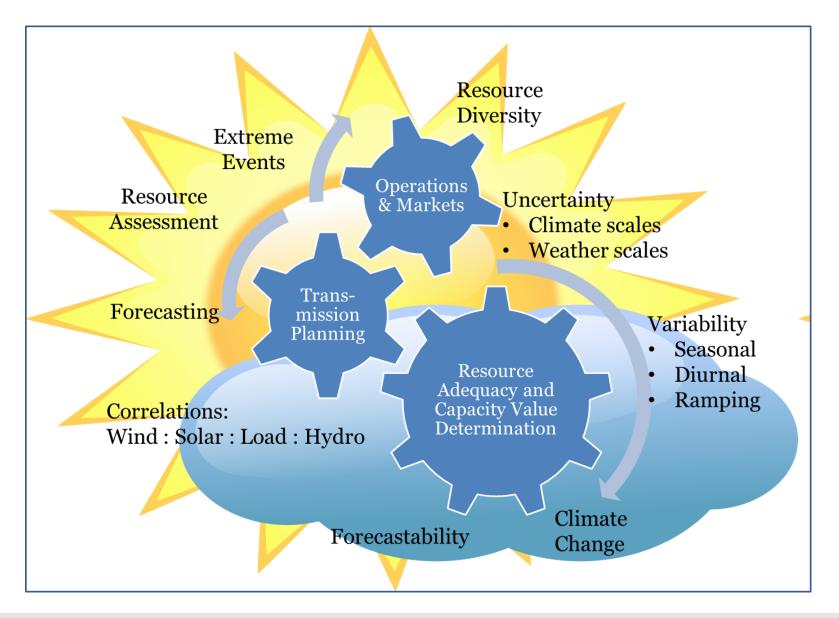
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- Generation
- Responsive load
- Possible curtailment of wind power
- Required amount depends on the variability and forecast accuracy of wind and load during the relevant time period





Broader Use of Meteorology in Ops/Planning







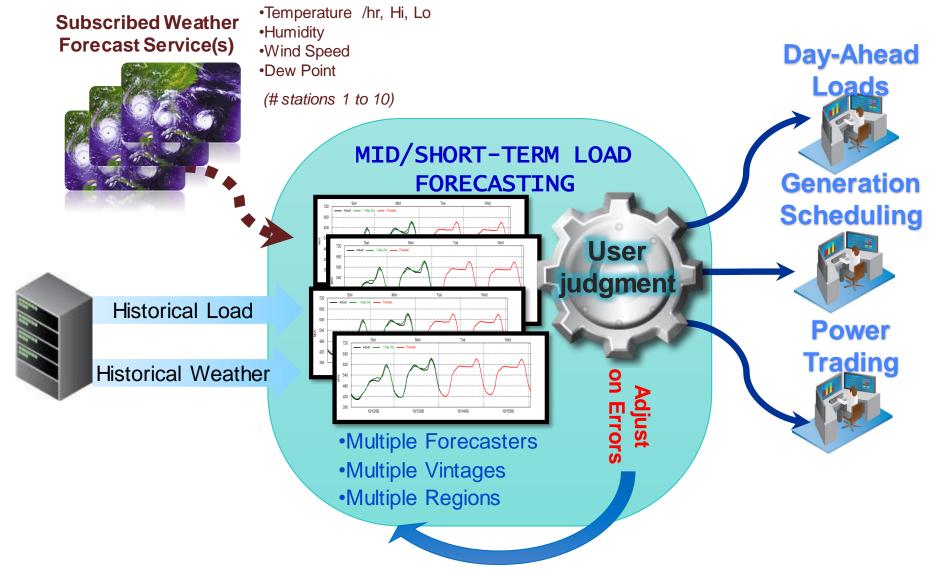


How are forecasts developed and used – data and models





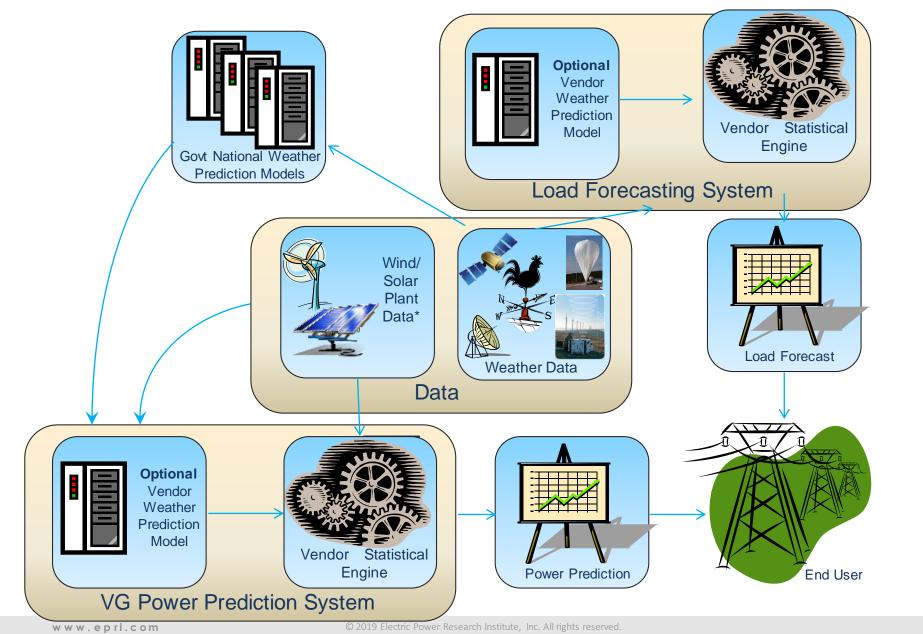
Short-term Load Forecasting Typical Setup



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Adding VER - Forecasting Electricity Supply and Demand





EPCI ELECTRIC POWER RESEARCH INSTITUTE

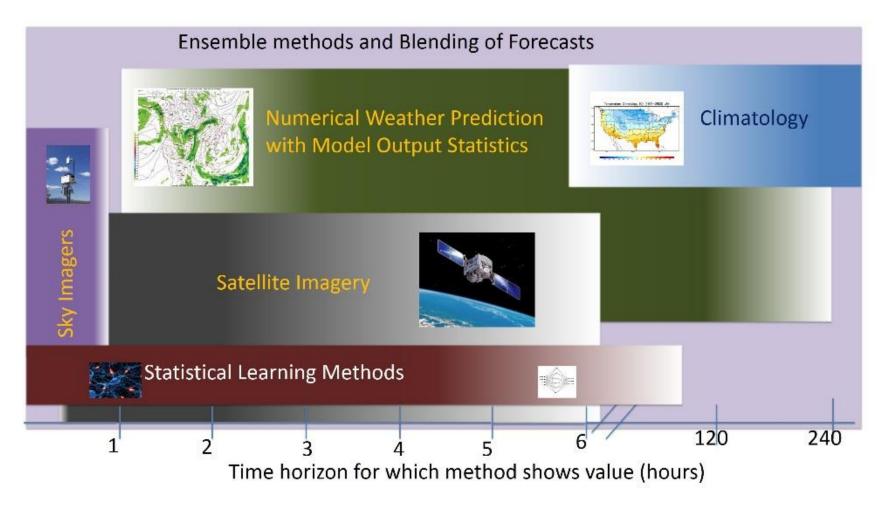
Summary: Types of forecasts, their applications, and methods for producing them

	Type of Forecast	Time Horizon	Key Applications	Methods
Generation	Intra-hour	5-60 min	Regulation, real-time dispatch market clearing	Statistical, persistence.
	Short term	1-6 hours ahead	Scheduling, load-following, congestion management	Blend of statistical and NWP models
	Medium term	Day(s) ahead	Scheduling, reserve requirement, market trading, congestion management	Mainly NWP with corrections for systematic biases
	Long term	Week(s), Seasonal, 1 year or more ahead	Resource planning, contingency analysis, maintenance planning, operation management	Climatological forecasts, NWP
Decision support	Ramp forecasting	Continuous	Situational awareness, Curtailment	NWP and statistical
	Load forecasting	Day ahead, hour-ahead, intra-hour	Congestion management, demand side management	Statistical

Source: http://greeningthegrid.org/resources/factsheets/copy_of_ForecastingWindandSolarGeneration.pdf



Solar Forecasting Methods



A .Tuohy, J. Zack, S.E. Haupt, J. Sharp, M. Ahlstrom, S. Dise, E. Grimit, C.Möhrlen, M.Lange, M.G.Casado, J. Black, M. Marquis, and C. Collier, "Solar Forecasting", IEEE Power & Energy Magazine, Nov/Dec 2015

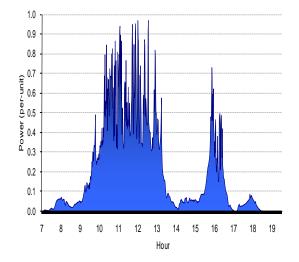


Solar vs Wind Forecasting

- Observability
 - Solar is better observed in short time scales \rightarrow cloud detection
 - Potentially more difficult day ahead to catch cloud formation
- Variability
 - Solar plants more variable in small sizes
 - But, over larger system, may be less variable

Location

- Windy sites tend to be more windy than average
- Sunny sites tend to be less cloudy
- System Operations
 - Quasi-linear relationship versus cubed for wind
 - Well known clear sky day relationship



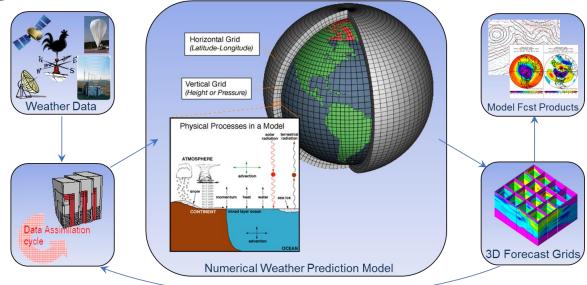


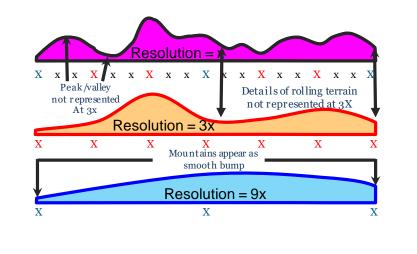
Wind/Solar Forecasts versus Load Forecasts

- Electricity demand follows a regular pattern that is modulated by weather...mainly temperature
- Variable generation forecasts are defined by weather
- Wind and irradiance harder to forecast than temperature and exhibit more spatial and temporal variability
- Thus, we should expect load forecasts to exhibit considerably lower errors.

Numerical Weather Modeling

- Solves equations for evolution of the atmosphere to predict weather
 - Typical resolutions of 40km to 1km, 30 to 140 horizontal levels
 - Mapped onto 3D grid, with intervals of 15 minutes to one hour
- Several important factors:
 - Initial conditions data assimilation and boundary conditions
 - Parameterizations
 - Resolution
 - Model numerics
 - Uncertainty quantification









Observational Data

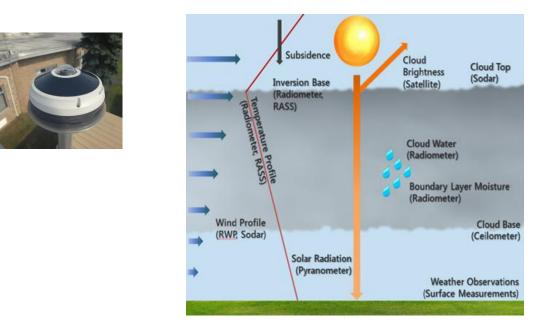
On-site facility data

- Metadata (configuration, static data)
- Archived and real time power and status data
- Meteorological data
- Satellite data
 - Latest GOES satellites and equivalent around the world provide high resolution information
- Sky imagers
 - Potentially useful for short term, though still in R&D for most applications
- Offsite observations
 - Meteorology stations placed upstream

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Includes lidar, sodar, ceilometer, radiometer, etc.



