

# Operating Reserve Demand Curve, Scarcity Pricing and Intermittent Generation

## Lessons from the Texas ERCOT Experience

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# Introduction

# Background

- 80's & 90's: privatization/liberalization of electricity sector
- But after the 2000s many markets faced reliability issues

Resource adequacy problem or “missing-money” problem

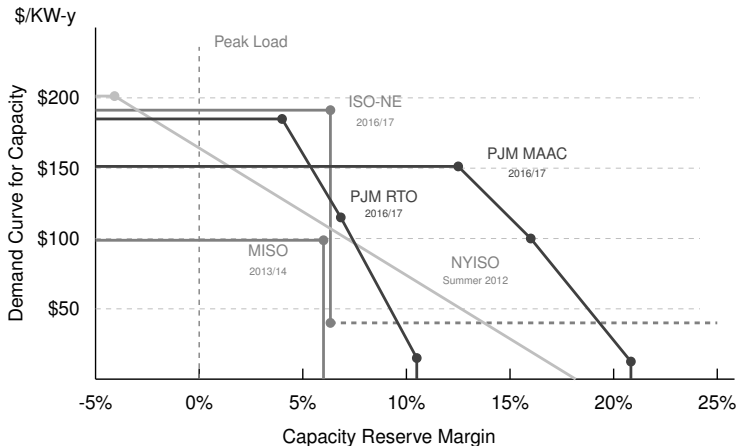
- Competitive electricity markets weren't providing right incentives to build adequate capacity to guarantee reliability
  - Price Caps
  - Negative prices (mostly due to wind subsidies)
  - Regulatory uncertainty

# Background

- To solve the resource adequacy problem, many countries and regions implemented **Capacity Mechanisms**
  - Capacity Adequacy Requirements
  - Centralized Capacity Markets
- Where? ISO-NE, PJM, NYISO, etc.
- **Goal:** provide sufficient and clearly perceivable long-term price signals (lower investment risk)  
(Hoschle et al 2017, *EE*)

# Background

Some examples of demand curves for capacity in the U.S.  
(Spees et al. 2013, *EEEP*)



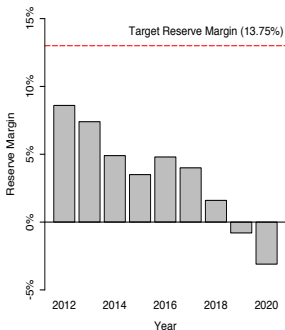
# Background

- Another potential solution to the resource adequacy problem is to introduce a **futures market for wholesale electricity**
- Where? Singapore (since April 2015)
- **Goal:** lower the risk of investment in capacity (stable long-term price signals)  
(see Wolak 2017, *Working Paper*)

# Meanwhile, in Texas...

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- The Texas ERCOT market was facing the Resource Adequacy Problem



- However, no capacity mechanisms were implemented, nor a futures market for wholesale electricity



# Background

- Alternatively, the regulator implemented in June 2014...

## Operating Reserve Demand Curve (ORDC) system

- Mechanism to provide adequate investment incentives through scarcity pricing system

# Background

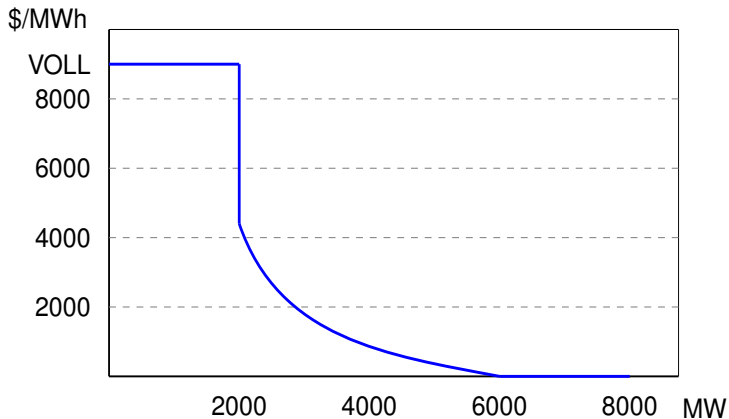
- Alternatively, the regulator implemented in June 2014...

