



Wind variability and impact on markets

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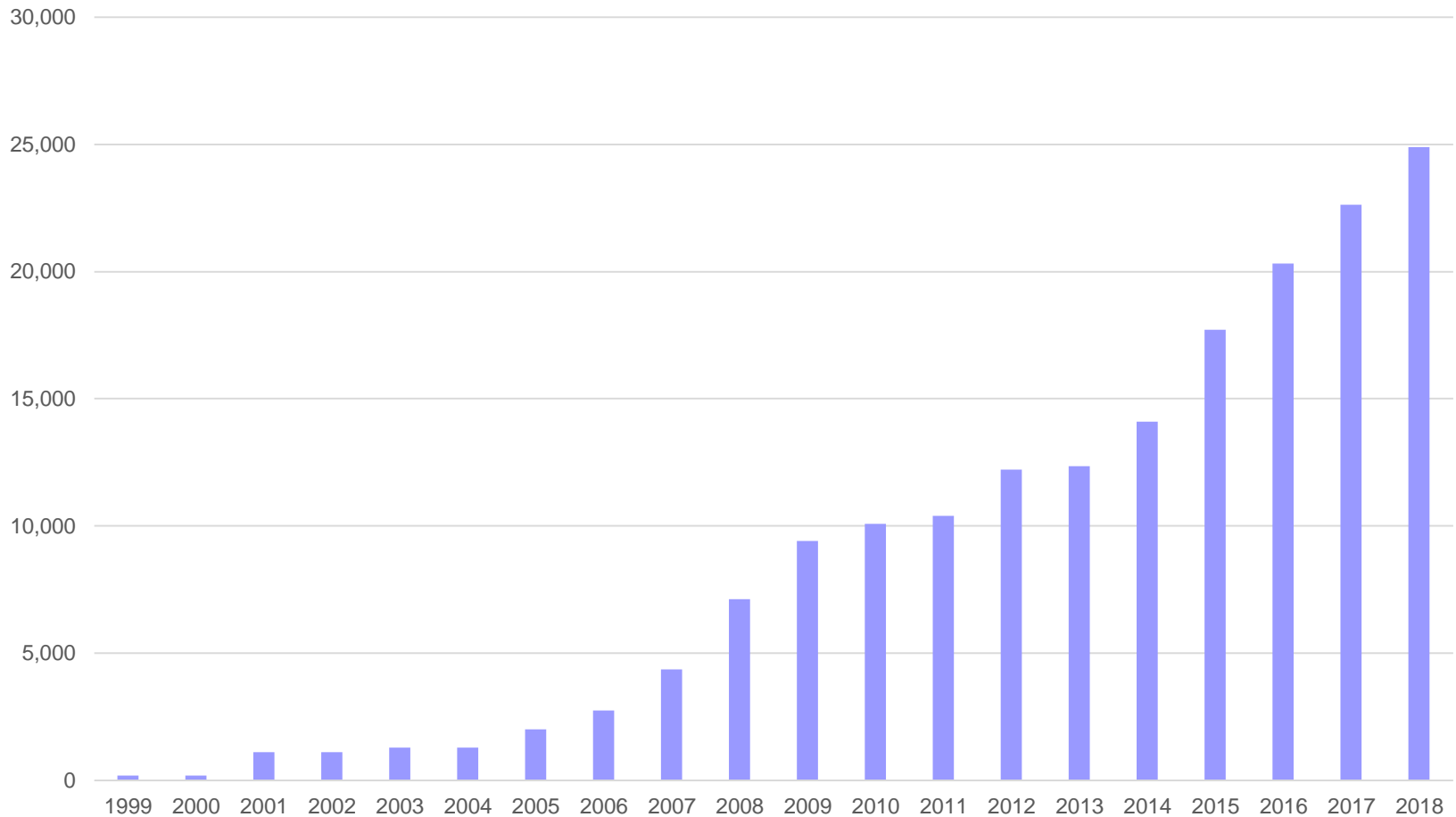
Outline

- Growth of wind in Texas,
- Challenges under high levels of wind,
- Comparison of Texas wind penetration to rest of US,
- Comparison of: West Texas/ERCOT; Denmark/EU; and, South Australia/Australia,
- Texas as microcosm of high wind challenges,
- Statistical modeling to understand challenges under high penetration,
- Generalized dynamic factor model and Kolmogorov spectrum,
- Scaling of wind power and wind power variability,
- Implications for electricity systems and organized wholesale markets,
- Conclusion.



Texas has experienced remarkable wind growth.

Wind generation capacity in Texas (MW, end of year)



Source: USDOE 2019.



Challenges under high levels of wind integration.

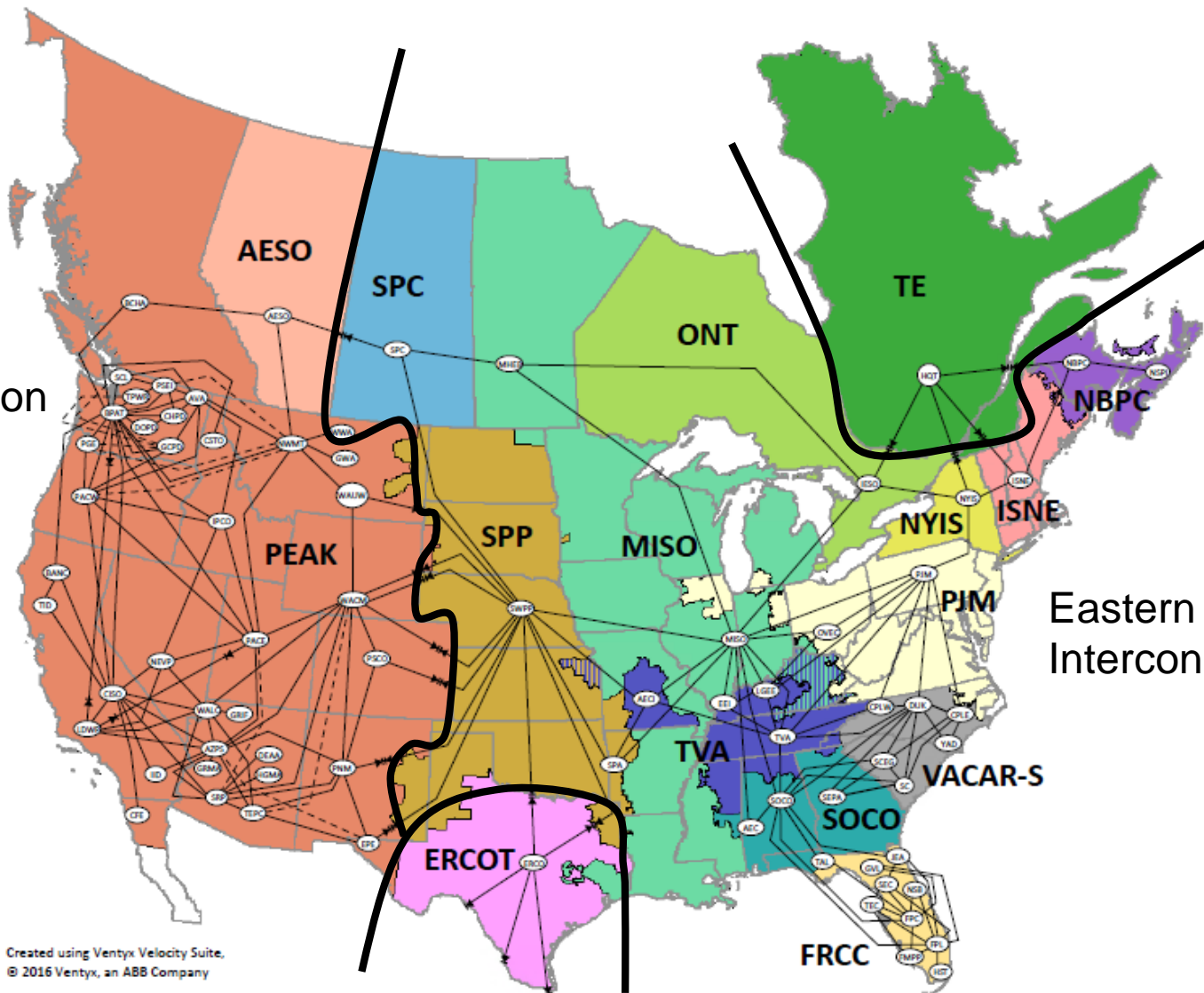
- Typical time of daily peak of US *inland* wind production coincides with daily minimum of electrical load and *vice versa*:
 - Difference between load and wind (“net load”) must be supplied by other resources.
- Variability of wind production:
 - Changes in supply-demand balance must be compensated by other resources.
- With higher wind penetrations, timing and variability become more critical.



Measurement of wind penetration.

