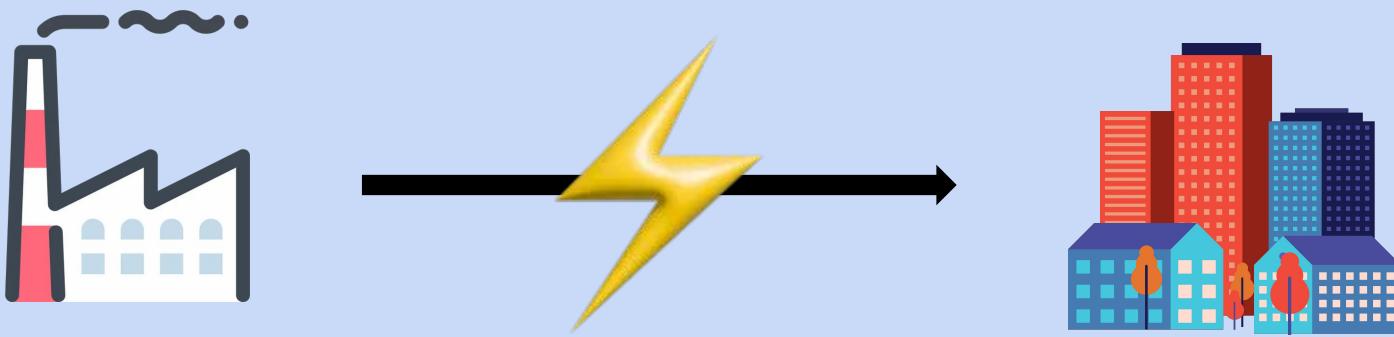
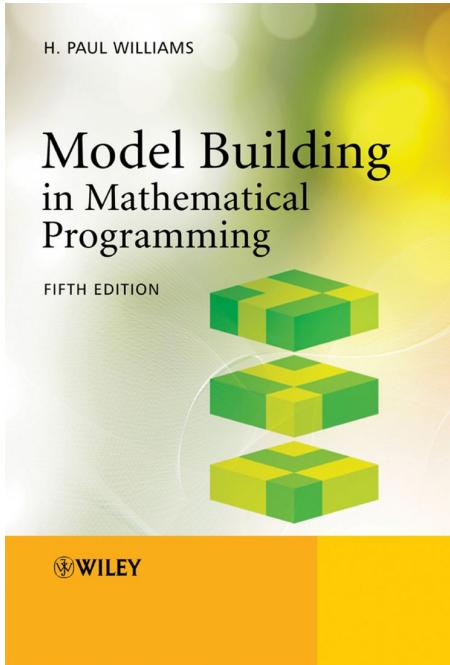


Stochastic Tariff Rates Optimization Model



Inaccurate Deterministic Problem



	Minimum level	Maximum level	Cost per hour at minimum	Cost per hour per megawatt above minimum	Cost
Type 1	850 MW	2000 MW	1000	2	2000
Type 2	1250 MW	1750 MW	2600	1.30	1000
Type 3	1500 MW	4000 MW	3000	3	500

12 p.m. to 6 a.m.	15 000 MW
6 a.m. to 9 a.m.	30 000 MW
9 a.m. to 3 p.m.	25 000 MW
3 p.m. to 6 p.m.	40 000 MW
6 p.m. to 12 p.m.	27 000 MW

Outline of the presentation

Stochastic Optimization Model

Dataset Generation

Scenarios Reduction

Results

Limits

Stochastic Optimization Model

Decision Variables

- $n_{i,t}$: Number of generating units of type i working in period t (integer, non-negative).
- $s_{i,t}$: Number of generators of type i started up in period t (integer, non-negative).
- $x_{i,t}^\xi$: Total output rate from generators of type i in period t under scenario ξ (non-negative).