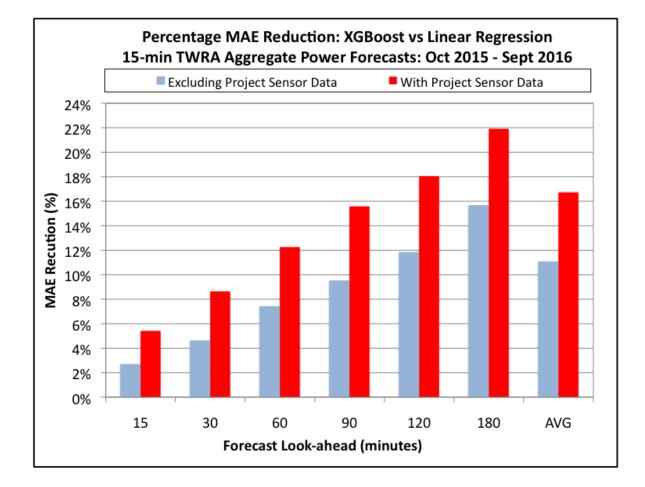


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Statistical Modeling

- Statistical methods used extensively throughout process
- Machine learning used for correcting model bias and other post processing of models
- Very short term forecasting (less than one hours) often uses statistical approaches
- Load forecasting has traditionally used neural networks and similar with temperature forecast for load forecast up to days ahead
- With increased prevalence and improvements in AI methods, new techniques can show significant improvement versus older methods

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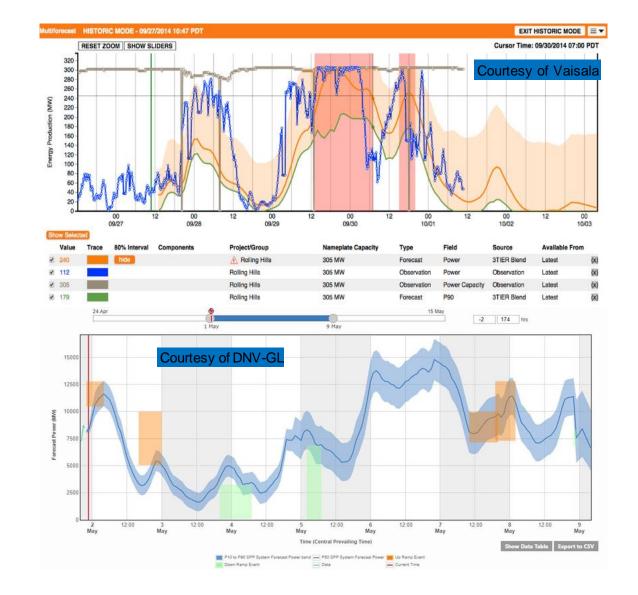


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Examples of Forecast Products

- Deterministic 'best guess' forecast
 - Most typical forecast
 - Easiest to use and understand
- Probabilistic forecast
 - Confidence intervals around deterministic forecast
 - Scenario based forecasts
 - Used for awareness or to support decisions
- Event based forecast
 - Shows high risk periods for awareness
- Situational awareness
 - Figures of the services territory with weather patterns
 - Regional generation spread, etc.

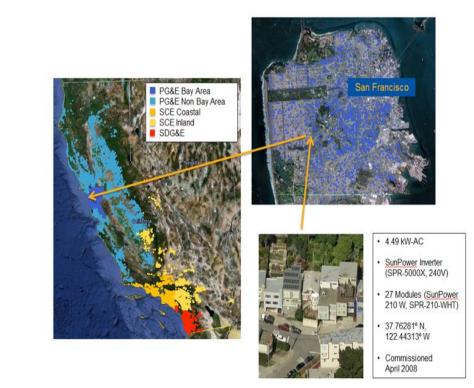
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Challenges for Load Forecasting with Distributed PV

- Requires accurate solar forecasting as an input to load forecasting
 - Could be direct modifier or included in the load forecast
- May not have all data needed
 - PV production in real time
 - Info about weather conditions
 - Design specs of PV
- Relative importance of data will vary based on
 - Geographical conditions,
 - PV penetration and type,
 - Size of system operator's territory
 - Type of utility/ISO and their load forecast methodology



From Clean Power Research for CA





Evaluating Forecasting For Operations





Forecast Skill/Value Assessment

- Need to understand forecast performance for multiple reasons:
 - What is expected uncertainty of my forecast tomorrow, based on past forecasts?
 - Average or extreme errors under different conditions need to be understood
 - Which of two or more forecast providers performs better?
 - How do I compare forecasts in a meaningful, fair way?
 - If I am planning to integrate wind/solar in the future, what type of uncertainty may be seen?
 - What type of operating reserves should I be planning on carrying?
 - How should I simulate uncertainty when planning for the future?
- EPRI-CPS Energy and EPRI-Southern Company solar forecasting trial examined multiple solar forecasts in Texas to better understand above questions

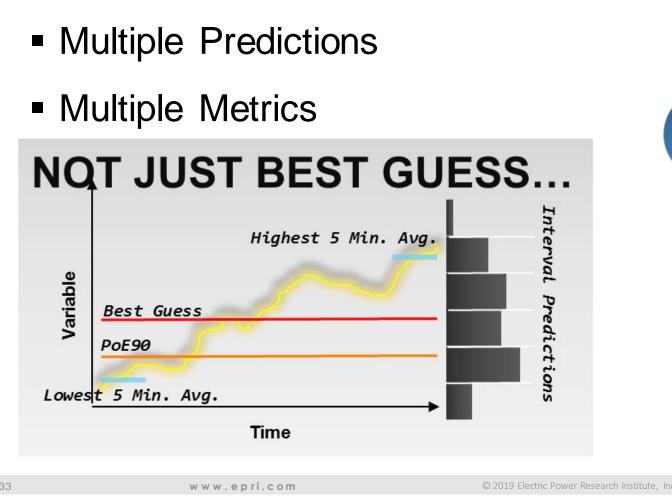


Trial Description – Key Facts

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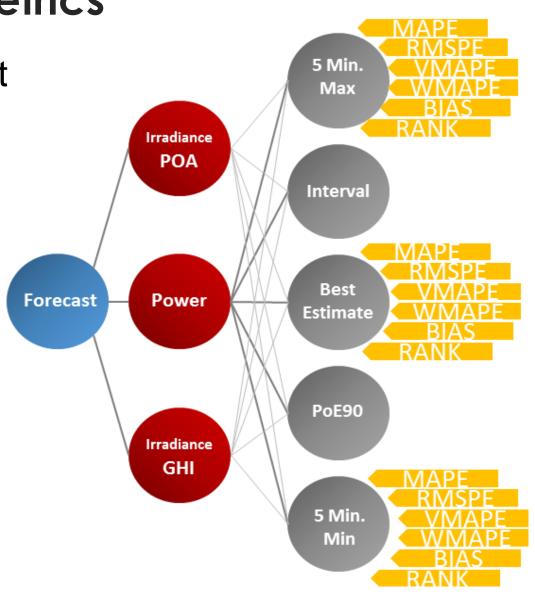






Standard Forecast Format & Metrics

- Standard XML based forecast format
- Multiple Variables



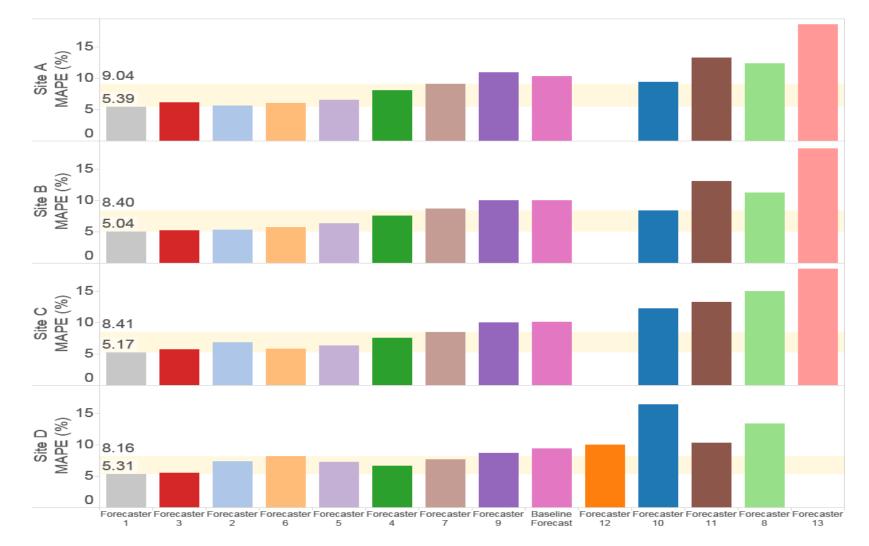
prediction

metric

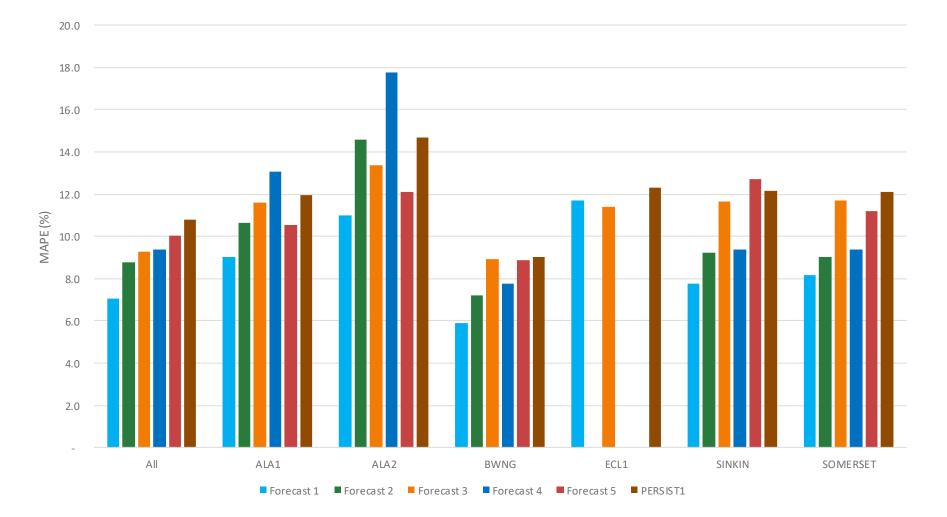
variable



Average MAPE – Southern Company Sites



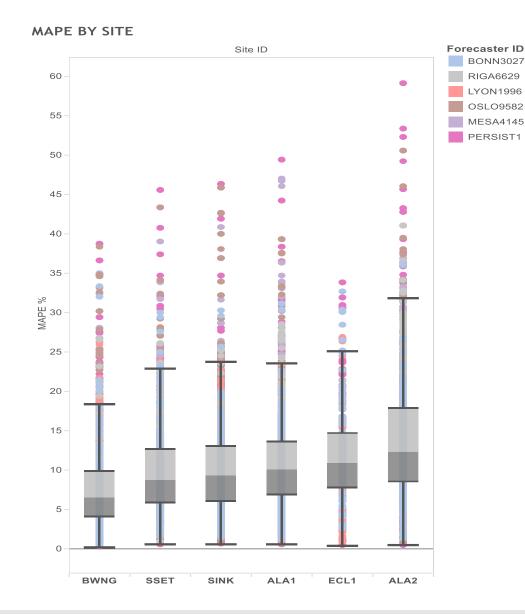
Summary Metrics – Mean Absolute Error – CPS Energy



Performance varies widely for different forecasters and across solar farms, due to data availability and technology



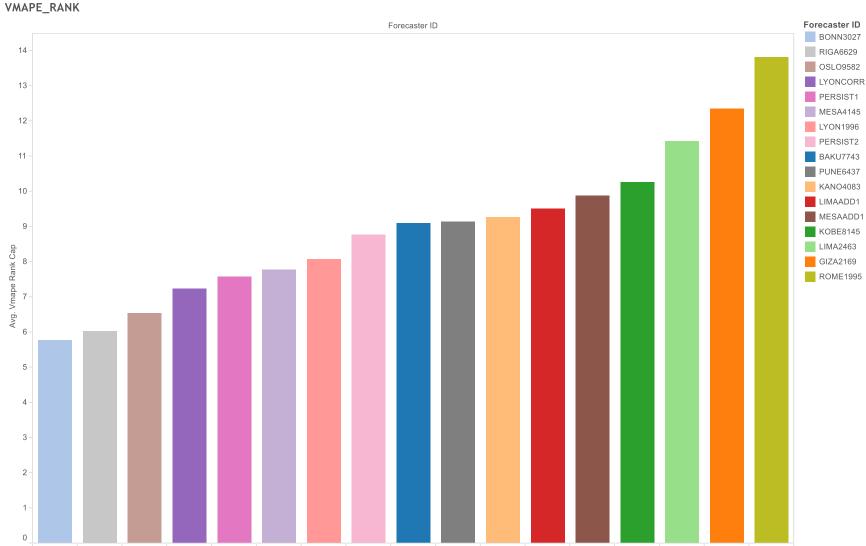
Daily performance by site



- Each dot represents performance in one day
- Shows even the best can be pretty far off in any day
- Sites with dual axis tracking seem to be more difficult to forecast
- While persistence does OK some days, on others it can have largest error