- Important metrics of penetration are wind as a fraction of load energy or power in “balancing area” or in interconnection.
- Contiguous US has tens of balancing areas and three interconnections:
 - Western,
 - Eastern,
 - Electric Reliability Council of Texas (ERCOT), most of Texas, smallest of US interconnections, peak load around 74.5 GW. ■5

Balancing Areas and Interconnections.



Western Interconnection

Eastern Interconnection

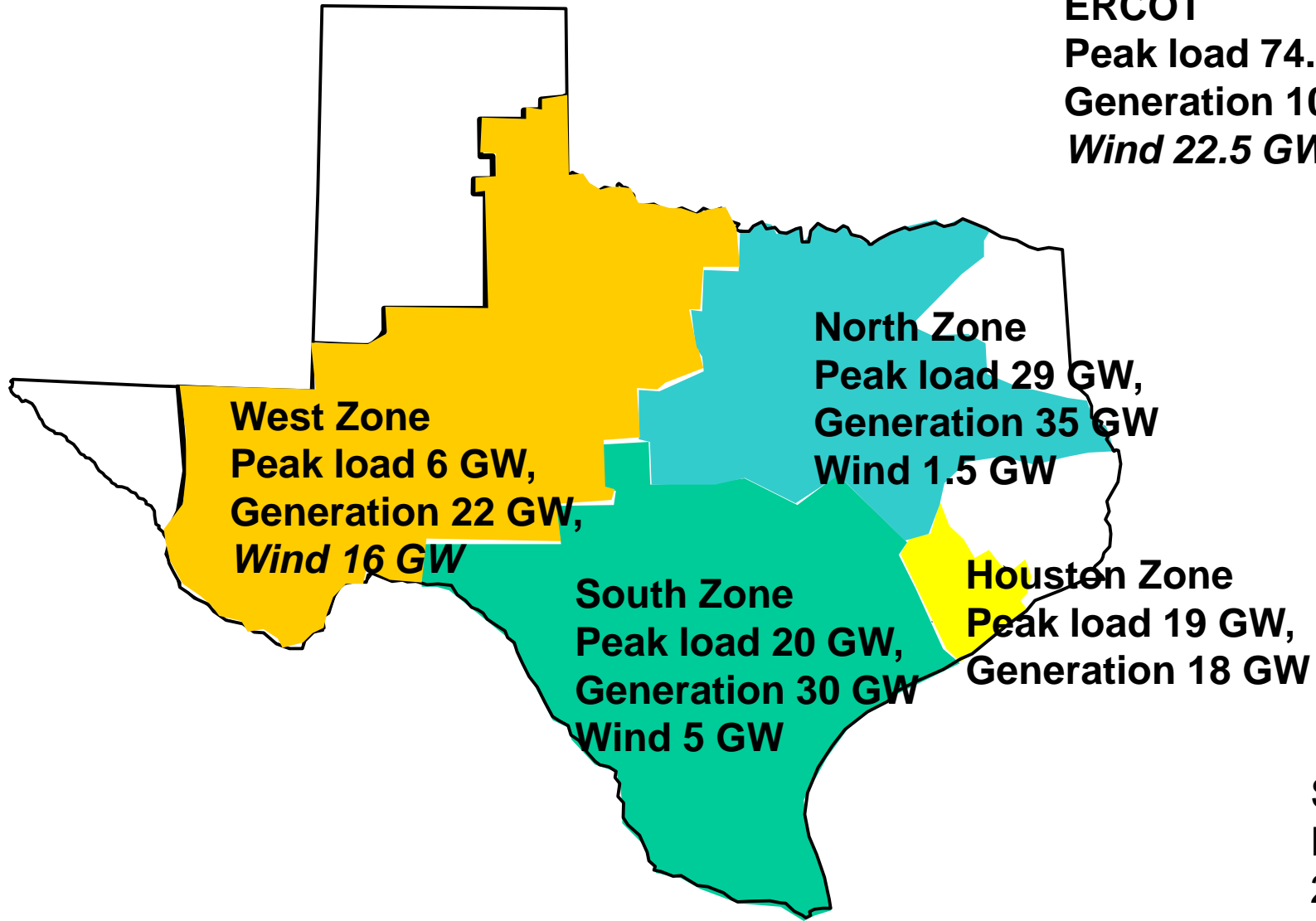


Comparison of Texas and ERCOT to rest of US.

- Wind provided **6.5%** of electricity by energy in 2018 in **US** (AWEA, 2019).
- Wind provided **16%** of electricity by energy in 2018 in **Texas** (AWEA, 2019).
- Wind provided **19%** of electricity by energy in 2018 in **ERCOT** (Potomac, 2019).
- ERCOT has, *by far*, the greatest wind penetration of the three US interconnections, and largest of any *large* balancing area.



Most ERCOT wind is in West Texas zone.





Comparison of West Texas/ERCOT to Denmark/EU, and South Australia/Australia

- The International Energy Agency highlights that Denmark and South Australia are in “Phase 4” of renewable integration, or even more advanced, (IEA, 2018), requiring “advanced technologies to ensure reliability.”
- Australia and US are in “Phase 2,”
- European Union is in “Phase 3.”



Comparison to Denmark.

- Denmark noted for high wind penetration.
- Denmark has two AC electrical networks:
 - Eastern Danish power system is part of the Nordic interconnection (peak load around 63 GW in 2015, NordREG, 2016),
 - Western Danish power system is part of the Continental Western European system (peak load around 530 GW, ENTSO-E, 2016).
 - 600 MW HVDC link between them.
- Even the Eastern Danish system *alone* is integrated into a system with peak load nearly as high as ERCOT.



West Texas zone vs Denmark: Area and generation capacity.

- West Texas zone area about 2.5 times Denmark area.
- Total installed power generation capacity in **West Texas** zone around **22 GW**, (compares to around **16 GW** in **Denmark**, (ENTSO-E, 2019)),
- So *total* Danish generation capacity is somewhat *smaller* than the total West Texas zone generation capacity.



West Texas vs South Australia Area and generation capacity.

- West Texas zone area about one tenth of South Australia area:
 - Stability issues more pressing in SA,
- Total installed power generation capacity in **West Texas** zone around **22 GW**, (compares to around **6 GW** in **South Australia**, (AEMO, 2018)),
- So South Australia generation capacity is significantly *smaller* than the total West Texas zone generation capacity.



West Texas vs Denmark and South Australia: Wind capacity.

- Wind power generation capacity in **West Texas** zone around 16 GW, **72%** of total installed generation capacity, compares to:
 - 9.5 GW of wind, **60%** of installed generation capacity, in **Denmark** (ENTSO-E, 2019), and
 - 1.8 GW of wind, 1 GW of solar, under **50%** of installed generation capacity in **South Australia** (AEMO, 2018),
- Total wind capacity higher in West Texas than Denmark or South Australia, and higher as %.



West Texas vs Denmark and South Australia: Wind energy.

- Annual wind energy production in West Texas zone as a fraction of electric energy consumption in **West Texas** around **100%**, compares to:
 - under **60%** in **Denmark**, (ENTSO-E, 2019), and
 - under **40%** in **South Australia**, (AEMO, 2018).
- But these are all misleading statistics, since West Texas, Denmark, and South Australia are each embedded in much larger interconnections!



ERCOT vs EU and Australia.

- Annual wind energy production in ERCOT as a fraction of electric energy consumption in **ERCOT** around **19%**, compares to:
 - around **11%** in **EU**, (ENTSO-E, 2019), and
 - around **7%** in **Australia**, (CEC, 2019).
- Overall renewable penetration in EU (32%, (ENTSO-E, 2019)) and Australia (21%, (CEC, 2019)) higher than ERCOT:
 - Due to hydro and solar.



ERCOT is microcosm of high wind challenges.

- Large amount of wind capacity:
 - Largest capacity of any US state,
- Small interconnection:
 - Smallest of three US interconnections,
- Significant wind production off-peak:
 - Due to West Texas wind,
 - Coastal wind better correlated with demand.



ERCOT is microcosm of high wind challenges.

- West Texas wind resources far from load centers:
 - Most transmission constraints thermal contingency, but some related to voltage or steady-state or transient stability,
 - Australian system may have more significant stability constraints.
- Little flexible hydroelectric generation:
 - Unlike Eastern and Western US, Europe, and Australia.



Wind power modeling to better understand challenges.

■ Big data flavor:

- Roughly 100 wind farms in ERCOT,
- Relevant issues at timescales from sub-minute to multi-year,
- One year of 1-minute data from 100 farms is around 50 million measurements,
- Understanding inter-year variability requires multi-year data sets.



