

Stochastic OPF in presence of Renewable

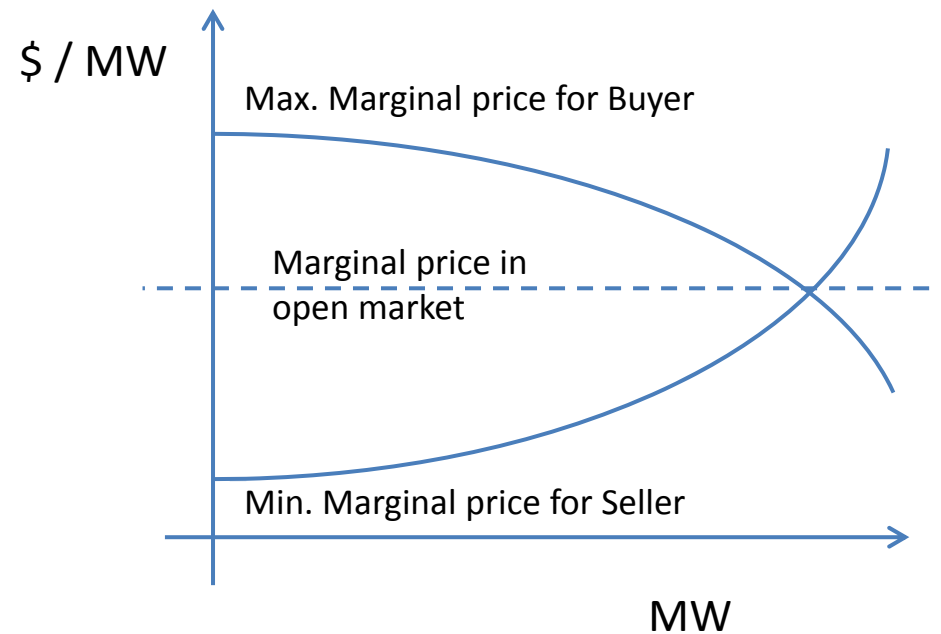
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EE292K Intelligent Energy Project

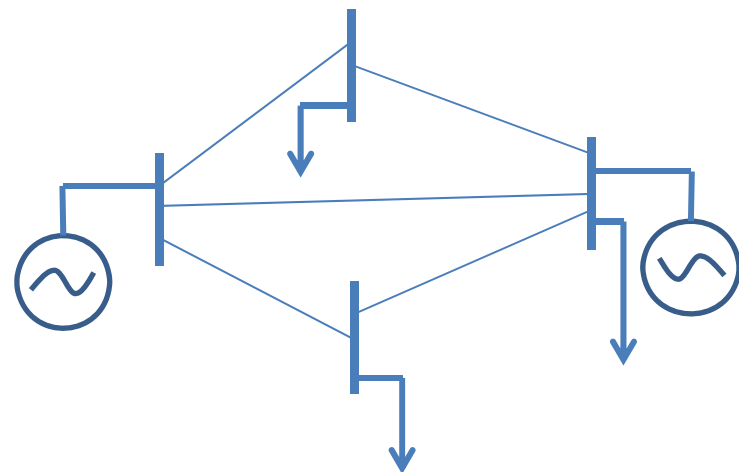
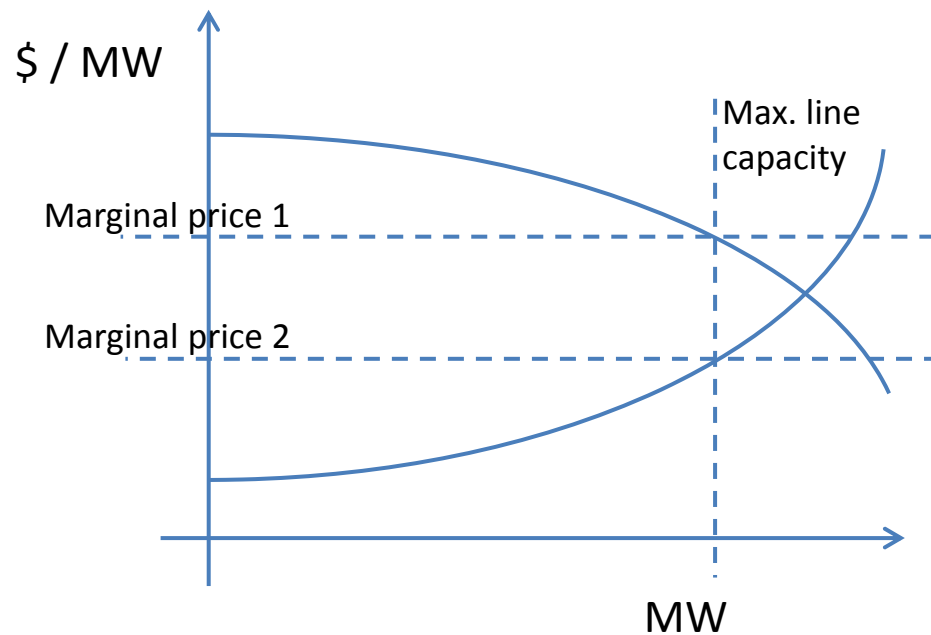
Market