Average Ranking for Variable Periods



BONN3027 RIGA6629 OSLO9582 LYONCO. PERSIST1 MESA4145 LYON1996 PERSIST2 BAKU7743 PUNE6437 KANO4083 LIMAADD1 MESAAD. KOBE8145 LIMA2463 GIZA2169 ROME1995



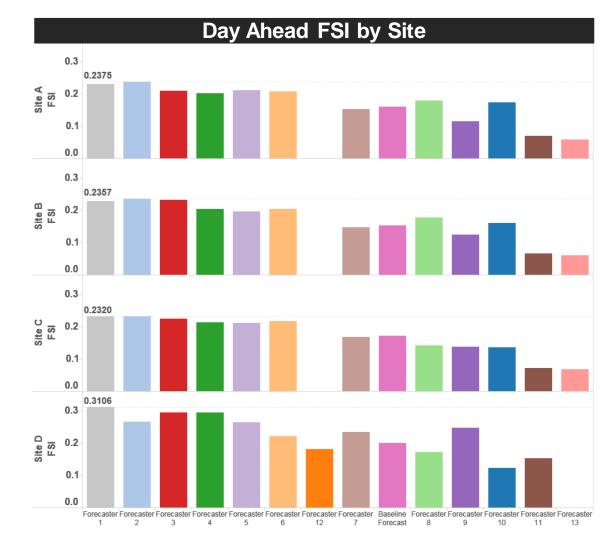
Forecast Skill Index - FSI

Used in previous trial to combine multiple accuracy assessments.

FSI includes:

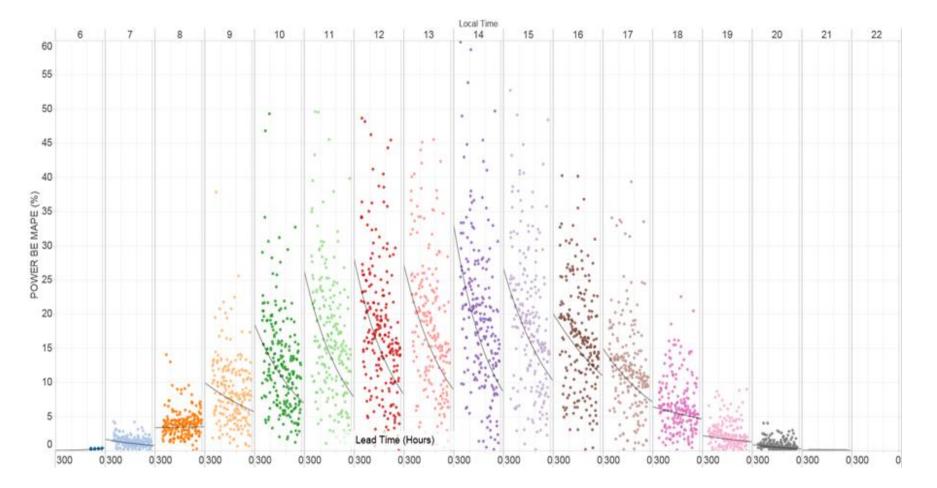
- MAPE (value, rank)
- RMSPE
- MAPE during variable periods
- MAPE weighted by time of day

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Performance of forecast by time of day and look ahead



More info: Solar Power Forecasting for Grid Operations: Evaluation of Commercial Providers, Electric Power Research Institute, December 2017. Palo Alto, CA. 3002012135



General Approaches for analyzing value

Simulation

- Production simulation to examine impact of improved forecasts or reduction in risk/reserves
- Can look at future higher penetration systems
- Captures the interaction of wind/solar and system costs

Historical

- Examine forecast value based on prices or system lambda
- Example "What is the value of an improved forecast in my trading decisions?"
- Straightforward for lower penetrations, but may not allow interaction between wind/solar and costs to be considered

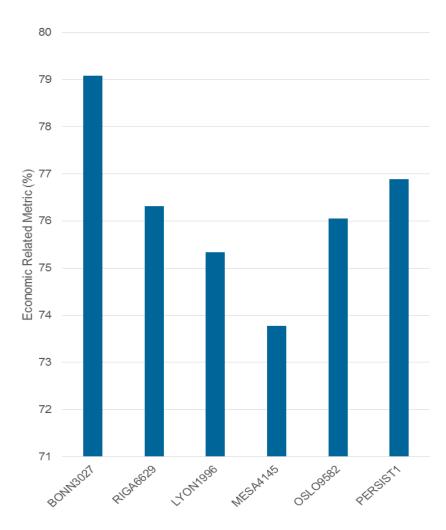
Methodology for system value assessment and baseline forecast performance (California solar project)

- Determine a methodology to capture the value of improved forecasts in economic and reliability terms
 - Based on discussion with stakeholders
 - Potential metrics could include reduced production cost, reduced reserve requirements, improved utility procurement, improved distribution switching operations, line rating increase for transmission, etc.
- Calculate the values of a forecast improvement based on prevailing system conditions in 2017 and 2018
 - Where appropriate, capture future value through simulations using the Power System Optimizer (PSO) production cost tool and datasets that represent future grid conditions in California

How do you currently value forecasts? Looking for input



CPS Energy Economic Based Metric



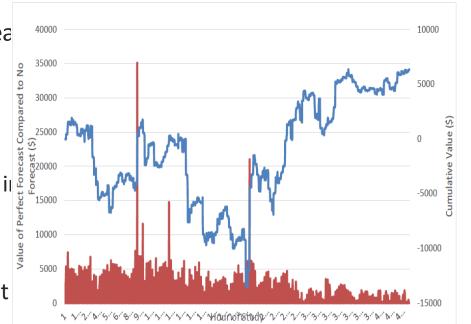
- Developed by CPS Energy to support trial
- Based on price spread between day ahead and real time
- Should reflect periods of high risk rather than specific spreads
 - We don't know what spread is in advance just want to reduce risk
- Weighted by time of day and normalized to DA-RT
- 100% is best
- Same forecast always wins...
- OSLO does better in many metrics that look at value or variability compared to others, so may be reason for high value
- Persistence also does well

Shows some link for value as market participant/purchasing, not for ops



Historical Analysis of Value

- Normally takes the form of using prices or marginal costs to understand additional revenue or reduction in generation costs
- Difference between prices or energy costs in day ahead and rea time to produce value of knowing more in day ahead
- Finding: More related to the actual prices analyzed and their profiles than the value of forecasts
 - E.g. if DA always higher price or cost, then it may be better to bid in at 100% and then make up difference in real time
 - Doesn't capture penalties and increased risk as penetration increases
 - May be able to instead put a value on error and quantify how that changes with improved forecast



Additional value from perfect forecast in DA vs RT

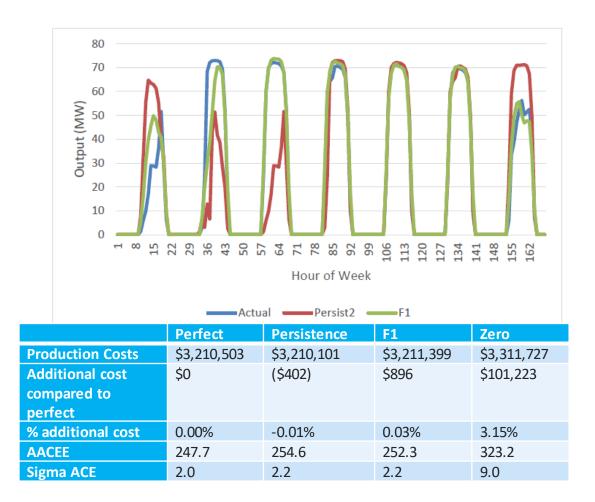
Not likely to be particularly useful manner to assess forecast performance

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Simulation based metrics – simple case study

- Used FESTIV tool
 - 1 week data for IEEE 118 bus
 - 3 forecast types examined
- Value in using any forecast compared to none at all
- Not much difference among skilled forecasts and simple persistence method
 - Maybe some reliability implication but not very clear
- System used and operational policies are likely to impact
 - Time resolution used, look ahead, resources startup times, etc.
- More detailed study of particular system should be performed



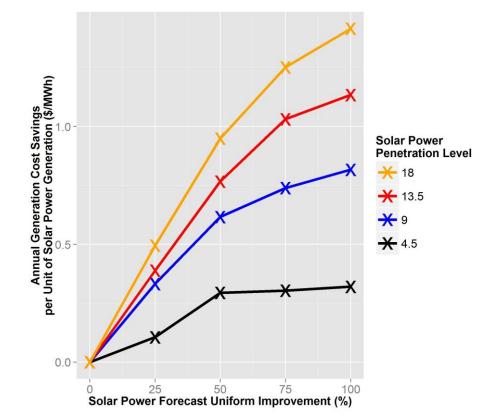
Value of improved day ahead forecasts not always easy to measure



Example: the value of day-ahead solar power forecasting improvements in ISO-New England

- NREL study for different regions of the country
- Based on simulations, improving solar forecasting for ISO-NE:
 - Lowers generation from fast-start and lower efficiency plants
 - Reduces ramping of all generators
 - Decreases solar power curtailment
 - Lowers start and shutdown costs
 - ...resulting in a reduction in overall operational electricity generation costs

Source: Brancucci et al. (2015). The Value of Day-Ahead Solar Power Forecasting Improvement (Draft).



Improving day-ahead state-of-the-art solar power forecasts in the ISO-NE system with 13.5% solar penetration decreases annual variable electricity generation costs by US



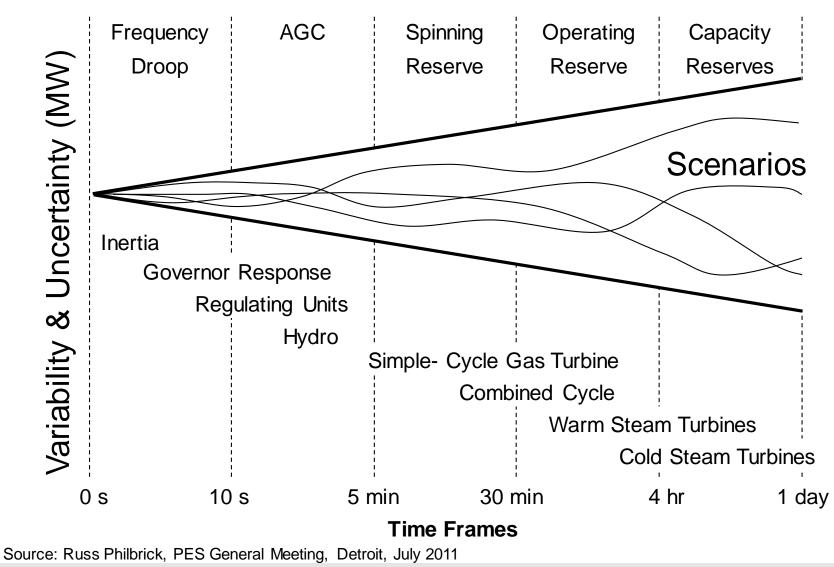


Use of Operational Forecasting in Planning





Background: Time Frames and Uncertainty



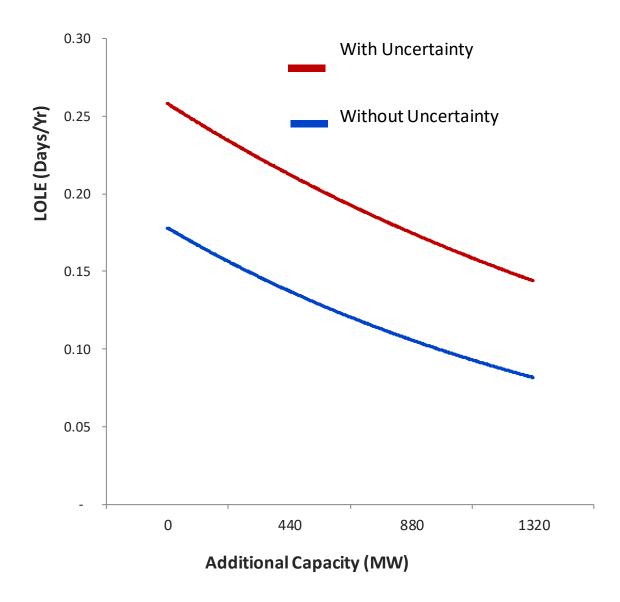
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Why is modeling of uncertainty in planning studies important?