Objective Function

Minimize the total cost:

$$\begin{aligned} & \min \quad \text{First-stage cost} + \text{Expected second-stage cost} \\ & \min \quad \sum_{i \in \mathcal{I}} \sum_{t \in \mathcal{T}} (E_{i,t} n_{i,t} + F_i s_{i,t}) + \sum_{\xi \in \Xi} \pi^\xi \sum_{i \in \mathcal{I}} \sum_{t \in \mathcal{T}} C_{i,t} (x_{i,t}^\xi - m_i n_{i,t}) \end{aligned}$$

Stochastic Optimization Model

Constraints

$$s_{i,t} \geq n_{i,t} - n_{i,t-1}, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}$$

$n_{i,0} = n_{i,T}$, (periodic boundary condition)

$$n_{i,t} \leq N_i, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}$$

Turn on reactors

$$\sum_{i \in \mathcal{I}} x_{i,t}^\xi \geq D_t^\xi, \quad \forall t \in \mathcal{T}, \xi \in \Xi$$

Number of reactors upper bound

$$\sum_{i \in \mathcal{I}} M_i n_{i,t} \geq 1.15 D_t^\xi, \quad \forall t \in \mathcal{T}, \xi \in \Xi$$

Flow meets the demand

$$m_i n_{i,t} \leq x_{i,t}^\xi \leq M_i n_{i,t}, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, \xi \in \Xi$$

Secure possibly higher demand

Flow can be produced

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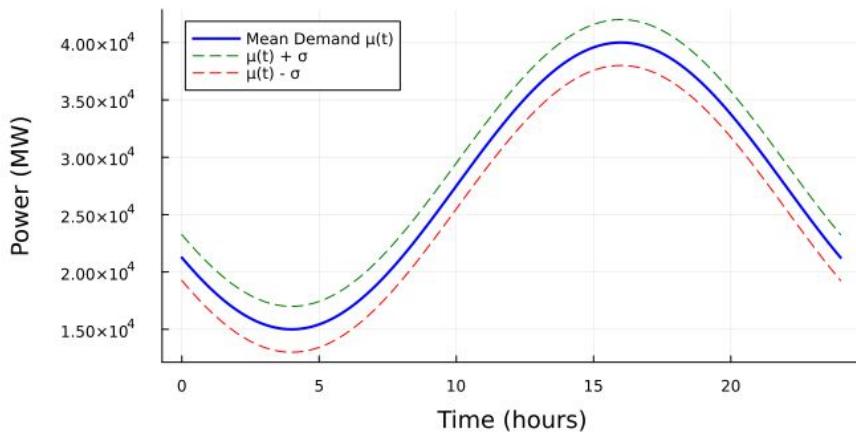
A''

Dataset Generation

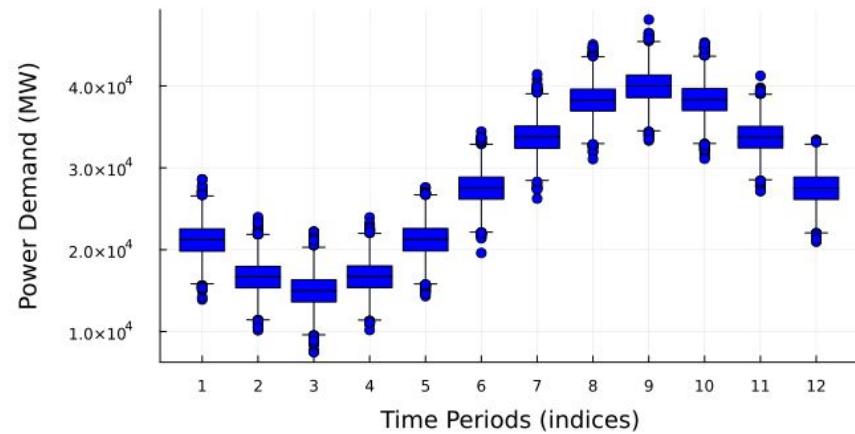
$$D_t \sim \mathcal{N}(\mu(t), \sigma^2)$$

$$\mu(t) = B + A \sin(\omega t + \varphi)$$

City Energy Consumption Over 24 Hours



Simulated Power Demand Through Day Time
Over 500 weeks



A''