- Buyer's economic utility function
- Seller's cost function
- Open market: Maximize Social Welfare (economic benefit \$\$)
- Market Price



Electricity Market

- Electrical Energy is bought and sold
- Players – Generator, Load, IGO
- Physical constraints
 - Transmission Line Capacity
 - Power Flow



Optimal Power Flow

$f(P_1, P_2, \dots, P_n)$ Objective function to minimize

$\underline{V}_k \leq |V_k| \leq \overline{V}_k$ Voltage magnitude limits

$L_{ki} \leq l_{ki}$ Line loss (heat) limits

$\underline{P}_k \leq P_k \leq \overline{P}_k$ Power generation/consumption limit

$P_k = \sum |V_k| |V_i| (G_{ki} \cos \theta_{ki} + B_{ki} \sin \theta_{ki})$ Power Flow

Renewables

- Clean energy generation – reduce carbon footprint
- Low cost to produce (even negative)
- Variable and Unreliable
- Challenges in optimal power flow problem

System Model

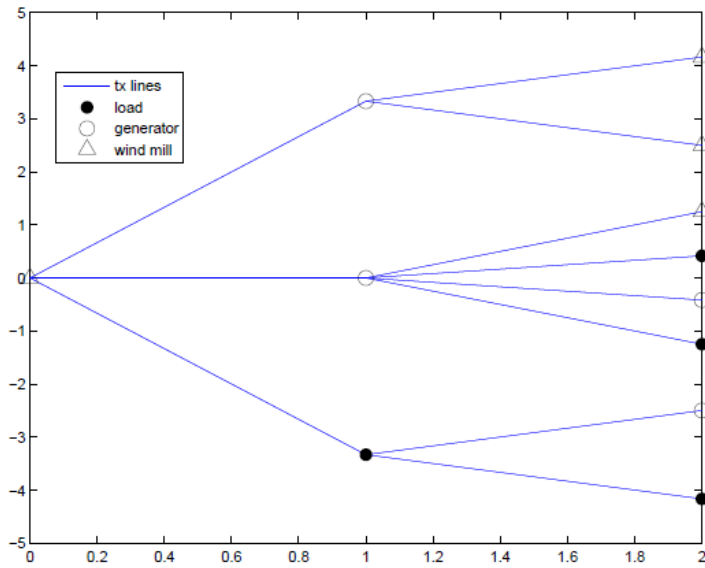


Figure 1: Power system network with $N = 12$ busses (indexed from top to bottom and left to right) and $J = 11$ transmission lines, $B^L = \{4, 8, 10, 12\}$, $B^{NR} = \{2, 3, 9, 11\}$, $B^R = \{1, 5, 6, 7\}$