- Example from 2013 study using SERVM production cost tool with operational uncertainty modeled
- Shows that reliability metrics calculated using perfect foresight may be underestimating the impact of uncertainty
- Other impacts will also be important
 - Cycling of generation and costs of flexibility
 - Use of energy storage and demand response
 - Other reliability metrics

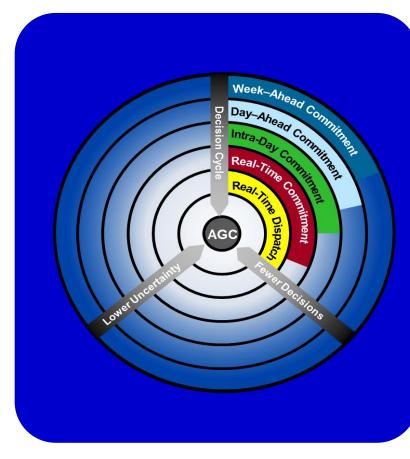






Simulating Operations: Production Cost / Operations Simulation Models State of the Art

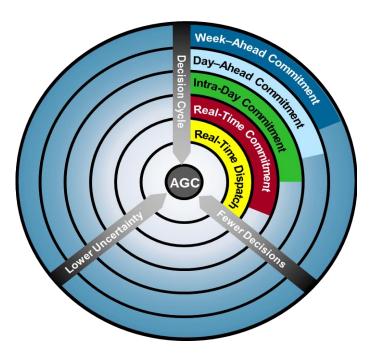
- Simulate at least two-stages of scheduling process
 - Day-ahead and real-time
- Include network security constraints
 - DC-OPF SCUC, SCED, model
- Detailed generating unit constraints
 - Ramp rate, min on/off time, start-up time
 - Advanced: Hydro reservoir and cascade constraints, combined cycle modes, energy storage optimization
- Annual simulations, hourly or down to five-minutes
- Outputs: Production costs, marginal costs, power flows, generation production, load shed or reserve shortage violations





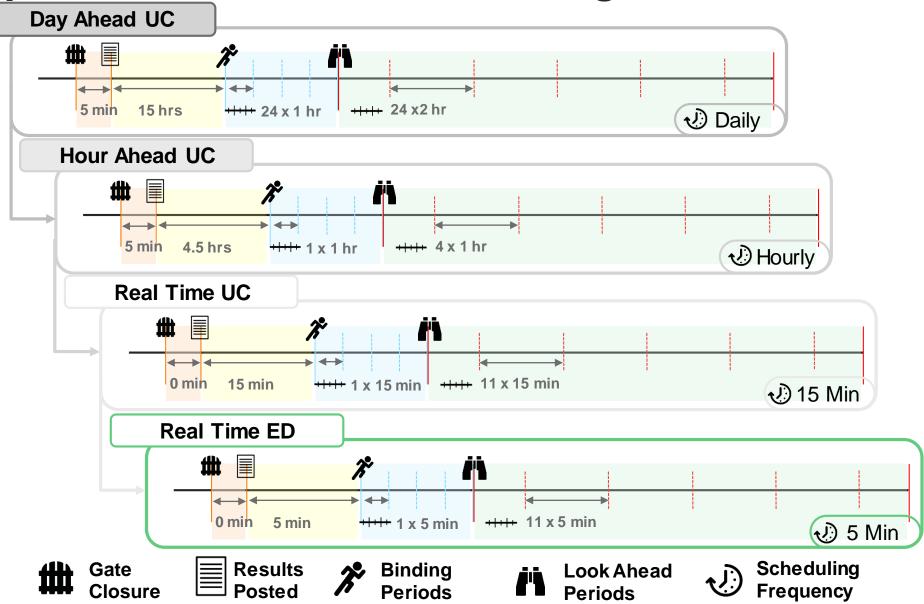
Multiple cycles in modeling

- Historical simulation tools did not capture the multi-cycle multi-decision, multitimescale approach of current steady-state electricity operations
 - As horizon approaches real-time, uncertainty impact decreases
 - As timescale resolution approaches zero, variability impact decreases
 - As horizon approaches real-time, fewer decisions are available to the operator
- Simulation tools that model multiple cycles can more realistically represent the impacts of variability and uncertainty and the mitigation strategies for those impacts





Multi-cycle, Multi-scenario modeling







Conclusions

- Wind and solar forecasts are widely used and understood for system operations
- Various models and methods, with large amounts of data, are used to produce forecasts
- Important to understand the performance of forecasts when considering how they should be used
- Simulation tools for studying renewable integration should use forecasts to capture short term uncertainty



Together...Shaping the Future of Electricity







Electricity Market Design 101

Erik Ela Principal Technical Leader

ESIG Meeting June 4, 2019 Denver, CO

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Agenda

- Electricity Markets Quick Primer
- EPRI Electricity Market Operations Research Area
- Key Market Design Research Questions
- Program 39: Electricity Market Design Reference Guide
- Key Topic Deep Dive
 - Reliability Services Valuation in Electricity Markets
 - Electric Power Supply Resilience
 - Price Formation

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A Market Like No Other

- Electrical energy cannot be stored (no inventory)
 - It can be converted to other forms of energy and stored but with additional capital and efficiency losses (e.g., pumped hydro plants)
- Energy is generated and consumed at almost the exact same time
- Electricity must be transported to consumers at the speed of light often from far distances



A Market Like No Other

- Laws of physics will dictate where power will go, who will get it, and how much of it will be lost along the way
 - If the road is full of trucks, you can't deliver anymore supply, and you can't use a different road
- Rules and economics will influence how resources are deployed, subject to technical constraints
- There are many different ways to supply it, but the consumer is indifferent as to the end product
 - Some suppliers have large capital costs and low variable costs, others are the opposite (price highly volatile even throughout day)
- Consumers rarely, if ever, are exposed to true costs. Therefore, and for other reasons, market failure is rampant



Independent System Operators/ Regional Transmission Operators (ISO/RTO)

Job #1 - System operating reliability

- Real time balancing
- Generator and transmission scheduling
- Ensure long-term reliability
- Coordinate/facilitate long-term planning for generation and transmission

Operate and administer markets

- Day-ahead energy
- Real-time energy
- Ancillary services
- Transmission rights
- Capacity markets

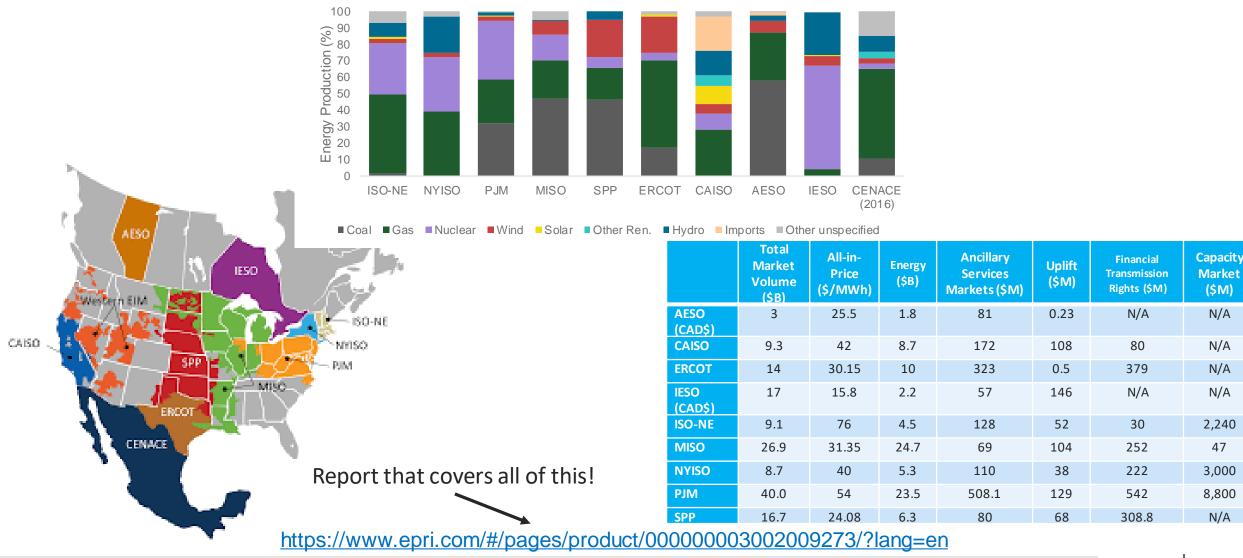
Not involved on retail transactions





Organized Electricity Markets in North America

Nine Independent System Operators (ISOs) and Regional Transmission Organizations (RTOs) serve 66% of consumers in the U.S. and more than 50% of Canada's population. See the ISO/RTO Council website: www.isorto.org.





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N/A

N/A

N/A

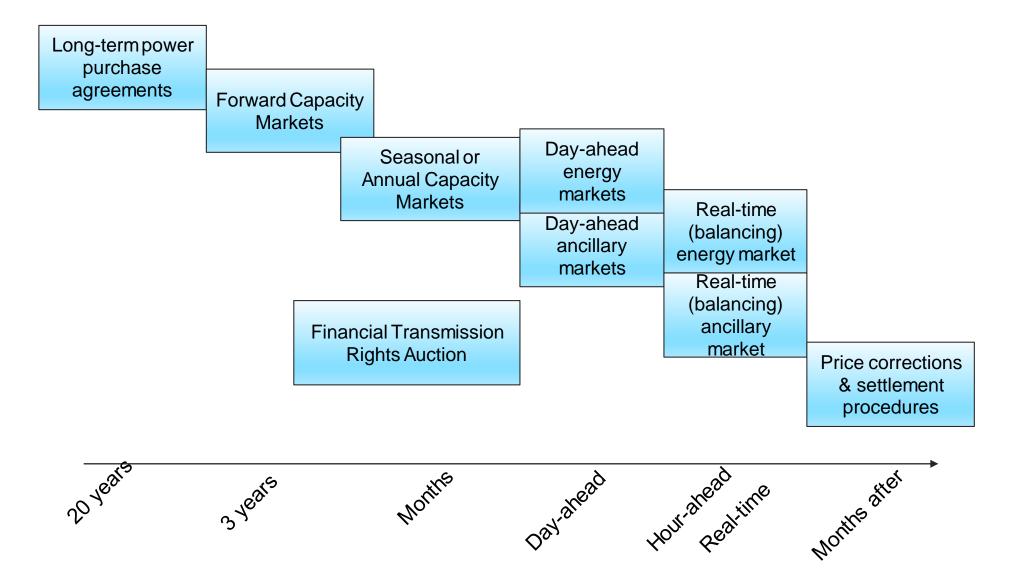
2,240

47 3,000

8,800

N/A

Electricity Market Timelines





Energy Markets

Bid caps, capacity markets

Bid caps with capacity markets (Most U.S.)
Caps equal to VOLL, no capacity markets (ERCOT, AEMO)

Forward energy settlement

Two-settlement (U.S.)Real-time markets (AEMO)

Locational granularity

• Nodal pricing (U.S.)

• Zonal pricing (Europe)

Multi-part bidding

• 3-part bids (U.S.)

