

Pricing Energy Storage for Social-Welfare Maximization

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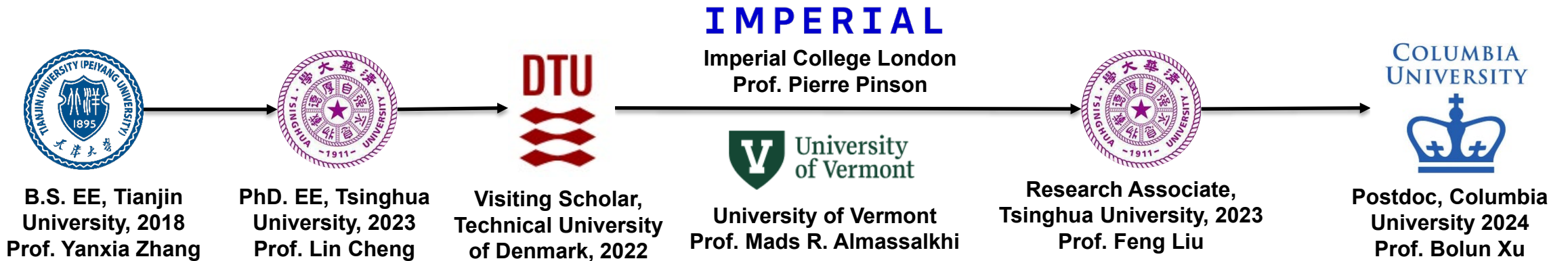
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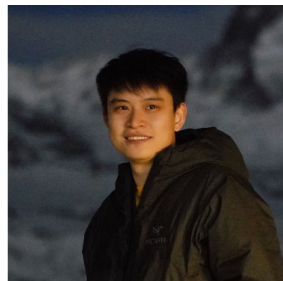
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Research Interest: Data-Driven Modeling, Optimization under Uncertainty and Market Design for Power System with Generalized Energy Storage Resources



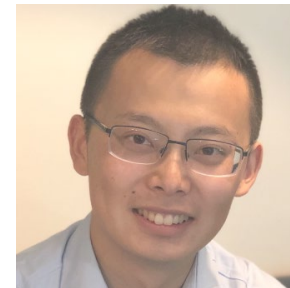
Group Member:



Ningkun (Nik) Zheng



Neal Ma



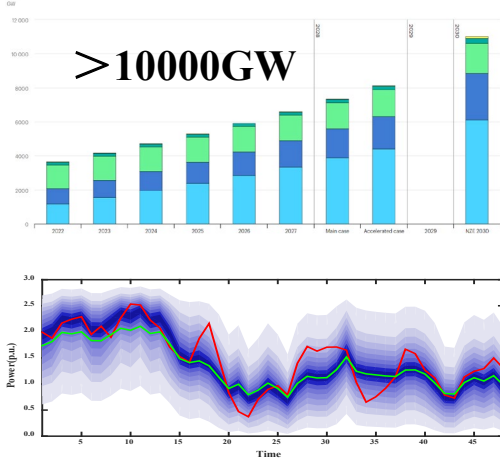
Bolun Xu

Group Homepage:

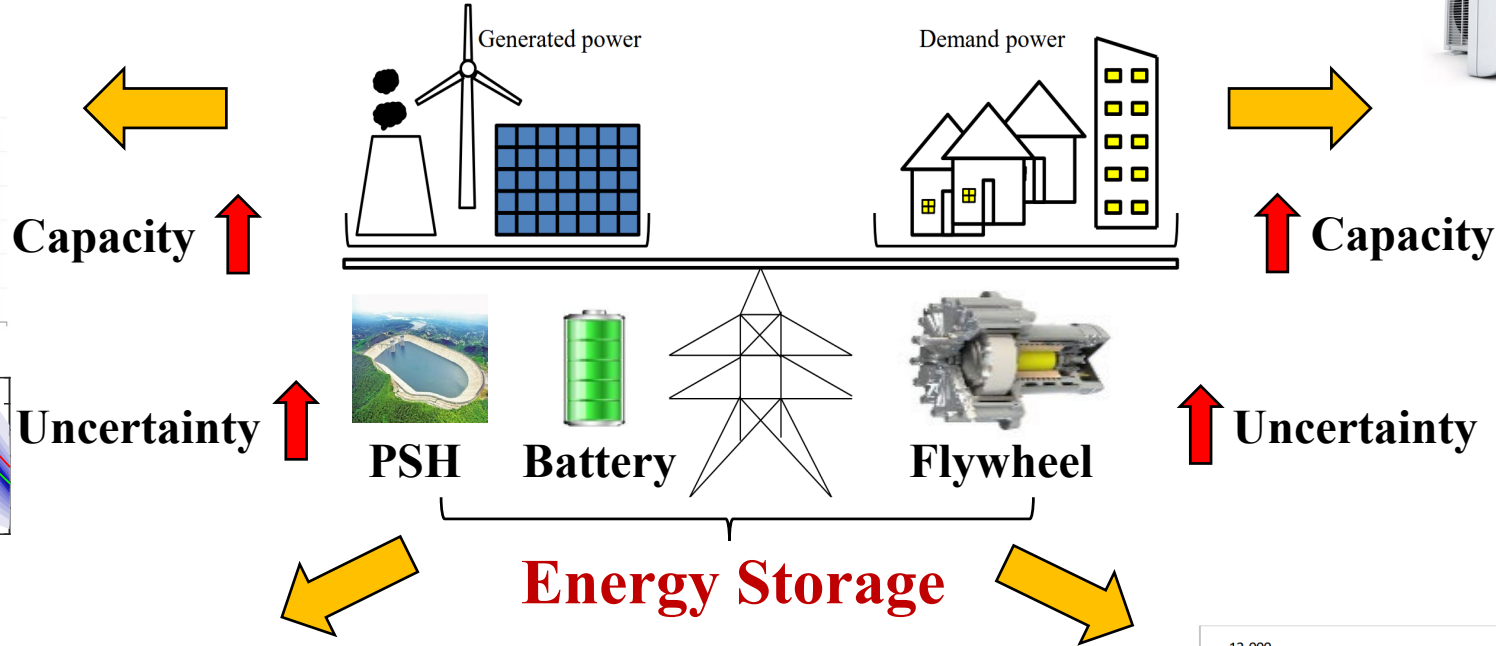
<https://bolunxu.github.io/>

Background—Role of Energy Storage

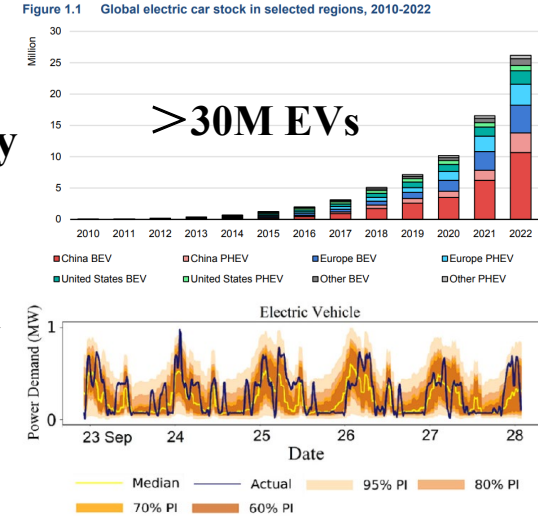
Renewables



Power Supply-Demand Balance



Flexible Loads

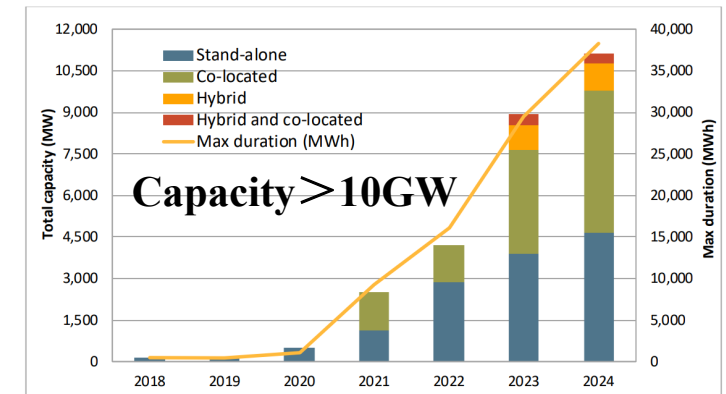


Multiple Services

- Energy Shifting (Price Arbitrage)
- Frequency Regulation
- Capacity Market...

