

# **Electricity Markets and Renewable Energy: United States vs. Europe**

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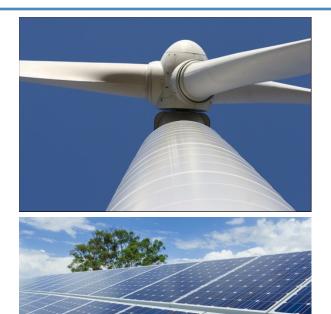


#### ETH, Zurich, January 27, 2017



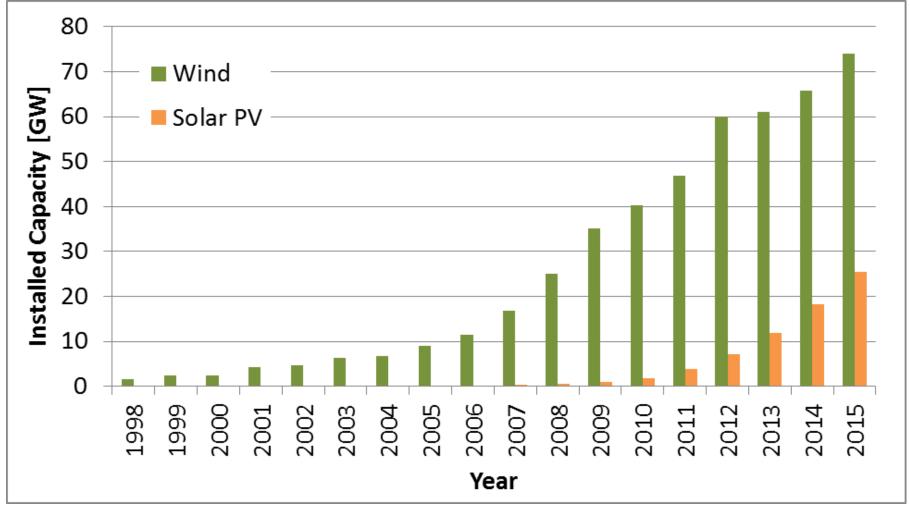
## **Outline**

- Background
  - Renewable energy penetration levels
  - Electricity market operations
  - Core research questions
- Addressing Renewables in Short-term Operations
  - Operating reserve demand curves
- Addressing Renewables in Long-term Planning
   Generation Expansion and Revenue Sufficiency
- Concluding Remarks



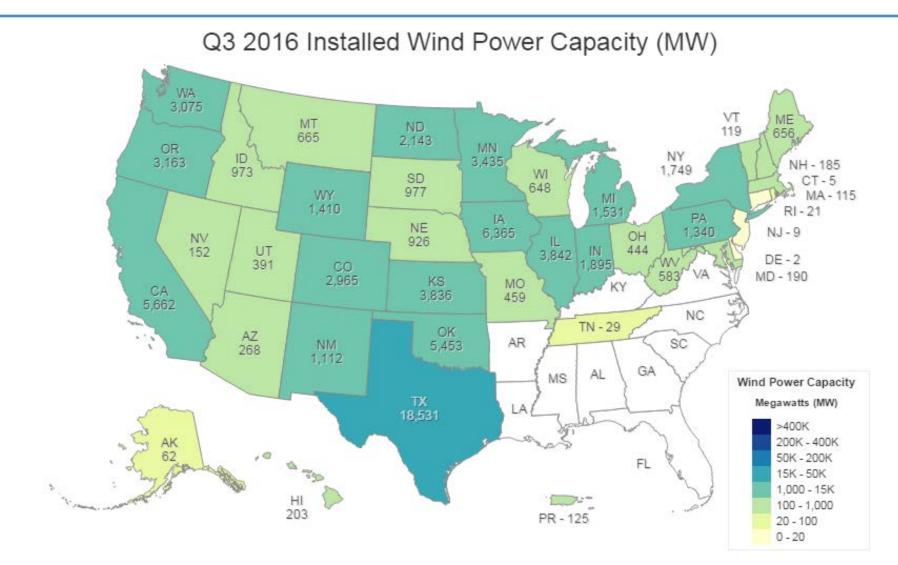


### Rapid Growth in U.S. Renewable Energy Capacity



2015: Wind/solar meet 5%/1% of U.S. electricity load

## Wind Power Capacity in the United States



#### Total Installed Wind Capacity: 75,714 MW

Source: American Wind Energy Association Q3 2016 Market Report

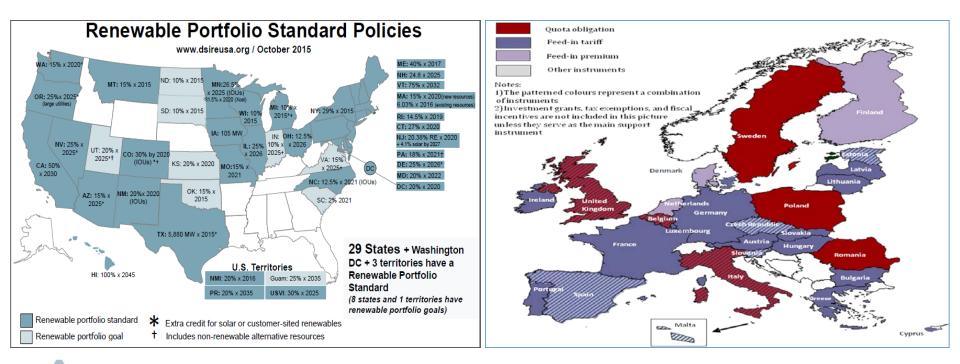
## **Subsidy Schemes for Renewables**

#### United States (13.5% renewables in 2014)

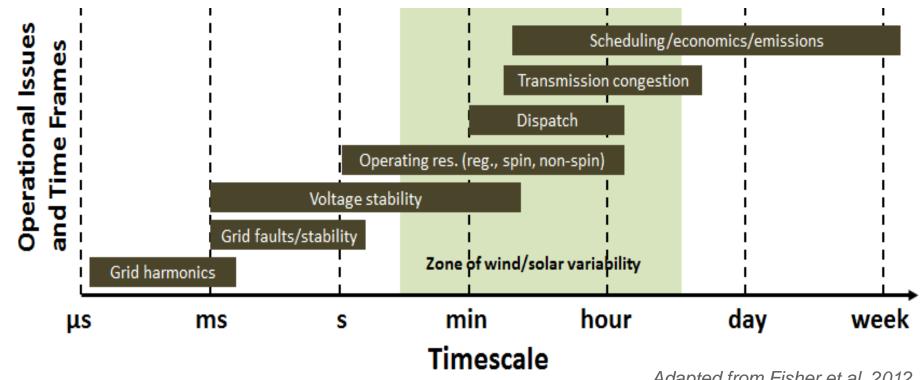
- Direct support for renewables
  - Production/Investment tax credits (Federal)
  - Renewable portfolio standards/goals (State)
- Climate change policies
  - EPA's Clean Power Plan (stayed by Supreme Court)
  - Regional cap and trade programs

#### **Europe** (27% renewables in 2014)

- Direct support for renewables
  - Feed-in tariffs for wind and solar
    - Fixed to premium tariffs and auctions
  - Green certificates
- Climate change policies
  - European Emissions Trading System (ETS)



## **Operating the Power Grid is Complex**



Adapted from Fisher et al. 2012

Handling uncertainty in a cost-effective manner is at the core of the problem.