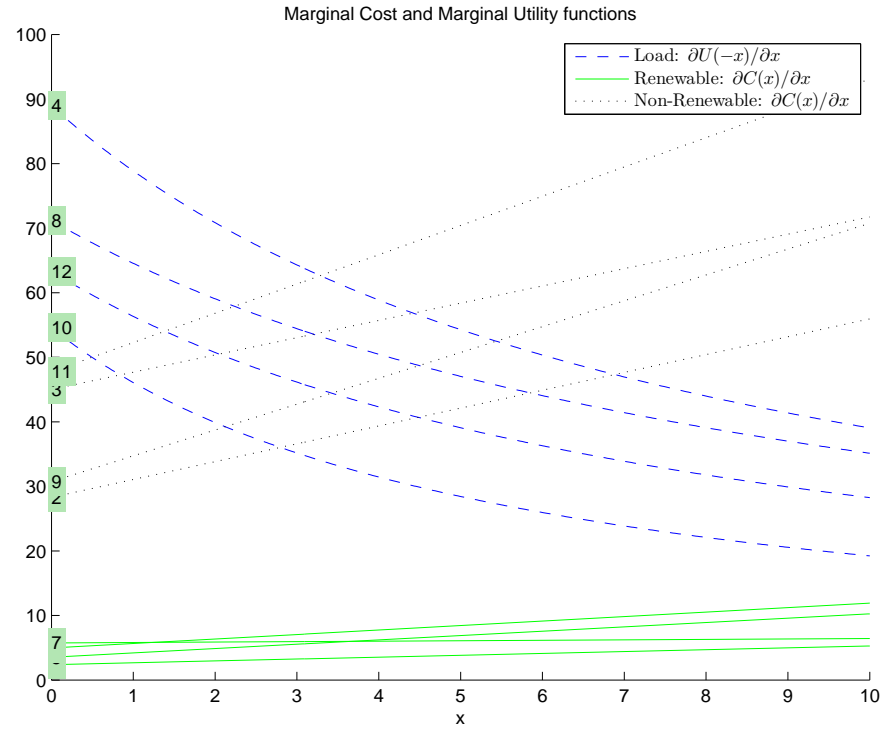


Figure 2: Marginal Cost/Utility functions for generators/loads, labels at start of the curve represent the bus number.

Deterministic Approach

- Maximize social welfare

$$\underset{\mathbf{V}, \mathbf{P}}{\text{minimize}} \quad \sum_{i=1}^N C_i(P_i)$$

- Constraints

- Power flow
- Line loss
- Voltage levels
- Renewable Power

$$\text{subject to:} \quad P_i = \sum_{k=1}^N |V_i||V_k|(G_{ik}\cos\theta_{ik} + B_{ik}\sin\theta_{ik}), \quad i = 1, \dots, N$$

$$L_i(\mathbf{V}) \leq l^{max}, \quad i = 1, \dots, J$$

$$V_{min} \leq |V_i| \leq V_{max}, \quad i = 1, \dots, N$$

$$P_i \leq P_i^{max}, \quad i \in B^R$$

$$P_i \leq 0, \quad i \in B^L$$

- Not a convex problem

$$L_{mn}(\mathbf{V}) = |V_m - V_n|^2 g_{mn}$$

- Linear approximation

How to solve

- Change of Variable
- Convex optimization
- Easy to solve (cvx)
- Restricted to tree topology

$$\mathbf{W} = \mathbf{V}\mathbf{V}^H$$

minimize
 \mathbf{W}, \mathbf{P}

subject to:

$$\sum_{i=1}^N C_i(P_i)$$

$$\mathbf{P} = \Re(\text{diag}(\mathbf{W}\mathbf{Y}^H)) \longleftrightarrow \mu$$