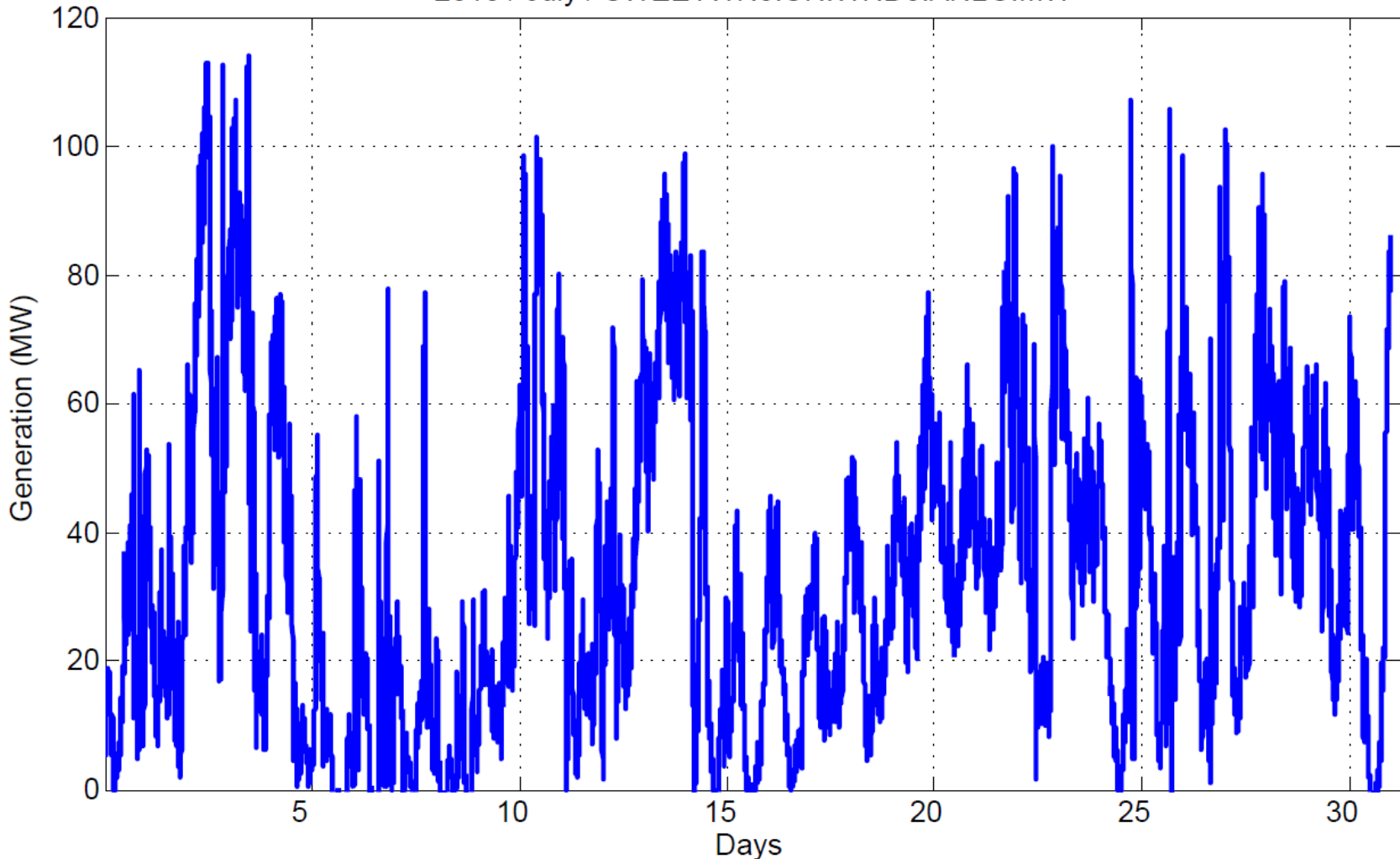
Wind power modeling to better understand challenges.

- Use statistical techniques:
 - relationship between time/season of maximum wind production and time of maximum load,
 - variability of wind and scaling of variability,
 - implications for needed flexibility in “residual” thermal system that provides for net load.
- Modeling issues:
 - Intermittency,
 - Correlation between wind and load.



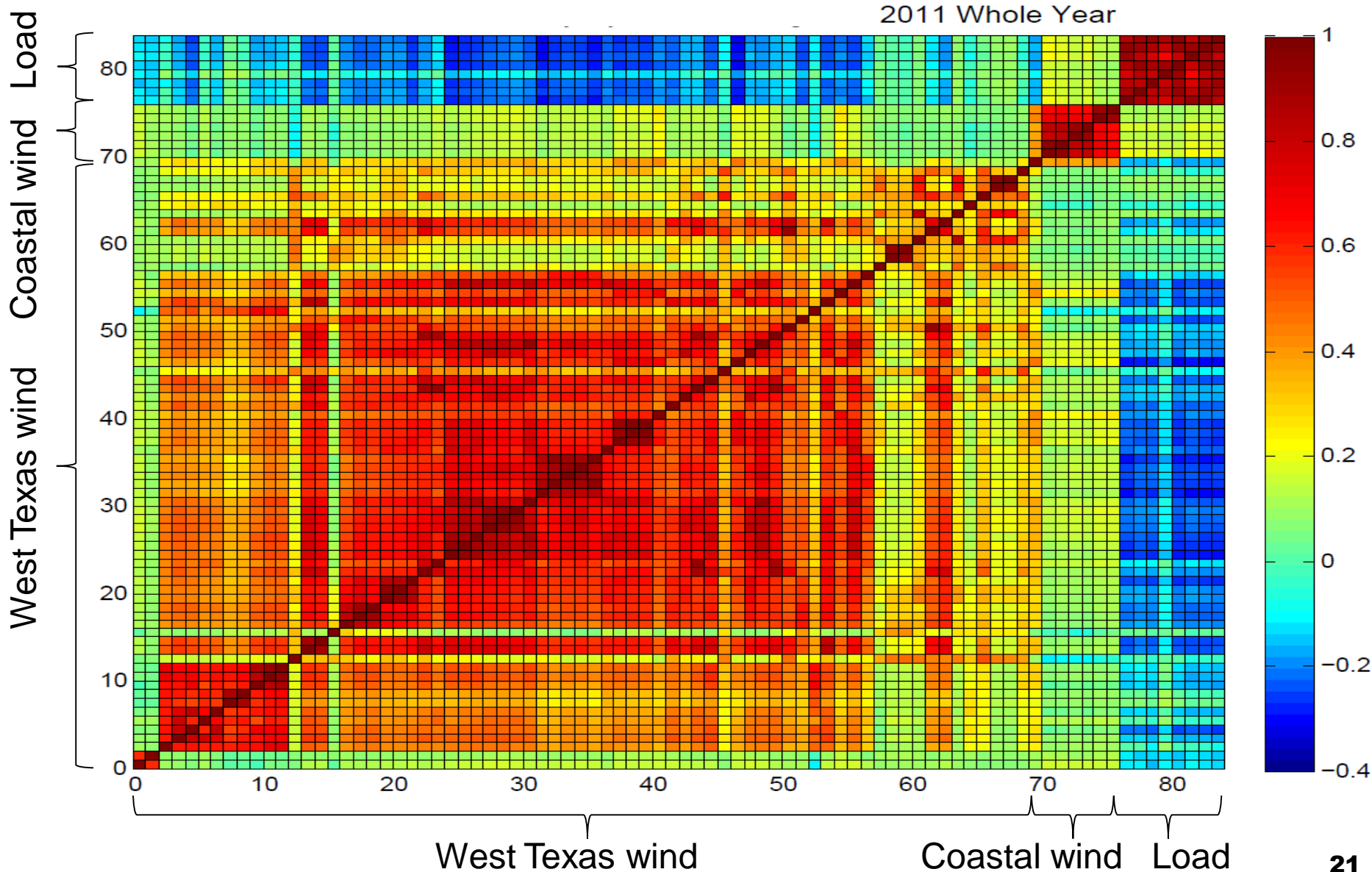
Intermittent wind power production.

2010 / July / SWEETWN3.UN.WND3.ANLG.MW





Correlation of ERCOT wind and load.