## Operating Reserve Demand Curve (ORDC) system

- Mechanism to provide adequate investment incentives through scarcity pricing system
- The basic idea
  - Generators holding additional reserves are compensated if reserves are low
  - The **compensation** is **added** to the **real-time** price

# Background: Capacity Markets

The Operating Reserves Demand Curve (Hogan 2005)



# Our research questions

- Does the ORDC provide clear long-term price signals (lower investment risk)?
  - Linked to the Real-Time market
  - Growing penetration of intermittent renewable generation
- Evaluate these issues based on the experience of the Texas ERCOT market over the past two years
- Why is this important?
  - Recently implemented in the Mexican reform
  - Some European markets are considering ORDC

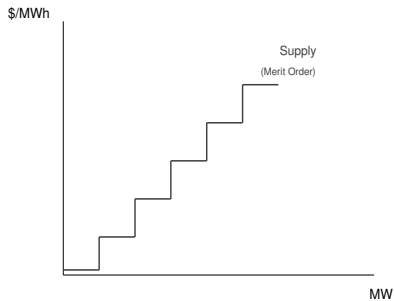
# Stylized Model

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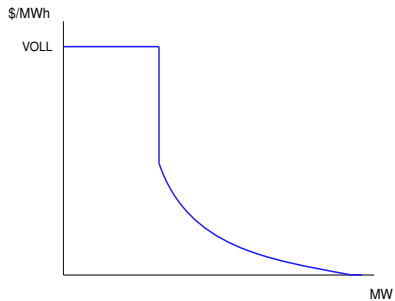
- In the presence of ORDC, generators have two options
  - Wholesale electricity market (RTM), or
  - ORDC

Data

# Stylized model

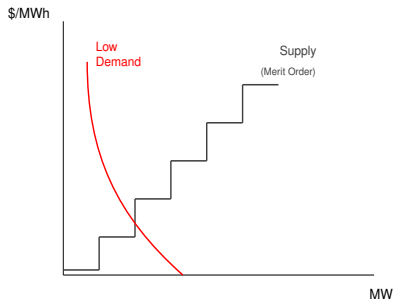


Wholesale market

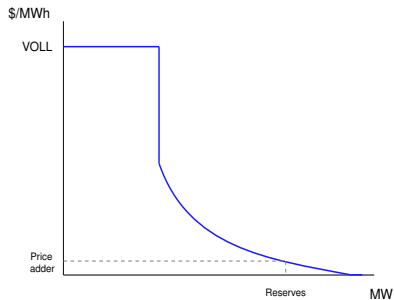


ORDC

# Stylized model



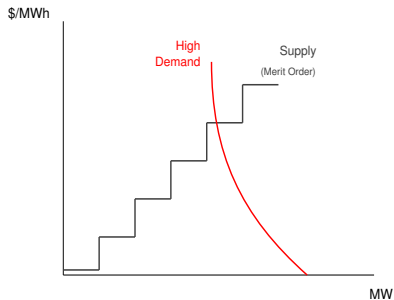
Wholesale market



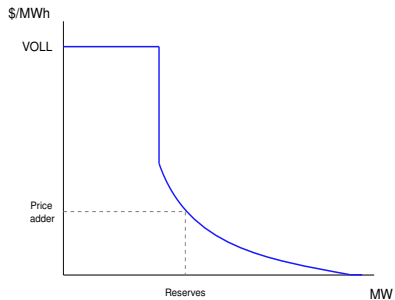
ORDC



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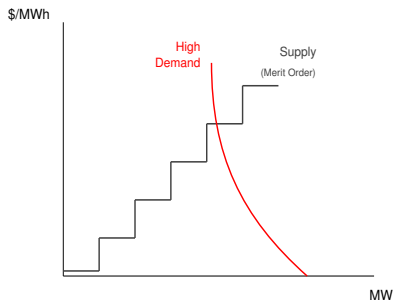