## The General Structure of Electricity Markets

#### United States

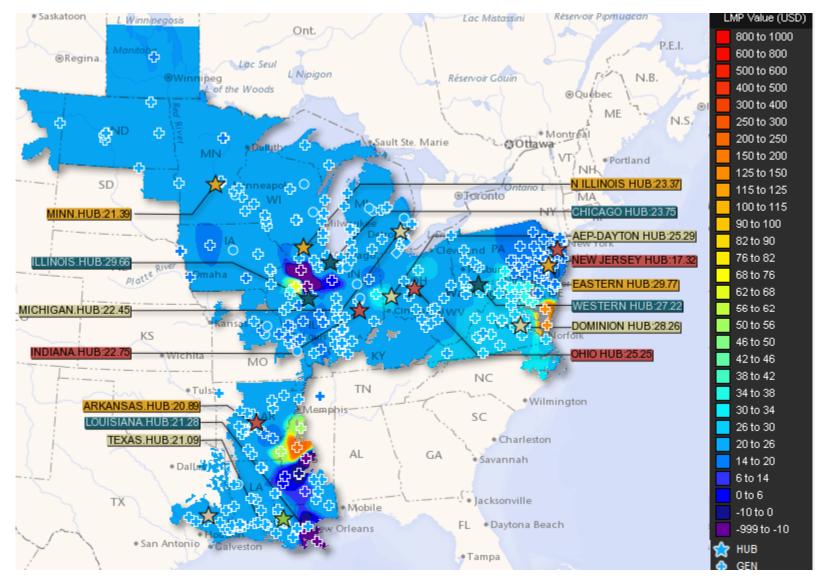
- Build into existing system operators (ISOs)
  - Emphasize physics of the power system
  - Short-term system operation
  - ISOs do not own transmission system
- Market design elements (United States)
  - Day-ahead market (ISO hourly)
  - Real-time market (ISO 5 min)
  - Complex bids/ISO UC
  - Locational marginal prices
  - Co-optimization of energy and reserves

- Europe
  - Introduced new power exchanges (PXs)
    - Emphasize markets and economics
    - Includes long-term contracts
    - TSOs typically own transmission system
  - Market design elements (Europe)
    - Day-ahead market (PX)
    - Real-time balancing (TSO)
    - Simple bids/generator UC
    - Zonal pricing/market coupling
    - Sequential reserve and energy markets

- Variable Renewable Energy (VER)
  - "Dispatchable" VER

- Variable renewable energy (VER)
  - VER as "must-take"

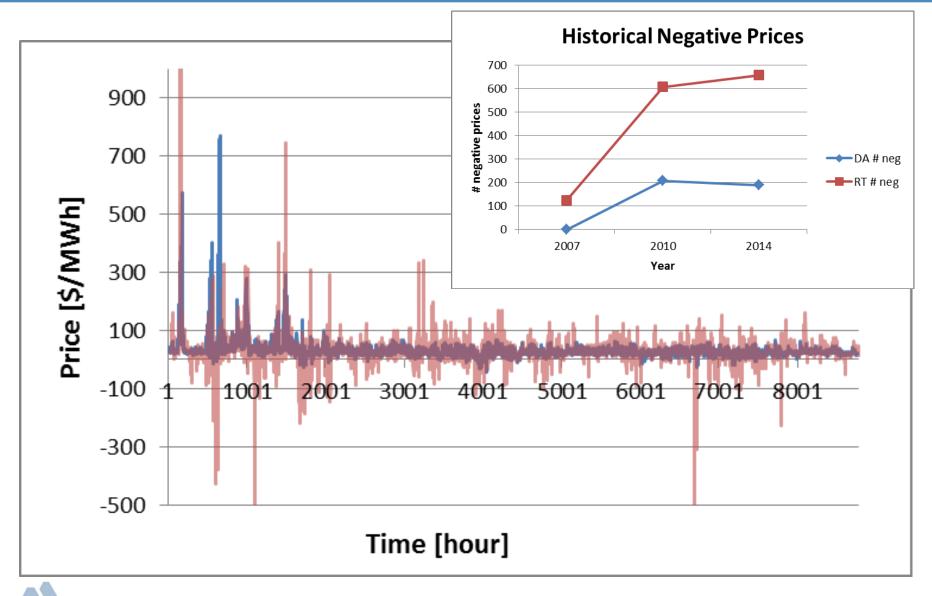
### **Congestion Management and Locational Marginal Prices (US)**



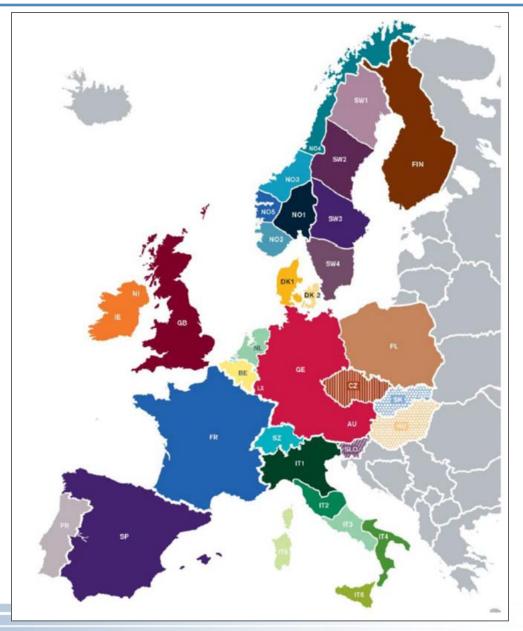
Nov 18 2015, 8.30am

www.miso-pjm.com

### Wind Power Influences Electricity Prices Today: Day-ahead and Real-time Prices at Node in Illinois (2014)

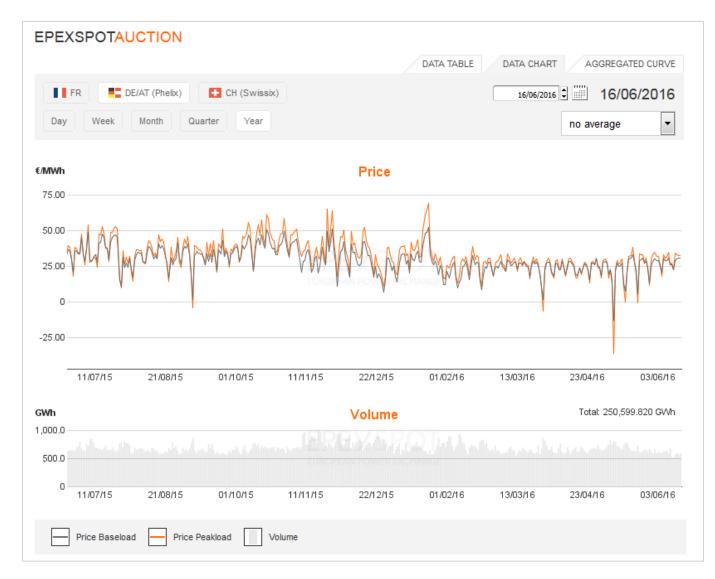


### Price Zones in European Day-Ahead Markets



Courtesy: Hans Auer, TU Wien

### **Electricity Prices in Central-Western Europe 2016**



Central-Western Europe: EPEXSPOT (2016)

