

Data-Driven Do-Not-Exceed Limits

Feng Qiu ¹

A joint work with Zhigang Li ² and Jianhui Wang¹

¹Energy Systems Division
Argonne National Laboratory

²Tsinghua University
Beijing, China

FERC Software Conference
June 23, 2015

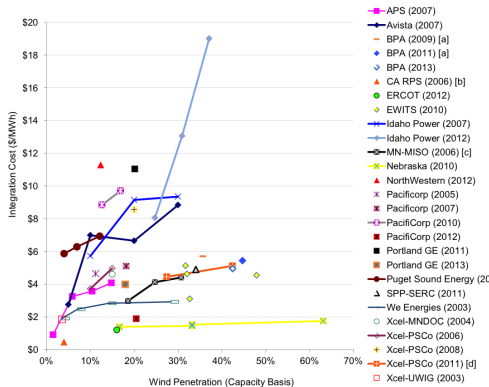
Outline

- 1 Introduction – Wind Integration
- 2 DNE Limits and Possible Enhancement
- 3 Data-Driven Wind Dispatch Range Determination
- 4 Numerical Demonstration



Wind Integration Status

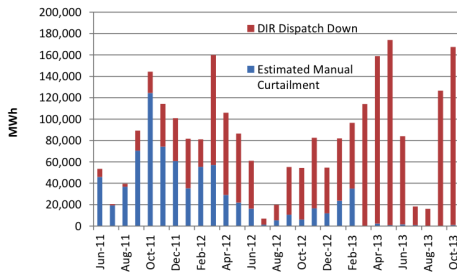
- ▶ Wind penetration in U.S.: 1.5% (2008) → 4.5% (2013) ... → 20% (2030)
- ▶ Wind integration costs
 - ▶ Non-dispatchable
- ▶ Have to improve wind dispatchability through better scheduling !



Wind Integration Cost (Source: EERE 2013)

Wind Dispatchability

- ▶ Traditionally a non-dispatchable resource
 - ▶ Curtailment occurs due to congestions or security reasons
 - ▶ Curtailment is implemented in an ex post fashion, endangering system reliability
- ▶ Proactive approaches to improve wind dispatchability
 - ▶ Dispatchable Intermittent Resource (DIR) protocol (MISO)



- ▶ Do-Not-Exceed (DNE) Limits (ISO New England)



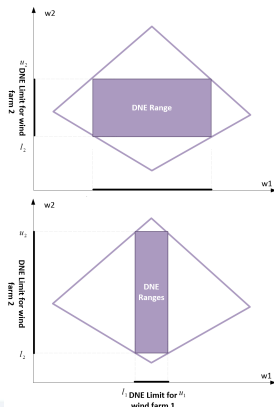
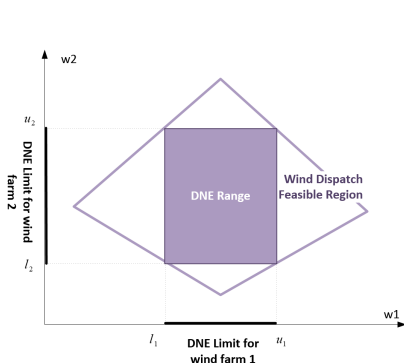
A Variable Resource Dispatch Framework – DNE Limits

- ▶ Proposed by ISO New England [Zhao, Zheng, and Litvinov (2015)]
- ▶ DNE procedure
 - 1 Determine the *dispatch base point* based on hours-ahead wind forecasting
 - 2 Calculate the *maximal ranges* of power output for each wind farm, based on security analysis with reserve levels given by the dispatch base point
 - 3 Wind farms follow these ranges as a dispatch guidance
- ▶ Benefits
 - ▶ A dispatch framework for variable resources
 - ▶ DNE limits are simple for execution



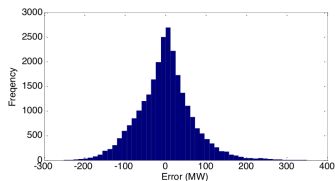
DNE Limits Selection

- ▶ Allocation of dispatch ranges in the dispatchable regions
- ▶ Maximize the weighted circumference of the box
- ▶ LMPs are used to weight the ranges [Zhao, Zheng, and Litvinov, 15']
 - ▶ A market perspective: $\sum_{i \in N} LMP_i \times (u_i - l_i)$
- ▶ Which wind farm should get larger limits? what other criteria for range determination?