Wholesale market

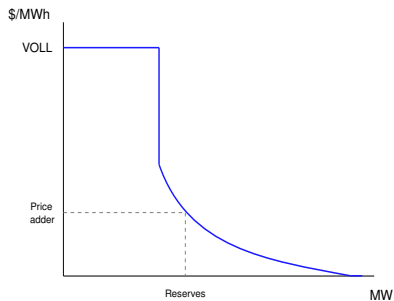


ORDC

# Stylized model



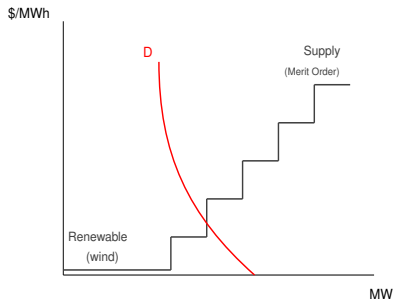
Wholesale market



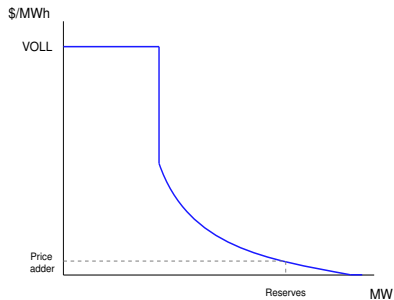
ORDC

Hypothesis #1: ORDC price increases as electricity demand increases (*ceteris paribus*)

# Stylized model with renewable generation (wind)

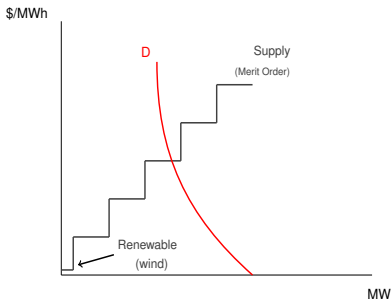


Wholesale market

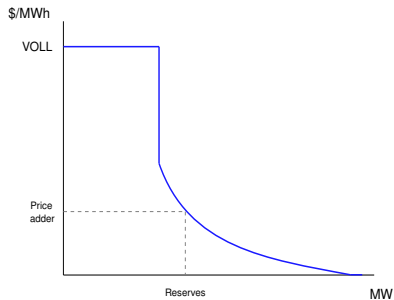


ORDC

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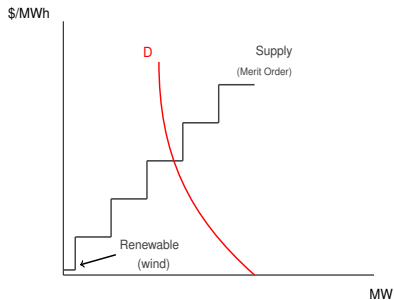


Wholesale market

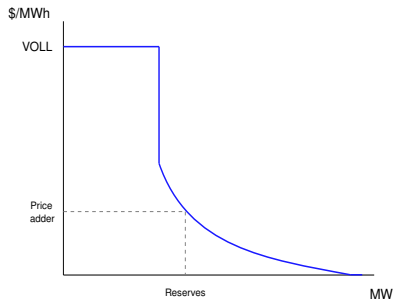


ORDC

# Stylized model with renewable generation (wind)



Wholesale market



ORDC

Hypothesis #2: ORDC price decreases as wind generation increases (*ceteris paribus*)

# Empirical model and results

# Baseline model

- Unit of observation: an hour ( $t$ )

$$\log(RTORPA)_t = \alpha_1 \log(Load)_t + \alpha_2 \log(Wind)_t + \lambda_t + \varepsilon_t$$

- Dependent variable
  - $RTORPA_t$ : ORDC real time price adder at  $t$  (\$/MWh)
- Explanatory variables
  - $Load_t$ : ERCOT's load at  $t$  (MW)
  - $Wind_t$ : ERCOT's wind generation at  $t$  (MW)
  - $\lambda_t$ : seasonal dummies

# Data and summary statistics (highlights)

- Unit of observation: an hour ( $t$ )
  - Time: January 2015 to December 2016
    - Data is missing for the last week of 2016
  - Source: Texas ERCOT data
- The variables are all stationary Tests
  - ADF test performed with and without trend and drift



# The determinants of ORDC prices

	(1)	(2)	(3)	(4)	(5)
	OLS	OLS	OLS	OLS	OLS
log Load	0.283*** (0.0202)	0.274*** (0.0232)	0.279*** (0.0234)	0.279*** (0.0232)	0.564*** (0.0348)
log Wind	-0.0859*** (0.00494)	-0.0812*** (0.00492)	-0.0814*** (0.00493)	-0.0805*** (0.00473)	-0.0829*** (0.00472)
Hour dummy		✓	✓	✓	✓
Day of Week dummy			✓	✓	✓
Day of month dummy				✓	✓
Month dummy					✓
r2	0.0731	0.105	0.107	0.111	0.134
N	17373	17373	17373	17373	17373
vce	robust	robust	robust	robust	robust