?

Pricing



Background—Pricing of Energy Storage



Cost Recovery



Conventional Generator

Market Design

VS



Energy Storage



Small Storage

Lose Money
Without Intelligence

Bid:

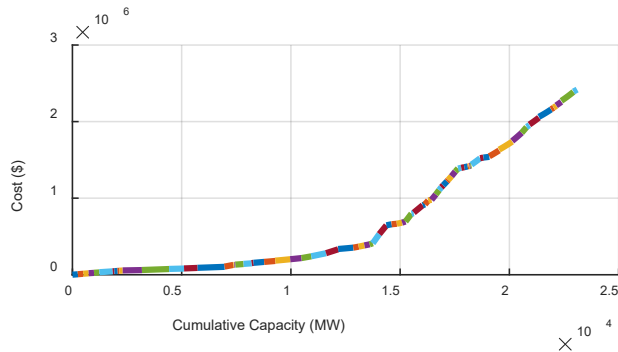
Marginal Cost

Opportunity Cost

Bid Generation:

Fuel Cost Curve

Price Prediction based
on Private Model



$$Q_{t-1}(e_{t-1} | \hat{\lambda}_t) := \max_{p_t, b_t} \hat{\lambda}_t (p_t - b_t) - c p_t + V_t(e_t)$$

$$V_{t-1}(e_{t-1}) := \mathbb{E}_{\hat{\lambda}_t} \left[Q_{t-1}(e_{t-1} | \hat{\lambda}_t) \right] \quad \text{Profit Maximization}$$

Market Clearing:

Marginal Energy Price

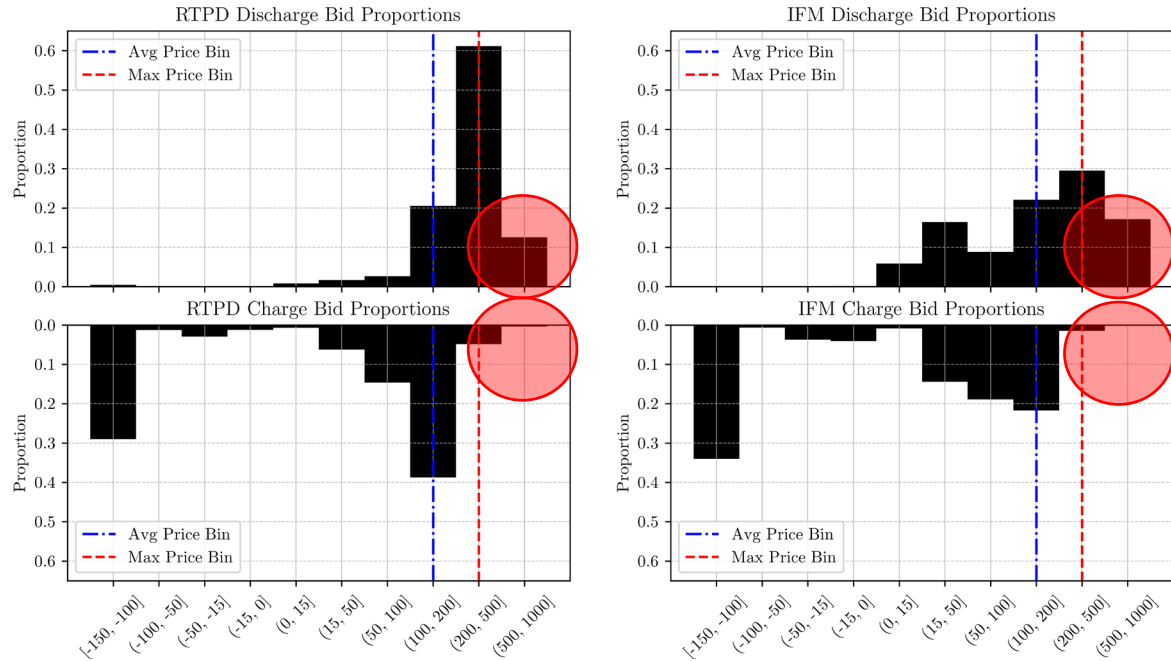
No Default Bid or Price

Background—Pricing of Energy Storage



Large Storage: Capacity Withholding (Extreme High Bid: \$500-1000/MWh!)

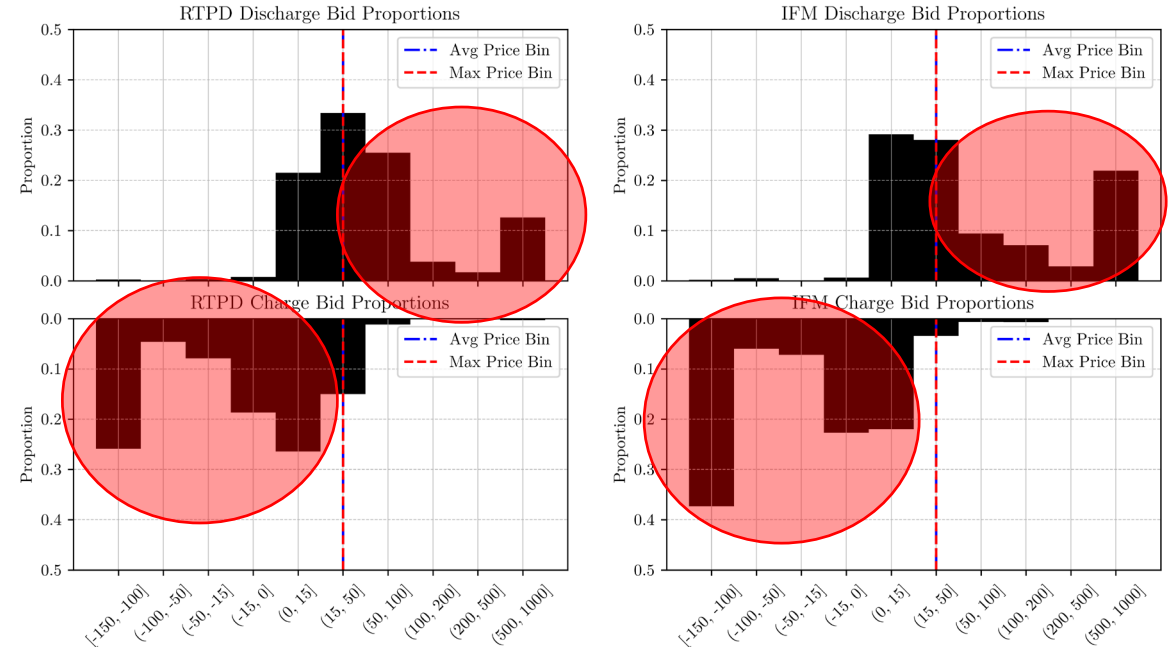
RTPD and IFM Charge/Discharge Proportions for January 16, 2024



Normal Price Day

Default Bid?

RTPD and IFM Charge/Discharge Proportions for May 15, 2024



Price Spike Day

<https://www.caiso.com/library/daily-energy-storage-reports>

Background—Pricing of Uncertainty

Deterministic Formulation

$$\min \sum_t [\sum_i G_i(g_{i,t}) + \sum_s M_s(p_{s,t})] \quad (1a)$$

$$[\lambda_t]: \sum_i g_{i,t} + \sum_s p_{s,t} - \sum_s b_{s,t} = D_t, \forall t \in \mathcal{T} \quad (1b) \quad \text{Energy Price}$$

$$[\theta_{s,t}]: e_{s,t+1} - e_{s,t} = -p_{s,t}/\eta_s + b_{s,t}\eta_s, \forall t \in \mathcal{T}, \forall s \in \mathcal{S} \quad (1c)$$

$$[\pi_t]: \sum_i \varphi_{i,t} + \sum_s \psi_{s,t} \geq d_t, \forall t \in \mathcal{T} \quad (1d) \quad \text{Reserve Price}$$

$$[\underline{\nu}_{i,t}, \bar{\nu}_{i,t}]: \underline{G}_i \leq g_{i,t} \leq \bar{G}_i - \varphi_{i,t}, \forall t \in \mathcal{T}, \forall i \in \mathcal{G} \quad (1e)$$

$$[\underline{k}_{i,t}, \bar{k}_{i,t}]: -RD_i \leq g_{i,t} - g_{i,t-1} \leq RU_i, \forall t \in \mathcal{T}, \forall i \in \mathcal{G} \quad (1f)$$

$$[\underline{\alpha}_{s,t}, \bar{\alpha}_{s,t}]: 0 \leq b_{s,t} \leq \bar{P}_s, \forall t \in \mathcal{T}, \forall s \in \mathcal{S} \quad (1g)$$

$$[\underline{\beta}_{s,t}, \bar{\beta}_{s,t}]: 0 \leq p_{s,t} \leq \bar{P}_s - \psi_{s,t}, \forall t \in \mathcal{T}, \forall s \in \mathcal{S} \quad (1h)$$

$$[\underline{L}_{s,t}, \bar{L}_{s,t}]: (\psi_{s,t} + p_{s,t})/\eta_s + \underline{E}_s \leq e_{s,t}, \quad (1i)$$

$$e_{s,t} \leq \bar{E}_s - b_{s,t}\eta_s, \forall t \in \mathcal{T}, \forall s \in \mathcal{S}$$