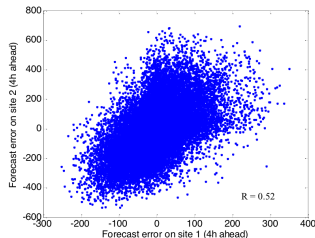


Discrepancy Between Forecast and Dispatch

- ▶ Actual wind power dispatched could deviate from base point because of
 - ▶ Point forecast
 - ▶ Forecasting errors
 - ▶ Forecasting bias
 - ▶ Curtailments due to congestions or security reasons
- ▶ Deviations may be correlated

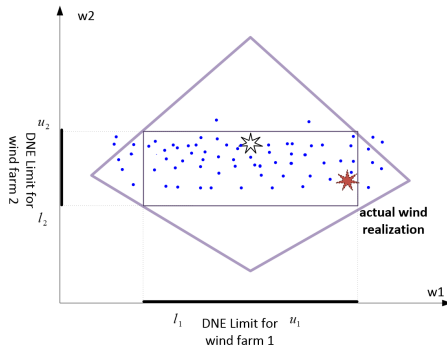
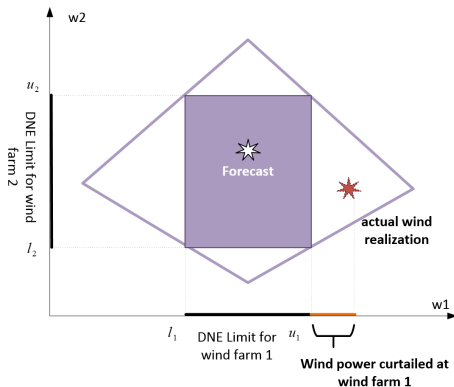


Source: NREL



DNE Limits Under Uncertainties

- ▶ Original DNE limits might not be effective in capturing the uncertainties
- ▶ Can we design DNE limits in another way to reduce curtailments?



A Data-Driven Dispatch Range Determination

- ➊ Using data to understand uncertainties
 - ▶ Data
 - ▶ Wind power forecasting from each wind farm
 - ▶ Observed wind power dispatch from each wind farm
 - ▶ Statistic features: mean, variance, covariance
 - ▶ Possible distribution functions
- ➋ Determine wind dispatch ranges considering the uncertainties

Goal of the Data-Driven DNE Limits

Maximize the probability that wind realization is within the DNE limits

$$\begin{aligned} \max_{\ell, u} \quad & \mathbb{P}\{\ell \leq \tilde{w} \leq u\} \\ \text{s.t.} \quad & \ell + (1 - v)u \in \mathcal{D} \quad \forall v \in [0, 1]^n \end{aligned}$$

$\mathcal{D} := \{w \in \mathbb{R}_+^n : \exists p \in \mathbb{R}_+^q \text{ s.t. } Ap + Bw \leq c\}$, called wind dispatchability set
 p : conventional generation dispatch; recourse variables.



Characterizing Uncertainties

- ▶ Perfect information
 - ▶ Statistic models
- ▶ Limited information
 - ▶ Moments approximation [Scarf (1958), Vandenberghe et al. (2007)]
 - ▶ $\mathcal{P}_1 := \{\xi \in \mathbb{R}^n : \mathbb{E}[\xi] = \mu; \mathbb{E}[\xi\xi^\top] = \delta\}$
 - ▶ $\mathcal{P}_2 := \{\xi \in \mathbb{R}^n : (\mathbb{E}[\xi] - \mu)^\top \Lambda^{-1}(\mathbb{E}[\xi] - \mu) \leq \gamma_1; \mathbb{E}[(\xi - \mu)(\xi - \mu)^\top] \leq \gamma_2\}$
 - ▶ Density function approximation [Pardo (2006), Jiang and Guan (2015)]
 - ▶ A family of distribution functions that is not “far” from a reference distribution function
 - ▶ Sample average approximation [Shapiro (2003), Shapiro and Nemirovski (2005)]
 - ▶ True distribution is replaced by the empirical one