• Single bids (Europe, AEMO)



Locational Marginal Prices Overview

 LMP definition: incremental operating cost to serve the next increment of demand at a particular location and time

- LMP =

[Marginal Cost of Energy] +

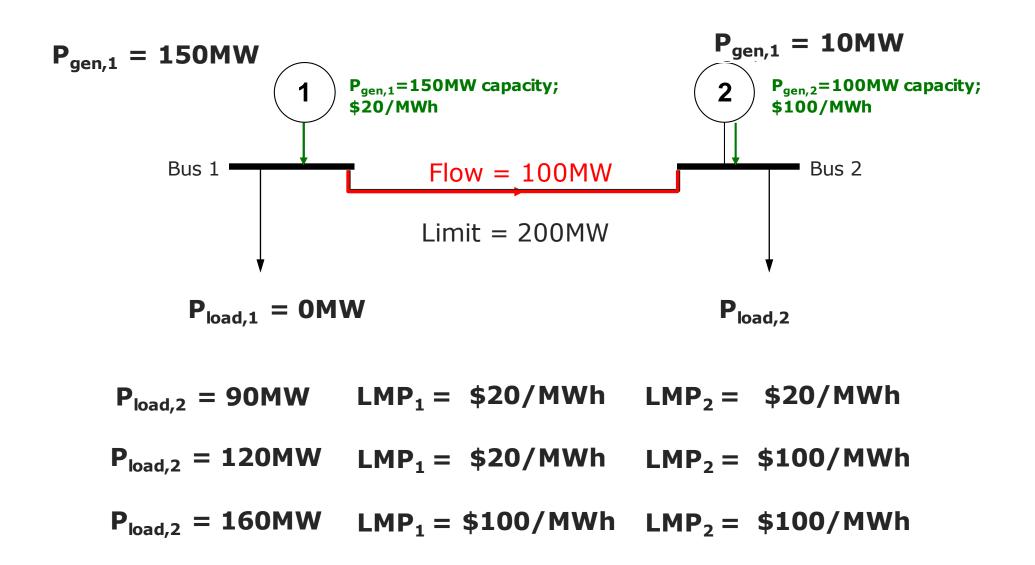
[Marginal Cost of Transmission Congestion] +

[Marginal Cost of Transmission Losses]

- Mathematically: Partial derivative of objective value (production cost) divided by partial derivative of demand (solved through dual solution of economic dispatch problem)
- Locational granularity: Suppliers nodal; demand zonal
- Temporal granularity: Day-ahead hourly; Real-time 5-min
 - Real-time: Settled hourly in PJM, ISO-NE, MISO until FERC Order 825, otherwise 5minute settlements



LMP in action





Two Settlement System

Market engine: Security-constrained unit commitment (SCUC)



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Day-Ahead Market (DAM)

•Financial

•Hourly pricing, day solved simultaneously

•Generally predictable pricing

•Generators offer amounts they are willing to generate for, loads bid in amount of load needed

•Commit enough generation to meet tomorrow's demand plus a certain amount of reserves

•LMPs created and generators are paid for energy schedule regardless of actual conditions

•About 95% of all energy in twosettlement markets is sold in the DAM Market engine: Security-constrained economic dispatch (SCED)

Real-Time Market (RTM)

Physical

•Five-minute pricing, market solved usually every five-minutes at a time

•RTM will make up differences by dispatching units to meet actual load

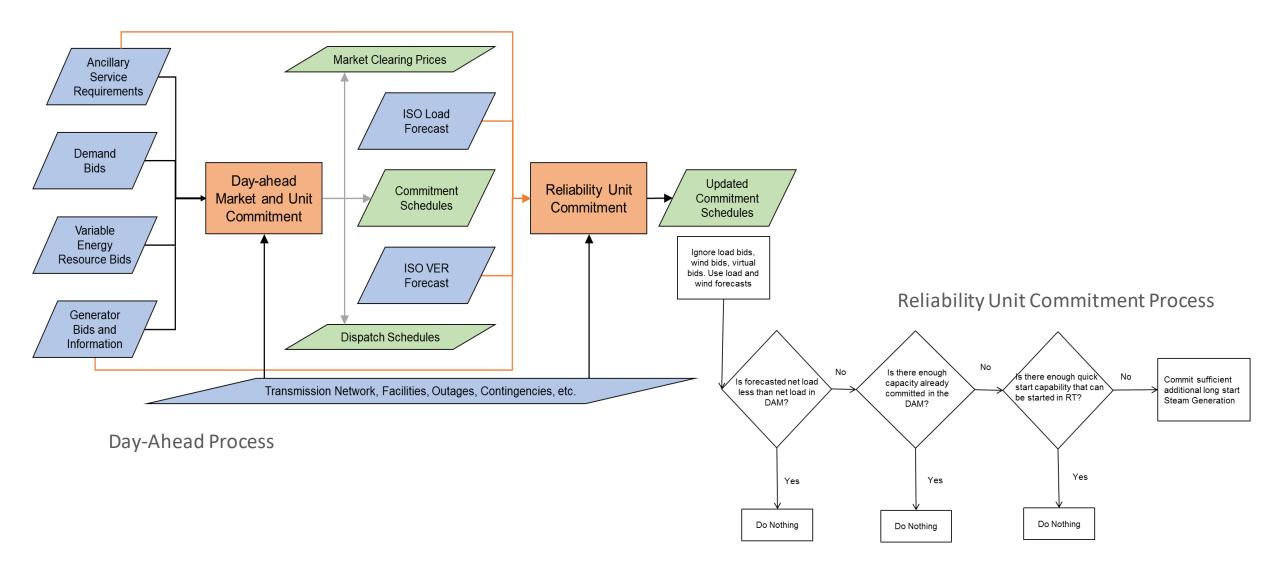
•Supply resources will get paid RTM LMP if they produce above their DA schedule or will pay RTM LMP if they produce below their DA schedule

•Payments are based on actual generation not schedules

•Generally, same constraints are used in the RTM as DAM, with the exception that generator starts are not common



Day-Ahead Commitment Process





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Virtual Trading

- Virtual trading: Market participants with or without physical supply or load buying/selling/transacting in the day-ahead market and buying/selling/transacting energy back in the real-time market who earn revenue off arbitrage
- Virtual trading now allowed in all markets all markets now have two (or more) settlement systems (SPP 2014)
- Hedge or speculate
- Any entity regardless of physical asset (just credit requirements)



- Paired virtual "spread" trading transactions allowed in PJM and ERCOT, being evaluated in other ISOs
- Objectives of virtual trading
 - Convergence of day-ahead and real-time prices and day-ahead commitment
 - Increased competition and liquidity in market
 - Hedging of physical assets
 - Limit supply- and demand-side market power by other physical asset holders
 - Show market designers market modeling flaws??
- Markets have different locational requirements (PJM all nodes, others only gen nodes and load zones, NYISO only load zones)



Uplift and make-whole payments

- ISO guarantees a generator following ISO directions to recover all bid-in operating costs
 - Mitigation and verification may be required for bid-in costs excessively high
- When the price for energy does not cover all costs, additional side payments (makewhole payments or revenue sufficiency guarantee) are paid to individual generators
- Typically happens because no-load costs and start-up costs do not influence the energy price paid
- Also for reliability must run resources
- Uplift usually allocated to loads, but also to generators that deviate from schedule
- Additional uplifts also occur
 - Day-ahead profit assurance
 - Price volatility make-whole payment
 - Alternative fuel make-whole





Ancillary Services Markets





Ancillary Services (FERC)

1. Scheduling, system control and dispatch:

Provided by the ISO or RTO, not applicable to this discussion.

2. Reactive supply and voltage control from generation service:

Generally supplied as a cost-based service. Lost Opportunity Costs provided if applicable.

3. Regulation and frequency response service:

Today, <u>regulation</u> is typically supplied and priced <u>by dynamic markets</u> in ISO/RTO regions. It is used to correct ACE. However, <u>frequency</u> <u>response</u>, as defined by the droop response of governors autonomously responding to frequency is generally <u>not included</u> in any dynamic markets <u>nor is it given cost-based rates</u>.

4. Energy imbalance service:

Energy imbalance is typically the service of the real-time markets balancing out the imbalance from the forward markets, and therefore is **priced by the real-time energy markets**. However, some areas have **implemented or proposed new market mechanisms** to support **ramp products** for energy imbalance.

5. Operating reserve – synchronized reserve service:

This service is typically supplied and priced by **<u>dynamic markets</u>** in ISO/RTO regions.

6. Operating reserve – supplemental reserve service:

This service is typically supplied and priced by **<u>dynamic markets</u>** in ISO/RTO regions.

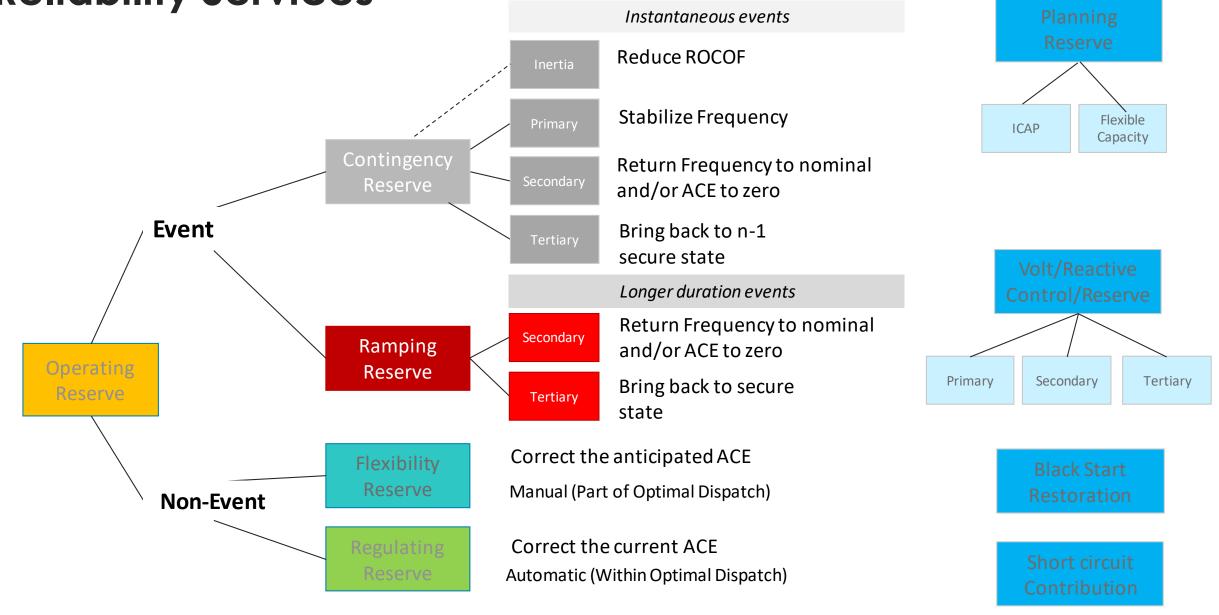
E. Ela, B. Kirby, N. Navid, and J.C. Smith, "Effective ancillary services market designs on high wind power penetration systems," In Proc. IEEE PES General Meeting, July 2012.