- Price and Dispatch Decoupled with Uncertainty
- Reserve Price ≈ 0 , Lose Profit for Reserve Provision
- ✓ Incorporate Uncertainty into Pricing and Dispatch

Probabilistic Formulation

□ Stochastic Optimization

- × Revenue adequacy and cost recovery for each scenario
- × Scenario-Based Price, No Default Price
- ✓ Market Analysis × Market Clearing

□ Robust Optimization

- ✓ Conservative

□ Chance-Constrained Optimization

- ✓ Tractable
- ✓ Scalable
- ✓ Control the Risk

$$\mathbb{P}\left(a_i(\mathbf{x})^T \boldsymbol{\xi}(\mathbf{x}) \leq b_i(\mathbf{x})\right) \geq 1 - \epsilon$$

[1] J. Kazempour, P. Pinson, and B. F. Hobbs, "A stochastic market design with revenue adequacy and cost recovery by scenario: Benefits and costs," *IEEE Transactions on Power Systems*, vol. 33, no. 4, pp.3531–3545, 2018.

[2] Y. Dvorkin, "A chance-constrained stochastic electricity market," *IEEE Transactions on Power Systems*, vol. 35, no. 4, pp. 2993–3003, 2019.

Problem Formulation & Preliminary

Two-Stage Chance-Constrained Economic Dispatch

$$\min \mathbb{E} \sum_t [\sum_i G_i(g_{i,t} + \varphi_{i,t} \mathbf{d}_t) + \sum_s M_s(p_{s,t} + \psi_{s,t} \mathbf{d}_t)] \quad (1a)$$

$$[\lambda_t]: \sum_i g_{i,t} + \sum_s p_{s,t} - \sum_s b_{s,t} = D_t, \forall t \quad (1b)$$

$$[\theta_{s,t}]: e_{s,t+1} - e_{s,t} = -p_{s,t}/\eta_s + b_{s,t}\eta_s, \forall t, \forall s \quad (1c)$$

$$[\pi_t]: \sum_i \varphi_{i,t} + \sum_s \psi_{s,t} = 1, \forall t \quad (1d)$$

$$0 \leq \varphi_{i,t}, \psi_{s,t} \leq 1, \forall t, \forall i, \forall s \quad (1e)$$

$$[\underline{\nu}_{i,t}, \bar{\nu}_{i,t}]: \mathbb{P}(G_i \leq g_{i,t} + \varphi_{i,t} \mathbf{d}_t \leq \bar{G}_i) \geq 1 - \epsilon, \forall t, \forall i \quad (1f)$$

$$[\underline{\kappa}_{i,t}, \bar{\kappa}_{i,t}]: -RD_i \leq g_{i,t} - g_{i,t-1} \leq RU_i, \forall t, \forall i \quad (1g)$$

$$[\underline{\alpha}_{s,t}, \bar{\alpha}_{s,t}]: 0 \leq b_{s,t}, \mathbb{P}(b_{s,t} - \psi_{s,t} \mathbf{d}_t \leq \bar{P}_s) \geq 1 - \epsilon, \forall t, \forall s \quad (1h)$$

$$[\underline{\beta}_{s,t}, \bar{\beta}_{s,t}]: 0 \leq p_{s,t}, \mathbb{P}(p_{s,t} + \psi_{s,t} \mathbf{d}_t \leq \bar{P}_s) \geq 1 - \epsilon, \forall t, \forall s \quad (1i)$$

$$[\underline{\iota}_{s,t}, \bar{\iota}_{s,t}]: \mathbb{P}((\psi_{s,t} \mathbf{d}_t + p_t)/\eta_s \leq e_{s,t} \leq \bar{E}_s - (b_{s,t} - \psi_{s,t} \mathbf{d}_t)\eta_s) \geq 1 - \epsilon, \forall t, \forall s \quad (1j)$$

$$b_{s,t} p_{s,t} = 0, \forall t, \forall s \quad (1k)$$

- First-Stage: Pre-dispatch $g_{i,t}, p_{s,t}, b_{s,t}, e_s, \varphi_{i,t}, \psi_{s,t}$
- Second-Stage: Re-dispatch $\varphi_{i,t} \mathbf{d}_t, \psi_{s,t} \mathbf{d}_t$

- ✓ Opportunity Pricing for Storage
- ✓ Default Bid and Benchmark the Market Power of Storage

Deterministic Reformulation

$$\mathbb{E}(G_i(g_{i,t} + \varphi_{i,t} \mathbf{d}_t)) = \sum_{j=0}^n C_j \sum_{k=0}^j \binom{j}{k} g_{i,t}^{j-k} \varphi_{i,t}^k \mathbb{E}(\mathbf{d}_t)^k$$

$$\mathbb{E}(M_s(p_{s,t} + \psi_{s,t} \mathbf{d}_t)) = M_s(p_{s,t} + \psi_{s,t} \mu_t)$$

$$\mathbb{E}(\mathbf{d}_t)^k = \sum_{0 \leq j \leq k} \binom{k}{j} \mu_t^{k-j} \sigma_t^j (j-1)!!$$

$$\underline{G}_i \leq g_{i,t} + \varphi_{i,t} \hat{\mathbf{d}}_t, g_{i,t} + \varphi_{i,t} \tilde{\mathbf{d}}_t \leq \bar{G}_i$$

$$b_{s,t} - \psi_{s,t} \hat{\mathbf{d}}_t \leq \bar{P}_s, p_{s,t} + \psi_{s,t} \tilde{\mathbf{d}}_t \leq \bar{P}_s$$

$$(\psi_{s,t} \tilde{\mathbf{d}}_t + p_{s,t})/\eta_s \leq e_{s,t}, e_{s,t} \leq \bar{E}_s - (b_{s,t} - \psi_{s,t} \hat{\mathbf{d}}_t)\eta_s$$

$$\hat{\mathbf{d}}_t = \mu_t - F^{-1}(1 - \epsilon)\sigma_t, \tilde{\mathbf{d}}_t = \mu_t + F^{-1}(1 - \epsilon)\sigma_t$$

Expectation

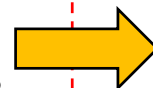


Energy Price

Storage Price

Reserve Cost

Chance-Constraints



Complementary Constraints



- Solve MILP with Binary Variables
- Substitute Binary Variables with Solution
- Resolve LP

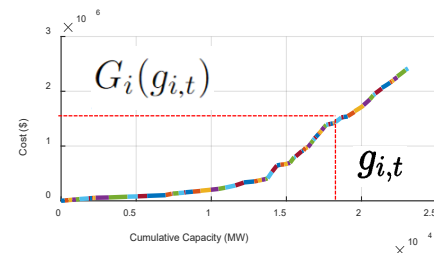
Robust Competitive Equilibrium!