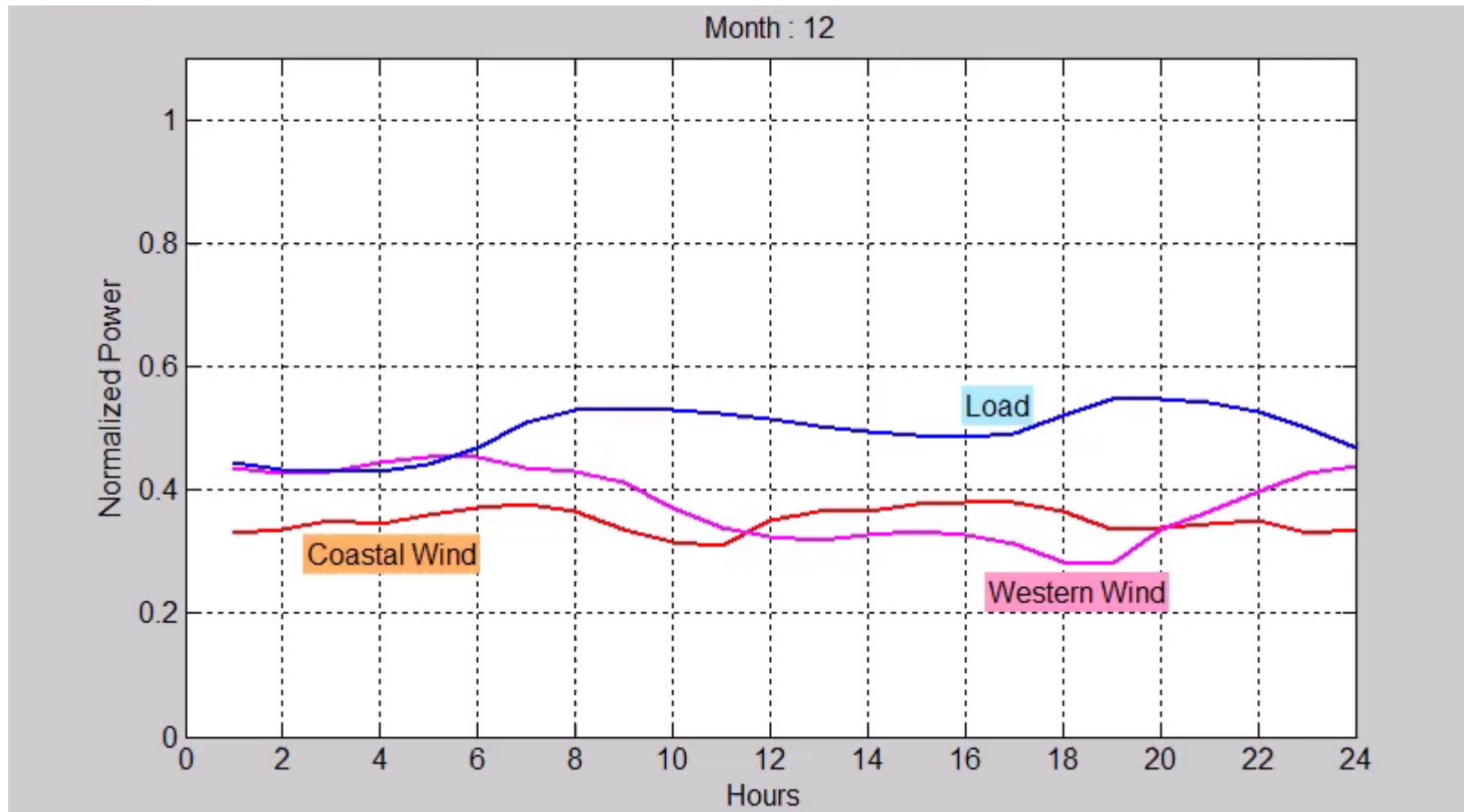


Statistical wind power model.

- Model wind power and load as sum of (slowly varying) diurnal **periodic** component plus **stochastic** component.
- Use “generalized dynamic factor model” (GDFM, Forni et al., 2005) for stochastic:
 - Decompose stochastic into sum of “**common**” component and “**idiosyncratic**” component.
 - Common component for wind and load powers expressed in terms of fewer underlying independent stochastic processes, the “factors,”
 - Idiosyncratic component different for each farm.



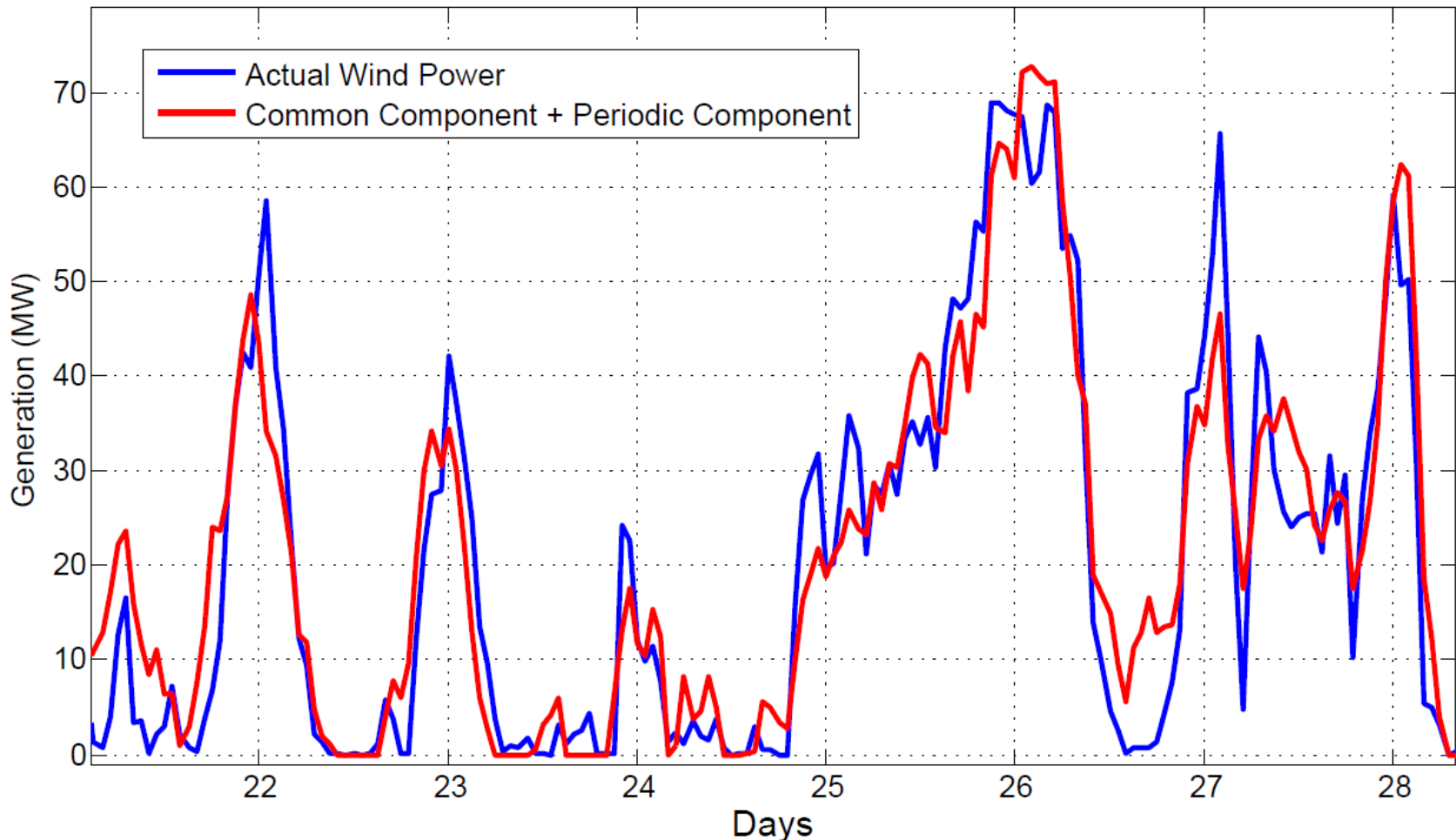
Diurnal periodic component slowly varies over year.





Periodic plus common accounts for most variation.