Dataset Generation

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Outline of the presentation

Stochastic Optimization Model

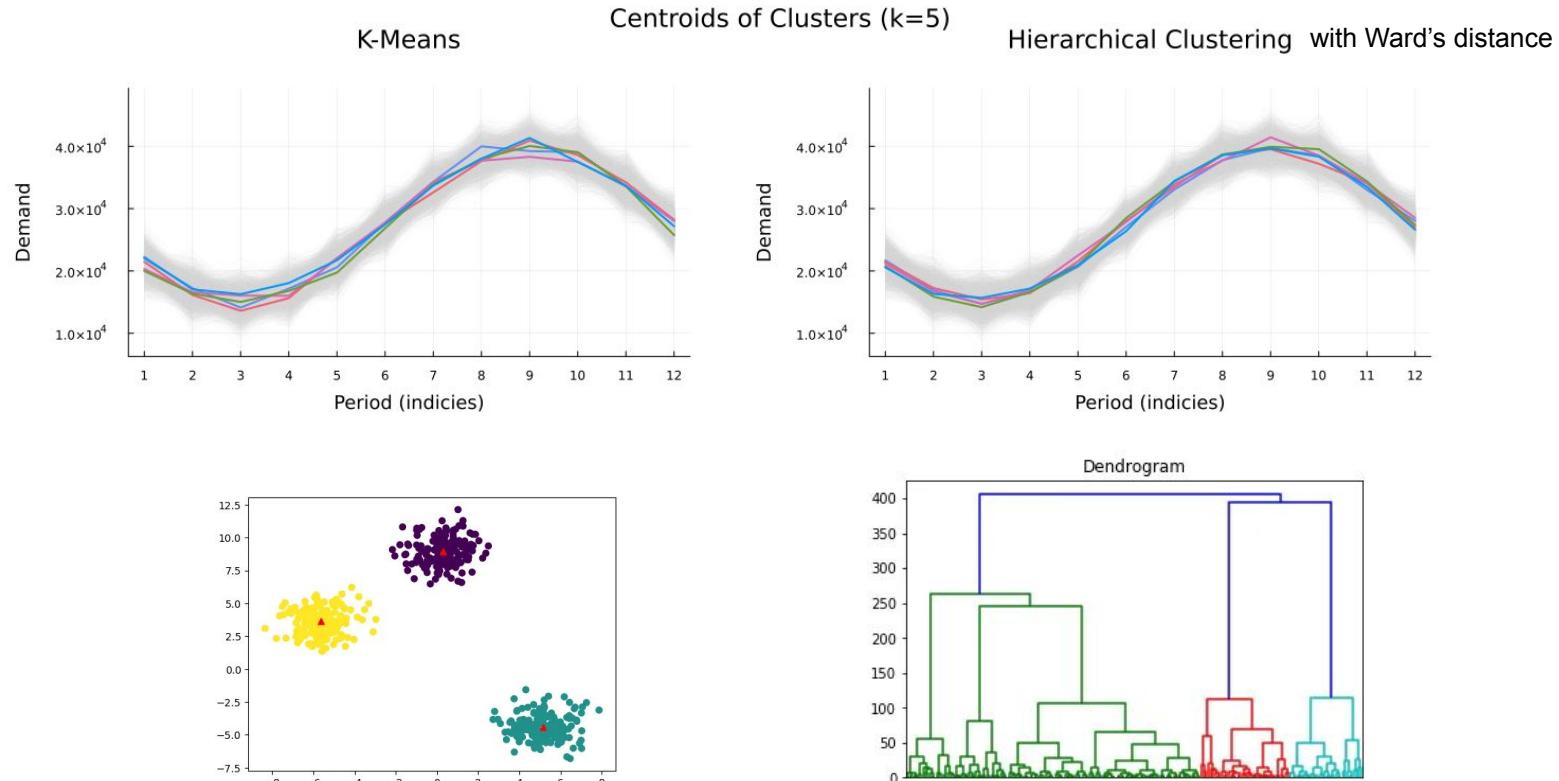
Dataset Generation

Scenarios Reduction

Results

Limits

Scenarios Reduction



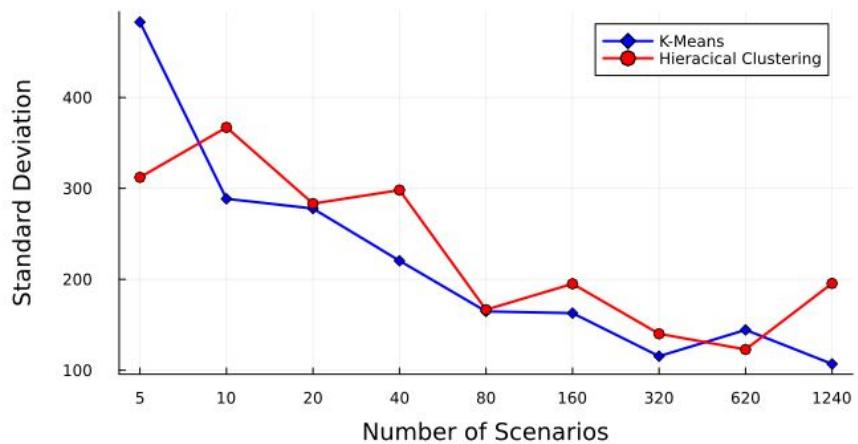
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Scenarios Reduction

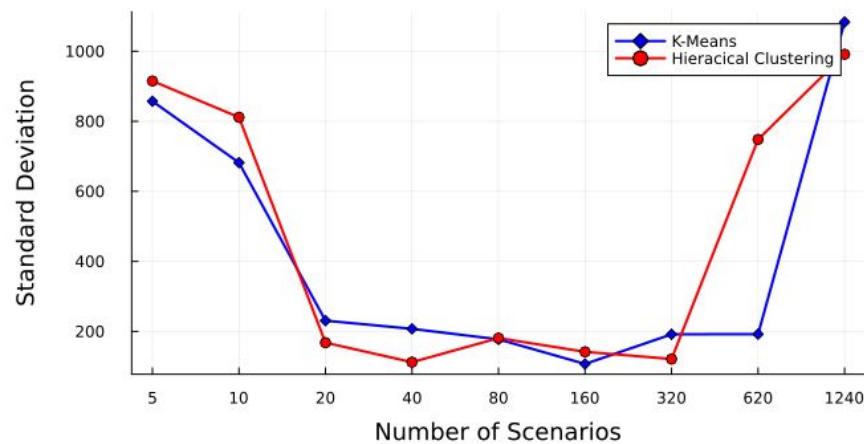
