

# **Workshop COOPS 2017 in Milano**

## **Bidding System Design for Multiperiod Electricity Markets**



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# Contents

**Problem formulation**

## ▶ Part I: What is bidding system design?

- ▶ relation between **market clearing problem** and **convex optimization**
- ▶ relation between **bidding curves** and **cost functions**
- ▶ difficulty in bidding system design for **multiperiod markets**

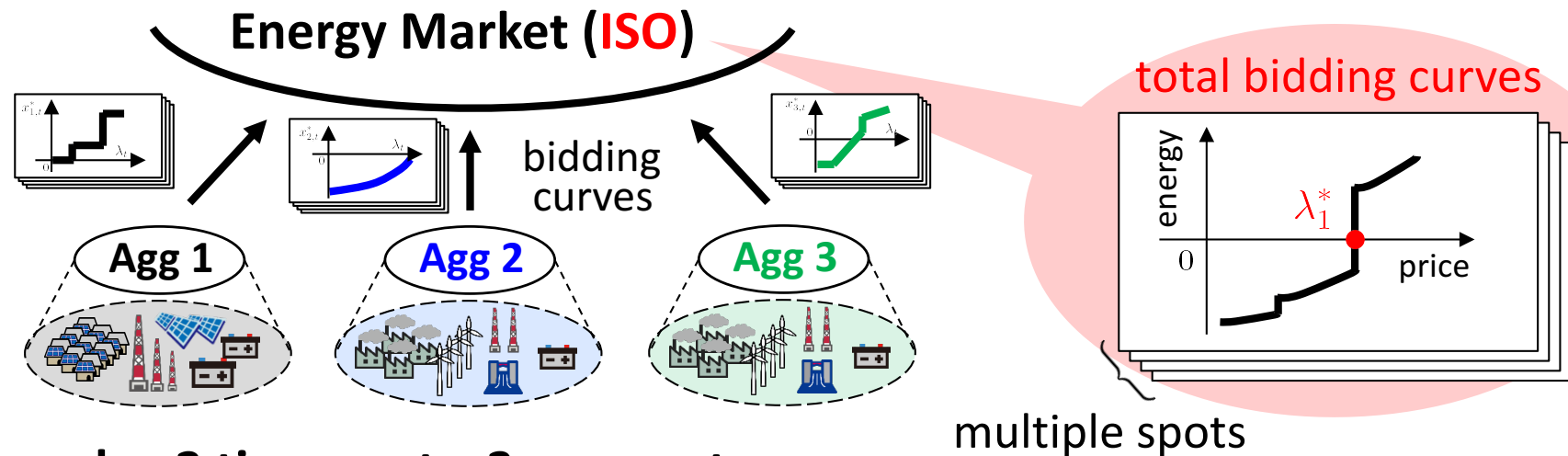
## ▶ Part II: How to design a bidding system for multiperiod markets?

- ▶ basis transformation compatible with **energy shift market**
- ▶ **sequential market clearing scheme**
- ▶ numerical examples

**An approximate  
solution method**



# Day-Ahead Energy Markets



**Example** 2 time spots, 3 aggregators

Market Results	Aggregator 1 (producer)	Aggregator 2 (consumer)	Aggregator 3 (prosumer)	Clearing Price
Spot 1 (AM)	150 [kWh]	-250 [kWh]	100 [kWh]	10 [yen/kWh]
Spot 2 (PM)	100 [kWh]	-50 [kWh]	-50 [kWh]	5 [yen/kWh]

Decision variables:  
Balanced

$$x_1^* = \begin{pmatrix} 150 \\ 100 \end{pmatrix} \quad x_2^* = \begin{pmatrix} -250 \\ -50 \end{pmatrix} \quad x_3^* = \begin{pmatrix} 100 \\ -50 \end{pmatrix}$$

$$\lambda^* = \begin{pmatrix} 10 \\ 5 \end{pmatrix}$$

**Market clearing:** Find “desirable”  $\lambda^*$  &  $(x_\alpha^*)_{\alpha \in \mathcal{A}}$  such that  $\sum_{\alpha \in \mathcal{A}} x_\alpha^* = 0$



# Market Clearing as Optimization

Market Results	Aggregator 1 (producer)	Aggregator 2 (consumer)	Aggregator 3 (prosumer)	Clearing Price
Spot 1 (AM)	150 [kWh]	-250 [kWh]	100 [kWh]	10 [yen/kWh]
Spot 2 (PM)	100 [kWh]	-50 [kWh]	-50 [kWh]	5 [yen/kWh]

$\lambda^*$

**Profit of Agg  $\alpha$ :**  $J_\alpha(x_\alpha^*; \lambda^*) = \underbrace{\langle \lambda^*, x_\alpha^* \rangle}_{\text{income}} - \underbrace{F_\alpha(x_\alpha^*)}_{\text{cost}}$   
 (selfish objective function)

✓ See later how to determine  $F_\alpha(x_\alpha)$  (especially for **prosumer!**)

**Social profit:**  $\sum_{\alpha \in \mathcal{A}} J_\alpha(x_\alpha^*; \lambda^*) = \left\langle \lambda^*, \sum_{\alpha \in \mathcal{A}} x_\alpha^* \right\rangle - \sum_{\alpha \in \mathcal{A}} F_\alpha(x_\alpha^*)$   
 (social objective function)  $= 0$  **social cost**

**Social profit maximization = Social cost minimization**

$$\min_{(x_\alpha)_{\alpha \in \mathcal{A}}} \sum_{\alpha \in \mathcal{A}} F_\alpha(x_\alpha) \quad \text{s.t.} \quad \sum_{\alpha \in \mathcal{A}} x_\alpha = 0$$