Theoretical Analysis

- **Proposition 1: Physical Cost Origins:** (1) Marginal Cost of Cleared Generator & Storage (2) Storage Efficiency

$$\theta_{s,t} = H_i(g_{i,t} + \varphi_{i,t} \mathbf{d}_t) / \eta_s \quad \text{Charging} \quad \partial \mathbb{E} G_i(g_{i,t} + \varphi_{i,t} \mathbf{d}_t) / \partial g_{i,t} = H_i(g_{i,t} + \varphi_{i,t} \mathbf{d}_t)$$

$$\theta_{s,t} = \eta_s (H_i(g_{i,t} + \varphi_{i,t} \mathbf{d}_t) - M_s) \quad \text{Discharging}$$

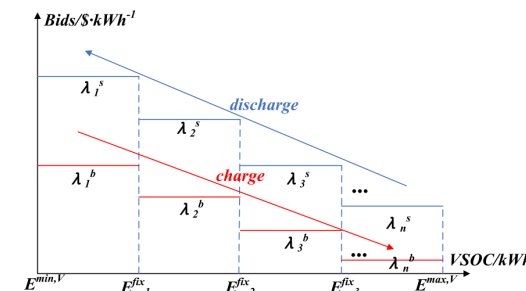


- **Proposition 2: Convex Opportunity Price:** Monotonically Decreases with Storage SoC

$$\partial \theta_{s,t} / \partial e_{s,t} \leq 0 \quad \text{Quadratic/Super-Quadratic Generation Cost Function} \quad \text{SoC-Dependent Bid}^{[1]}$$

- **Proposition 3: Uncertainty-aware Price:** Monotonically Increases with Uncertainty

$$\partial \theta_{s,t} / \partial \sigma_t \geq 0 \quad \text{Quadratic/Super-Quadratic Generation Cost Function} \quad \mathbb{E}(\theta_{s,t}(\mathbf{d}_t)) > \theta_{s,t}(\mathbb{E}(\mathbf{d}_t))$$

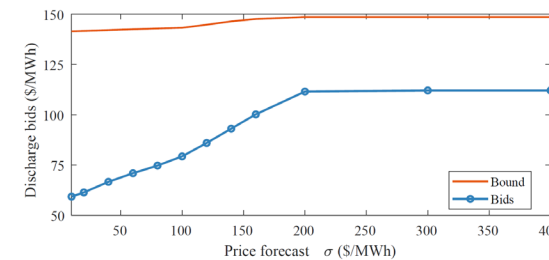


- **Theorem: Constrained and Bounded Price**

$$\theta_{s,t-1} = \frac{\eta_s}{\tilde{d}_t} (\theta_{s,t} \eta_s \tilde{d}_t + \lambda_t (\frac{\tilde{d}_t}{\eta_s^2} - \tilde{d}_t) + \pi_t - M_s \mu_t) \quad \text{Charging} \quad \text{Linear Relationship with Energy and Reserve Price}$$

$$\theta_{s,t-1} = \frac{1}{\eta_s \tilde{d}_t} (\theta_{s,t} \tilde{d}_t + \lambda_t (\eta_s^2 \tilde{d}_t - \tilde{d}_t) + \pi_t + M_s (\tilde{d}_t - \eta_s^2 \tilde{d}_t - \mu_t)) \quad \text{Discharging}$$

Uncertainty-Aware and Bounded Bid^[2]



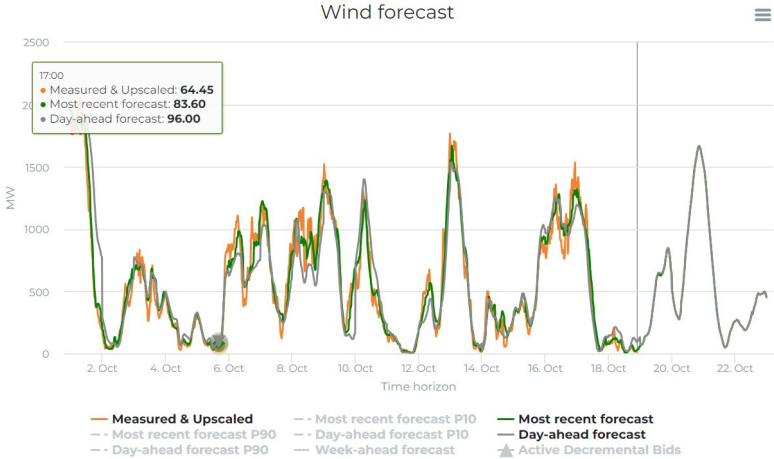
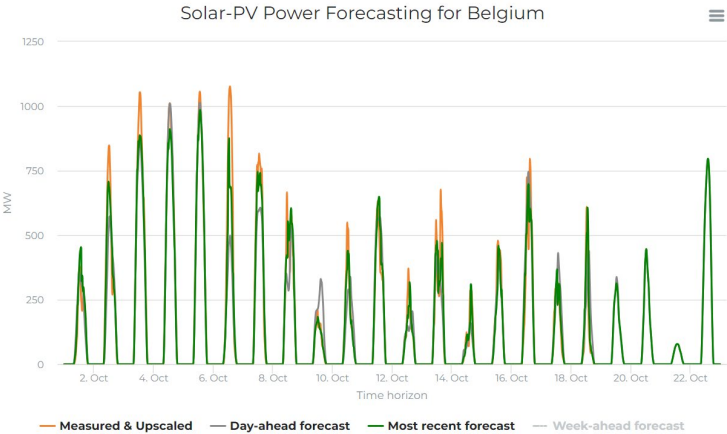
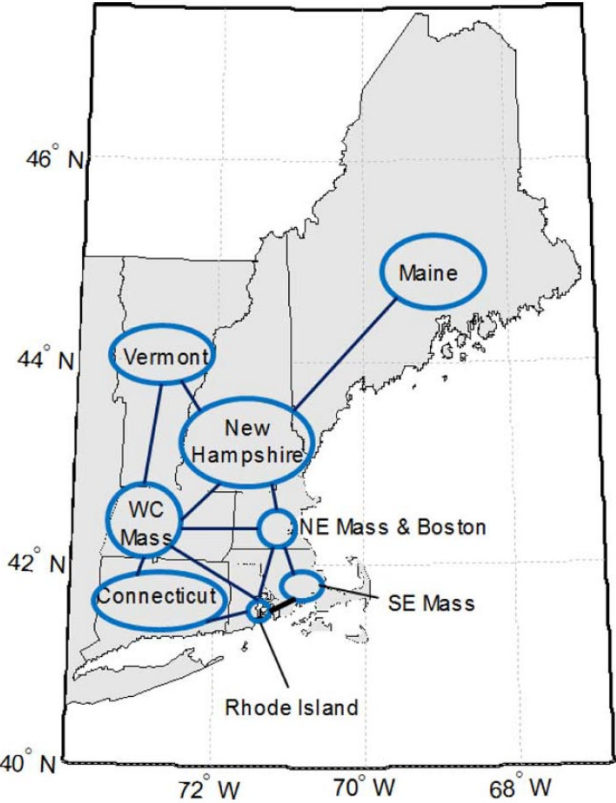
[1] N. Zheng, X. Qin, D. Wu, G. Murtaugh and B. Xu, "Energy Storage State-of-Charge Market Model," *IEEE Transactions on Energy Markets, Policy and Regulation*, vol. 1, no. 1, pp. 11-22, March 2023.

[2] Q., Xin, I. Lestas, and B. Xu. "Economic Capacity Withholding Bounds of Competitive Energy Storage Bidders." *arXiv preprint arXiv:2403.05705* (2024).

Case Study—Test System

- ISO-NE 8-Zone Test System

- **76 Generators: 23.1 GW**
- **Load: 13 GW**
- **Renewables: Wind and Solar^[1], 10%-90% of Total Generation Capacity**
- **Uncertainty: Elia Historical Data^[2]**
- **Multiple Storages: 10%-60% of Total Generation Capacity; 4-8-12 hr duration; 0.8, 0.85, 0.9, 0.95 One-Way Efficiency**
- **Coding^[3]: MatLab and solved by Gurobi 11.0 solver, Intel Corei9-13900HX @ 2.30GHz with RAM 16 GB.**



[1] D. Krishnamurthy, W. Li, and L. Tesfatsion, “An 8-zone test system based on iso new england data: Development and application,” *IEEE Transactions on Power Systems*, vol. 31, no. 1, pp. 234–246, 2015