Jan - March / 2013 / #5 Wind Farm / NDFactor: 20





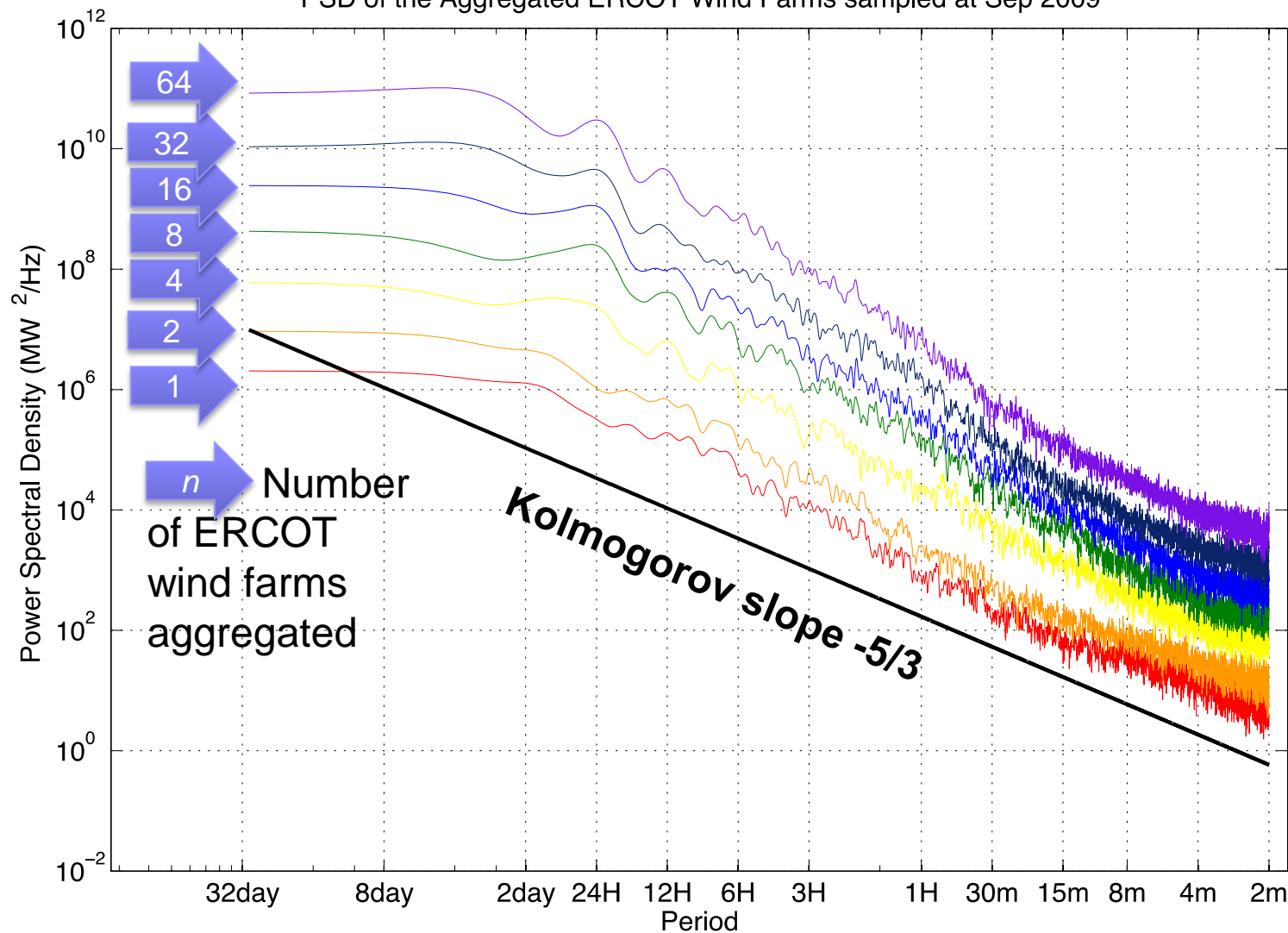
Kolmogorov slope of wind power spectrum

- A. N. Kolmogorov mainly known to electrical engineers through contributions to understanding of stochastic processes.
- Related contributions in turbulent flow crossed over to electrical engineering community through Apt (2007).
- Kolmogorov used dimension analysis to predict that power spectral density of wind power would have characteristic roll-off of slope $-5/3$.
- Verified in Apt (2007).



Wind power spectrum.

PSD of the Aggregated ERCOT Wind Farms sampled at Sep 2009

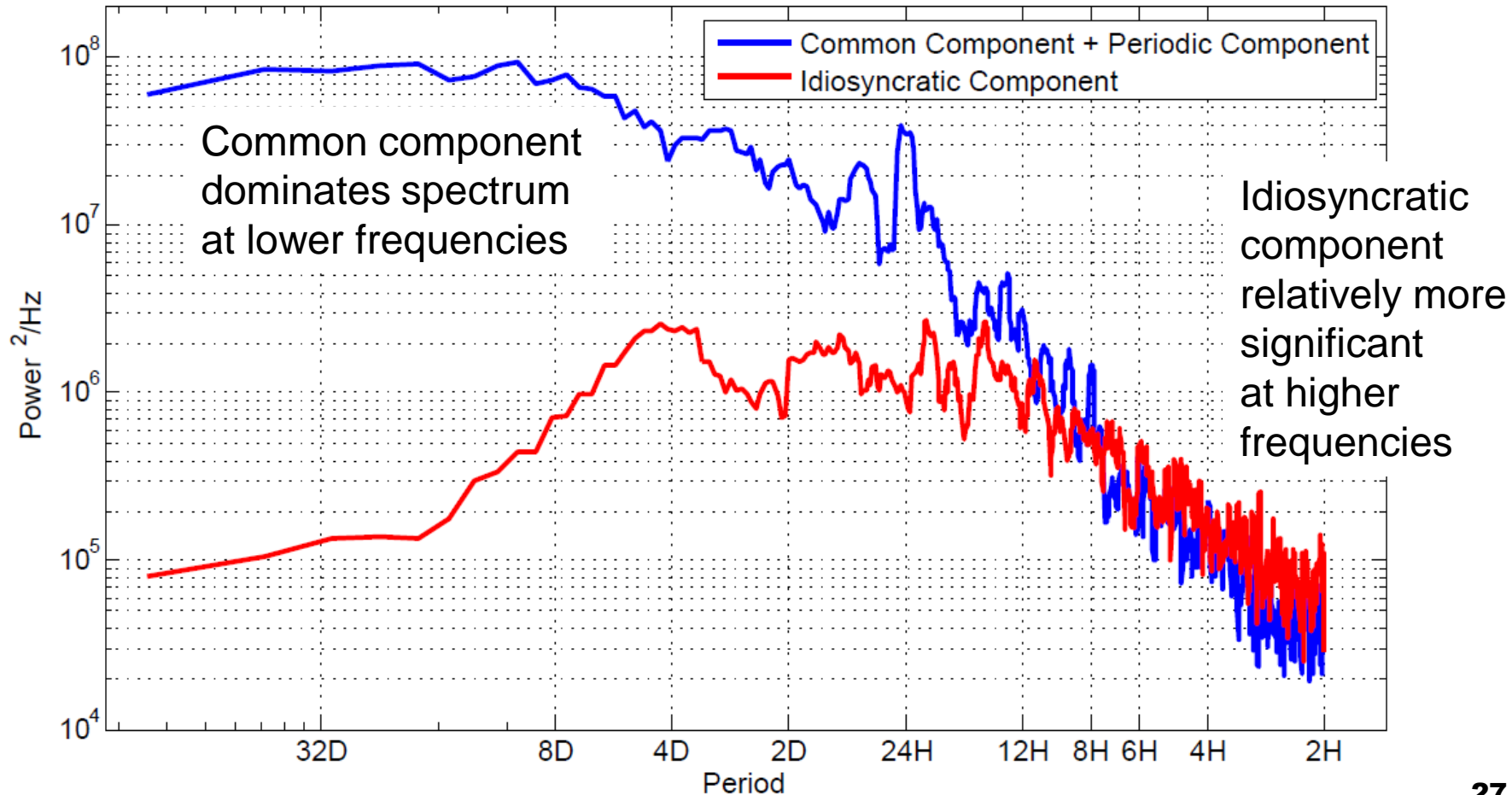


Source:
ERCOT
data,
Analysis
based on
Apt (2007).



Spectrum of common and idiosyncratic components.