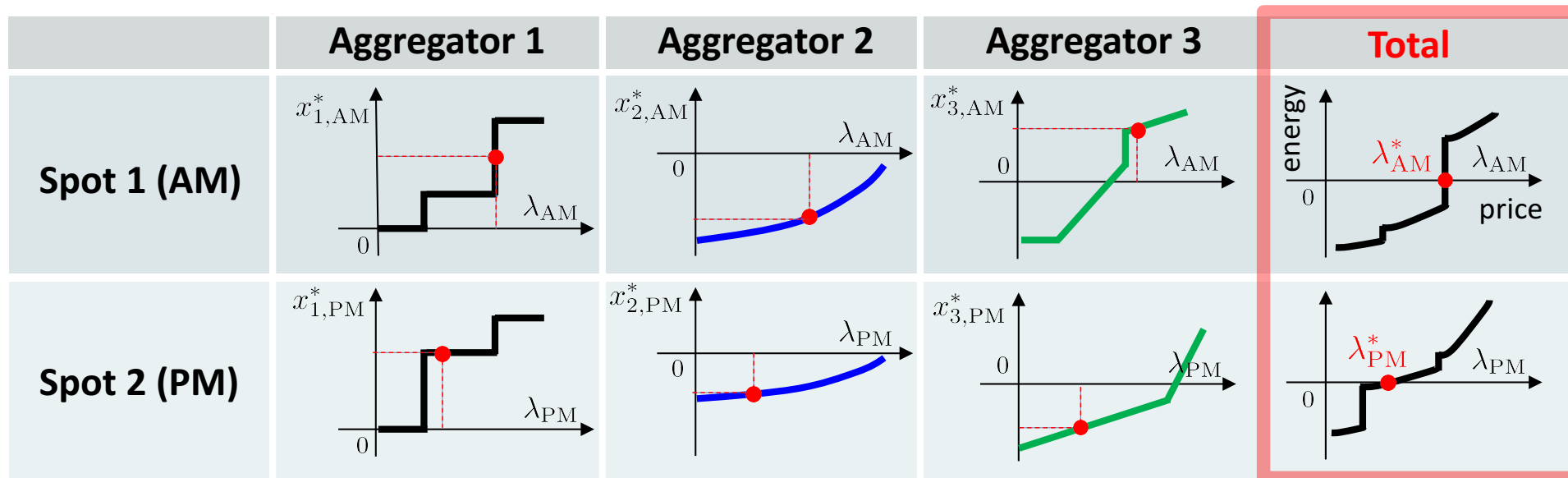


**primal:**  $(x_\alpha^*)_{\alpha \in \mathcal{A}}$   
**dual:**  $\lambda^*$



# Bidding Curves for Market Clearing

Suppose that **bidding curves for each spot** are submitted to ISO



ISO can find each **clearing price** and **balancing amounts** as **crossing points of (total) bidding curves**

**But...** Are such crossing points really solutions of

social cost minimization:  $\min_{(x_\alpha)_{\alpha \in \mathcal{A}}} \sum_{\alpha \in \mathcal{A}} F_\alpha(x_\alpha) \quad \text{s.t.} \quad \sum_{\alpha \in \mathcal{A}} x_\alpha = 0 \quad ??$





# Brief Summary: Bidding System Design

**Socially optimal market clearing problem:**

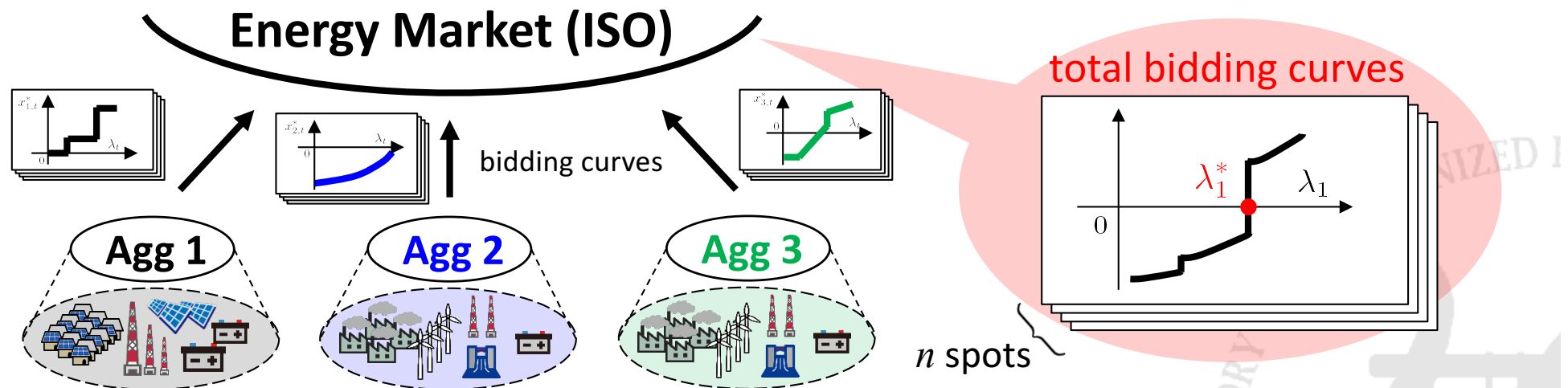
$$\text{Find } \begin{cases} \text{primal: } (x_\alpha^*)_{\alpha \in \mathcal{A}} \\ \text{dual: } \lambda^* \end{cases} \text{ solving } \min_{(x_\alpha)_{\alpha \in \mathcal{A}}} \sum_{\alpha \in \mathcal{A}} F_\alpha(x_\alpha) \quad \text{s.t.} \quad \sum_{\alpha \in \mathcal{A}} x_\alpha = 0$$