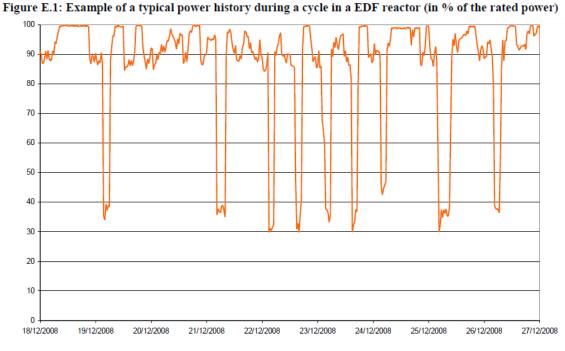
## **Distributed Generation and End-User Tariffs**

- "Net metering" a very hotly debated topic
- Example: Residential customer bills in Boston and Vienna
  - Annual consumption: 5000 kWh



### Nuclear Power: A Source of Flexibility?

- Nuclear shut-downs
  - Primarily due to economics (US)
  - Public resistance (Europe)
- Importance of nuclear flexibility with increasing renewable penetration levels
  - United States: Nuclear energy is currently baseload
  - Europe: Flexible nuclear operations (e.g. France, Germany)



Source: NEA, 2011

### **Core Research Questions**

- How to plan and operate power systems with increasing shares of renewable generation?
- How to design electricity markets to provide adequate incentives for market participants?
- How to facilitate the transition to low-carbon power and energy systems?
- Develop improved analytics for evolving electricity markets and energy systems
- Methods
  - Power systems engineering
  - Energy economics and policy
  - Optimization: Mathematical programming
  - Decision theory
  - Agent-based simulations
  - Game theory

*If the Only Tool You Have is a Hammer, then Every Problem Starts Looking Like a Nail...* 



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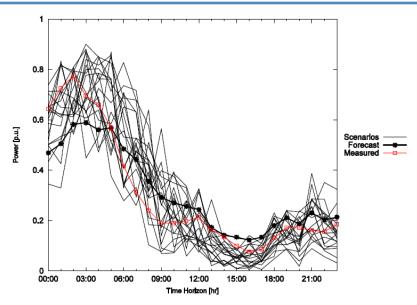






## Improved Market Operations with Renewables

- How do we best address increasing uncertainty and variability in system/market operations?
- Forecasting of wind and solar power
  - Importance of estimating forecast uncertainty (e.g. conditional kernel density estimation)
- Stochastic unit commitment
  - Minimizes expected cost across uncertainties
  - Implicit operating reserves
  - Most studies show significant benefits
  - Several challenges for real-world implementation
    - Computational aspect
    - Pricing and market implementation
  - Limited industry applications so far
- Improved operating reserve strategies
  - Dynamic/probabilistic reserves
    - New reserve categories



Bessa et al. Renewable Energy, 40(1): 29-39, 2012.

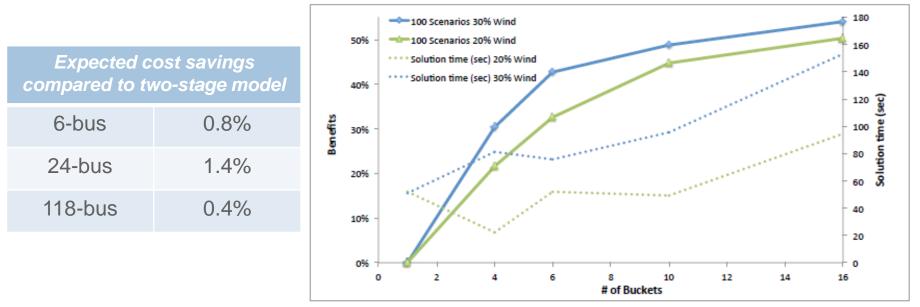
Study	Stochastic UC Cost Savings
Gröwe et, al. (1995)	1.6%
Takriti et. al. (1996, 2000)	0.4-4.0%
Tuohy et. al. (2008)	0.6%
Wang et. al. (2008)	1.3%
Pappala et. al. (2009)	2.8-3.8%
Ruiz et. al. (2009, 2010)	0.8-1.8%
Constantinescu et. al. (2011)	1.0%
Wang et al. (2011)	2.9 %
Zhou et al. (2013)	1.7 %
Papavasilliou and Oren (2013a,b)	1.9-5.4%

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Δ

### An Improved Stochastic Unit Commitment Formulation

- Traditional two-stage stochastic unit commitment model
  - Unit commitment decisions are the same across all scenarios.
- New improved approach
  - Unit commitment decisions depend on wind forecast level ("bucket") and time segments, i.e., more flexible solutions, approaching a multi-stage formulation.



Expected cost benefit\* and solution time (6-bus)

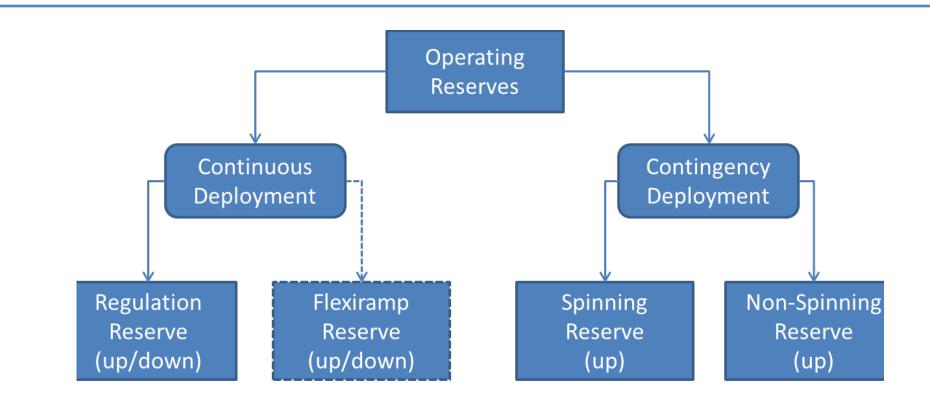
\* Percent of multi-stage benefit on 2-stage.