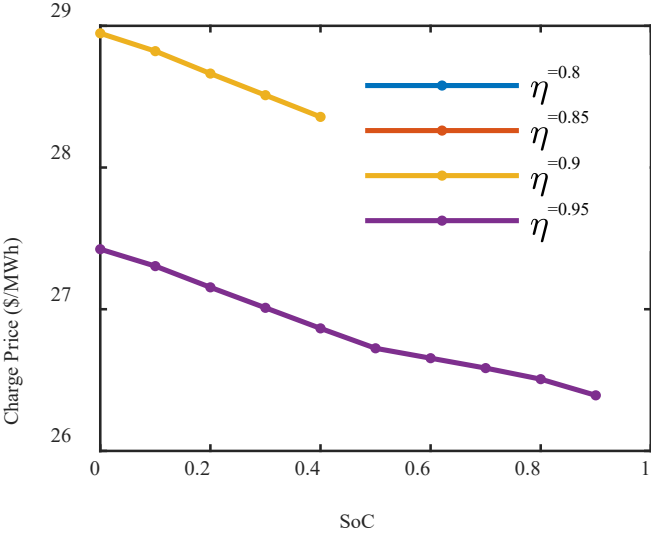
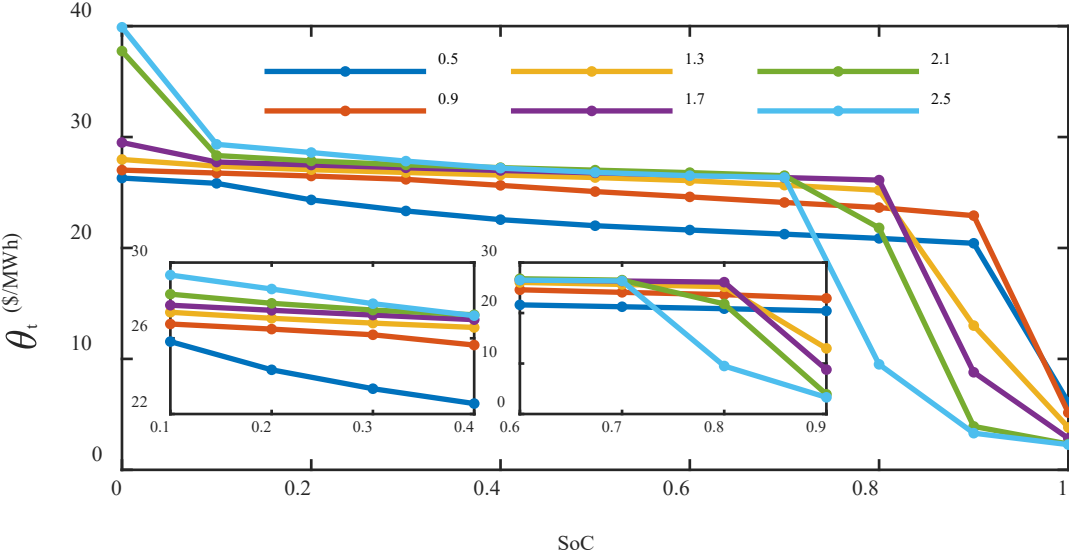
[2] Elia, “Forecast error data from elia,” 2024. [Online]. Available: <https://www.elia.be/en/grid-data>.

[3] Code and Data: https://github.com/thuqining/Storage_Pricing_for_Social_Welfare_Maximization

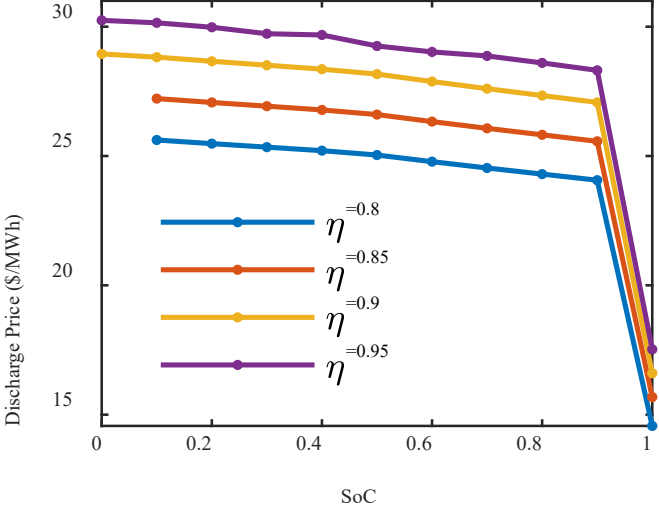
Case Study—Simulation Results

- **Convex** Opportunity Price Function $\theta(\text{SoC})$
- **Monotonically Decreasing**

- **Higher Efficiency: Higher Discharge Price, Lower Charge Price**



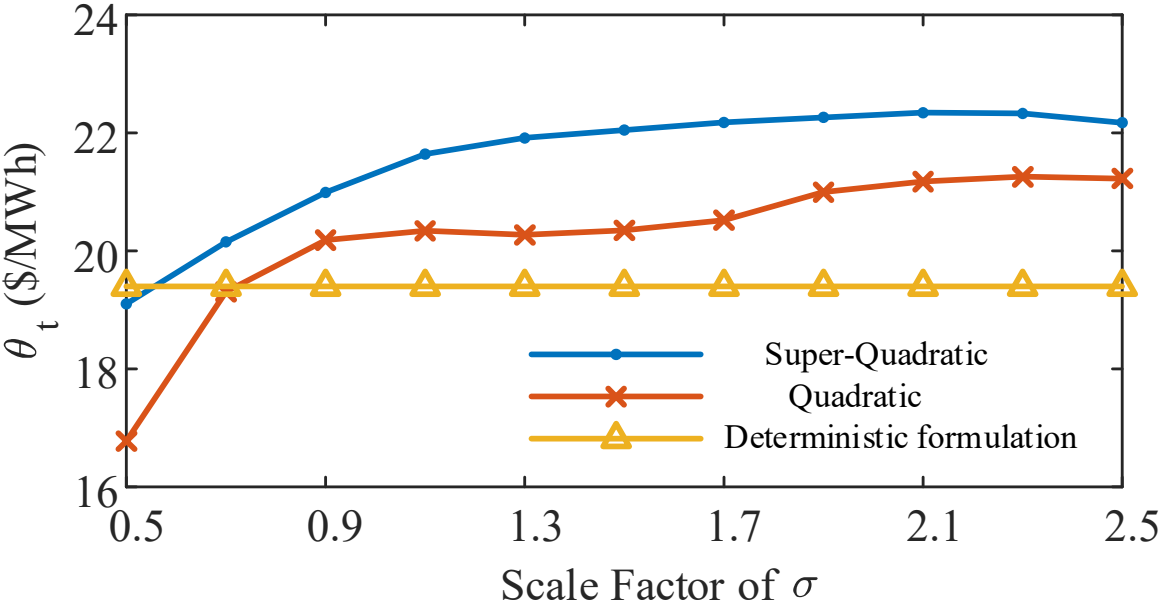
(a)