**Q1: What is a reasonable cost function  $F_\alpha(x_\alpha)$  ??**

- ✓ Prosumption  $x_\alpha$  should be a mixture of generators, batteries, renewables etc

**Q2: Is it possible to construct bidding curves from  $F_\alpha(x_\alpha)$  ??**

- ✓ Social cost should be minimized with  $(x_\alpha^*)_{\alpha \in \mathcal{A}}$  and  $\lambda^*$  found as crossing-points



- ✓ Bidding system design = Distributed algorithm design under pre-specified ISO operation



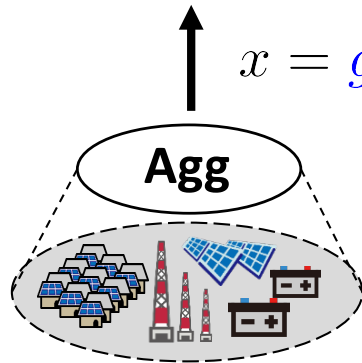
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- ▶ **Part I: What is bidding system design?**
  - ▶ relation between **market clearing problem** and **convex optimization**
  - ▶ relation between **bidding curves** and **cost functions**
  - ▶ difficulty in bidding system design for **multiperiod markets**
  
- ▶ **Part II: How to design a bidding system for multiperiod markets?**
  - ▶ basis transformation compatible with **energy shift**
  - ▶ **sequential clearing scheme for energy shift markets**
  - ▶ numerical examples



# Prosumption Cost function



$$x = g - l + \eta^{\text{out}} \delta^{\text{out}} - \frac{1}{\eta^{\text{in}}} \delta^{\text{in}}$$

Internal decision variables

$\left\{ \begin{array}{l} \text{generated power } g \\ \text{battery charge/discharge } \delta^{\text{in}}, \delta^{\text{out}} \end{array} \right.$

✓ **Constraints:**  $g \in \mathcal{G}, \delta \in \mathcal{D}$   
(e.g.  $0 \leq g \leq \bar{g}$ )

**Given**

Generation cost  $G(g)$   
 Battery usage cost  $D(\delta)$

**【Theorem】** If  $G(g)$  and  $D(\delta)$  are both convex, then

$$F(x) = \min_{(g, \delta) \in \mathcal{F}(x)} \left\{ G(g) + D(\delta) \right\} \text{ is convex with respect to } x$$

where  $\mathcal{F}(x) := \left\{ (g, \delta) \in \mathcal{G} \times \mathcal{D} : x = g - l + \eta^{\text{out}} \delta^{\text{out}} - \frac{1}{\eta^{\text{in}}} \delta^{\text{in}} \right\}$

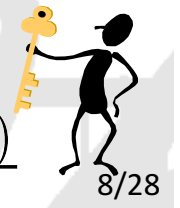
**Example 2 time spots (AM/PM) Constants:**  $l_{\text{AM}} = 50, l_{\text{PM}} = 10, \eta^{\text{out}} = \eta^{\text{in}} = 1$

$$F(x_{\text{AM}}, x_{\text{PM}}) = \min_{g, \delta} \left\{ G(g) + D(\delta) \right\} \text{ s.t. } \begin{pmatrix} x_{\text{AM}} \\ x_{\text{PM}} \end{pmatrix} = \begin{pmatrix} g_{\text{AM}} - 50 + \delta_{\text{AM}}^{\text{out}} - \delta_{\text{AM}}^{\text{in}} \\ g_{\text{PM}} - 10 + \delta_{\text{PM}}^{\text{out}} - \delta_{\text{PM}}^{\text{in}} \end{pmatrix}$$

$g, \delta$  not unique!

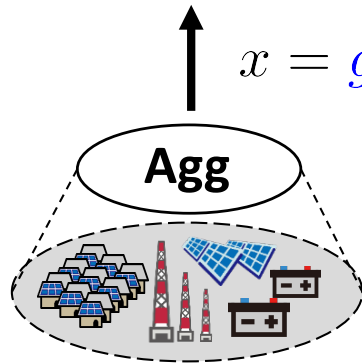
✓  $x = 0$  : supply-demand balance inside aggregator

**Optimal strategy for energy resources ensures convexity of  $F(x)$**





# Prosumption Cost function



$$x = g - l + \eta^{\text{out}} \delta^{\text{out}} - \frac{1}{\eta^{\text{in}}} \delta^{\text{in}}$$

Internal decision variables

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where  $\mathcal{F}(x) := \left\{ (g, \delta) \in \mathcal{G} \times \mathcal{D} : x = g - l + \eta^{\text{out}} \delta^{\text{out}} - \frac{1}{\eta^{\text{in}}} \delta^{\text{in}} \right\}$

✓ **Uncertain renewables** can be handled as **robust optimization** like:

$$F(x) = \max_{p \in \mathcal{P}} \min_{(g, \delta) \in \mathcal{F}(x, p)} \left\{ G(g) + D(\delta) \right\}$$

where  $\mathcal{P}$  is a scenario set of renewable generation

on-going work

(More interesting to see **how magnitude of uncertainty affects economics!**)