C. Uckun, A. Botterud, J. Birge, "An Improved Stochastic Unit Commitment Formulation to Accommodate Wind Uncertainty," *IEEE Transactions on Power Systems, 31 (4), 2507-2517, 2016.* 

## A Short History of Operating Reserves with Renewable Energy

Application	Method	Reference
Probabilistic reserve estimates with renewable energy	Meet probabilistic reliability target assuming Normal distribution for renewable energy forecast error	Söder 1993 Doherty and O'Malley 2005 Makarov et al. 2008 2009
	Reserves in UC to minimize total system cost, including expected load shedding, assuming Normal distribution for forecast errors	Ortega-Vazquez et al. 2009
	Reserve evaluation based on <b>probabilistic wind</b> power forecasts	Matos Bessa 2011 Zhou and Botterud 2014
	Operating reserve demand curves	Hogan 2005 ERCOT 2013b Zhou and Botterud 2014
	New reserve products (flexi-ramp)	Wang, Hobbs 2014, 2015
	Reserves in integration studies and industry developments	Ackerman et al. 2007 Ela et al. 2011 Holttinen et al. 2013 Enernex 2010 GE Energy 2010 Lew et al. 2013 Mills et al. 2013 ERCOT 2013a ERCOT 2013b Ellison et al. 2012 Ela et al. 2014

## **Evolving U.S. Market for Operating Reserves**

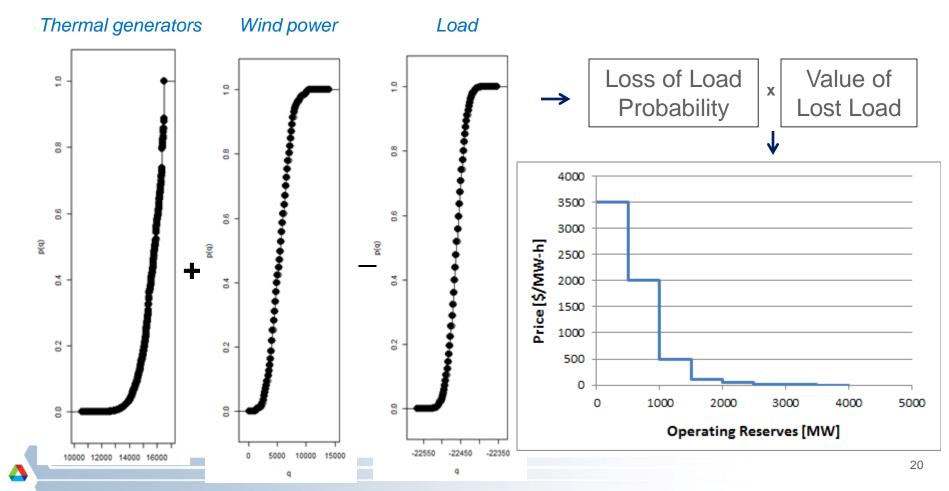


- Regulation reserve: Pay for performance
- "Flexi-ramp reserves" to ensure sufficient ramping capability available in real-time
  - California ISO (CAISO)
  - Midcontinent ISO (MISO)

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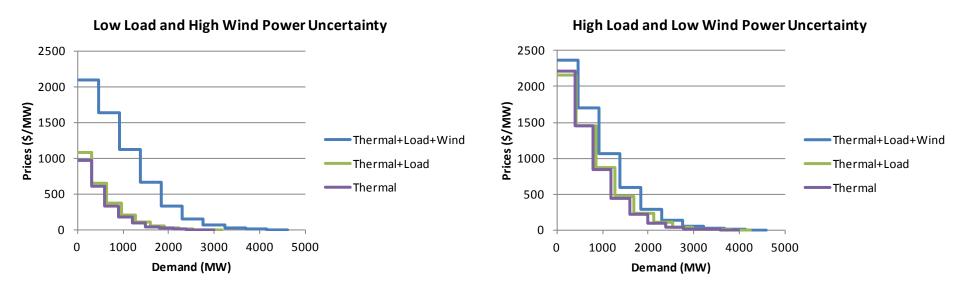
## A Dynamic Operating Reserves Demand Curve (ORDC)

- Consider the uncertainties from load and supply
  - Probabilistic wind power forecast based on Kernel Density Estimation (Bessa et al. 2012)
- Estimate the risk of supply shortage for system
- Link the expected cost of this risk to the price to pay for reserves (Hogan 2005)



## **ORDCs Depend on the Wind Power Forecast**

- Operating reserve demand curves for more efficient market pricing
  - Demand for reserves is dynamic and varies by situation (e.g. wind power forecast uncertainty)



## **Unit Commitment with ORDCs**

**Objective:** 

$$\operatorname{Min} \sum_{t,i} \left[ fc_{t,i} + C(\operatorname{ens}_t) + \operatorname{sc}_{t,i} \right] \left( -\sum_t pfr_t \right) \quad \begin{array}{l} \text{Expected utility of} \\ \text{operating reserves} \end{array}$$

#### **Constraints:**

- (1) Load-generation balance:  $\sum_{i} p_{t,i} + wg_t + ens_t = D_t \forall t$
- (2) Wind generation dispatch  $wg_t \leq W_t \ \forall t$

(3) Generation  $p_{t,i} = PMIN_i \times u_{t,i} + \sum_k delta_{t,i,k} \quad \forall t, i$   $p_{t,i} \leq pbar_{t,i} \quad \forall t, i$   $pbar_{t,i} \leq PMAX_i \times u_{t,i} \quad \forall t, i$ 

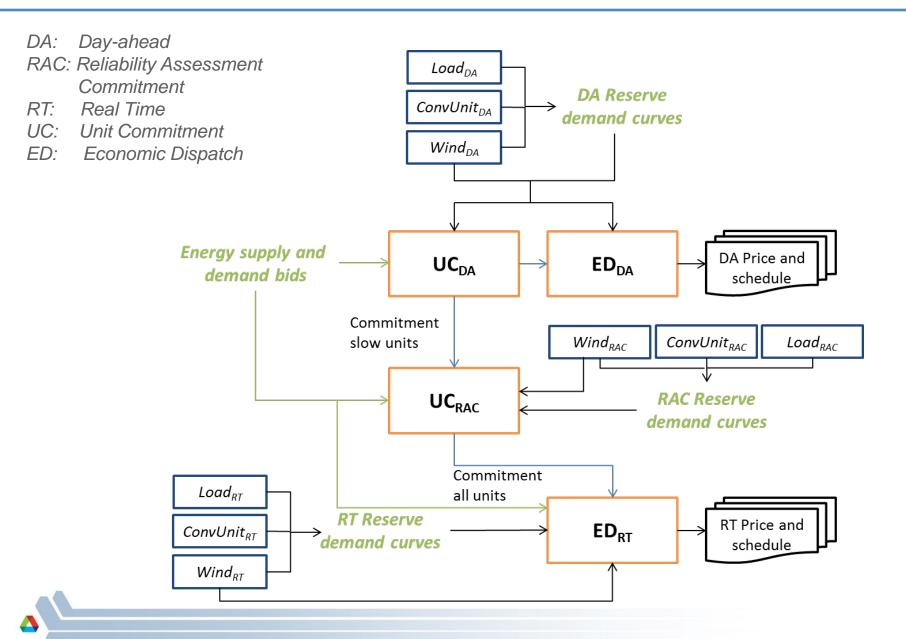
(4) Energy block  $delta_{t,i,k} \leq PR_{i,k} \quad \forall i, k$  (5) Spin/non-spin reserve balance  $pbar_{t,i} - p_{t,i} \ge sr_{t,i} + nsrn_{t,i} \quad \forall t, i$   $nsrn_{t,i} \le NSRRESPONSETIME \times MSR_i \times u_{t,i} \quad \forall t, i$   $nsrf_{t,i} \le QSC_i \times (1 - u_{t,i}) \quad \forall t, i$   $deltasrd_{t,k} \le SRDCR_k \quad \forall t$   $\sum_i sr_{t,i} \ge SRRATIO \times \sum_k deltsrd_{t,k} \quad \forall t$   $\sum_i (nsrn_{t,i} + nsrf_{t,i}) \ge NSRRATIO \times \sum_k deltsrd_{t,k} \quad \forall t$  $sr_{t,i} \le SRRESPONSETIME \times MSR_i \times u_{t,i} \quad \forall t, i$ 