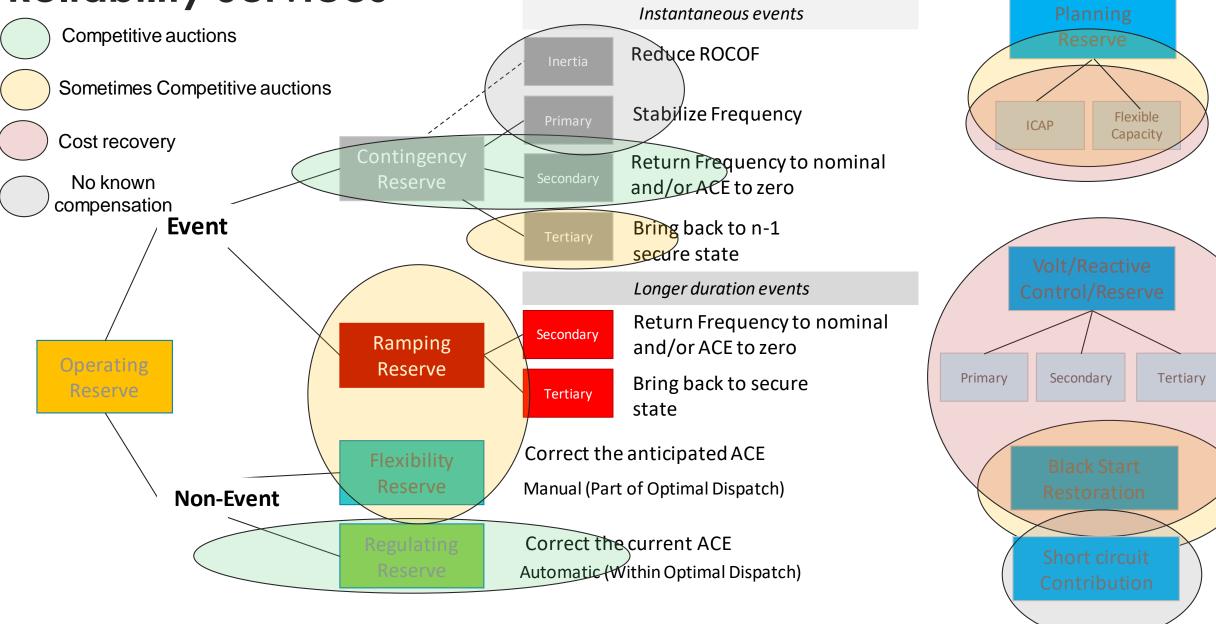


Reliability Services





Reliability Services



EPCI ELECTRIC POWER RESEARCH INSTITUTE

Reliability Service Compensation

- Some reliability attributes are not currently incentivized:
- Sometimes auctions and market-based pricing for certain services may be impractical
- Prioritization of market design and software changes also key
 - It cost money to develop, discuss, test, implement, and administer new designs

Reasons why a market product may not be implemented	Example	
Too complex to design (e.g., software complexity)	Volt/VAR support	
Too specific to certain local areas (little to no competition)	Volt/VAR support	
System inherently has more than sufficient amounts of the service	Synchronous Inertia	
Costs for the service may be small, so cost of administrating market product may outweigh benefits	Black start (restoration) service	
A specific resource requirement rather than a system-wide need	Low Voltage Ride Through	

The examples are used for illustrative purposes only and the reason may not be necessarily true for each example in each region.



Ancillary Service Markets

Relationship with Energy Markets

• Co-optimization and Lost Opportunity Costs (U.S., AEMO) • Sequential and/or contract-based (Europe)

Two-settlement

Two-settlement (most U.S.)

Day-ahead only (ERCOT), Real-time only (ISO-NE)

Price hierarchy and product substitution

• Exact hierarchy varies

Shortage pricing

• High prices set when reserve are short through single step, multi-step, or full demand curves

Ancillary Services typically paid clearing price for capacity, then paid energy price when deployed. Regulation also paid a mileage price for amount of movement.

74

Co-optimization

	Energy Cost	Capacity	Reserve Cost
Gen1	10\$/MWh	100 MW	1 \$/MWh
Gen2	20\$/MWh	100 MW	5 \$/MWh
Gen3	25 \$/MWh	150 MW	15 \$/MWh

Load = 250 MW	Reserve Requirement = 50 MW
---------------	-----------------------------

Sequential

	Energy Schedule
Gen1	100 MW
Gen2	100 MW
Gen3	50 MW
Total	250 MW

Co-optimized

	Energy Schedule	Reserve
Gen1	100 MW	0 MW
Gen2	50 MW	50 MW
Gen3	100 MW	0 MW
Total	250 MW	50 MW

Lost opportunity Cost

	Energy Cost	Capacity	Ramp Rate
Gen1	10\$/MWh	100 MW	1 MW/min
Gen2	20\$/MWh	100 MW	5 MW/min
Gen3	25 \$/MWh	150 MW	8 MW/min

Load = 250 MW	Reserve Requirement = 50 MW, 5-
	minute response required

	Energy Schedule	Reserve	Energy
Gen1	100 MW	0 MW	07
Gen2	90 MW	10 MW	Reserve
Gen3	60 MW	40 MW	Reserve
Total	250 MW	50 MW	

Energy Price = 25/MWh

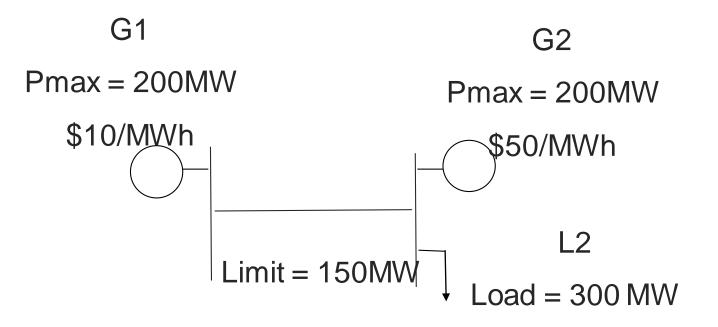
Reserve Price = 5/MWh



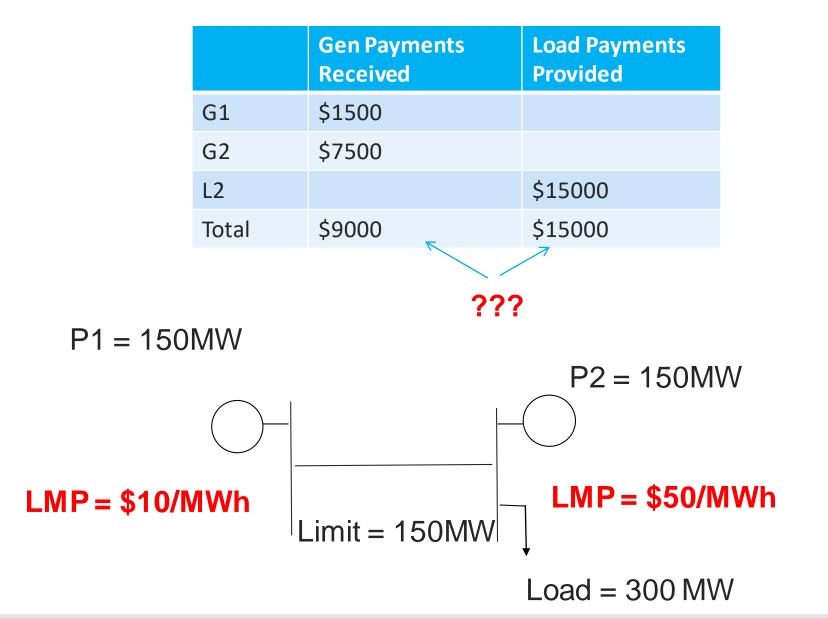














- Physical transmission rights (PTR) cannot allow an efficient optimal scheduling of the supply resources on the transmission system
- PTR transitioned to Financial Transmission Rights (FTR) in organized electricity markets
- When transmission congestion is present in the energy market, there is more money collected from loads than is paid to generators
- Market Participants bid on rights to these moneys through FTR auctions, then "financially" owning the transmission line where congestion occurs
- Do not have to physically own line
- Does not impact scheduling in energy market
- Congestion amounts based on day-ahead congestion, funds from day-ahead congestion
 - Some ISOs guarantee full funding to FTR holders, others do not



FTR market designs

	ISO-NE	NYISO	PJM	MISO	SPP	ERCOT	CAISO	IESO
Name	FTRs	Transmission Congestion Contracts (TCCs)	FTRs	FTRs	Transmission Congestion Rights (TCRs)	Congestion Revenue Rights (CRRs)	CRRs	Transmission Rights (TR)
Mechanism for allocating auction revenues	Point-to-point Auction Revenue Rights (ARRs)	Auction revenues allocated after the auction	Point-to-point ARRs	ARRs	ARRs	Auction revenues allocated after the auction	CRRs directly allocated prior to the auction	Revenues go to the transmission rights clearing account
Auction period for short-term FTRs	Annual and monthly	6-month and monthly	Long-term (3-year), annual, and monthly	Annual and monthly	Annual and monthly	Semi-annual and monthly	Seasonal and monthly	Annual and monthly
Options	No	No	Yes	No	No	Yes	No	No
Classes	Peak and off- peak	Single	Peak, off-peak, and full day	Peak and off- peak	Peak and off- peak	Peak weekday, Peak week-end, off-peak	Peak and off- peak	Single
FTR funding shortfall procedure		Full funding through uplift	FTR shortfalls under-funded	FTR shortfalls under-funded			FTR shortfalls under-funded	Full funding through clearing account





Capacity Markets





Missing Money Example

- Peaking unit is used 20 hours per year
- In those 20 hours, it sets the price based on its offer typically based on its variable cost (fuel and O&M)
- There is no additional rent to use to recover the capital cost that is not reflected in the energy market bid
- When the ISO is short on capacity or reserve in the energy market, it may set the price at scarcity prices
- These prices are higher than the peaker's variable cost and can help recover fixed capital cost
- Regulators often do not allow these scarcity prices to go too high
- Nor do they allow shortage conditions to occur too frequently
- If the peaker is to remain, how does it earn its revenue?

Other means of RA and capital cost recovery

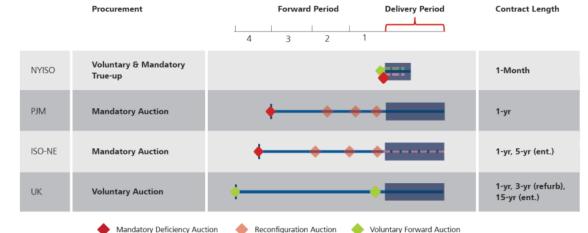
	Regulated Planning (Customers Bear Most Risk)			Market Mechanisms (Suppliers Bear Most Risk)		
	Regulated Utilities	Administrative Contracting	Capacity Payments	LSE RA Requirement	Capacity Markets	Energy-Only Markets
Examples	SPP, BC Hydro, most of WECC and SERC	Ontario	Spain, South America	California, MISO (both also have regulated IRP)	PJM, NYISO, ISO-NE, Brazil, Italy, Russia	ERCOT, Alberta, Australia's NEM, Scandinavia
Resource Adequacy Requirement?	Yes (Utility IRP)	Yes (Administrative IRP)	Yes (Rules for Payment Size and Eligibility)	Yes (Creates Bilateral Capacity Market)	Yes (Mandatory Capacity Auction)	No (RA not Assured)
How are Capital Costs Recovered?	Rate Recovery	Energy Market plus Administrative Contracts	Energy Market plus Capacity Payments	Bilateral Capacity Payments plus Energy Market	Capacity plus Energy Markets	Energy Market ORDC High Scarcity

Pfeifenberger and Spees (2009), Review of alternative market designs for resource adequacy.



Capacity Markets

- Capacity Markets Loads will pay resources strictly based on capacity to meet future resource adequacy needs and reliability "guarantees"
 - Day-ahead must offer requirement
- A reserve margin is used for uncertainties with forced outages or unexpected high demand
 - Typical numbers: 12-18%
- If a peak demand is 10,000 MW then the operator must make sure it has at least 11,500 MW of capacity installed with a 15% reserve margin
 - If capacity is scarce and more capacity needs to be built, the price for capacity will be higher
- These markets can give peaking generators enough money to recover fixed costs that they cannot get in the energy market



G. Anstey and M. Schönborn, "The British Capacity Market: Reflections on a Visible Hand," Energy Market Insights, no. 11, Dec. 2014.