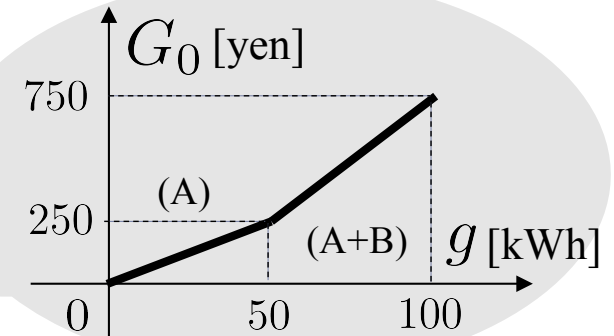


# Derivation of Bid Functions

**Example 2 time spots**  $l_{AM} = 50, l_{PM} = 50$

Generators & loads:  $\begin{pmatrix} x_{AM} \\ x_{PM} \end{pmatrix} = \begin{pmatrix} g_{AM} - l_{AM} \\ g_{PM} - l_{PM} \end{pmatrix}$

Spec of generators:  $\begin{cases} (A) & 0 \sim 50 \text{ [kWh]} & 5 \text{ [yen/kWh]} \\ (B) & 0 \sim 50 \text{ [kWh]} & 10 \text{ [yen/kWh]} \end{cases}$



$\begin{cases} \text{Generation cost: } G(g_{AM}, g_{PM}) = G_0(g_{AM}) + G_0(g_{PM}) & \text{additively decomposable} \\ \text{Feasible generator outputs: } 0 \leq g_{AM} \leq 100 & 0 \leq g_{PM} \leq 100 & \text{disjoint} \end{cases}$

$$\max_{x \in \mathcal{X}} J(x; \lambda) = \max_{x_{AM} \in [-50, 50]} \{ \lambda_{AM} x_{AM} - G_0(x_{AM} + 50) \} + \max_{x_{PM} \in [-50, 50]} \{ \lambda_{PM} x_{PM} - G_0(x_{PM} + 50) \}$$

**decomposable!**



## Bid functions

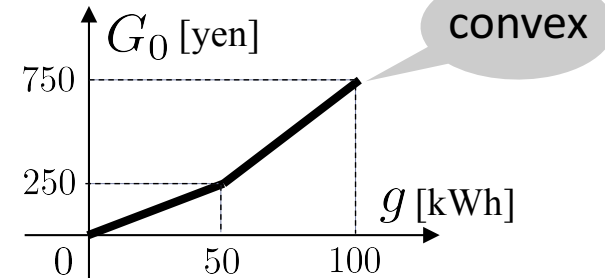
$$x_t^*(\lambda_t) = \arg \max_{x_t \in [-50, 50]} \{ \lambda_t x_t - G_0(x_t + 50) \}, \quad t \in \{AM, PM\}$$



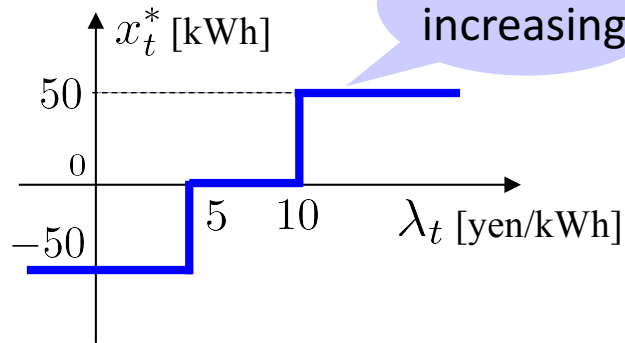
# Mathematics behind Bid Functions

## Bid functions (period-specific)

$$x_t^*(\lambda_t) = \arg \max_{x_t \in [-50, 50]} \{ \lambda_t x_t - G_0(x_t + 50) \}$$



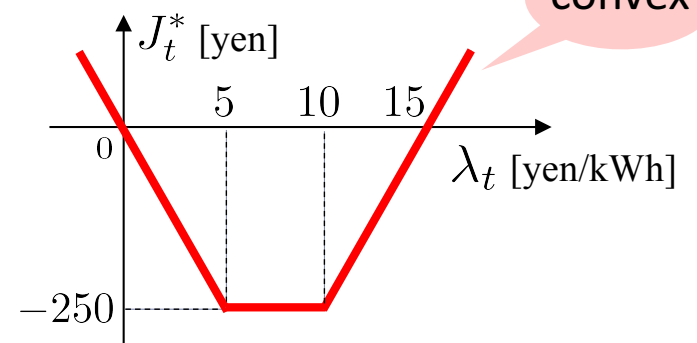
### Bidding curve



$$x_t^*(\lambda_t) = \partial J_t^*(\lambda_t)$$



### Maximum profit



$$J_t^*(\lambda_t) = \max_{x_t \in [-50, 50]} \{ \lambda_t x_t - G_0(x_t + 50) \}$$

## Legendre transform of cost function



**【Legendre transformation】** (convex conjugation) ✓  $\overline{\overline{F}} = F \iff F : \text{convex}$

$$\overline{F}(\lambda) := \sup_{x \in \mathcal{X}} \{ \langle \lambda, x \rangle - F(x) \}$$



# Multiperiod Bid Function

Generation cost:  $G(g_{AM}, g_{PM}) = G_0(g_{AM}) + G_0(g_{PM})$

Feasible generator outputs:  $0 \leq g_{AM} \leq 100 \quad 0 \leq g_{PM} \leq 100$

Ramp rate limit **(Added)**:  $-10 \leq g_{AM} - g_{PM} \leq 10$  **temporally correlated!**