6) Utility from reserve demand  $pfr_t = \sum_k SRDC_k \times deltsrd_{t,k}$ ,  $\forall t$ 

(7) Shut down and start up

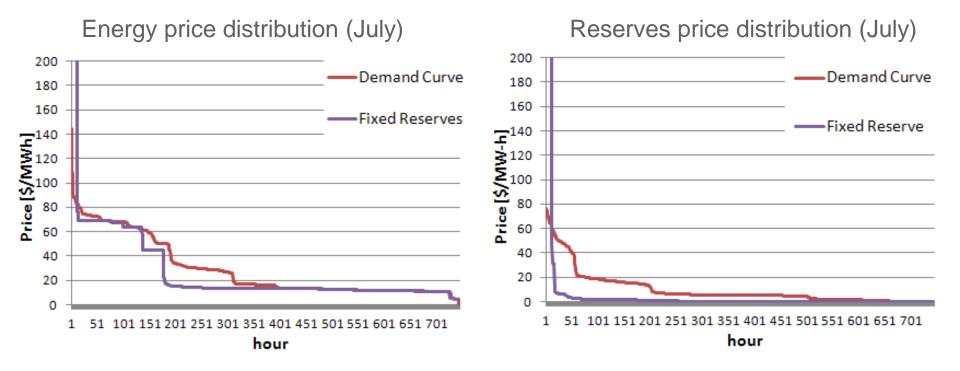
- (8) Minimum up and startup
- (9) Minimum down and shutdown
- (10) Ramp up and Ramp down

## Simulating Electricity Market Operations



## **OR Demand Curves: Implications for Market Prices**

Results from simulation of co-optimized electricity market (Illinois case)



- Benefits of a dynamic demand curve for operating reserves
  - Gives higher prices for energy and reserves in most hours, fewer extreme price spikes
  - Stabilizes revenue stream for thermal generators
  - Better reflects wind power forecast uncertainty in prices
  - Adds demand flexibility to the scheduling and dispatch

### **ORDCs have been Introduced in the ERCOT Market**

ERCOT (Texas) implemented an ORDC in the real-time market June 1<sup>st</sup> 2014

- Price Adders for Online and Offline Reserves
- Online Reserve Price Adder (RTORPA) is added to the LMP

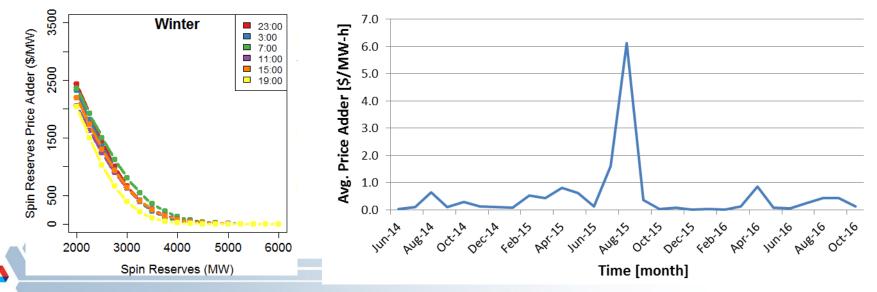
$$VOLL \qquad LOLP$$

$$VOLL \qquad VOLL \qquad VOLP$$

$$RTORPA = P_{S} = v * 0.5 * \pi_{S}(R_{S}) + P_{NS}$$

$$RTOFFPA = P_{NS} = v * (1 - 0.5) * \pi_{NS}(R_{SNS})$$

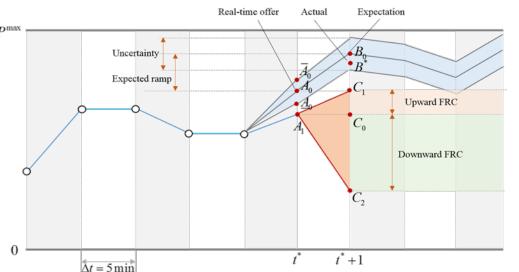
- "The price after the addition of RTORPA to LMPs approximates the pricing outcome of Real-Time energy and Ancillary Service co-optimization since RTORPA captures the value of the opportunity cost of reserves based on the defined ORDC." ERCOT 2014



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## Wind Power Providing Flexible Ramp Capacity (FRC)

- Wind turbines are capable adjusting their active power output at the rate of 0.05-0.25 p.u./s.
- By operating at a sub-optimal operating point, the wind power producers (WPPs) are capable of offering ramping service.
- Though the WPPs' opportunity cost of providing FRC is high, it is economic if frequent commitment of fast-start unit can be avoided.



Including the WPPs in the real-time market as FRC providers

$$\hat{f} = \sum_{t \in T} \sum_{i \in FG} u_{i,t} c_i^{SU} + \sum_{s \in S} \rho_s \left[ \Delta t \sum_{t \in T} \sum_{i \in G} C_i \left( p_{i,t,s}^g \right) + VOLL \cdot \sum_{t \in T} \Delta l_{t,s} + \omega \left( \Delta r_t^u + \Delta r_t^d \right) + \sum_{t \in T} \sum_{j \in W} \left( \pi_j^u r_{j,t,s}^{w,u} + \pi_j^d r_{j,t,s}^{w,d} \right) \right]$$

Considering the impact of uncertainty

#### Compensating the WPPs for providing FRCs

$$\begin{split} \sum_{i \in G} r_{i,t,s}^{g,u} + \sum_{j \in W} r_{j,t,s}^{w,u} \geq & \left[ \hat{L}_{t+1,s} - \sum_{i \in G^{II}} \left( \Delta p_{i,t+1,s}^{g,su} - \Delta p_{i,t+1,s}^{g,sd} \right) - \sum_{j \in W} \hat{p}_{j,t+1,s}^{w} \right] - \left( L_{t,s} - \Delta l_{t,s} - \sum_{j \in W} \hat{p}_{j,t,s}^{w} \right) + \xi^{u} : \beta_{t,s}^{u}, \forall t, \forall s \\ \sum_{i \in G} r_{i,t,s}^{g,d} + \sum_{j \in W} r_{j,t,s}^{w,d} \geq & - \left[ \hat{L}_{t+1,s} - \sum_{i \in G^{II}} \left( \Delta p_{i,t+1,s}^{g,su} - \Delta p_{i,t+1,s}^{g,sd} \right) - \sum_{j \in W} \hat{p}_{j,t+1,s}^{w} \right] + \left( L_{t,s} - \Delta l_{t,s} - \sum_{j \in W} \hat{p}_{j,t,s}^{w} \right) + \xi^{d} : \beta_{t,s}^{d}, \forall t, \forall s \end{split}$$