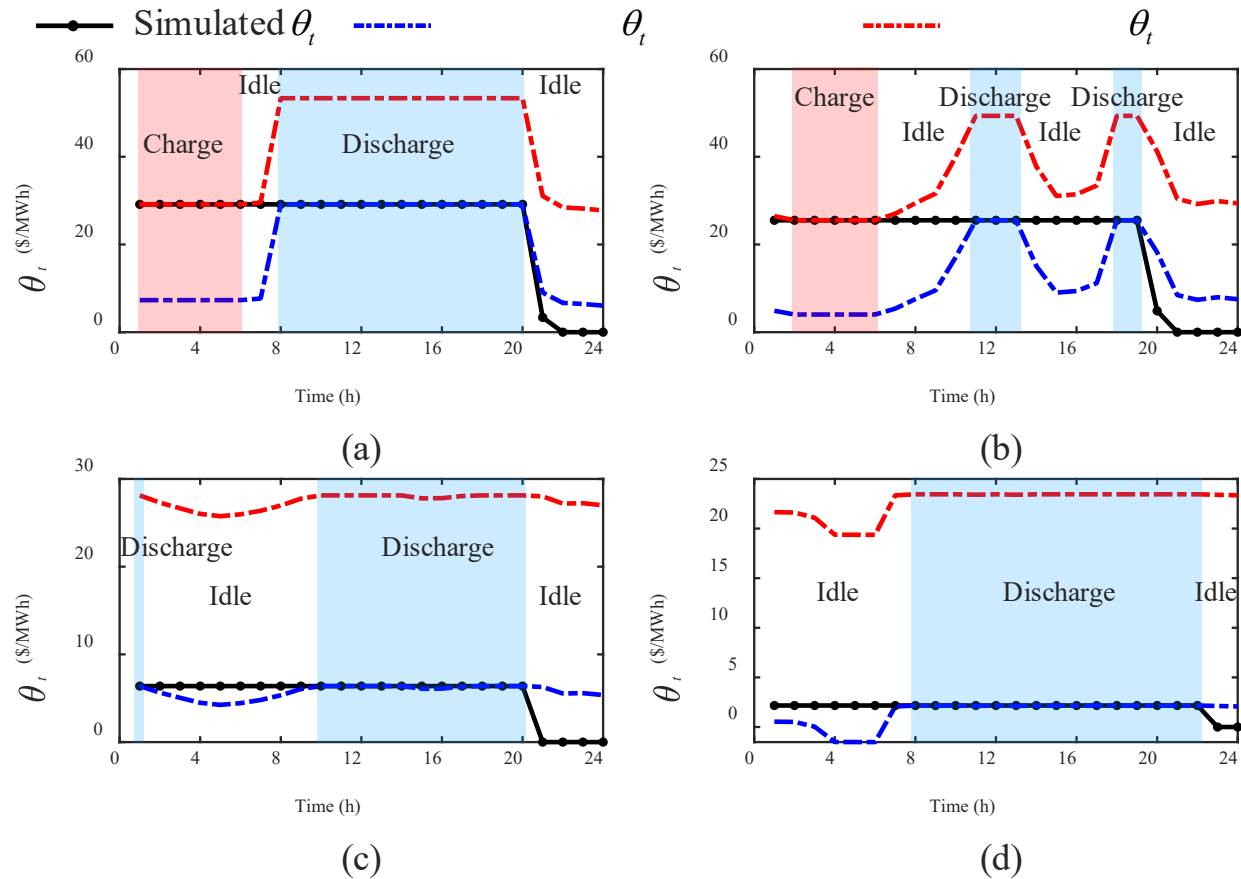
(b)

Case Study—Simulation Results

- **Uncertainty-Aware Price Function $\theta(\sigma)$**
- **Proposed Formulation: Monotonically Increasing**
- **Deterministic Formulation: Fixed**



- **Constrained and bounded Price**
- **Anticipated Default Bid**



Case Study—Comparative Results

- Benchmark Comparison
- Storage Profit Maximization^[1,2]

$$V_{t-1}(e_{t-1}) = \max_{p_t, b_t} \lambda_t(p_t - b_t) - Mp_t + V_t(e_t)$$

$$e_{t+1} - e_t = -p_t/\eta + b_t\eta$$

$$0 \leq b_t \leq \bar{P}, 0 \leq p_t \leq \bar{P}$$

$$0 \leq e_t \leq \bar{E}$$

$$p_t = 0, \lambda_t < 0$$

$$O_t(p_t) = \frac{\partial(Mp_t + V_t(e_t))}{\partial p_t} = M + \frac{1}{\eta}v_t(e_{t-1} - p_t/\eta)$$

$$B_t(b_t) = \frac{\partial(V_t(e_t))}{\partial b_t} = \eta v_t(e_{t-1} + b_t\eta)$$

$$\min \sum_i \sum_t G(g_{i,t}) + O_t(p_t) - B_t(b_t)$$

Opportunity Value Function

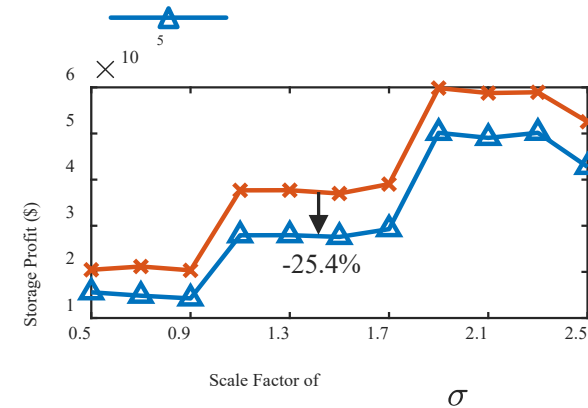


Storage Bid

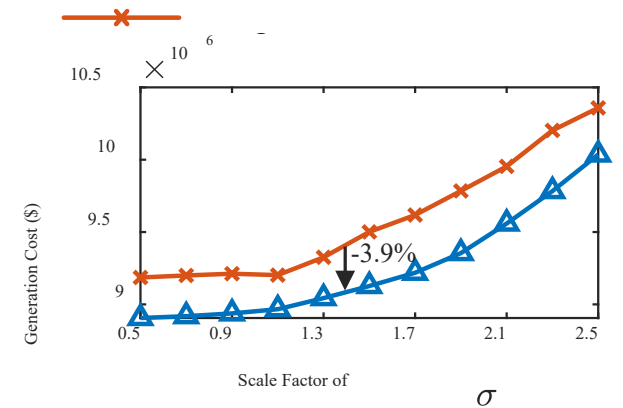


Market Clearing

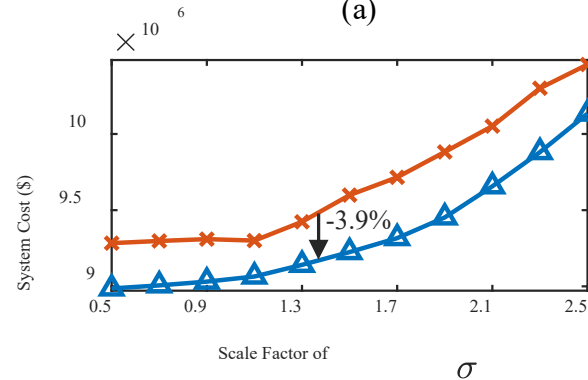
- **Enhance social welfare**, reduce conventional generator production and consumer payment; **Sacrifice** storage margins
- **Electricity Payment Decrease by 17%**
- **Storage Profit Reduces by 0.5% (based on Electricity Payment)**



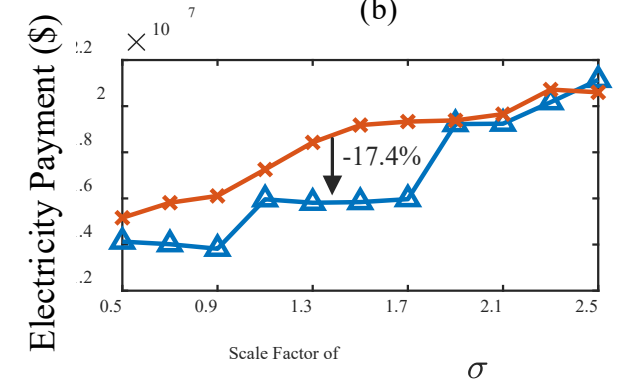
(a)



(b)



(c)



(d)

[1] X. Qin, I. Lestas, and B. Xu, "Economic capacity withholding bounds of competitive energy storage bidders," arXiv preprint arXiv:2403.05705, 2024

[2] N. Zheng, X. Qin, D. Wu, G. Murtaugh and B. Xu, "Energy Storage State-of-Charge Market Model," IEEE Transactions on Energy Markets, Policy and Regulation, vol. 1, no. 1, pp. 11-22, March 2023.