## Multiperiod bid function

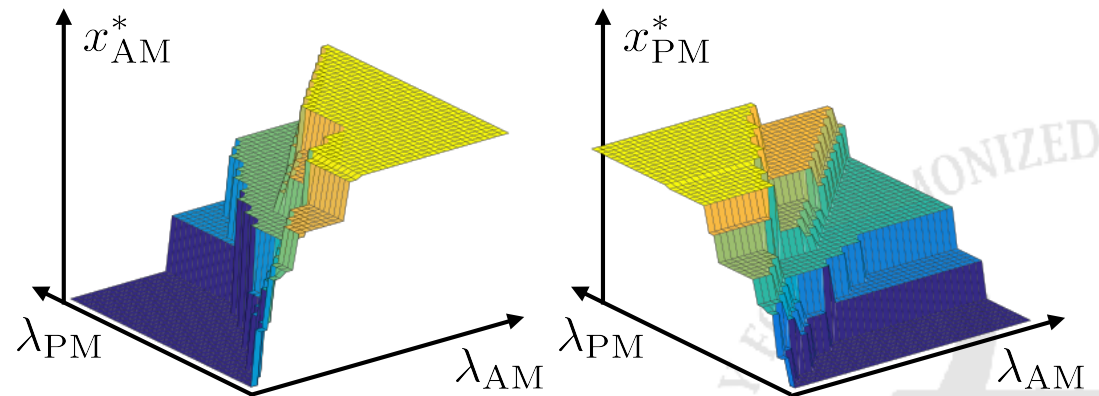
**indecomposable!**

$$x^*(\lambda) = \arg \max_{x \in \mathcal{X}} \{ \langle \lambda, x \rangle - F(x) \} = \begin{pmatrix} x_{AM}^*(\lambda_{AM}, \lambda_{PM}) \\ x_{PM}^*(\lambda_{AM}, \lambda_{PM}) \end{pmatrix}$$

✓  $x^* = \partial \bar{F}$  : monotone increasing

~~bidding curves~~ →

**bidding hyperplanes**



Multiperiod bid function is **not compatible with current bidding system**



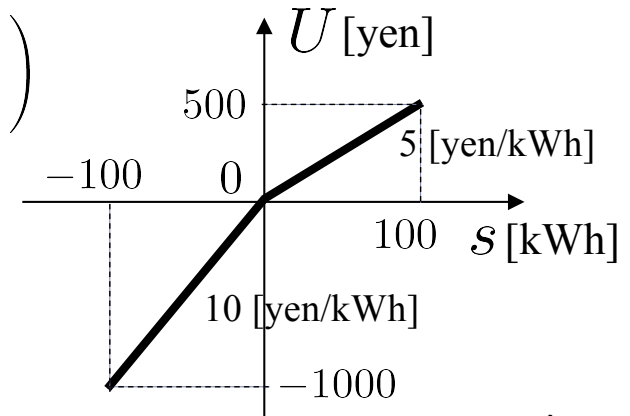
# Separability of Multiperiod Bid Function

**Example Battery aggregator**  $\begin{pmatrix} x_{AM} \\ x_{PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$

SOC constraints:  $\delta \in \mathcal{D}_{SOC}$  **not disjoint!**

Cost function based on **utility of final SOC**:

$$D(\delta) = -U(s_{fin}(\delta)) \quad s_{fin}(\delta) = s_0 + \sum_{t \in \{AM, PM\}} (\delta_t^{in} - \delta_t^{out})$$



**not additively decomposable!**



**【Lemma】** The multiperiod bid function is *separate* iff the cost function is *additively decomposable* and its domain is *disjoint*

$$\text{i.e. } x^*(\lambda) = \begin{pmatrix} x_1^*(\lambda_1) \\ \vdots \\ x_n^*(\lambda_n) \end{pmatrix} \iff F(x) = \sum_{t=1}^n F_t(x_t), \quad x \in \mathcal{X}_1 \times \dots \times \mathcal{X}_n.$$

**Negative fact!!** Traditional bidding curves available **just in very special cases**



# Brief Summary: Bidding System Design

**Socially optimal market clearing :**  $\min_{(x_\alpha)_{\alpha \in \mathcal{A}}} \sum_{\alpha \in \mathcal{A}} F_\alpha(x_\alpha) \quad \text{s.t.} \quad \sum_{\alpha \in \mathcal{A}} x_\alpha = 0$

✓ Bidding system design = Distributed algorithm design under pre-specified ISO operation

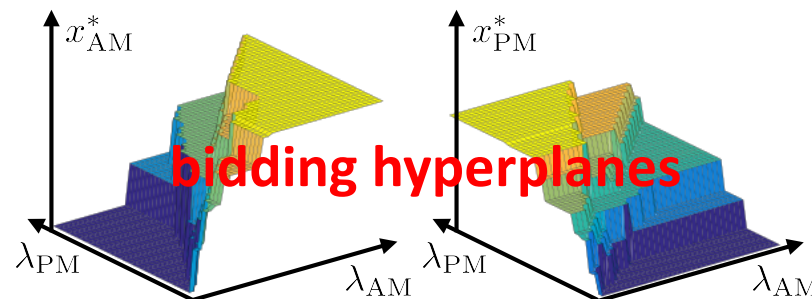
**【Theorem】**  $F(x) = \min_{(g, \delta) \in \mathcal{F}(x)} \{G(g) + D(\delta)\}$  is convex

where  $\mathcal{F}(x) := \{(g, \delta) \in \mathcal{G} \times \mathcal{D} : x = g - l + \eta^{\text{out}} \delta^{\text{out}} - \frac{1}{\eta^{\text{in}}} \delta^{\text{in}}\}$