**Overall FRC supply** 

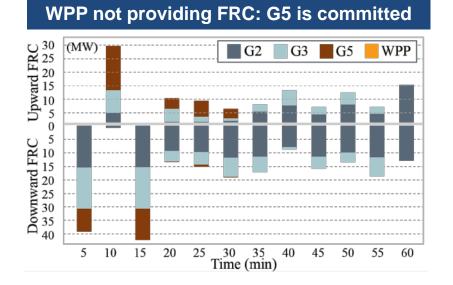
#### **Certain system-wide variation**

#### Margin for uncertain variation

Chen et al., IEEE Trans. Power Systems, in press.

## Wind Power Providing Flexible Ramp Capacity (FRC)

In the case system, by utilizing the WPPs' capability of providing FRC, the commitment of expensive fast-start G5 is effectively reduced. Meanwhile the WPPs are compensated for providing the service, the overall system cost is reduced for most of the cases.

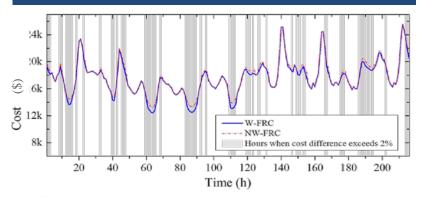


Expected running hours (h) 250 216 216 216 216 216 216 W-FRC 200NW-FRC 143.75 150 110 119.5 10050 G1 G2 G3 G4 G5 Units

The commitment of G5 is effectively reduced

WPP providing FRC: G5 is not committed (MW) G2 G3 G5 WPP WPP provides-upward FRC 0 5 Downward FRC 10 15 20 25 30 WPP provides downward FRO 35 40 5 10 15 20 25 30 35 45 50 55 60 40Time (min)

System cost is reduced for most of the cases



Chen et al., IEEE Trans. Power Systems, in press.

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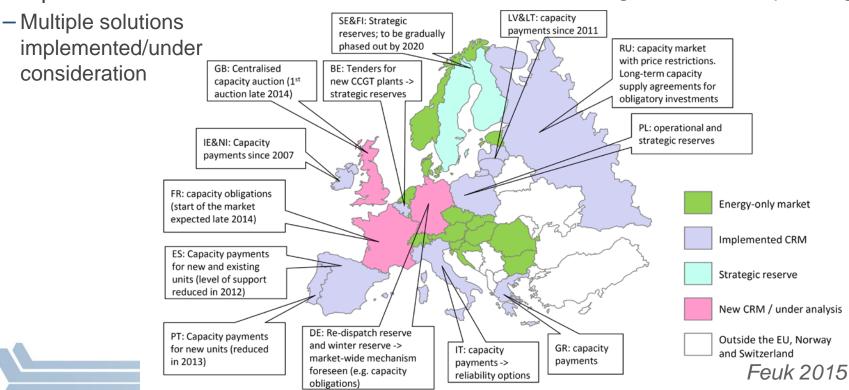
## **Resource Adequacy and Capacity Mechanisms**

Open questions

Europe

- Do we need specific resource adequacy mechanisms?
- Do we need to incentivize capacity with specific attributes?

- United States
  - Capacity markets
    - PJM, NE-ISO, NYISO, MISO
  - Capacity obligations
    - CA-ISO
  - Energy Only
    - ERCOT/Texas
  - Integrated resource planning



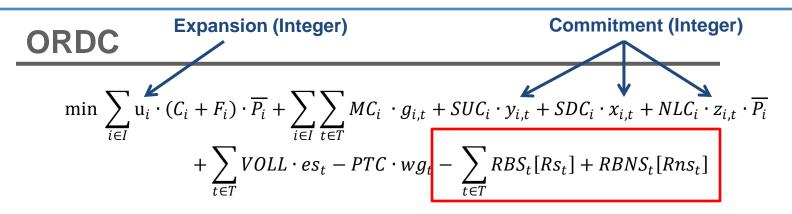
### Generation Expansion and Capacity Adequacy in Texas (ERCOT market)

- Model three different market polices to value reserves, energy and capacity
  - -1) Operating Reserves Demand Curve
    - ERCOT curves, but co-optimization
  - -2) Fixed Reserves Scarcity Pricing
    - Used in most U.S. markets
    - We assume:
      - \$100/MW-h spin-up
      - \$500/MW-h total reserve
  - -3) Capacity Payments
    - \$40/kW-year
    - No reserve scarcity pricing
- Case Study of ERCOT market in Texas
   4 thermal unit types (Nuclear, Coal, NGCC, NGCT)

  - 2013 ERCOT wind and load profile
  - -2024 total load projection (15% growth)
  - Wind varies from 10% to 40% of total demand

Parameter	Value
Peak Load (MW)	77,471
Existing Generation Capacity (MW)	73,380
Nuclear	4,400
Coal	19,500
NGCC	43,600
NGCT	5,880
Maximum Wind Resource Capacity Factor	33.0%

### **Centralized Generation Expansion: Formulation**



**Reserve Scarcity Pricing (FRSP) / Capacity Payment (CP)** 

$$\min \sum_{i \in I} u_i \cdot (C_i + F_i - CP) \overline{P_i} + \sum_{i \in I} \sum_{t \in T} MC_i \cdot g_{i,t} + SUC_i \cdot y_{i,t} + SDC_i \cdot x_{i,t} + NLC_i \cdot z_{i,t} \cdot \overline{P_i} \\ + \sum_{t \in T} ESC \cdot es_t + SRSC \cdot srs_t + NRSC \cdot nrs_t \\ \sum_{i \in I} rs_{i,t} + wr_t + rss_t = RRs_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + RRns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = RRs_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = Rns_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = Rns_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = Rns_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = Rns_t + Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = Rns_t \forall t \in T \\ \sum_{i \in I} (rs_{i,t} + rns_{i,t}) + wr_t + rnss_t = Rns_t \forall t$$