Case Study—Comparative Results

● Benefit **Scales Up** with Increased Renewable & Storage Integration

▪ Sensitivity Analysis of Storage Capacity

✓ 4-hr storage:
20% storage capacity (-18%)



60% storage capacity (-27%)

✓ 20% storage capacity:
4-hr storage (-18%)



12-hr storage (-23%)

▪ Sensitivity Analysis of Renewable Capacity

✓ 30% renewable capacity (-18%)



✓ 50% renewable capacity (-21%)



✓ 70% renewable capacity (-22%)

Case Study—Comparative Results

● Uncertainty Realization & Risk-Aversion

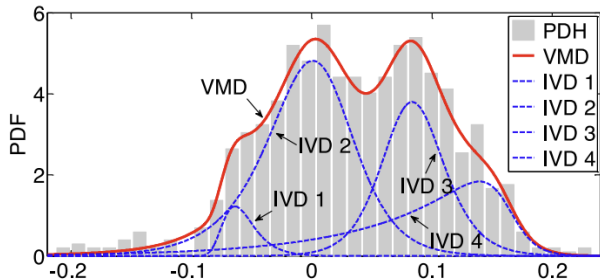
(1) Normal Distribution

$$F^{-1}(1 - \epsilon) = \Phi^{-1}(1 - \epsilon)$$

(2) Distributionally Robust

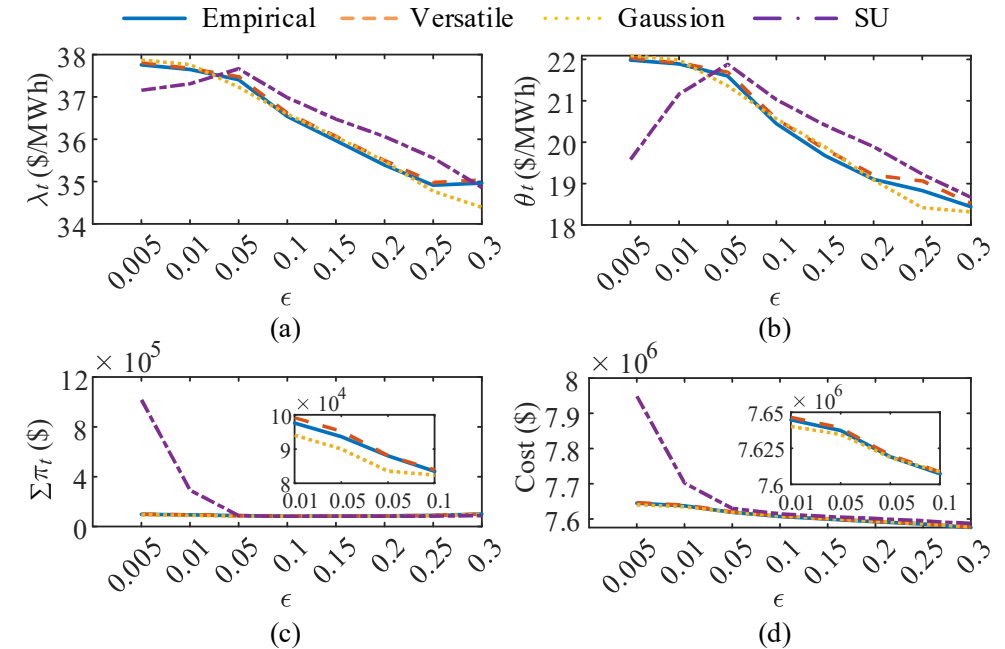
Type & Shape	$F^{-1}(1-\epsilon)$	ϵ
No Distribution Assumption (NA)	$\sqrt{(1-\epsilon)}/\epsilon$	$0 < \epsilon \leq 1$
Symmetric Distribution (S)	$\sqrt{1/2\epsilon}$	$0 < \epsilon \leq 1/2$
	0	$1/2 < \epsilon \leq 1$
Unimodal Distribution (U)	$\sqrt{(4-9\epsilon)}/9\epsilon$	$0 < \epsilon \leq 1/6$
	$\sqrt{(3-3\epsilon)/(1+3\epsilon)}$	$1/6 < \epsilon \leq 1$

(3) Versatile Distribution^[1]



[1] Z.-S. Zhang, Y.-Z. Sun, D. W. Gao et al., “A versatile probability distribution model for wind power forecast errors and its application in economic dispatch,” IEEE Transactions on Power Systems, vol. 28, no. 3, pp. 3114–3125, 2013.

- **More Risk-Aversion** → **Higher Cost & Prices**
- **Versatile Distribution** Best Fits the Netload Forecast Error



Distributions	Energy Price	Storage Price
Versatile	0.08	0.12
Gaussian	0.23	0.20
SU	0.48	1.00

Compared with Empirical Result
Lower RMSE: Better Fitting

Case Study—Comparative Results

● Computational Efficiency

Generator Number	1	76	76	76	76	76	76	76	76	76
Storage Number	1	1	5	10	50	100	500	1000	5000	10000
Time1 (s)	0.08	0.42	0.57	0.78	5.63	11.86	370.00	424.45	>3000	>3000
Time2 (s)	0.04	0.35	0.46	0.59	0.74	1.16	2.85	5.32	33.34	72.60

 **Scalable!**

Time1: Complementary Constraints to Prevent Storage from Simultaneous Charging and Discharging

Time2: Relaxation of Complementary Constraints^[1]

$$E^r(\mathbf{P}_c^r, \mathbf{P}_d^r) = \mathbf{1}_T E_0 + \eta_c \mathbf{A} \mathbf{P}_c^r - \frac{1}{\eta_d} \mathbf{A} \mathbf{P}_d^r$$

$$\mathbf{0} \leq \mathbf{1}_T E_0 + \eta_c \mathbf{A} \mathbf{P}_c - \frac{1}{\eta_d} \mathbf{A} \mathbf{P}_d$$

$$\mathbf{E}_{\max} \geq \mathbf{1}_T E_0 + \eta \mathbf{A} (\mathbf{P}_c - \mathbf{P}_d)$$

$$\mathbf{0} \leq \mathbf{P}_c \leq \mathbf{1}_T P_{\max}$$

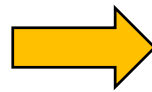
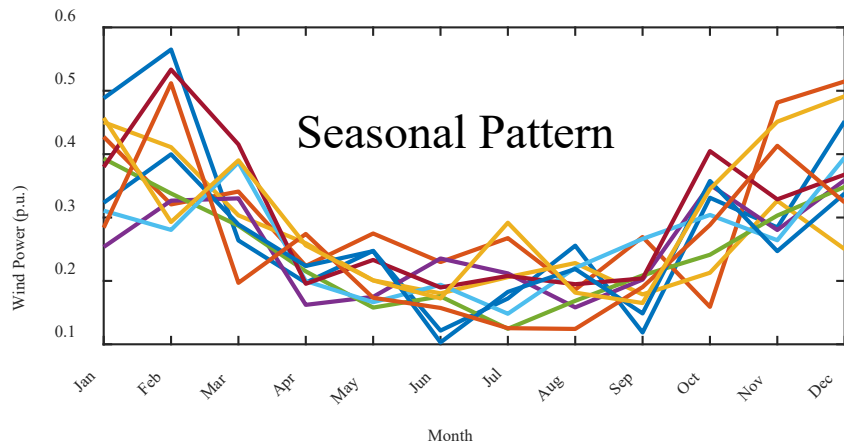
$$\mathbf{0} \leq \mathbf{P}_d \leq \mathbf{1}_T P_{\max}$$

$$\mathbf{P}_c + \mathbf{P}_d \leq \mathbf{1}_T P_{\max}$$

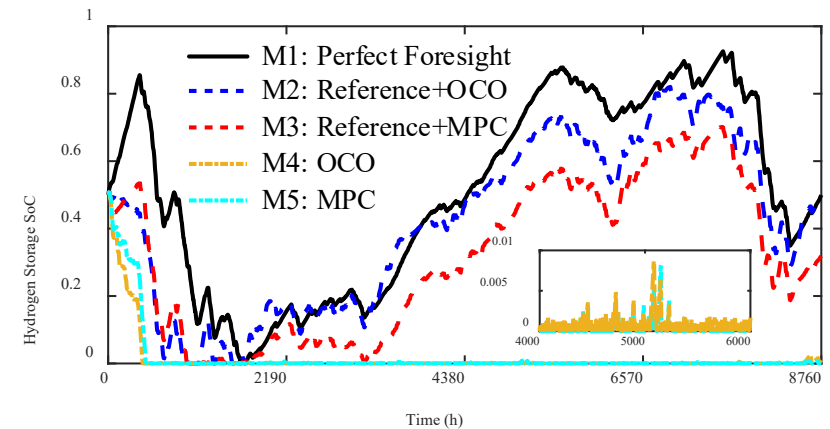
[1] N. Nazir and M. Almassalkhi, “Guaranteeing a physically realizable battery dispatch without charge-discharge complementarity constraints,” *IEEE Transactions on Smart Grid*, vol. 14, no. 3, pp. 2473–2476, 2021.

Conclusion & Future Work

- ✓ Two-Stage **Chance-Constrained** Pricing Framework—Default Bid, Benchmark Market Power of Storage, Integrated into Current Economic Dispatch
- ✓ Theoretical Analysis: **Convex, Uncertainty-Aware, Bounded and Anticipated**
- ✓ Simulations: Significantly **Enhance Social Welfare!**—CASIO^[1]
- ? Long-Duration Storage—Track the Seasonal Pattern^[3]



Annual Dispatch



[1] <https://www.caiso.com/notices/new-initiative-storage-bid-cost-recovery-and-default-energy-bids-enhancements-workshop-call-on-7-8-24>

[2] N. Qi, N. Zheng, and B. Xu. "Chance-constrained energy storage pricing for social welfare maximization." arXiv preprint arXiv:2407.07068.