$$L_i(\mathbf{W}) \leq l^{max}, i = 1, \dots, J \longleftrightarrow \lambda$$

$$V_{min}^2 \leq W_{ii} \leq V_{max}^2, i = 1, \dots, N$$

$$P_i \leq P_i^{max}, i \in B^R$$

$$P_i \leq 0, i \in B^L$$

$$\mathbf{W} \in \mathbb{S}_+^N$$

$$[\text{Rank}(\mathbf{W}) = 1]$$

Solution

- Locational Marginal Prices
- Congestion Cost
- Power Injections

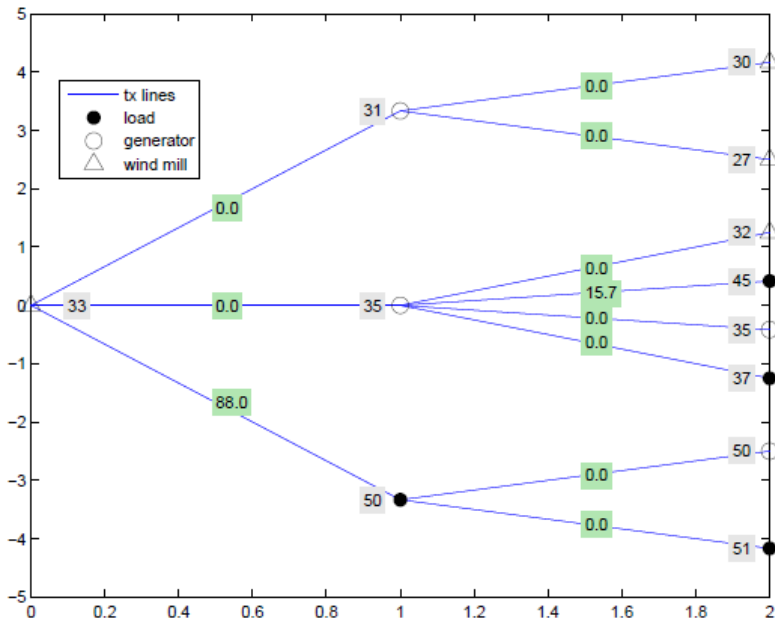


Figure 4: LMP are shown near the bus and congestion cost are shown near center of line

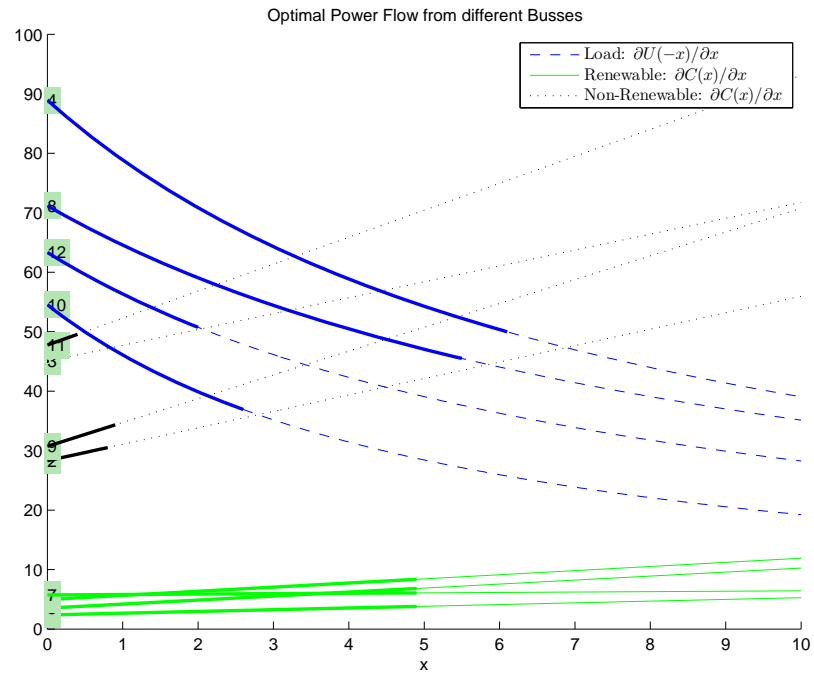


Figure 3: The power injection at various bus has been shown using dark lines on marginal curves.