**Multiperiod bid function:**

$$x^*(\lambda) = \arg \max_{x \in \mathcal{X}} \{\langle \lambda, x \rangle - F(x)\}$$

✓ monotone increasing  $x^* = \partial \bar{F} : \mathbb{R}^n \rightarrow \mathbb{R}^n$



**【Lemma】**

$$x^*(\lambda) = \begin{pmatrix} x_1^*(\lambda_1) \\ \vdots \\ x_n^*(\lambda_n) \end{pmatrix} \quad \text{separate}$$

additively decomposable

$$F(x) = \sum_{t=1}^n F_t(x_t), \quad x \in \mathcal{X}_1 \times \cdots \times \mathcal{X}_n \quad \text{disjoint}$$

**Bidding system design for multiperiod markets is not so simple!!**



# Contents

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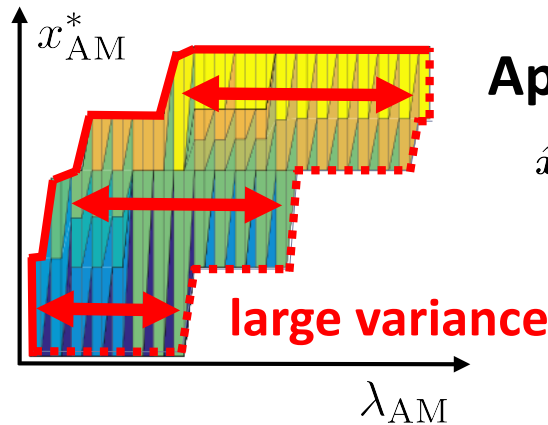
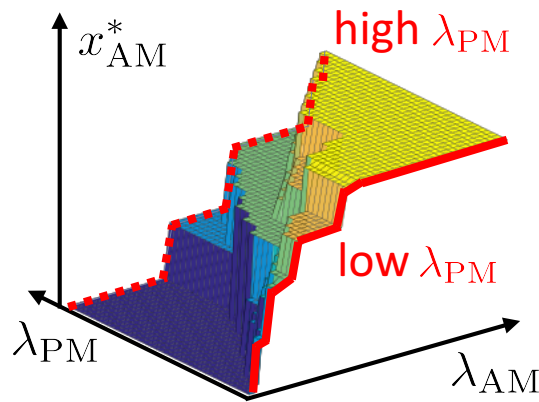
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# Ideas in Proposed Approach

## A) Basis transformation towards better approximation

## B) Approximation of bidding hyperplanes to bidding curves



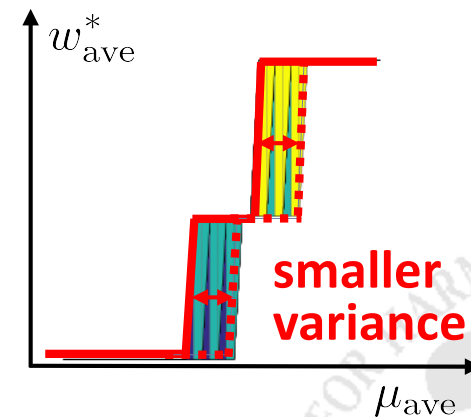
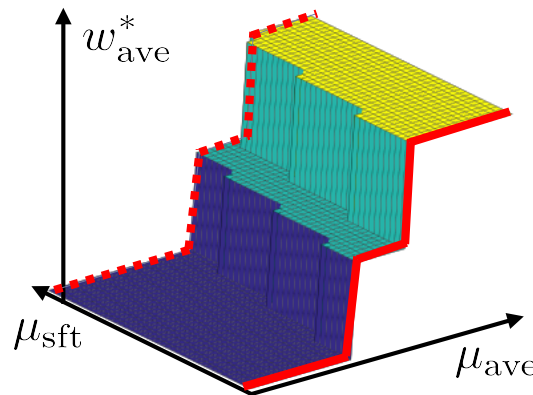
Approximate bidding curve

$$\hat{x}_{AM}^*(\hat{\lambda}_{AM}) = x_{AM}^*(\hat{\lambda}_{AM}, \lambda_{PM}) \Big|_{\lambda_{PM} = \hat{\lambda}_{PM}}$$

temporal correlation reduced!

$$\begin{pmatrix} x_{AM} \\ x_{PM} \end{pmatrix} \xrightarrow{T} \begin{pmatrix} w_{ave} \\ w_{sft} \end{pmatrix}$$

$$\begin{pmatrix} \lambda_{AM} \\ \lambda_{PM} \end{pmatrix} \xrightarrow{T} \begin{pmatrix} \mu_{ave} \\ \mu_{sft} \end{pmatrix}$$



## C) Sequential market clearing scheme

✓ From optimization view: (A) preconditioning (B)-(C) updates of primal/dual variables

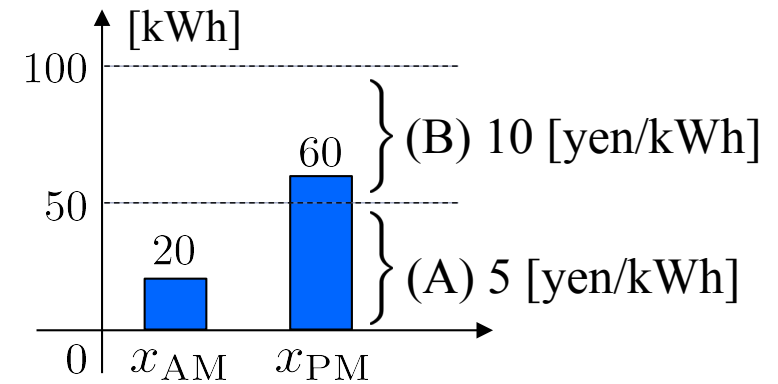


# Energy Shift: A Key Property of Batteries

**Agg 1 (generators):**  $\begin{pmatrix} x_{1,AM} \\ x_{1,PM} \end{pmatrix} = \begin{pmatrix} g_{AM} \\ g_{PM} \end{pmatrix}$