Introduction to Data-Driven Approaches

- ▶ A typical data-driven approach

- 1 Design an uncertainty set: all probability distributions that satisfy the definition, e.g., \mathcal{P}_1 and \mathcal{P}_2
- 2 Distributionally-robust optimization (a worst-case point of view)

$$\max_{\ell, u} \inf_{p \in \mathcal{P}} \mathbb{P}\{\ell \leq \xi \leq u\}$$

- ▶ Disadvantages

- ▶ Non-convex reformulations/approximations
- ▶ Overly conservative



Sample Average Approximation

- ▶ Actual wind power dispatched at wind farm i in the next time period

$$\tilde{w} = w^* + \tilde{e},$$

$w^* \in \mathbb{R}_+^n$: the base point; $\tilde{e} \in \mathbb{R}^n$: error vector

- ▶ Replace the real probability by $|S|$ samples (observations) $w^j = w^* + e^j$:

$$\mathbb{P}\{(\tilde{w} = w |_{w^* + e^j})\} = \frac{1}{|S|} \quad j \in S$$

- ▶ Approximate the probability:

$$\mathbb{P}\{\ell \leq \tilde{w} \leq u\} = \frac{\sum_{j \in S} \mathbb{1}\{\ell \leq w^j \leq u\}}{|S|}$$

$$\text{where } \mathbb{1}(\ell \leq w^j \leq u) = \begin{cases} 1 & \text{if } \ell \leq w^j \leq u \\ 0 & \text{otherwise} \end{cases}$$



Sample Average Approximation-Continued

- ▶ SAA reformulation

$$\max_{\ell, u} : \mathbb{P}(\ell \leq \tilde{w} \leq u)$$

\Downarrow

$$\max : \sum_j z_j$$

$$\text{s.t. } \ell - (1 - z^j) * M \leq w^j \leq u + (1 - z^j) * M \quad \forall j \in S$$

- ▶ Strong valid inequalities for SAA based on the *mixing set* results [Luedtke, Ahmed & Nemhauser (2010)] [Günlük & Pochet (2001)] etc.
- ▶ A mixed-integer linear programming (MILP) problem that can be readily solved by commercial solvers



Solution Approach: Delayed Constraint and Column Generation

- ▶ Master problem: SAA formulation + necessary dispatchability set constraints

$$\begin{aligned} \max : & \sum_j z_j \\ \text{s.t. } & \ell - (1 - z^j) * M \leq w^j \leq u + (1 - z^j) * M \quad \forall j \in S \\ & \ell, u \in \mathbb{R}_+^n, z^j \in \{0, 1\} \quad \forall j \in S \\ & Ap^k + B(\ell + (1 - v^k)u) \leq c \quad k = 1, 2, \dots \end{aligned}$$

k is the iteration number.¹

- ▶ Subproblem: Identify the most violated dispatchability constraints

$$\begin{aligned} \max_{v \in [0, 1]^n} \min_{s \in \mathbb{R}_+^n} & \bar{\mathbf{1}}^\top s \\ \text{s.t. } & Ap + B(\ell^k + (1 - v)u^k) - s \leq c \end{aligned}$$

- ▶ Observation: maximal violations always achieved by “corner” points of the “box”
- ▶ Solution: dualize the inner problem and convert it to a MILP problem

¹ Solving Two-stage Robust Optimization Problems Using a Column-and-Constraint Generation Method, B Zeng, L Zhao, Operations Research Letters 41 (5), 457-461

Case Study: 6-Bus System

► Configuration

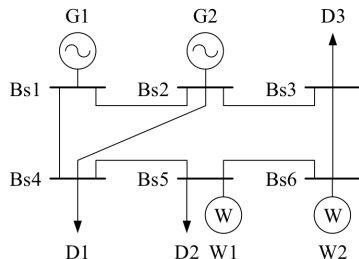
- 6 bus, 7 lines
- 2 thermal generators (250MW*2)
- 2 wind farms (100MW*2)
- System loads: 266MW-434MW

► Wind data (Eastern Wind Dataset by NREL¹⁾)

- Wind farm #1: site 3902 (W89.18, N41.68)
- Wind farm #2: site 3945 (W88.55, N40.49)