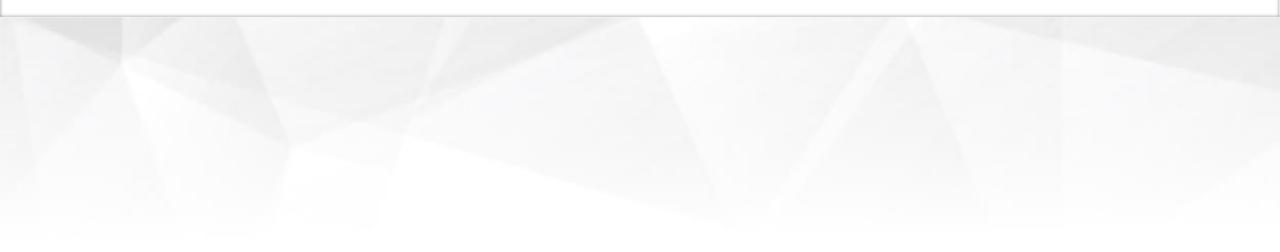


Capacity market designs

	РЈМ	ISO-NE	NYISO	MISO
Name	Reliability Pricing Model (RPM)	Forward Capacity Market (FCM)	Installed Capacity (ICAP) Market	Planning Resource Auction (PRA)
Sloped demand curve	Yes	Yes (only for system wide)	Yes	No
Maximum price	Net CONE	Net CONE	Net CONE	Net CONE
Minimum offer price rule	Yes	Yes	Yes	No
Number of locations	9 zones	3 zones	4 zones	10 zones
Must-offer requirement	DAM energy	DAM energy and RTM energy	DAM energy	DAM energy
Forward period	3 years ahead	3 years ahead	1 month prior	2 months prior
Commitment period	1 year	1 year	6 months	1 year
Product definition	Generic MW	Generic MW	Unforced capacity	Unforced capacity



Emerging Challenges





Key R&D Challenges

- More resources may be needed for fewer periods, but still needed
 - Do they have the long-term incentives to avoid retirement?
 - Will responsive demand and a smarter grid reduce the need for resources?
- Are there proper incentives to get the increased amount of flexibility on the system that is needed to maintain system reliability
 - Short-term: Attract available flexible resources to provide this flexibility
 - Long-term: Attract new or existing facilities to build or modify with flexibility attributes
- How are "essential reliability services" being incentivized and valued?
 - Services that used to be inherent that may be declining
 - Should these be incentivized through market products or otherwise?



Key R&D Challenges (2)

- What is the right price?
 - Non-Convexities, minimize uplift vs. marginal prices, out-of-market vs. in-the-market, alignment of pricing and reliability, market transparence vs. private information
- Should the market internalize policy-related externalities (such as technology favoring policies)?
 - Environmental, job preservation, fuel diversity, local tax
- Interfacing transmission/wholesale with distribution/retail
- Changing resource mixes have new characteristics: Technology agnostic vs. realism?
 - Energy storage technologies, demand response, renewables, distributed energy, natural gas
- Simplicity vs. complexity
 - Do stochastics, reactive power, multi-state configurations need to be part of the market clearing?
- Are there incentives for resources to withstand and recover from extreme events?



Market Evolution

FERC Order	Year Issued	Description
Order 831	2016	Offer Caps in Markets Operated by Regional Transmission Organizations and Independent System Operators; raises the price cap of energy from \$1,000/MWh to \$2,000/MWh
Order 825	2016	Aligns settlement periods with pricing periods, and updates triggers for scarcity pricing
Order 841	2018	Definition of Electric Storage Resource (ESR) and common requirements for the ESR market participation model
Order 842	2018	Order requiring all newly interconnecting resources to have capability to provide primary frequency response
Order 844	2018	Order directing rules for allocation of uplift costs and additional transparency measures for out of market procedures

- Western EIM continues expanding utilities
- DER Aggregation Market Participation Workshops
- Canadian Market Evolution (IESO Market Renewal, AESO Capacity Market)
- Fuel Security Market Designs (PJM, ISO-NE)
- Carbon Pricing (NYISO, PJM)



Common and Unique Initiatives

	Common or Similar Initiatives	Unique Initiatives
Energy Markets	Improved coordinated transaction scheduling (NYISO, ISO-NE, PJM, MISO, SPP, and CAISO) Combined Cycle Modeling Improvements (PJM, MISO, SPP) Virtual spread product (e.g., up-to-congestion, or point-to- point) (NYISO, MISO, CAISO) Pricing of fast start resources (ISO-NE, NYISO, MISO, CAISO)	Do-not-exceed limits for wind generation (ISO- NE) Flow control resources as market participants (MISO) Fifteen minute market (CAISO) Integrated day-ahead market (CAISO)
Ancillary Service Markets	Regulation pay-for performance improvements (PJM, SPP, CAISO) Ramp products (MISO, SPP, CAISO) Primary frequency response (MISO, ERCOT, CAISO)	Synchronous Inertia service (ERCOT) Gas system limitations on reserve provision (NYISO) Reserve product for voltage (MISO)
Financial Transmission Rights	 FTR funding issue resolution and shortfall allocation methods (PJM, MISO, CAISO) Long-term FTR (ISO-NE, MISO, CAISO) Incorporating transmission outages into FTR auctions (SPP, CAISO) 	Third-party FTR clearing (ISO-NE) Rights for PAR-controlled lines (NYISO)
Capacity Markets	Sloped demand curves (ISO-NE, MISO) Locational capacity market improvements (ISO-NE, NYISO, MISO) Performance incentives (ISO-NE, NYISO, PJM) Locational price hedging (NYISO, PJM, MISO)	Flexible capacity procurement (CAISO) Changing from annual to seasonal markets (MISO)



Thank You!!

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Advanced Use of Forecasting

Probabilistic Methods and Other Advanced Techniques

Aidan Tuohy

ESIG Forecasting and Markets Workshop June 4, 2019



 Image: Second system
 Image: Second system

 Image: Second



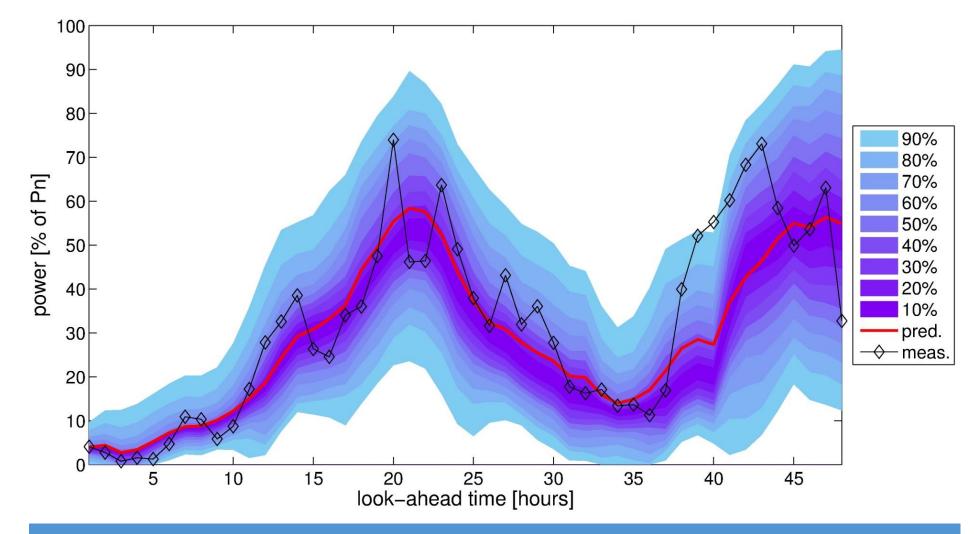
Agenda

Overview of probabilistic forecasts

Use of probabilistic information in operations

Other uses of forecasting – reserves, frequency response, DER

We can't forecast perfectly – so, use probabilistic forecasts



Probabilistic data, in useful form, can be used for more than awareness

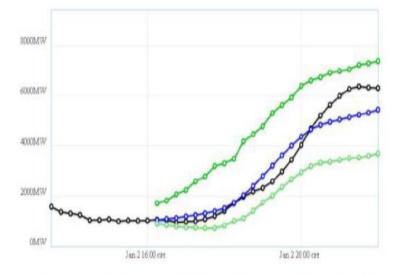
Source: Pierre Pinson, DTU, Denmark

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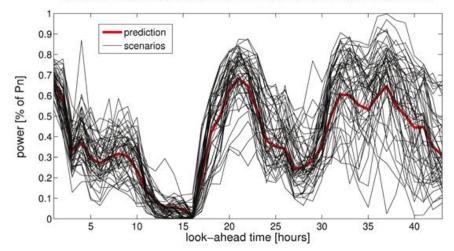


Why use probabilistic forecasts?

- Takes advantage of risk-based methods such as stochastic programming
- Captures outliers (assuming data is representative)
- Should be more economically efficient while can also be more reliable than traditional methods
- May give us better rationale for responding to some extreme events, e.g. extreme cold



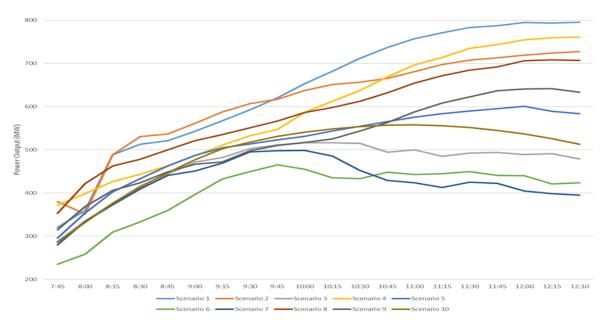
Source: David Maggio, Using probabilistic information in real life, 2011 UWIG Forecasting Workshop





Probabilistic Forecasting Methods

- EPRI studies and others have shown potential benefits of using probabilistic info.
 - WECC system study to examine potential use of this info
 - Showed that use of probabilistic forecasts can improve reliability and/or economics over current deterministic methods
 - Typically in order of few % (<5%)
- Research needs identified in area of probabilistic forecasting
 - Improved probabilistic forecasts that capture the uncertainty in a suitable narrow band
 - Operational tools that can use such forecasts to assist decisions, e.g. operating reserve requirements
 - Education on how these tools could be used and further developed
 - Identifying different use cases ISO and utility use cases



PV output scenarios

Probabilistic forecasts can improve operations and provide situational awareness - but we need to know how best to use them



Probabilistic Forecasting Applications

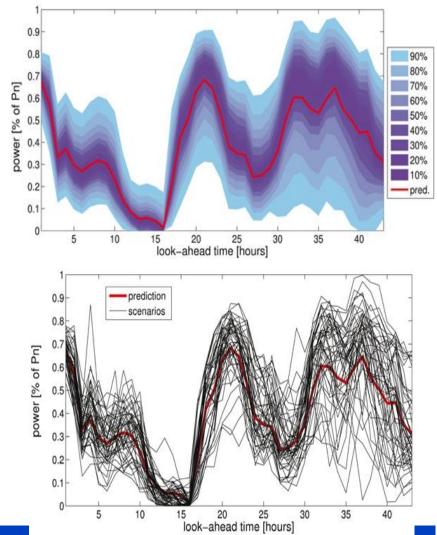
Application	Description	In use?
Stochastic Scheduling (STUC)	Committing or dispatching resources to meet multiple potential scenarios based on minimizing the expected cost	N/A
Robust optimization-based Unit commitment and dispatch	Similar to STUC, but only using the "worst case" scenario to ensure the decisions are robust towards those situations	N/A
Do Not Exceed Dispatch VER Limits	Using probabilistic forecasts to determine the likelihood of VER providing greater energy than expected that can exceed balancing capabilities or network limit violations	ISO-NE
Outage Planning	Using long-term probabilistic forecasts to determine whether it is optimal to take certain units (including non-renewable ones) or transmission facilities out for maintenance when it is likely they are not needed and economic losses are minimal	N/A
Energy Market Trading (E.g., virtual trading)	Using potential scenarios and probability of those scenarios to assist in taking financial positions within the market (either as a VER, load, or virtual trader). Longer term probabilistic forecasts can also be used for long-term markets, like financial transmission rights.	Yes
Level of frequency response	Using probabilistic forecasts to understand the level of frequency response (or synthetic inertia) a wind plant or collection of wind plants can provide	N/A

Others: Real-time contingency analysis, topology switching, reserve scheduling



Generation of Probabilistic VG Scenarios - Motivation

- With increasing levels of VG, operations will be more concerned with uncertainty than before → need tools that can manage this uncertainty
- Uncertainty can be characterized in a number of ways and is already provided by some vendors
 - Probabilistic information as a distribution
 - Confidence intervals or exceedance levels
 - Ensembles/multiple forecasts
 - Scenarios for how the future may unfold
- Operations or Integration study applications