Spec of gens:  $\begin{cases} (A) & 0 \sim 50 \text{ [kWh]} & 5 \text{ [yen/kWh]} \\ (B) & 0 \sim 50 \text{ [kWh]} & 10 \text{ [yen/kWh]} \end{cases}$

**Agg 2 (loads):**  $\begin{pmatrix} x_{2,AM} \\ x_{2,PM} \end{pmatrix} = \begin{pmatrix} -l_{AM} \\ -l_{PM} \end{pmatrix} = \begin{pmatrix} -20 \\ -60 \end{pmatrix}$

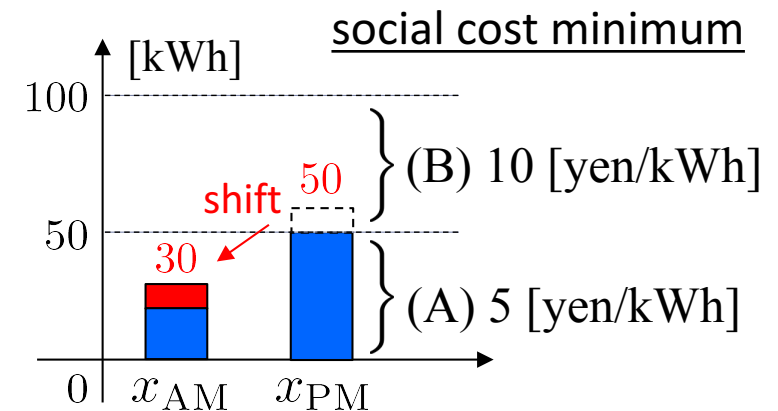


**Optimal price:**  $\begin{pmatrix} \lambda_{AM}^* \\ \lambda_{PM}^* \end{pmatrix} = \begin{pmatrix} 5 \\ 10 \end{pmatrix}$

**Agg 3 (batteries):**  $\begin{pmatrix} x_{3,AM} \\ x_{3,PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$

$\lambda_{PM}^* > \lambda_{AM}^* \longrightarrow \begin{cases} \text{PM: discharge (sell)} \\ \text{AM: charge (buy)} \end{cases}$

**New optimal price:**  $\begin{pmatrix} \lambda_{AM}^* \\ \lambda_{PM}^* \end{pmatrix} = \begin{pmatrix} 5 \\ 5 \end{pmatrix}$



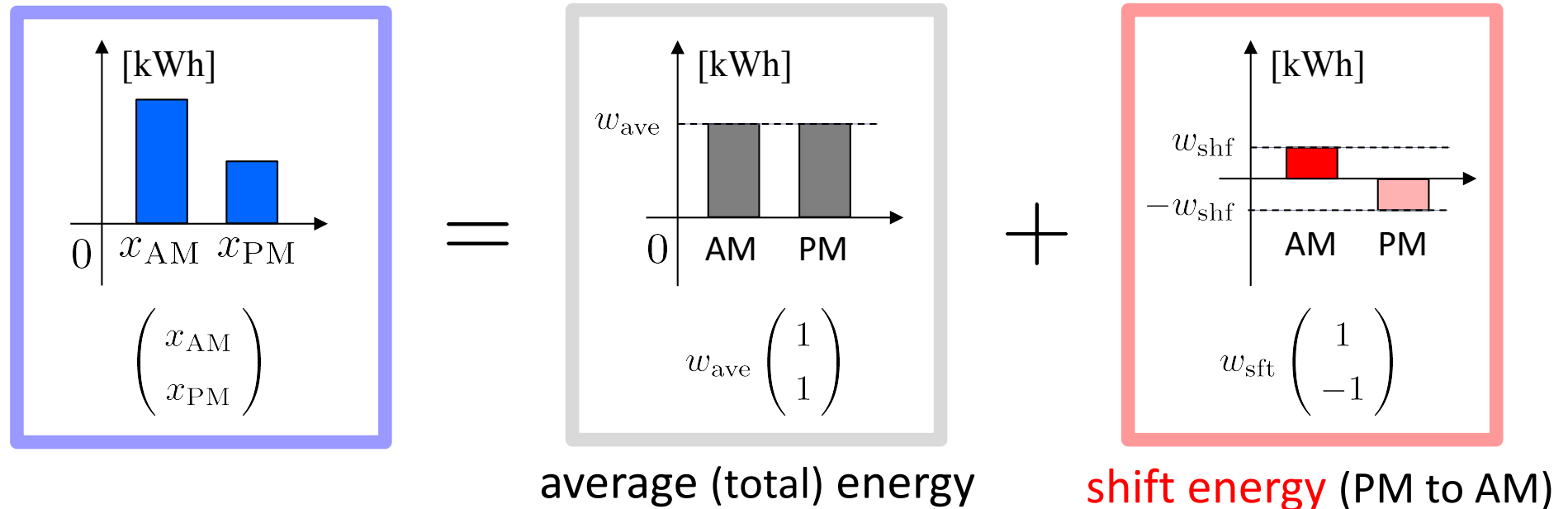
**Battery leads to price levelling-off!**