Stochastic Optimization

- Random variable
- Average social welfare \$
- Average line loss
- Knowledge of probability distribution

minimize
 $\mathbf{W}(\mathbf{X}), \mathbf{P}(\mathbf{X})$

subject to:

$$E_{\mathbf{X}} \left(\sum_{i=1}^N C_i(P_i(\mathbf{X})) \right)$$

$$\mathbf{P}(\mathbf{X}) = \Re(\text{diag}(\mathbf{W}(\mathbf{X})\mathbf{Y}^H))$$

$$E_{\mathbf{X}}(L_i(\mathbf{W}(\mathbf{X}))) \leq l^{max}, i = 1, \dots, J \longleftrightarrow \lambda$$

$$V_{min}^2 \leq W_{ii}(\mathbf{X}) \leq V_{max}^2, i = 1, \dots, N$$

$$P_i(\mathbf{X}) \leq P_i^{max}(\mathbf{X}), i \in B^R$$

$$P_i(\mathbf{X}) \leq 0, i \in B^L$$

$$\mathbf{W}(\mathbf{X}) \in \mathbb{S}_+^N$$

Online Learning

- Iterative in time
- Stochastic sub-gradient method
- ‘learns’ the distribution
- λ converges approx.

$$\underset{\mathbf{W}^t, \mathbf{P}^t}{\text{minimize}} \sum_{i=1}^N C_i(P_i^t) + \sum_{i=1}^J \lambda_i^t (L_i(\mathbf{W}^t) - l^{max})$$

$$\text{subject to: } \mathbf{P}^t = \Re(\text{diag}(\mathbf{W}^t \mathbf{Y}^H))$$

$$V_{min}^2 \leq W_{ii}^t \leq V_{max}^2, \quad i = 1, \dots, N$$

$$P_i^t \leq P_i^{max, t}, \quad i \in B^R$$

$$P_i^t \leq 0, \quad i \in B^L$$

$$\mathbf{W}^t \in \mathbb{S}_+^N$$

$$\lambda_i^{t+1} = \lambda_i^t + \frac{\alpha}{t} (L_i(\mathbf{W}^t) - l^{max})$$