Wind Farm #5 / Jan - March / 2013



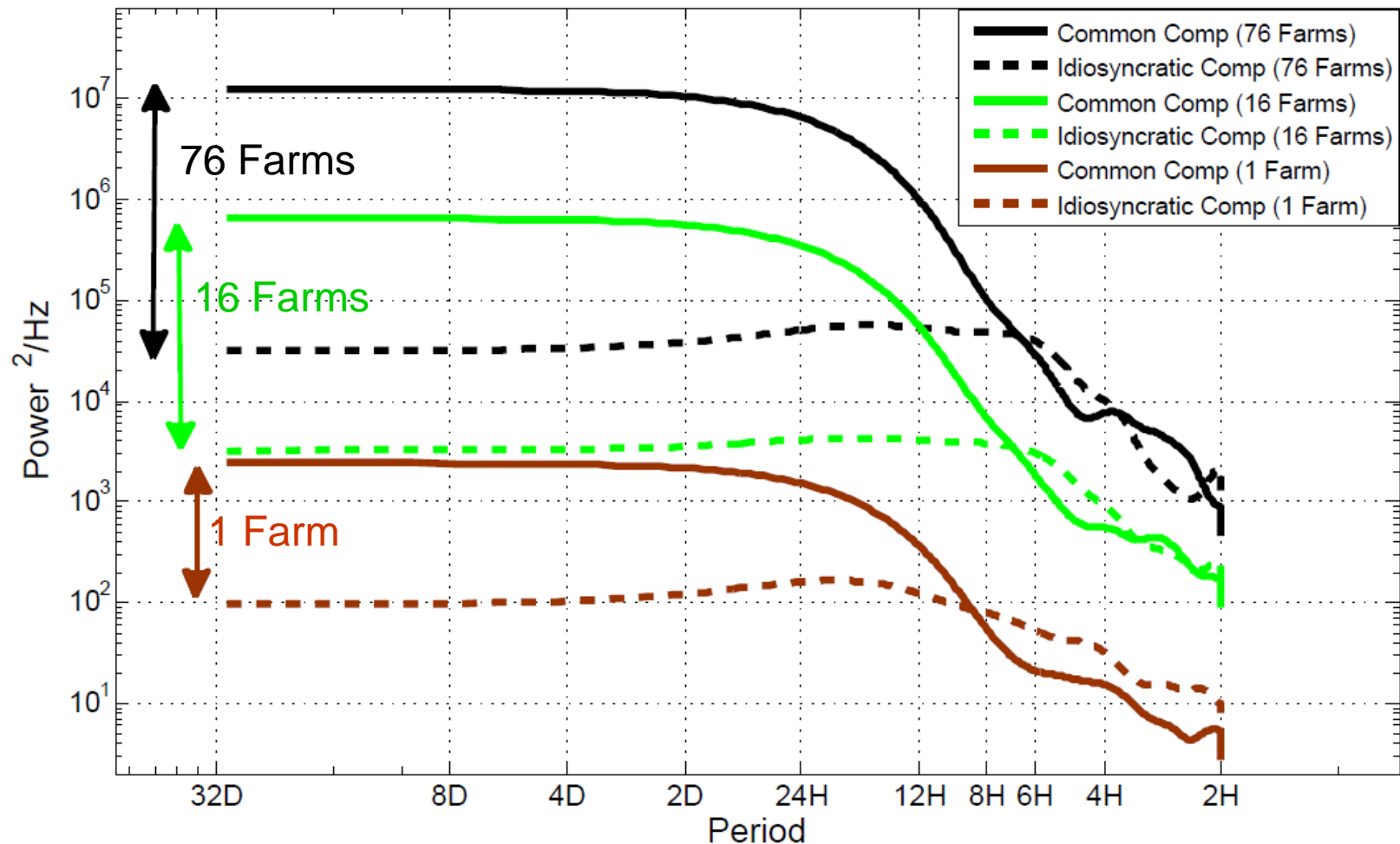


Scaling of wind power and wind variability.

- Intuitive that aggregating of wind over large areas should reduce relative variability.
- However, variability of each component scales differently with aggregation:
 - Periodic:
 - Scales approximately linearly with capacity,
 - Common stochastic:
 - Effects of underlying (weather) factors tend to add,
 - Idiosyncratic stochastic:
 - Weakly correlated between farms, so grows slowly.

Scaling of wind power and wind variability.

Total Common & Idiosyncratic Components / 2011



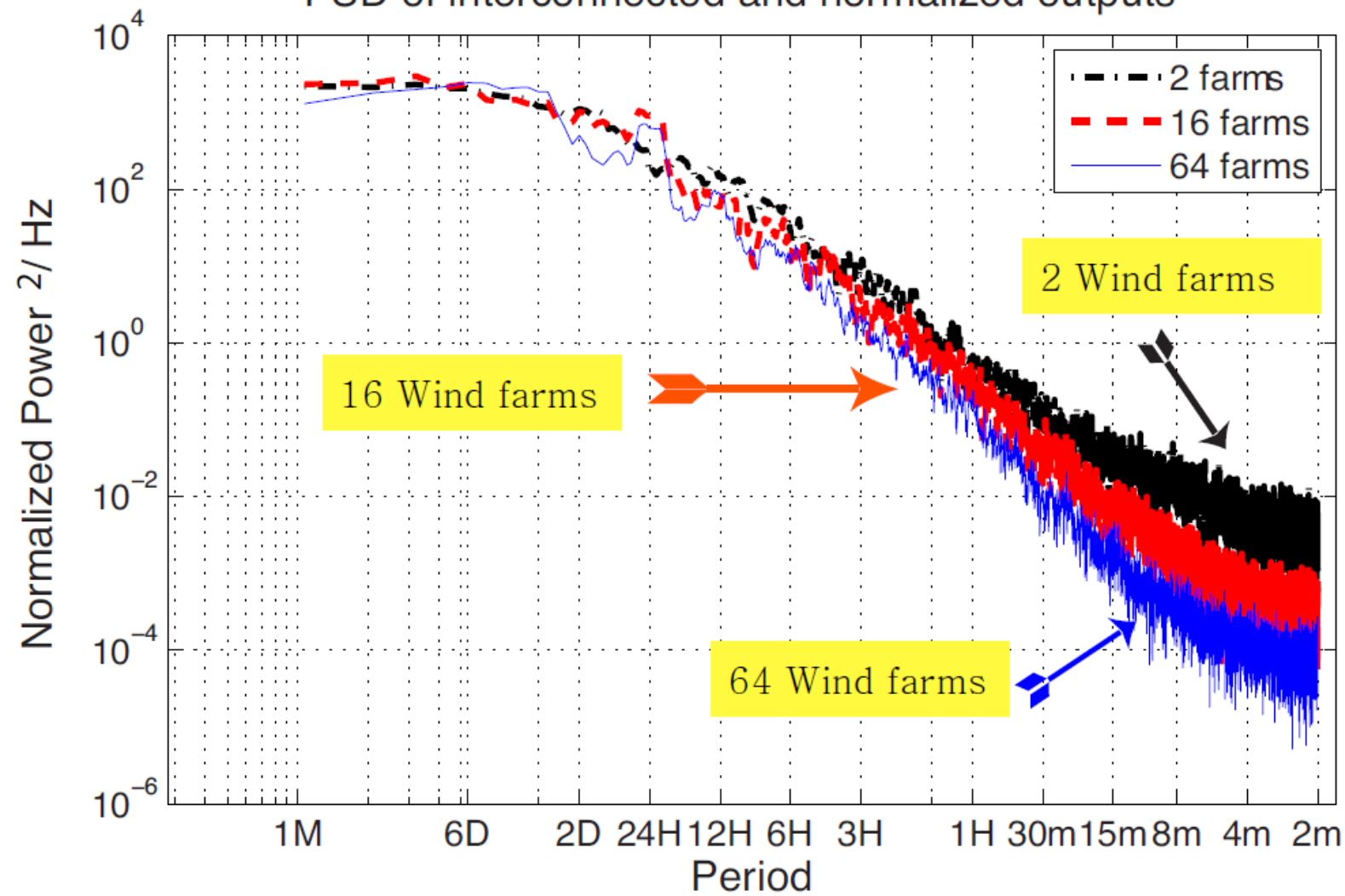


Scaling of wind power and wind variability.

- Higher frequency components of stochastic components grow more slowly with aggregation than lower frequency components:
 - Because idiosyncratic component grows slowly,
 - Aggregation reduces high frequency components relative to low frequency.
- Aggregation does not solve variability:
 - Diurnal periodic component,
 - Common stochastic component.

Scaling of wind power and wind variability.

PSD of interconnected and normalized outputs



Source:
based on
Apt (2007),
and Lee
and Baldick
(2014).