Standard errors in parentheses

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

# Robustness Checks (I)

- Potential endogeneity between prices and demand
  - Solution: IV regression
- We use weather as an instrument for ERCOT load
  - Daily data on the (squared) deviation from average temperatures
  - Data from the three major metropolitan areas in TX (Houston, Dallas and SA)
  - Extreme weather (hot and cold) highly correlated with load

# The determinants of ORDC prices (IV)

	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	IV	OLS	IV	OLS	IV
log Load	0.568*** (0.139)	0.544*** (0.127)	0.920*** (0.211)	0.830*** (0.217)	0.989*** (0.250)	0.856*** (0.234)
log Wind	-0.149*** (0.0335)	-0.149*** (0.0335)	-0.153*** (0.0345)	-0.153*** (0.0346)	-0.143*** (0.0334)	-0.143*** (0.0336)
Day of Week dummy					✓	✓
Day of month dummy					✓	✓
Month dummy			✓	✓	✓	✓
r2	0.0840	0.0839	0.139	0.139	0.184	0.184
N	724	724	724	724	724	724
vce	jackknife	jackknife	jackknife	jackknife	jackknife	jackknife

Standard errors in parentheses

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# Robustness Checks (II)

- Persistence of the effect of the some variables on ORDC prices
  - Solution: autoregressive distributed lag (ARDL) model
- Alternative model

$$\log(RTORPA)_t = \sum_{i=1}^l \alpha_i \log(RTORPA)_{t-i} + \sum_{j=0}^n \beta_j \log(Load)_t + \sum_{j=0}^n \gamma_j \log(Wind)_t + \lambda_t + \varepsilon_t$$

# The determinants of ORDC prices (ARDL)

	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	OLS	OLS	OLS	OLS	OLS
log Load	0.390*** (0.0292)	0.848*** (0.0587)	0.249*** (0.0518)	0.499*** (0.0697)	0.252*** (0.0521)	0.505*** (0.0706)
log Wind	-0.0662*** (0.0129)	-0.0764*** (0.0140)	-0.0489*** (0.0166)	-0.0549*** (0.0170)	-0.0489*** (0.0166)	-0.0550*** (0.0170)
(T-1) log RTORPA	0.756*** (0.0264)	0.750*** (0.0258)	0.990*** (0.0336)	0.970*** (0.0331)	0.986*** (0.0356)	0.965*** (0.0351)
(T-1) log Load	-0.332*** (0.0253)	-0.720*** (0.0551)	-0.127 (0.0930)	-0.253** (0.126)	-0.136 (0.0933)	-0.275** (0.128)
(T-1) log Wind	0.0440*** (0.0132)	0.0547*** (0.0143)	0.0458 (0.0300)	0.0450 (0.0300)	0.0469 (0.0300)	0.0465 (0.0300)
(T-2) log RTORPA			-0.310*** (0.0292)	-0.295*** (0.0287)	-0.297*** (0.0418)	-0.279*** (0.0411)
(T-2) log Load			-0.0474 (0.0480)	-0.0918 (0.0734)	-0.0407 (0.0480)	-0.0735 (0.0749)
(T-2) log Wind			-0.0246 (0.0167)	-0.0170 (0.0167)	-0.0259 (0.0167)	-0.0188 (0.0167)
(T-3) log RTORPA					-0.0131 (0.0245)	-0.0163 (0.0242)
Full set of dummies		✓		✓		✓
r2	0.604	0.621	0.642	0.654	0.642	0.654
N	17372	17372	17371	17371	17370	17370
vce	robust	robust	robust	robust	robust	robust