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Scenarios Reduction

In-Sample Stability



Out-Sample Stability



Elbow Criterion
=> $k = 20$ scenarios

A''

Outline of the presentation

Stochastic Optimization Model

Dataset Generation

Scenarios Reduction

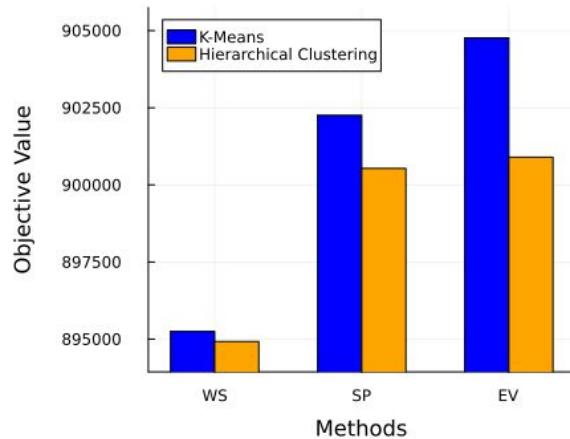
Results

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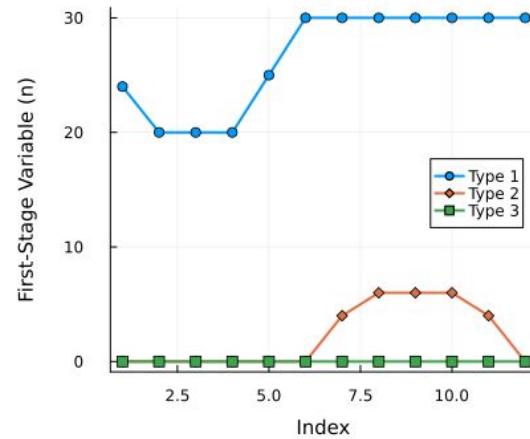
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Results