► Comparison

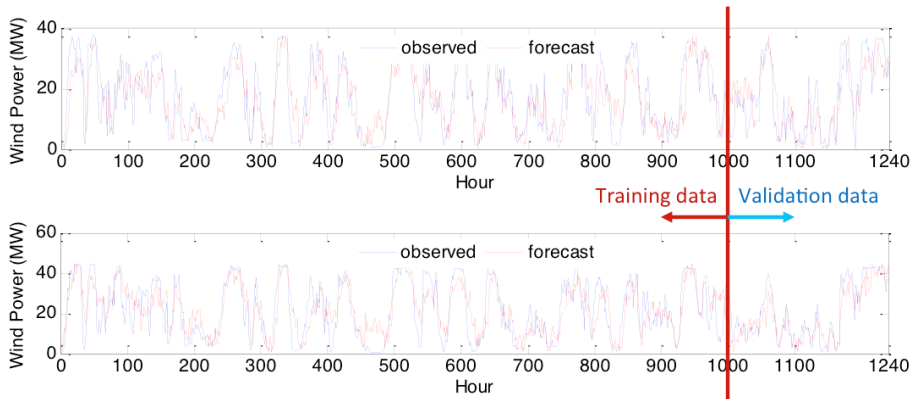
- Proposed method (D-DNE)
- Original method (LMP-weighted DNE)



¹ Available: http://www.nrel.gov/electricity/transmission/data_resources.html

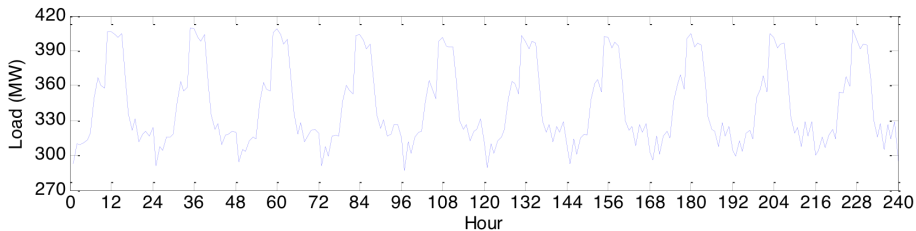
Case Study: A 6-Bus System

- ▶ Wind profiles (1240 hours, for training and validation)



Case Study: 6-Bus System

► Load profile



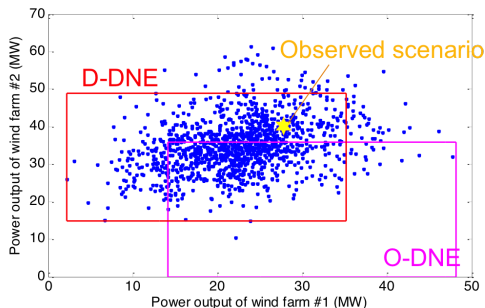
DNE Limits Comparison: Single Snapshot (t=240)

► D-DNE:

- Actual wind realization lies in the D-DNE box
- No wind curtailment

► O-DNE

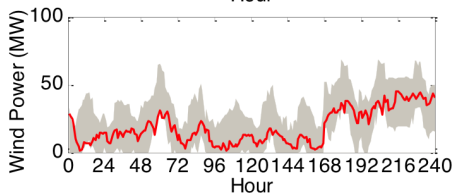
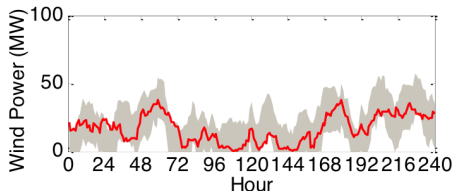
- Wind realization lies outside of the O-DNE box
- Wind curtailment occurs in wind farm #2



DNE Comparison: Multiple Snapshots

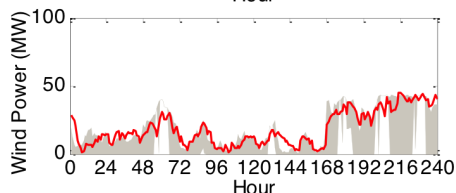
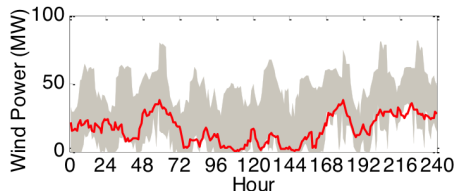
D-DNE