## **Centralized Generation Expansion: Formulation**

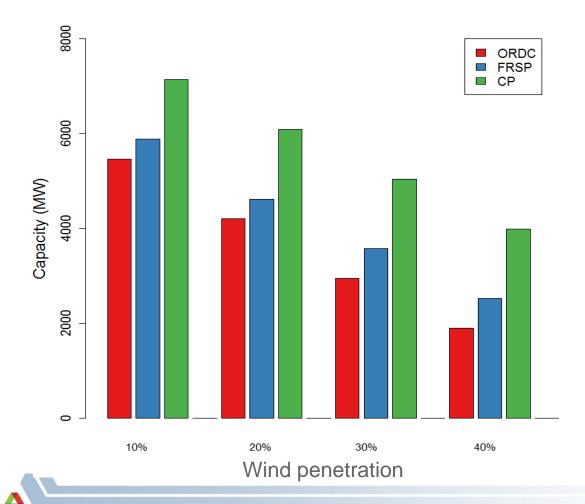
Load Balance				Unit Reserves
$\sum_{i \in I} g_{i,t} + wg_t + es_t = D_t$ Shado				
	w Price	$rs_{i,t} \leq$	$z_i \cdot \overline{P_i} \cdot SPR_i$	$\forall i \in I, s \in S, t \in T$
		$rns_{i,t} \leq (u_i \cdot$	$(-z_i) \cdot \overline{P_i} \cdot NSR_i$	$\forall i \in I, s \in S, t \in T$
Thermal Output				Wind Balance
$g_{i,t} + rs_{i,t} \le z_{i,t} \cdot \overline{O_i}  \forall i \in I, t \in I$	$\equiv T$		$wg_t + wr_t +$	$wc_t = W_t  \forall t \in T$
$g_{i,t} \geq z_{i,t} \cdot \underline{O_i}  \forall i \in I, t \in T$				
Ramping				Unit Commitment
$g_{i,t} \le g_{i,t-1} + z_{i,t} \cdot RU_i \qquad \forall i \in \mathcal{S}_{i,t-1}$	$\in I, t \in T \neq 1$	$Z_{i,t} =$	$= z_{i,t-1} + y_{i,t} - $	$x_{i,t}  \forall \ i \in I, t \in T \neq 1$
$g_{i,t} \geq g_{i,t-1} - z_{i,t-1} \cdot RD_i \qquad \forall i$	$t \in I, t \in T \neq 1$		$Z_i$	$u_i \leq u_i  \forall \ i \in I, t \in T$
			$x_{i,t}$ , $y_{i,t}$ ,	$z_{i,t} \geq 0 \ \forall \ i \in I, t \in T$

- Integer variables for expansion and commitment
- Significant reduction in computation time (up to 5000x\*)
- Enables solving for full year of operations (8760 hourly periods)

\* B. Palmintier and M. Webster, "Impact of unit commitment constraints on generation expansion planning with renewables," in 2011 IEEE Power and Energy Society General Meeting, 2011, pp. 1–7.

## **Results: Capacity Expansion**

- Only new NGCT capacity is developed
  - CP results in most new capacity
  - -ORDC and FRSP are comparable



ORDC – operating reserve demand curve FRSP – fixed reserve scarcity pricing CP – capacity payment

## **Results: Average Prices**

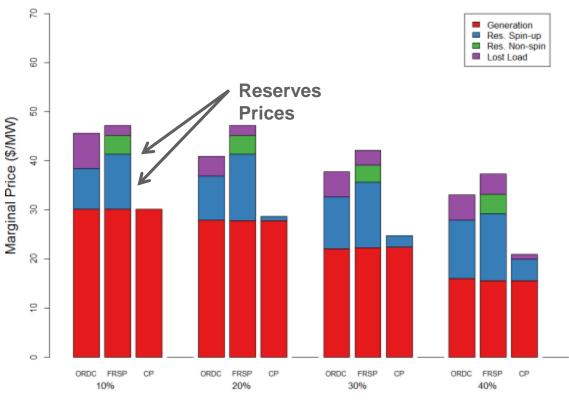
 Prices drop with increasing wind

### ORDC > CP

- CP has no reserves pricing mechanism
  - Lower prices
- More capacity under CP
  - Essentially no lost load

### FRSP > ORDC

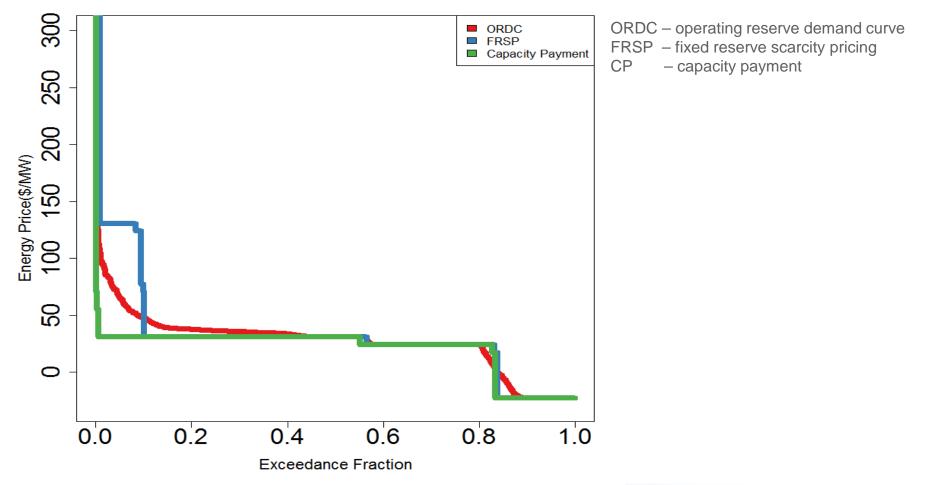
- Higher reserve prices
  - Scarcity price spikes
  - Mostly non-spin
- Less frequent lost load
  - Few hours, large price impact