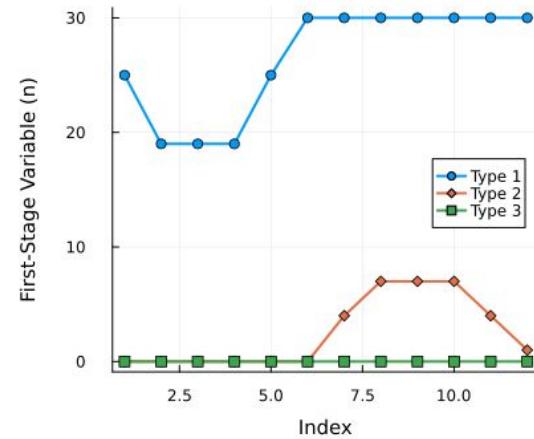
Comparison of WS, SP, and EV Solutions



K-Means Decisions n



H-Clust Decisions n



SP instead of EV
=> Between 500€ and 2500€ per day

Outline of the presentation

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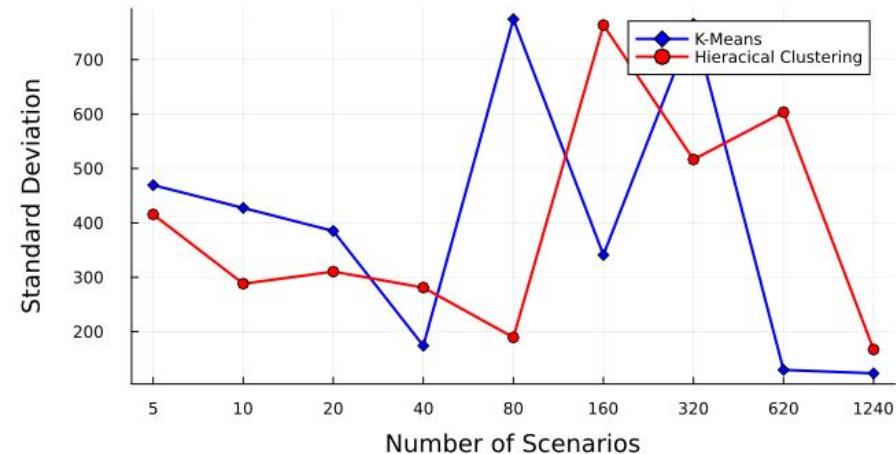
Results

Limits

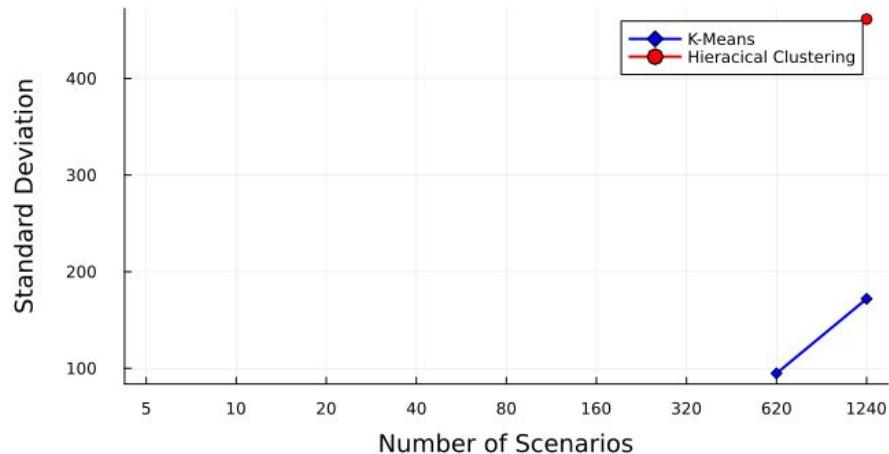
Limits

When we change a bit the
numbers of available reactors...

In-Sample Stability



Out-Sample Stability



A''

Limits

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Conclusion

- **Objective:** Incorporate stochasticity in a deterministic model
- **Key Features:** K-means & Hierarchical clustering
- **Improvements:**
 - ◆ More tolerant with infeasible solutions
 - ◆ Explore Sample Average Approximation
 - ◆ Study real datasets