87.9% wind realizations are covered



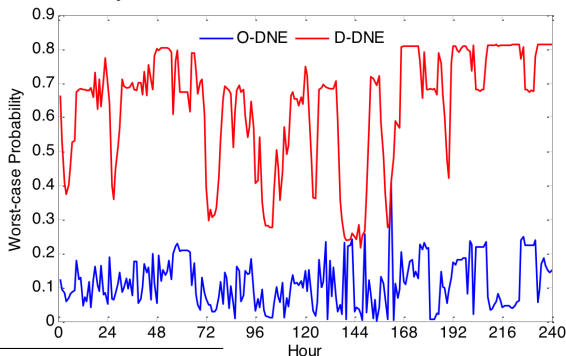
O-DNE

37.5% wind realizations are covered



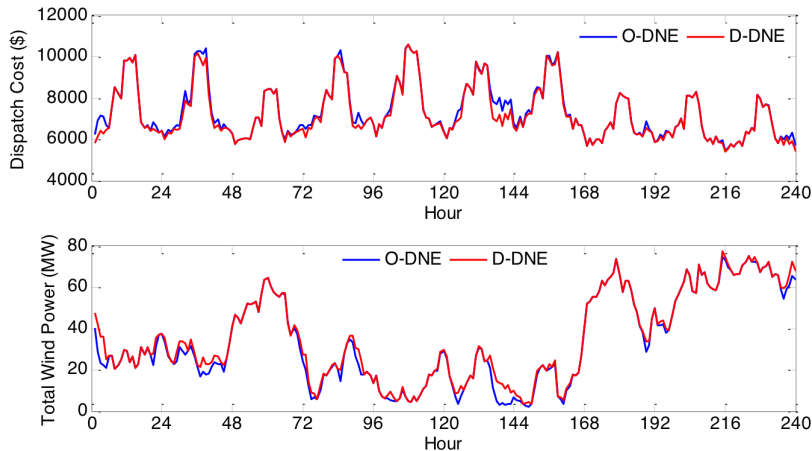
Wind Curtailment Probability Under Worst-Case Probability

- ▶ Goal: To see how robust our DNE limits are in the worst-case “scenario”
- ▶ Measurement: The probability of “coverage” under the worst case scenario (probability)
- ▶ Approach: A lower bound on the probability given incomplete distribution information, i.e., only the first two moments as in \mathcal{P}_1 ¹



¹B. P. Van Parys, P. J. Goulart, and D. Kuhn, "Generalized Gauss inequalities via semidefinite programming," *Mathematical Programming*, 2015.

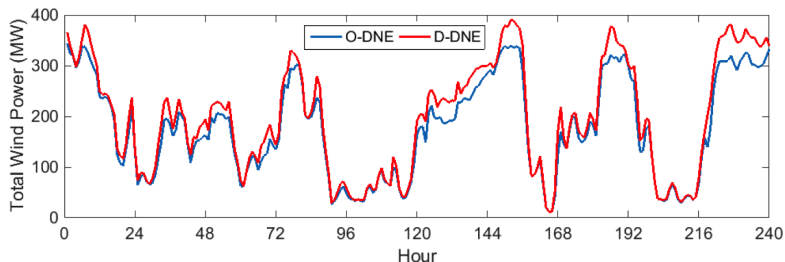
Economic Benefits



- ▶ D-DNE accommodates more wind power and incurs less dispatch cost

Algorithm Scalability: IEEE 118-Bus System

- Configuration
 - 118 bus, 186 lines, 76 conventional generators, 10 wind farms
 - Wind data: from 10 sites in Eastern Wind Dataset by NREL
- Total simulation time: 388s (240 runs)



Strengthened Formulation

6-Bus System (2 wind farms)

MP Iter.	Original Formulation			Strengthened Formulation		
	Root gap (%)	# nodes	Time (s)	Root gap (%)	# nodes	Time (s)
#1	97.8	7	0.73	42.7	0	0.26
#2	97.8	0	0.74	5.5	0	0.37
Term.	Optimal			Optimal		

IEEE 118-Bus System (10 wind farms)

MP Iter.	Original Formulation			Strengthened Formulation		
	Root gap (%)	# nodes	Time (s)	Root gap (%)	# nodes	Time (s)
#1	92.8	>1.4 M	>7200	75.1	0	1.01
#2	-	-	-	19.2	0	1.55
Term.	Time limit exceeded			Optimal		

- ▶ Computation efficiency is significantly improved
 - ▶ Exploit the structure of SAA
 - ▶ Only a few data points are dominant



Summary

- ▶ Data-Driven Do-Not-Exceed Limits
 - ▶ DNE limits considering uncertainties
 - ▶ Data-driven, requiring little knowledge
 - ▶ Computational efficiency and scalability
 - ▶ Improved wind utilization
- ▶ Possible future research
 - ▶ Multi time period
 - ▶ Resource redispatch



Thank you!

Comments?