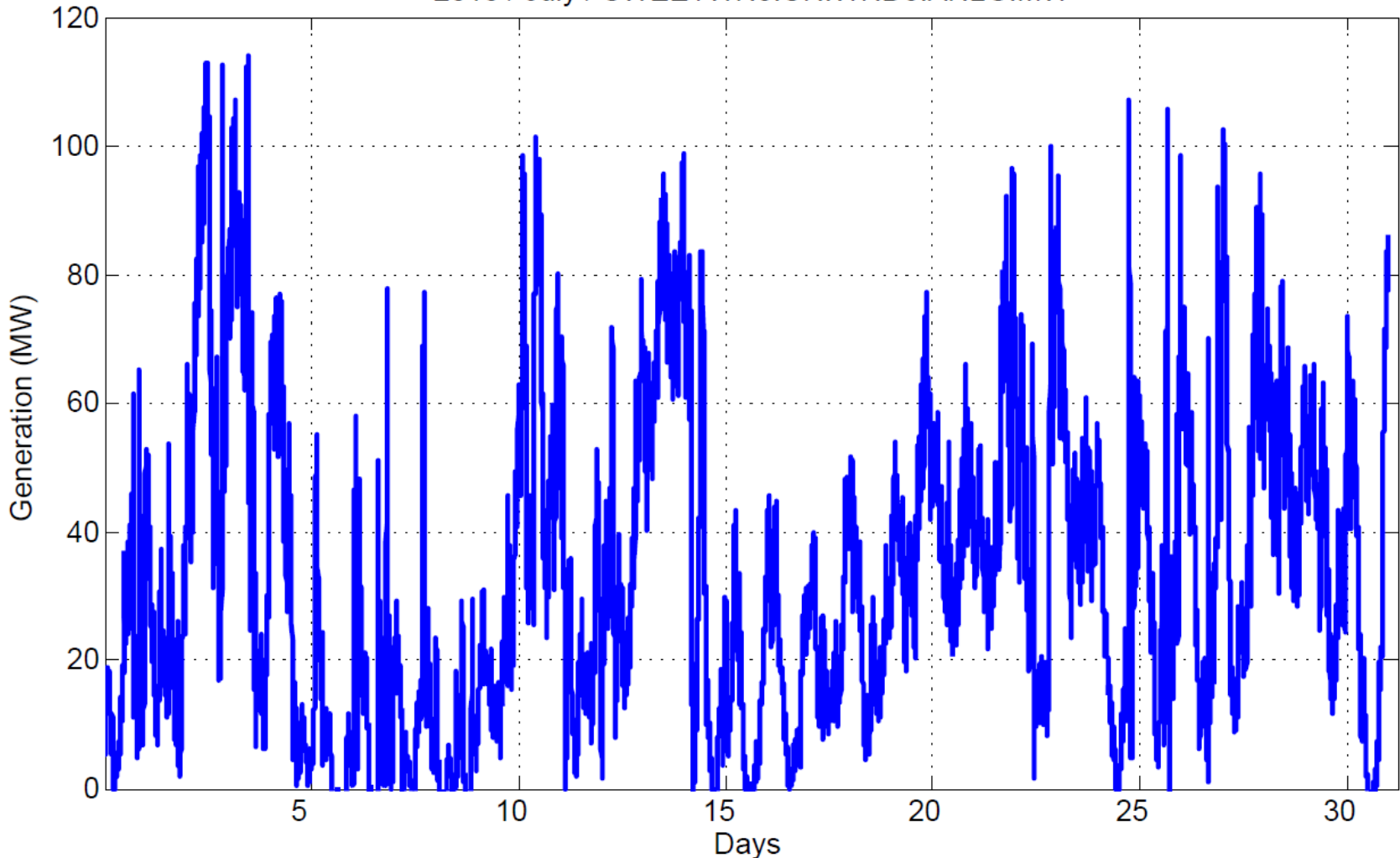
Scaling of wind power and wind variability

- Echoes observations in Katzenstein, Fertig, and Apt (2010):
 - Most reduction of variability is obtained by aggregating relatively few farms,
 - Still expect significant intermittency in total wind, even aggregating many farms in a region,
 - Intermittency only reduced further by aggregating over geographical scales that span different wind regimes:
 - Inland and coastal Texas wind.



Intermittent wind power production.

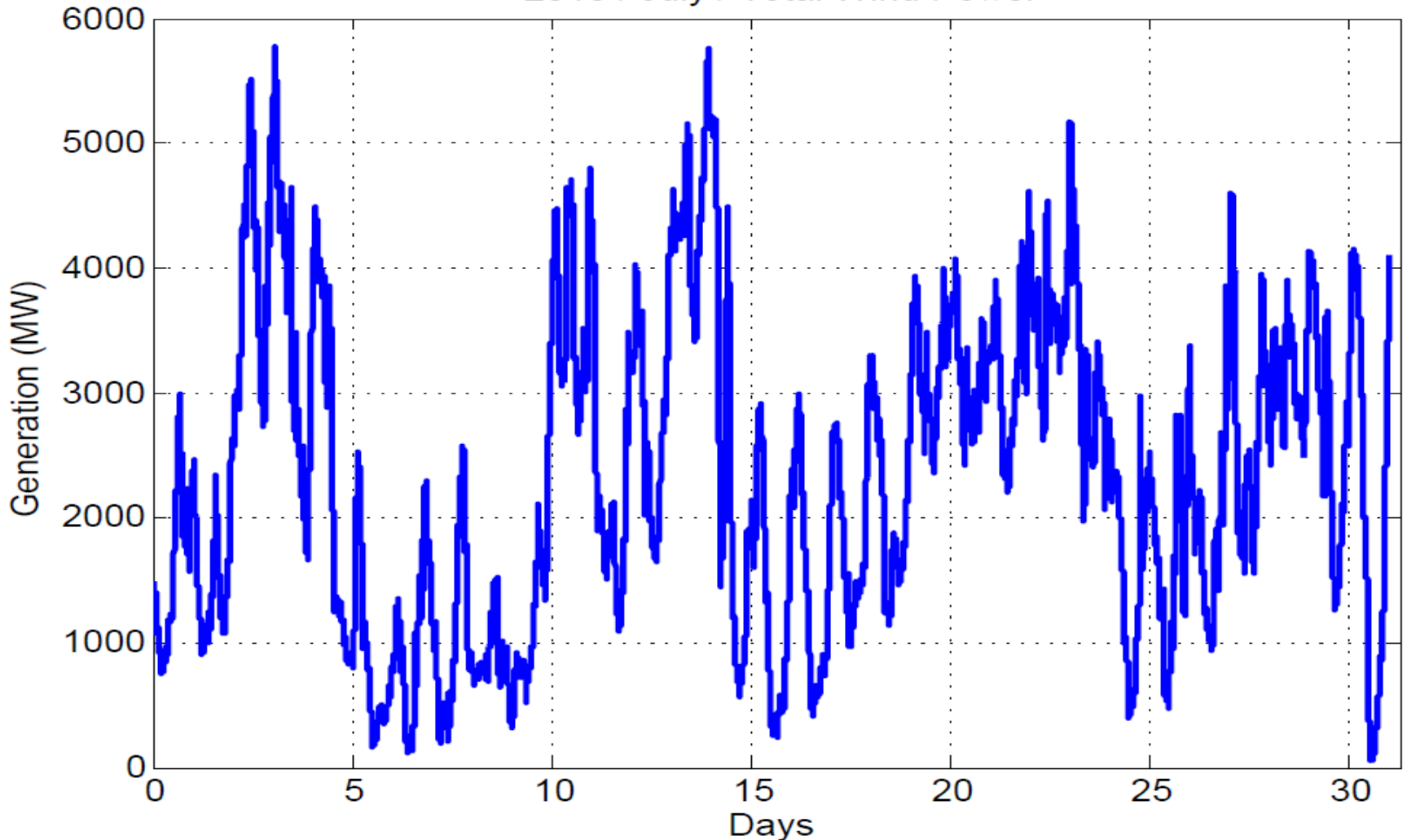
2010 / July / SWEETWN3.UN.WND3.ANLG.MW





Intermittent wind power production.

2010 / July / Total Wind Power





Implications for electricity systems.

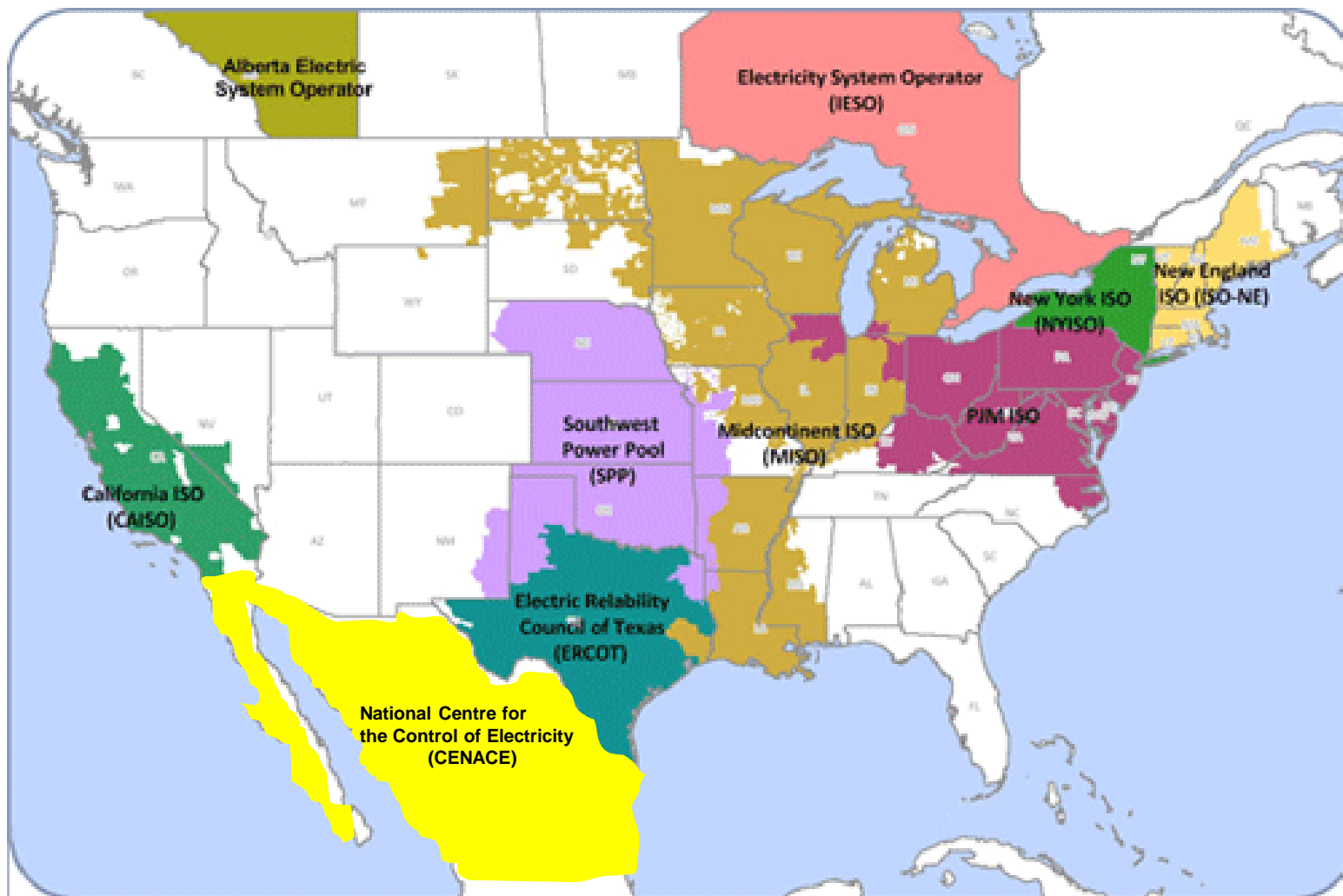
- Electricity supply must match load continuously (first law of thermodynamics),
- In short-term, variation between mechanical power and electrical load is compensated by inertia of electrical machines:
 - About 8 seconds of supply in inertia.
- Over longer time-frames, generators are instructed (“dispatched”) to adjust mechanical power to balance generation and load.
- Wind variability complicates balancing.