Results

- Line loss variation
- Locational Marginal Prices variation

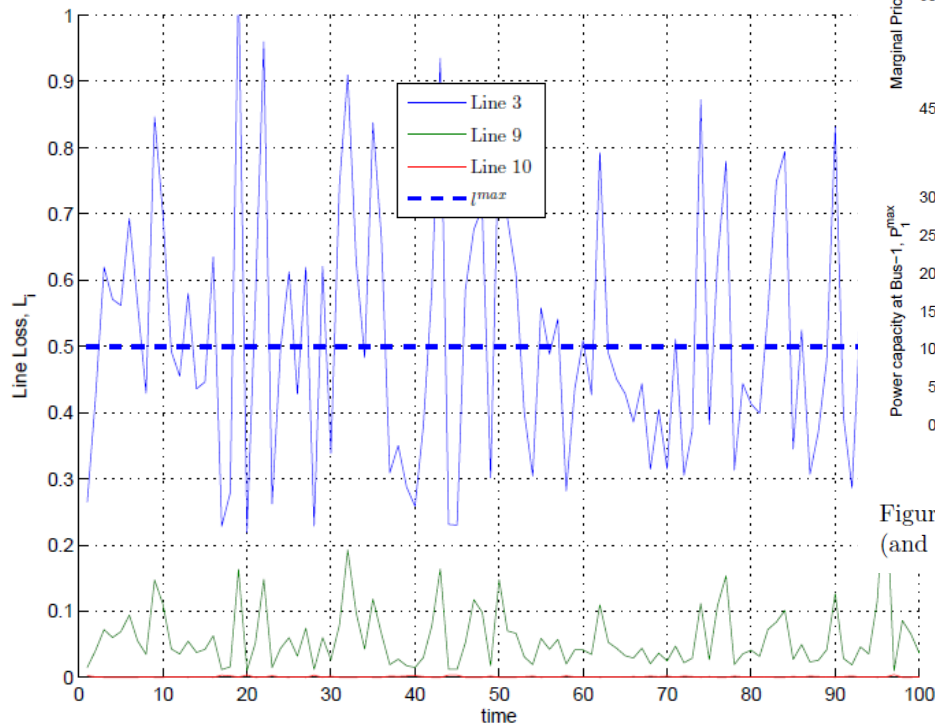


Figure 7: Variation in line loss with time

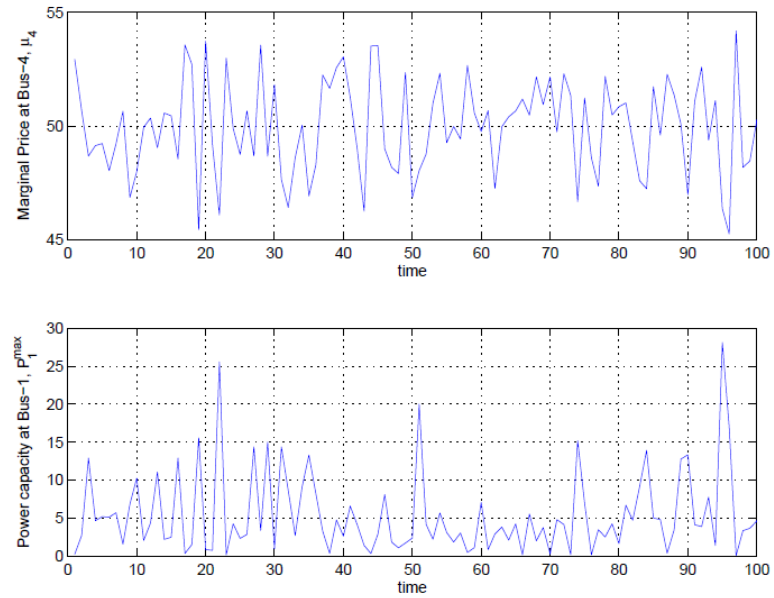


Figure 6: Figure shows the variation of LMP at bus-4 (load bus) with variations of wind (and hence P_1^{max})

Different distribution

- Optimal social welfare decreases with increase in standard deviation
- Congestion cost tends to increase