Probabilistic data, in useful form, can be used for more than awareness



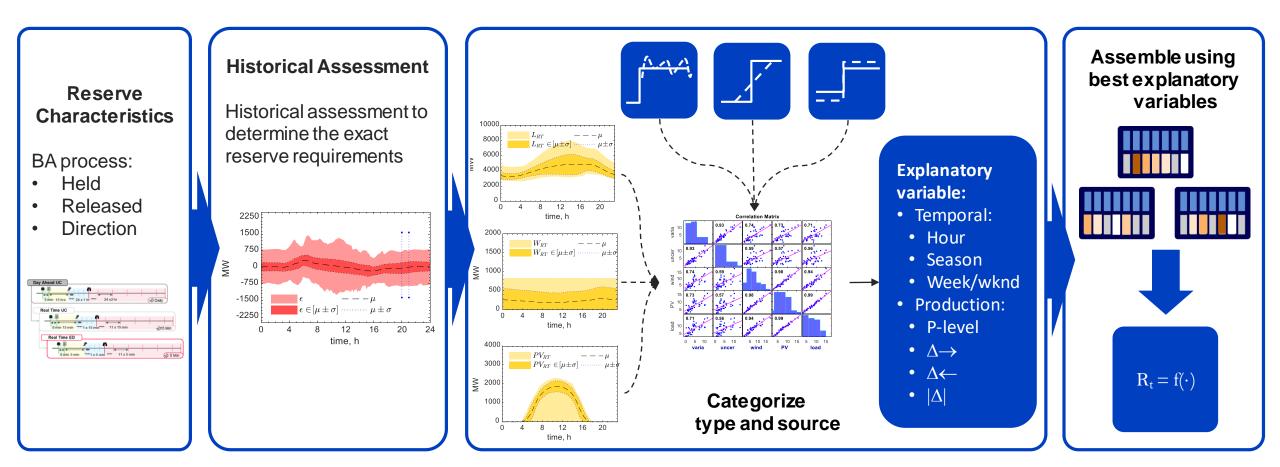


Other advanced forecasting uses





EPRI Dynamic Reserve Requirement Method



Reserve requirement is "forecasted," just like load or RES production

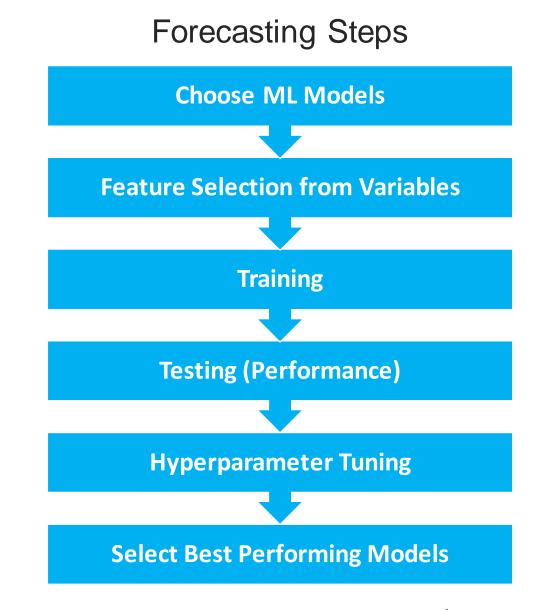


2018 Reserve Forecasting Improvements

 Current EPRI "Clustering Method" uses correlations across explanatory variables to create a lookup table of all reserve needs and impact sources (load, wind, solar); that aggregated produce the forecast for reserve requirements

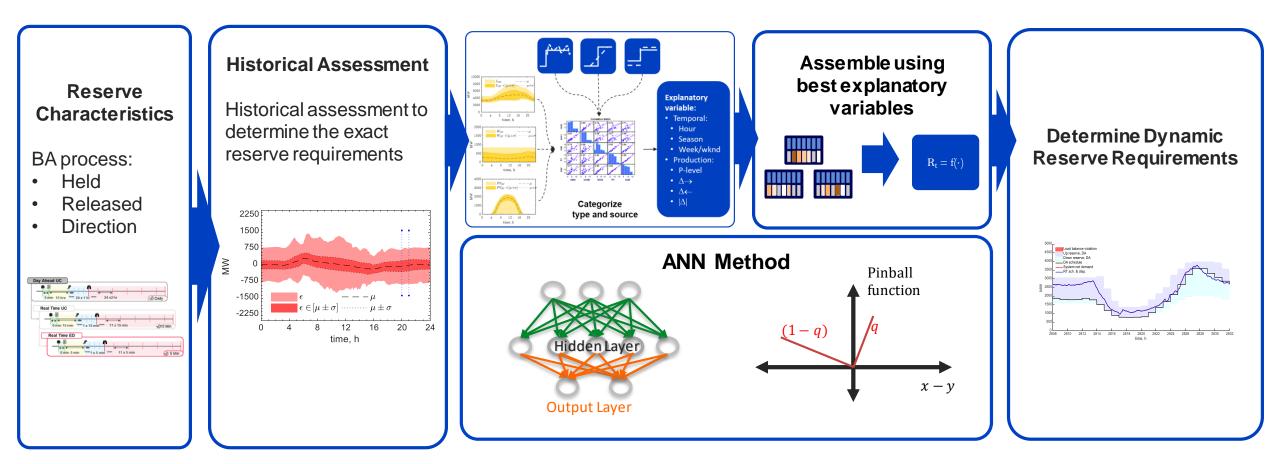
Motivation for 2018:

- Apply available machine-learning (ML) tools for reserve forecasting to improve performance over existing methods
- ML methods have been successfully applied to power system forecasting applications and exist in commercial tools
- Test, study performance, and implement ML methods in DynADOR





Dynamic Reserve Requirement Methods



Alternative Methods to Determine Dynamic Reserve Requirements



Related: HECO Study Overview and Conclusions

- Evaluate current, proposed (now existing), and recommended operating reserve requirement methods on Oahu as well as other operational recommendations
- Study impacts of higher VER
 - Utility scale: 108 MW to 287 MW; DPV: 291 to 564MW
- Simulation using FESTIV tool (commitment, dispatch, AGC)
- Metrics: Costs, ACE, frequency error, HECO compliance, head room risk
- Conclusions:
 - Cycling of mid-merit resources for balancing can provide economic benefits
 - However, other reasons for must-run status must be considered
 - Combined use of cycling and advanced dynamic reserve requirements can provide economic and reliability benefits
 - Estimated \$21-\$24M annual savings (4%) in addition to improved reliability
 - VER Curtailment during high ACE required on future system
 - Evaluate frequency responsiveness as part of reserve providers



EPRI DOE Solar Forecasting Project- Three Workstreams

- 3 Year Project, anticipated \$1,8M DOE funding, \$760k EPRI/utility cost share (\$110k from 173.05 over 3 years)
- A Forecasting Work Stream to develop and deliver probabilistic forecasts with targeted improvements for utility scale and behind-themeter (BTM) solar
- A Design Work Stream to identify advanced methods for managing uncertainty based on results from advanced scheduling tools

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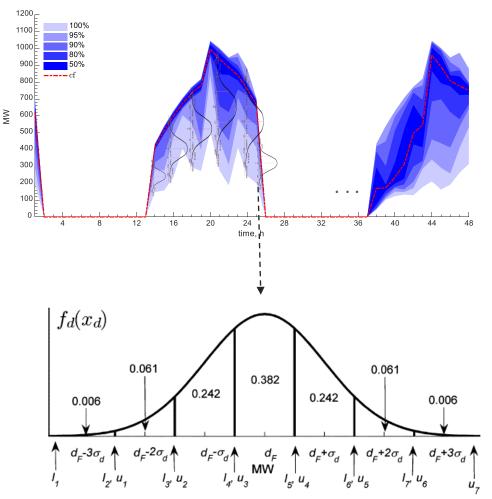
 A Demonstration Work Stream to develop and demonstrate a scheduling management platform (SMP) to integrate probabilistic forecasts and scheduling decisions in a modular and customizable manner





Probabilistic Forecasts in DynADOR

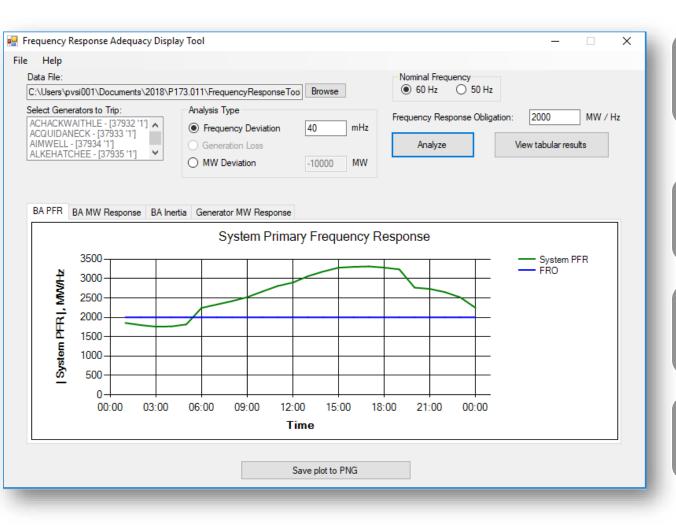
- Reserve determination using probabilistic forecasts
- Part of the DOE-funded OPTSUN project
- Determination:
 - For a given probabilistic forecast, determine the likelihood of net demand materialization
 - For each possible net demand combinations, determine dynamic reserves
 - The dynamic reserve requirements are given by the aggregation of the weighted individual reserve requirements



M. A. Ortega-Vazquez and D. S. Kirschen, "Estimating the Spinning Reserve Requirements in Systems with Significant Wind Power Generation Penetration," *IEEE Transactions on Power Systems*, Vol. 24, Issue 1, pp. 114-124, Feb. 2009.



Forecast Frequency Response



Calculate anticipated PFR and compare to FRO

• PFR actually depends on event

Response for a predefined frequency deviation (+/-)

Response for a predefined MW deviation (+/-)

Individual generator MW responses, as well as total load response.



FRADT: "What-if' Mode

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115





FRADT: Case Study

Forward Looking Analysis of Eastern Interconnection (EI) Primary Frequency Response

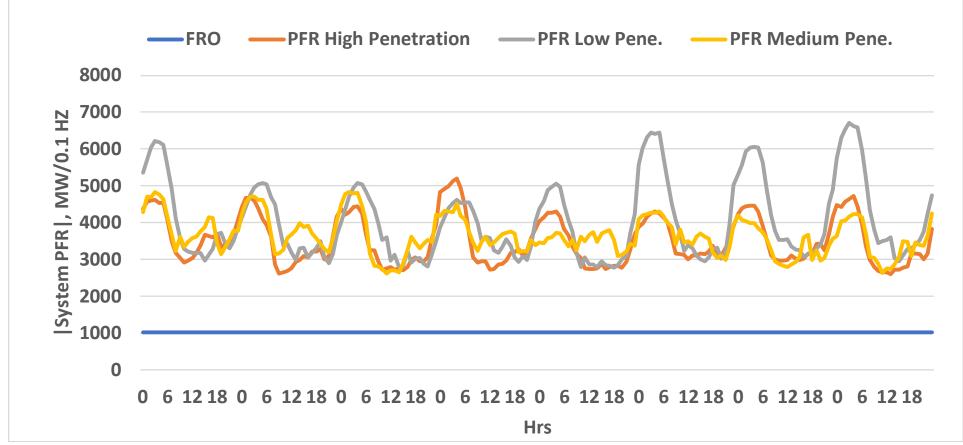
- A case study to show how the tool can be used at an interconnection level for forward looking analysis.
 - Aim of this study is to demonstrate the application. Results have not been vetted and should only be viewed as a way to analyze results of the tool, not as future predictions
 - Detailed insights on interconnection level results will be provided in 2019.
- Scenarios developed in NREL's Eastern Renewable Generation Interconnection Study (ERGIS) are used in this FRADT case study
 - More details on ERGIS study can be obtained from the NREL website <u>https://www.nrel.gov/grid/ergis.htm</u>; Report: <u>https://www.nrel.gov/docs/fy16osti/64472.pdf</u>





FRADT: Case Study

Application using ERGIS Data - El Primary Frequency Response for -0.1 HZ Frequency Deviation



The plot shows that EI has adequate PFR for -0.1 Hz frequency deviation, based on the assumptions made





Conclusions

- Advanced uses of forecasting are being developed in a few areas
 - Use of uncertainty information
 - Development of new operating practices (reserves, frequency response)
- Probabilistic information is beginning to be used for more than situational awareness

 Other operational processes can use VER forecasts and also concepts developed in forecasting area



Together...Shaping the Future of Electricity