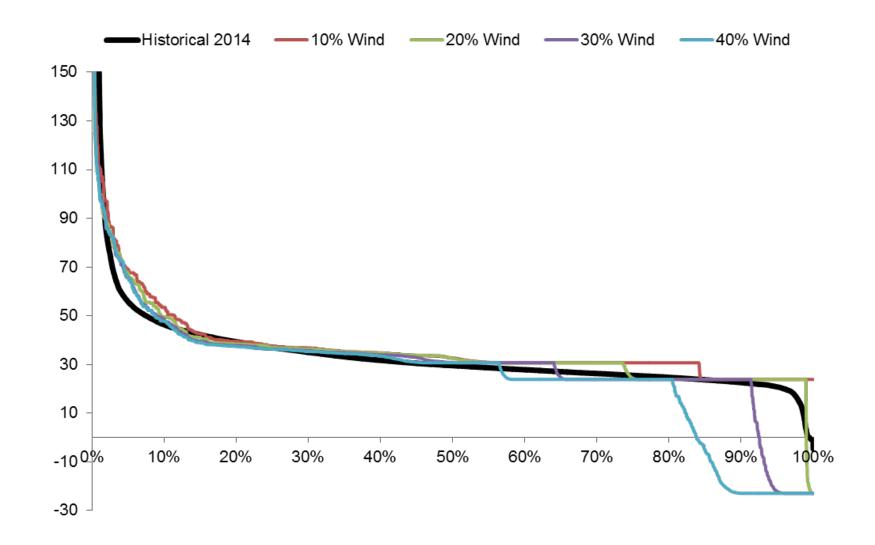
Wind penetration

## Results: Annual Price Exceedance Curve (40% wind)

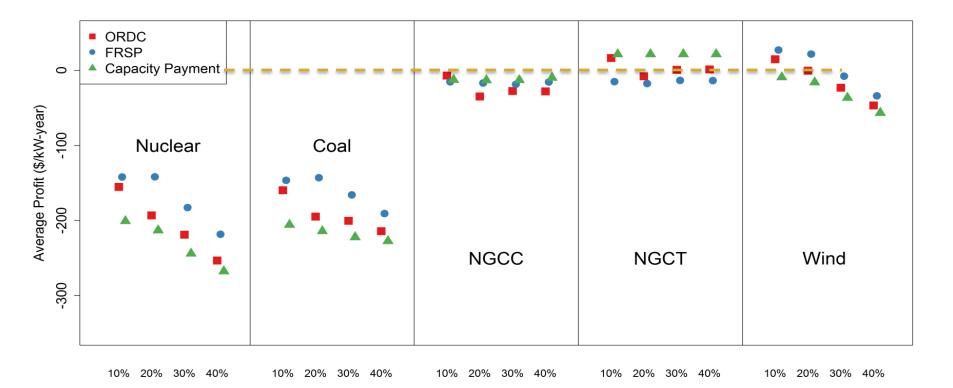
ORDC -> More continuous price spectrum



## **ORDC Historical Comparison**

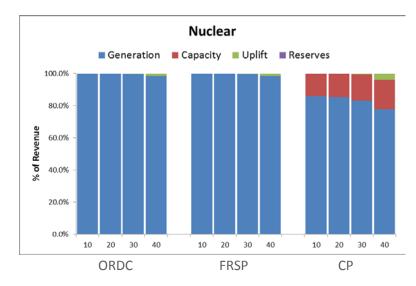


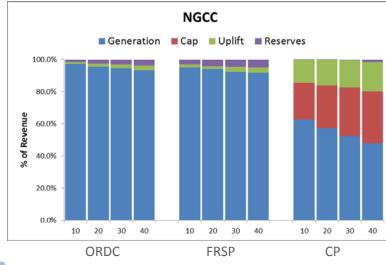
### **Results: Generator Profits**

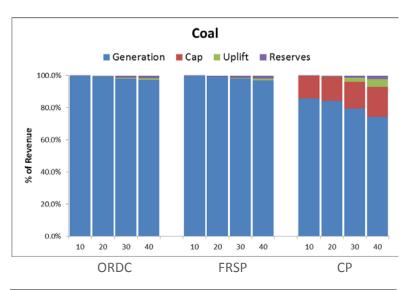


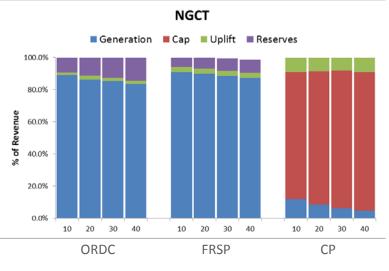
- Nuclear, Coal and Wind profits decrease with increasing wind
  - More exposed to lower off-peak prices
- Gas units (NGCT, NGCC) receive additional revenues from providing reserves
- The optimal investment choice (NGCT) breaks even under all mechanisms for all wind penetration levels

### **Revenue by Source for each Technology**









## **Towards Improved Electricity Markets**

 Several measures could improved the functioning of electricity markets (supporting energy only markets)

### **United States**

- Gradual removal of RES subsidies/ tax credits and introduction of CO2 pricing at national level
- Liquid markets for long-term contracts: hedging for both generators (low/negative prices) & customers (high/price spikes)
- Implementation of Intraday-market for balancing
- Higher time resolution of real-time market settlements (5 min)
- Revision of ancillary services markets (product definitions and quantities)
- Better coordination between ISOs

### Europe

- Gradual removal of RES subsidies and correct CO2 price signals
- Flow-based cross-border transmission capacity allocation and auctioning (improved market coupling of the different market zones)
- Price zones that better reflect congestion patterns
- "Imbalance netting" to avoid counteracting activations of control zones in frequency regulation
- Shortening time frames in the Intraday-market
- Dispatchable renewables
- Co-optimization of energy and reserves
- others

• others

## Electricity Market Design with Renewable Energy

- Review of current and proposed market designs
  - How to achieve capacity adequacy and revenue sufficiency in the long-run?
  - How to ensure and incentivize flexibility in short-run operations?



Technical Report NREL/TP-5D00-61765, Sept. 2014.

### Evolution of Wholesale Electricity Market Design with Increasing Levels of Renewable Generation

E. Ela,<sup>1</sup> M. Milligan,<sup>1</sup> A. Bloom,<sup>1</sup> A. Botterud,<sup>2</sup> A. Townsend,<sup>1</sup> and T. Levin<sup>2</sup>

<sup>1</sup> National Renewable Energy Laboratory <sup>2</sup> Argonne National Laboratory

## **Outline**

- Background
  - Renewable energy penetration levels
  - Electricity market operations
  - Core research questions
- Addressing Renewables in Short-term Operations
  - Operating reserve demand curves
- Addressing Renewables in Long-term Planning
   Generation Expansion and Revenue Sufficiency
- Concluding Remarks







## **Concluding Remarks**

- Electricity markets and renewable energy
  - Fundamental challenges the same in Europe and United States: uncertainty and variability
  - Implications for operations, planning, and markets
  - Flexibility is key, but solutions differ
  - More advanced electricity markets in the US, more support for renewables in Europe
  - Physical complexity vs. economic transparency in market design
- Many solutions to variable renewable energy integration challenges
  - Supply flexibility, demand response, energy storage
  - Forecasting, operational practices, market design
  - No silver bullet: Ideally, the most cost effective solutions should prevail
  - Lessons can be learned from both Europe and United States
    - Intraday markets, long-term markets (Europe -> United States)
    - Co-optimization energy/reserves, locational pricing, dispatchable renewables (United States -> Europe)



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## **Acknowledgements and References**

### Collaborators, including

– Zhi Zhou	Argonne National Laboratory
<ul> <li>Todd Levin</li> </ul>	Argonne National Laboratory
– Canan Uckun	Argonne National Laboratory
— Jianhui Wang	Argonne National Laboratory
– Hans Auer	TU Wien, Austria
<ul> <li>Andreas Fleischhacker</li> </ul>	TU Wien, Austria
– John R. Birge	University of Chicago
<ul> <li>Ricardo Bessa</li> </ul>	INESC TEC, Portugal
<ul> <li>Vladimiro Miranda</li> </ul>	INESC TEC, Portugal
<ul> <li>Runze Chen</li> </ul>	Tsinghua University, China

#### Main Sponsors

-U.S. DOE Energy Efficiency and Renewable Energy (EERE)

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### Additional Publications and Information

Wind power forecasting and electricity markets project: <u>http://ceeesa.es.anl.gov/projects/windpowerforecasting.html</u> Personal website: <u>http://botterud.mit.edu/</u>



# **Electricity Markets and Renewable Energy: United States vs. Europe**

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#### ETH, Zurich, January 27, 2017