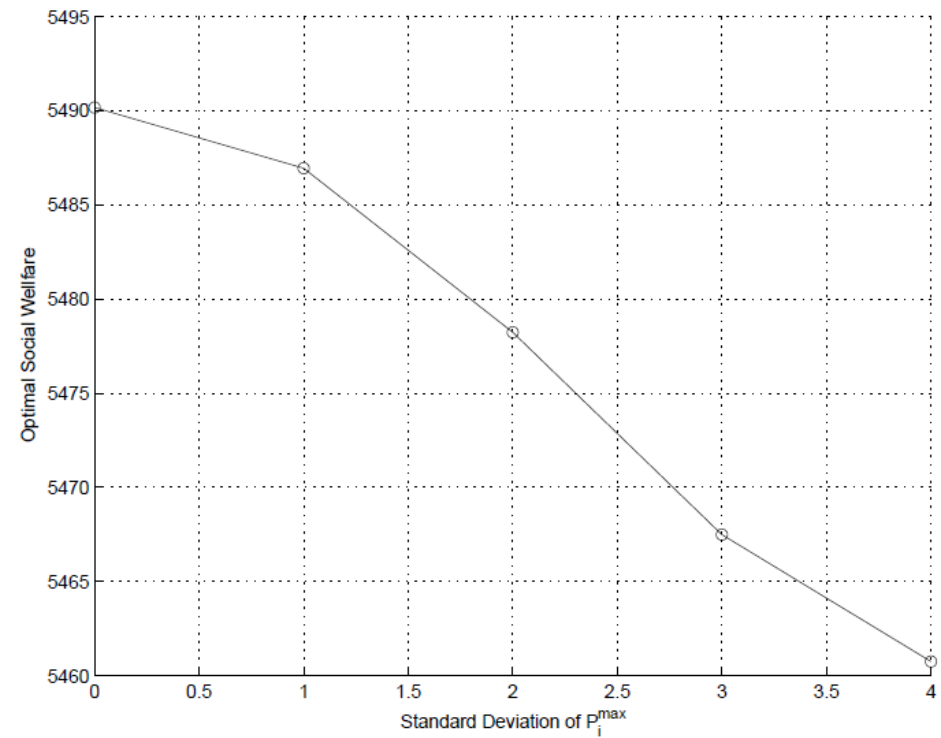
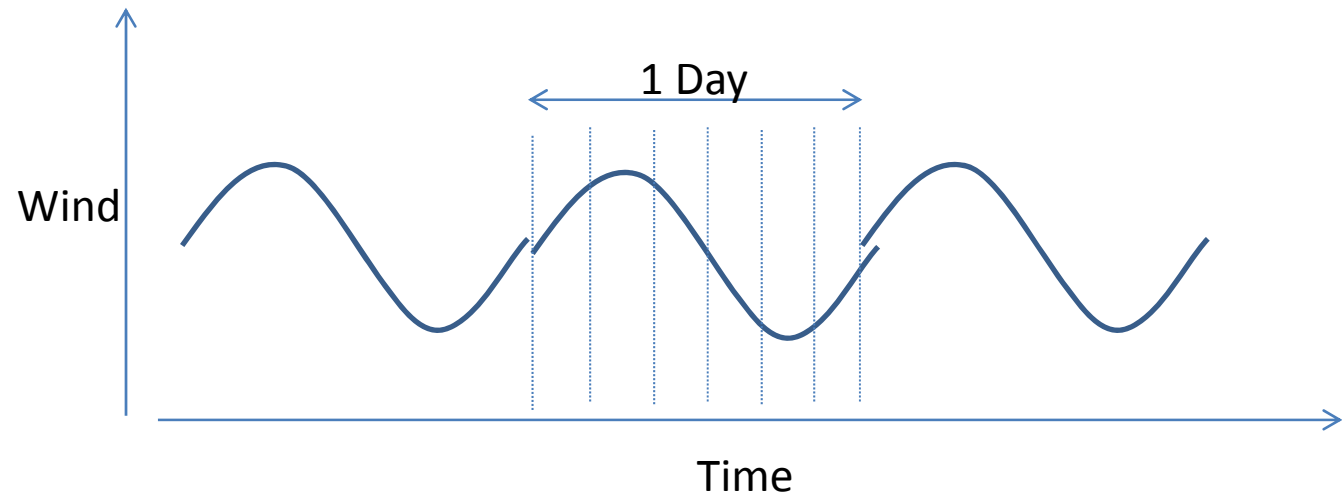


Figure 8: Effect of standard deviation of renewables on social welfare

Diurnal Wind Pattern

- Most variation captured by diurnal pattern
- Optimize over finite number of time slot
- Close to optimal



Rolling Horizon

$$\begin{aligned} & \underset{\mathbf{W}(t), \mathbf{P}(t)}{\text{minimize}} && \sum_{t=1}^T \sum_{i=1}^N C_i(P_i(t)) \\ & \text{subject to:} && \mathbf{P}(t) = \Re(\text{diag}(\mathbf{W}(t)\mathbf{Y}^H)) \quad t = 1, \dots, T \\ & && \frac{1}{T} \sum_{t=1}^T L_i(\mathbf{W}(t)) \leq l^{max} + \delta_{Li}, \quad i = 1, \dots, J \\ & && V_{min}^2 \leq W_{ii}(t) \leq V_{max}^2, \quad i = 1, \dots, N \\ & && P_i(t) \leq P_i^{max}(t) + \sum_{\tau=1}^{t-1} (P_i^{max}(\tau) - P_i(\tau)) + \delta_{Pi}, \quad i \in B^R, \quad t = 1, \dots, T \\ & && P_i(t) \leq 0, \quad i \in B^L, \quad t = 1, \dots, T \\ & && \mathbf{W}(t) \in \mathbb{S}_+^N, \quad t = 1, \dots, T \end{aligned}$$

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