“Organized” wholesale markets.

- About 60% of US electric power supply is sold through “organized” markets administered by Regional Transmission Organizations (RTOs) (USEIA, 2016).
- RTOs include Midcontinent, California, New England, New York, PJM, Southwest Power Pool, Electrical Reliability Council of Texas (ERCOT).
- Will focus on organized markets.

Organized wholesale markets in North America.



Source: www.ferc.gov



Organized wholesale markets.

- Dispatchable generation typically receives a target generation level every 5 minutes:
 - Ramp to this level over next 5 minute interval,
- Target generation level based on forecast of the load minus renewable production.
- Fluctuations within 5 minute intervals and error in forecast:
 - compensated by generation that responds to faster signals, “regulation ancillary service,”
 - more variability requires more regulation.



Organized wholesale markets.

- Scaling analysis implies that wind variability in 5 minute interval grows slowly with total wind:
 - Required amount of regulation ancillary service grows slowly with total wind capacity,
 - Needed regulation capacity in ERCOT still mostly driven by load variability,
 - Various changes to market design have enabled better utilization of regulation capacity.
- Variability over tens of minutes to hours to days:
 - Growing with wind.



Day-ahead market.

- Short-term forward market based on anticipation of tomorrow's conditions,
- Provides advance warning for “slow start” generators that require hours to become operational, “committed,”
- Wind forecasts can be poor day-ahead:
 - Implications if generator fleet is mostly slow start,
 - Necessitates commitment of significant capacity “just in case,” with implications for lower efficiency, increased emissions.



Real-time market.

- Arranges for 5 minute dispatch signals,
- Increasingly also represents commitment of “fast-start” generators through “lookahead dispatch” (not, yet, in ERCOT).
- Increasing availability of fast-start generators avoids commitment except when they are very likely to be needed.
- Large wind ramps and high off-peak wind can still be problematic if not enough installed and available flexible capacity to compensate for wind variability.



Must-take resources.

- In some markets, wind is “must take,” necessitating that other resources compensate for almost all wind variability.
- In ERCOT, Midcontinent, and some other areas, wind farms participate by offering into market and being dispatched within limits:
 - Just like all other generators,
 - Provides flexibility to RTO to curtail “economically,” with prices falling low, to zero, or even negative,
 - Arguably facilitated high level of wind in ERCOT.

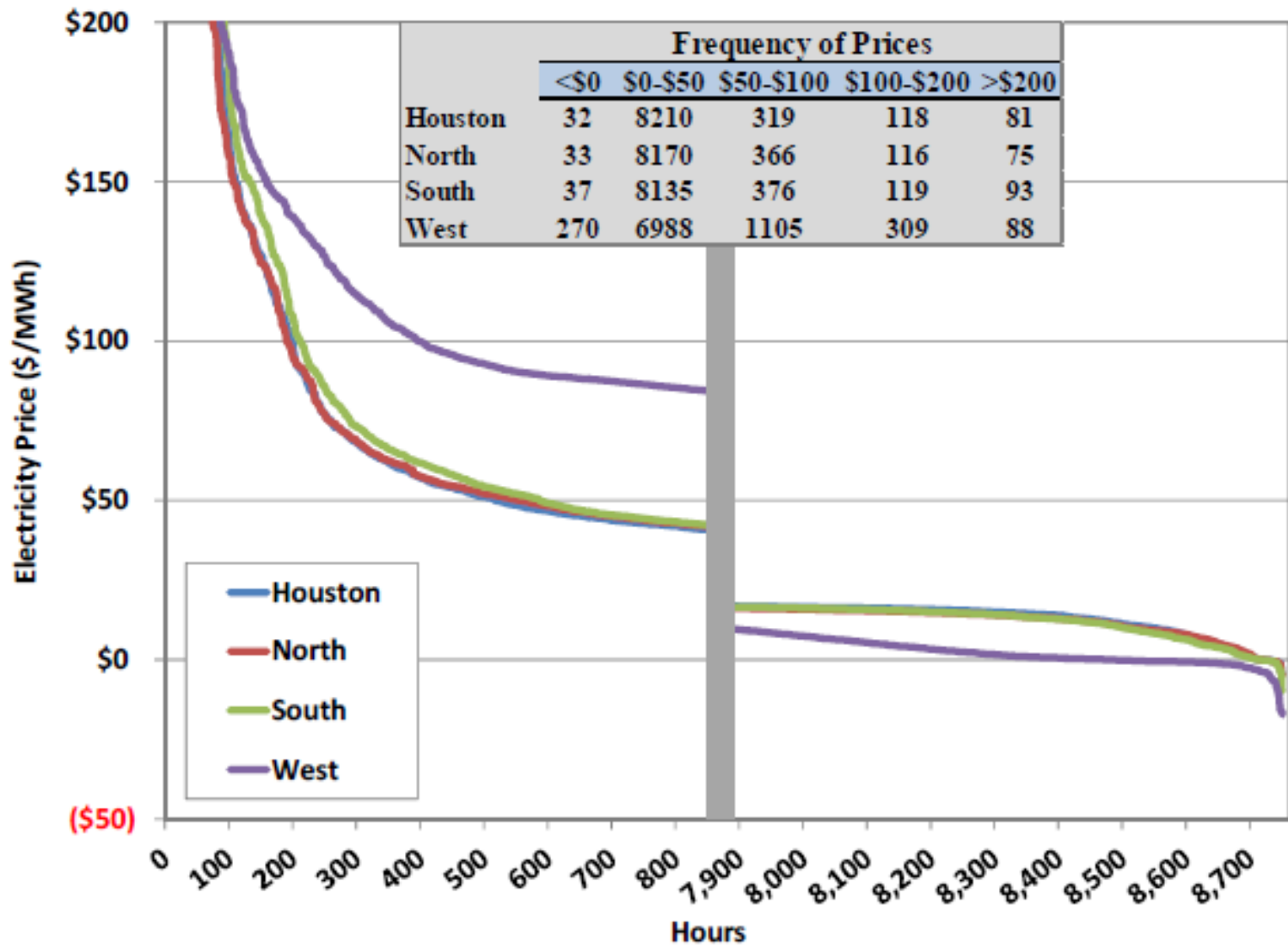


Diurnal periodic variation, intermittency, and markets.

- West Texas wind has peak production when load is low.
- When stochastic wind component adds to periodic peak but load is low, total wind production requires thermal generation to dispatch down or switch off.
- In market based approach to integrating wind, this results in low, zero, or even negative prices.



ERCOT price-duration curve in 2018.



Source: Potomac (2019), Figure 9.



Conclusion.

- Periodic component plus GDFM for stochastic component provides good match to statistics of empirical wind power data:
 - Periodic, common stochastic, and idiosyncratic stochastic components.
- Explains characteristics of aggregated wind production and scope for reduction of variability by aggregation.
- Markets with wind will experience times of low, zero, or negative prices.



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