Standard errors in parentheses

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

## The determinants of ORDC prices (ARDL) (IV)

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	OLS	OLS	OLS	OLS	OLS	OLS	est7	est8
log Load	0.706*** (0.226)	1.022*** (0.367)	1.028*** (0.271)	1.221*** (0.397)	0.734*** (0.240)	1.000*** (0.363)	1.170*** (0.286)	1.322*** (0.415)
log Wind	-0.154*** (0.0345)	-0.158*** (0.0313)	-0.144*** (0.0332)	-0.151*** (0.0311)	-0.155*** (0.0343)	-0.157*** (0.0318)	-0.146*** (0.0335)	-0.149*** (0.0315)
(T-1) log RTORPA	0.413*** (0.104)	0.418*** (0.0342)	0.398*** (0.102)	0.408*** (0.0346)	0.411*** (0.101)	0.417*** (0.0376)	0.404*** (0.0986)	0.410*** (0.0373)
(T-1) log Load	-0.388* (0.216)	-0.754** (0.376)	-0.450* (0.270)	-0.908** (0.418)	-0.523 (0.368)	-1.243** (0.596)	-0.929** (0.425)	-1.509** (0.640)
(T-1) log Wind	0.0690** (0.0303)	0.0690** (0.0311)	0.0534* (0.0296)	0.0567* (0.0311)	0.0475 (0.0343)	0.0519 (0.0377)	0.0432 (0.0339)	0.0475 (0.0369)
(T-2) log RTORPA					0.000387 (0.0606)	-0.00186 (0.0377)	-0.0150 (0.0628)	-0.0145 (0.0377)
(T-2) log Load					0.117 (0.245)	0.585 (0.397)	0.450 (0.285)	0.802* (0.456)
(T-2) log Wind					0.0324 (0.0279)	0.0336 (0.0315)	0.0216 (0.0277)	0.0239 (0.0314)
Full set of dummies			✓	✓			✓	✓
r2	0.242	0.239	0.314	0.311	0.244	0.240	0.317	0.315
N	723	723	723	723	722	722	722	722
vce	robust	unadjusted	robust	unadjusted	robust	unadjusted	robust	unadjusted

Standard errors in parentheses

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# Conclusions

# Conclusions & policy implications

- Many regions solved the “resource adequacy” problem using Capacity Mechanisms/Future Markets
  - Solve the “missing money” problem
  - Provide a stable framework for investors
  - Guarantee reliability thru long-term capacity commitments
- However, in the Texas ERCOT market, implemented an alternative mechanism to solve the “resource adequacy” problem



# Conclusions & policy implications

- ORDC: alternative mechanism to provide investment incentives thru scarcity pricing
  - Implemented in ERCOT in June 2014
- We show that ORDC prices mostly depend on:
  - Load shocks/weather shocks
  - Wind speed

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*Does the ORDC provide sufficient and clearly perceivable long-term price signals (lower investment risk)?*

# Thanks! Questions?

Your feedback is much appreciated: [rbajo@unav.es](mailto:rbajo@unav.es)

Extra slides

# Extra Slides

# Stationarity tests

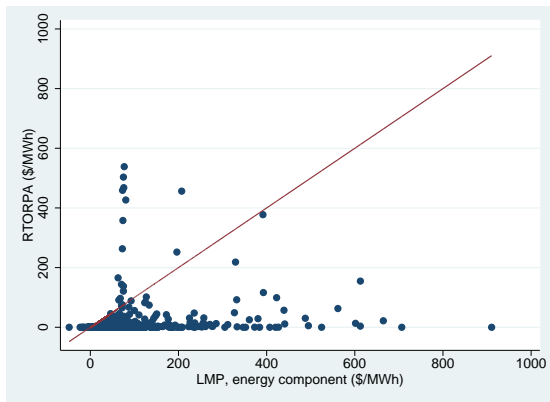
Variable	D-F test		D-F test (trend)		D-F test (drift)	
	t-statistic	p-value	t-statistic	p-value	t-statistic	p-value
log RTORPA	-15.114	0.000	-16.725	0.000	-16.593	0.000
log Wind output	-13.979	0.000	-14.522	0.000	-13.979	0.000
log ERCOT load	-5.007	0.000	-5.025	0.000	-5.007	0.000

**Note:** Critical values for rejection of hypothesis of unit root at the 1% confidence level are -3.430 for the “no drift/no trend” case, -3.960 for the trend case, and -2.332 for the drift case.

Return

# Electricity market prices versus ORDC prices

Figure: LMP versus Price adder (\$/MWh)



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