[3] N. Qi, K. Huang, Z. Fan, and B. Xu. "Long-term energy management for microgrid with hybrid hydrogen-battery energy storage: A prediction-free coordinated optimization framework." Applied Energy 377 (2025): 124485.

Thank You!

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Theoretical Analysis—Deterministic Framework

Deterministic Formulation

$$\min \sum_t [\sum_i G_i(g_{i,t}) + \sum_s M_s(p_{s,t})] \quad (1a)$$

$$[\lambda_t]: \sum_i g_{i,t} + \sum_s p_{s,t} - \sum_s b_{s,t} = D_t, \forall t \in \mathcal{T} \quad (1b)$$

$$[\theta_{s,t}]: e_{s,t+1} - e_{s,t} = -p_{s,t}/\eta_s + b_{s,t}\eta_s, \forall t \in \mathcal{T}, \forall s \in \mathcal{S} \quad (1c)$$

$$[\pi_t]: \sum_i \varphi_{i,t} + \sum_s \psi_{s,t} \geq d_t, \forall t \in \mathcal{T} \quad (1d)$$

$$[\underline{\nu}_{i,t}, \bar{\nu}_{i,t}]: \underline{G}_i \leq g_{i,t} \leq \bar{G}_i - \varphi_{i,t}, \forall t \in \mathcal{T}, \forall i \in \mathcal{G} \quad (1e)$$

$$[\underline{\kappa}_{i,t}, \bar{\kappa}_{i,t}]: -RD_i \leq g_{i,t} - g_{i,t-1} \leq RU_i, \forall t \in \mathcal{T}, \forall i \in \mathcal{G} \quad (1f)$$

$$[\underline{\alpha}_{s,t}, \bar{\alpha}_{s,t}]: 0 \leq b_{s,t} \leq \bar{P}_s, \forall t \in \mathcal{T}, \forall s \in \mathcal{S} \quad (1g)$$

$$[\underline{\beta}_{s,t}, \bar{\beta}_{s,t}]: 0 \leq p_{s,t} \leq \bar{P}_s - \psi_{s,t}, \forall t \in \mathcal{T}, \forall s \in \mathcal{S} \quad (1h)$$

$$[\underline{L}_{s,t}, \bar{L}_{s,t}]: (\psi_{s,t} + p_{s,t})/\eta_s + \underline{E}_s \leq e_{s,t}, \quad (1i)$$

$$e_{s,t} \leq \bar{E}_s - b_{s,t}\eta_s, \forall t \in \mathcal{T}, \forall s \in \mathcal{S}$$

- **Proposition 1: Physical Cost Origins: (1) Marginal Cost of Cleared Generator & Storage (2) Storage Efficiency**

$$\theta_{s,t} = H_i(g_{i,t})/\eta_s \quad \text{Charging}$$

$$\theta_{s,t} = \eta_s(H_i(g_{i,t}) - M_s) \quad \text{Discharging}$$

- **Proposition 2: Convex Opportunity Price: Monotonically Decreases with Storage SoC**

$$\partial \theta_{s,t} / \partial e_{s,t} \leq 0. \quad \text{Quadratic/Super-Quadratic Generation Cost Function}$$

- **Theorem: Constrained and Bounded Price**

$$\theta_{s,t} \geq \eta_s(\lambda_t - \pi_t - M_s), \theta_{s,t-1} = \lambda_t/\eta_s + \pi_t\eta_s \quad \text{Charging}$$

$$\theta_{s,t} = \eta_s(\lambda_t - \pi_t - M_s), \theta_{s,t-1} \leq \lambda_t/\eta_s + \pi_t\eta_s \quad \text{Discharging}$$

× **Uncertainty-aware Price**

Chance-constrained optimization

Deterministic Reformulation

$$\mathbb{P}(a_i(\mathbf{x})^T \boldsymbol{\xi} \leq b_i(\mathbf{x})) \geq 1 - \epsilon$$

$$a_i(\mathbf{x})^T \boldsymbol{\mu} + b_i(\mathbf{x}) + F^{-1}(1 - \epsilon) \sqrt{a_i(\mathbf{x})^T \boldsymbol{\Sigma} a_i(\mathbf{x})} \leq 0$$

(1) Gaussian Distribution

$$F^{-1}(1 - \epsilon) = \Phi^{-1}(1 - \epsilon)$$

(2) Distributionally Robust^[1]

Type & Shape	$F^{-1}(1 - \epsilon)$	ϵ
No Distribution Assumption (NA)	$\sqrt{(1 - \epsilon)/\epsilon}$	$0 < \epsilon \leq 1$
Symmetric Distribution (S)	$\sqrt{1/2\epsilon}$	$0 < \epsilon \leq 1/2$
	0	$1/2 < \epsilon \leq 1$
Unimodal Distribution (U)	$\sqrt{(4 - 9\epsilon)/9\epsilon}$	$0 < \epsilon \leq 1/6$
	$\sqrt{(3 - 3\epsilon)/(1 + 3\epsilon)}$	$1/6 < \epsilon \leq 1$
Symmetric & Unimodal Distribution (SU)	$\sqrt{2/9\epsilon}$	$0 < \epsilon \leq 1/6$
	$\sqrt{3}(1 - 2\epsilon)$	$1/6 < \epsilon \leq 1/2$
	0	$1/2 < \epsilon \leq 1$

(3) Versatile Distribution^[2]

$$F^{-1}(1 - \epsilon | a, b, c) = c - \ln\left(\left(1 - \epsilon\right)^{-1/b} - 1\right)/a$$

(4) Data-Driven Distributionally Robust^[3]

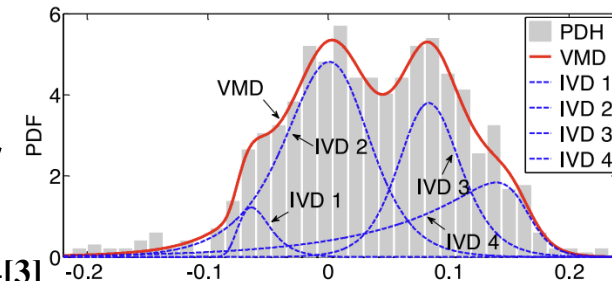
$$a_i(\mathbf{x})^T \boldsymbol{\mu}(\mathbf{x}) + F_x^{-1}(1 - \epsilon) \sqrt{a_i(\mathbf{x})^T \boldsymbol{\Sigma}(\mathbf{x}) a_i(\mathbf{x})} \leq b_i(\mathbf{x})$$

$$a_i(\mathbf{x})^T \boldsymbol{\mu}(\mathbf{x}) + \psi_K \mathbf{r}(\mathbf{x}) + \pi_K \bar{F}_x^{-1}(1 - \epsilon) \|\mathbf{y}\|_2 \leq b_i(\mathbf{x})$$

$$\sqrt{a_i(\mathbf{x})^T \boldsymbol{\Sigma} a_i(\mathbf{x})} \leq y_1, \sqrt{2\psi_K} \mathbf{r}(\mathbf{x}) \leq y_2$$

$$\psi_K = K^{(1/p - 1/2)}$$

$$\pi_K = \left(1 - \frac{4}{\epsilon} \exp\left(-\left(K^{1/p} - 2\right)^2 / 2\right)\right)^{-1/2}$$



[1] N. Qi, P. Pinson, M. R. Almassalkhi, et al., “Chance-constrained generic energy storage operations under decision-dependent uncertainty,” IEEE Transactions on Sustainable Energy, vol. 14, no. 4, pp. 2234–2248, 2023.

[2] Z.-S. Zhang, Y.-Z. Sun, D. W. Gao et al., “A versatile probability distribution model for wind power forecast errors and its application in economic dispatch,” IEEE Trans. on Power Systems, vol. 28, no. 3, pp. 3114–3125, 2013.

[3] N. Qi, P. Pinson, M. R. Almassalkhi, et al., “Capacity Credit Evaluation of Generalized Energy Storage Considering Endogenous Uncertainty,” IEEE Trans. on Power Systems (second review).