Energy market with explicit consideration of **energy shift??**



# Fourier-Like Basis Transformation

## Example 2 time spots



$$\begin{pmatrix} \lambda_{\text{AM}} \\ \lambda_{\text{PM}} \end{pmatrix} = \mu_{\text{ave}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + \mu_{\text{sft}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} \quad \left\{ \begin{array}{l} \mu_{\text{ave}} : \text{average (levelling-off) price} \\ \mu_{\text{sft}} : \text{energy shift price} \end{array} \right.$$

**Explicit consideration of levelling-off price & energy shift price**



**Energy shift market:**

$$\min_{(w_\alpha)_{\alpha \in \mathcal{A}}} \sum_{\alpha \in \mathcal{A}} H_\alpha(w_\alpha) \quad \text{s.t.} \quad \sum_{\alpha \in \mathcal{A}} w_\alpha = 0 \quad \left\{ \begin{array}{l} \text{primal: } (w_\alpha^*)_{\alpha \in \mathcal{A}} \\ \text{dual: } \mu^* \end{array} \right.$$

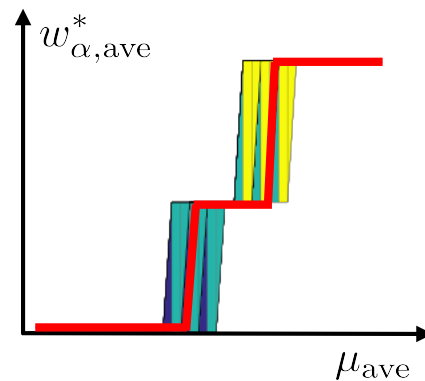
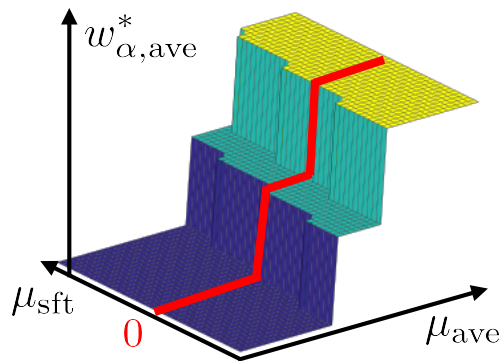


# Sequential Market Clearing

## Step 1) Market clearing of average energy amounts

Multiperiod bid function:  $\begin{pmatrix} w_{\alpha, \text{ave}}^*(\mu_{\text{ave}}, \mu_{\text{sft}}) \\ w_{\alpha, \text{sft}}^*(\mu_{\text{ave}}, \mu_{\text{sft}}) \end{pmatrix} = \arg \max_{w_{\alpha}} \{ \langle \mu, w_{\alpha} \rangle - H_{\alpha}(w_{\alpha}) \}$

**Approximate bid function:**  $\hat{w}_{\alpha, \text{ave}}^*(\hat{\mu}_{\text{ave}}) = w_{\alpha, \text{ave}}^*(\hat{\mu}_{\text{ave}}, \mu_{\text{sft}})|_{\mu_{\text{sft}}=0}$



assumption (premise) of price levelling-off

$$\begin{pmatrix} \lambda_{\text{AM}} \\ \lambda_{\text{PM}} \end{pmatrix} = \mu_{\text{ave}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + \mu_{\text{sft}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

→ ISO determines  $\hat{\mu}_{\text{ave}}^*$  &  $(\hat{w}_{\alpha, \text{ave}}^*)_{\alpha \in \mathcal{A}}$  by approximate bidding curves

## Step 2) Market clearing of **shift energy amounts**

**Approximate bid function:**  $\hat{w}_{\alpha, \text{sft}}^*(\hat{\mu}_{\text{sft}}) = \arg \max_{w_{\alpha, \text{sft}}} \{ \hat{\mu}_{\text{sft}} w_{\alpha, \text{sft}} - H_{\alpha}(\hat{w}_{\alpha, \text{ave}}^*, w_{\alpha, \text{sft}}) \}$

→ ISO determines  $\hat{\mu}_{\text{sft}}^*$  &  $(\hat{w}_{\alpha, \text{sft}}^*)_{\alpha \in \mathcal{A}}$  by approximate bidding curves

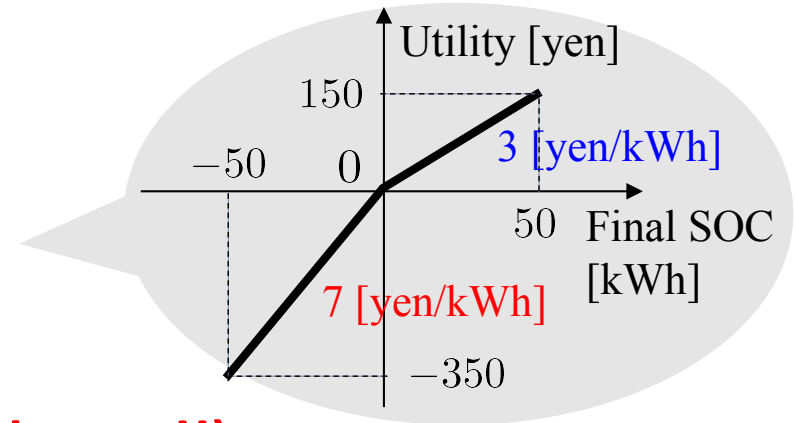


# Example: Sequential Market Clearing

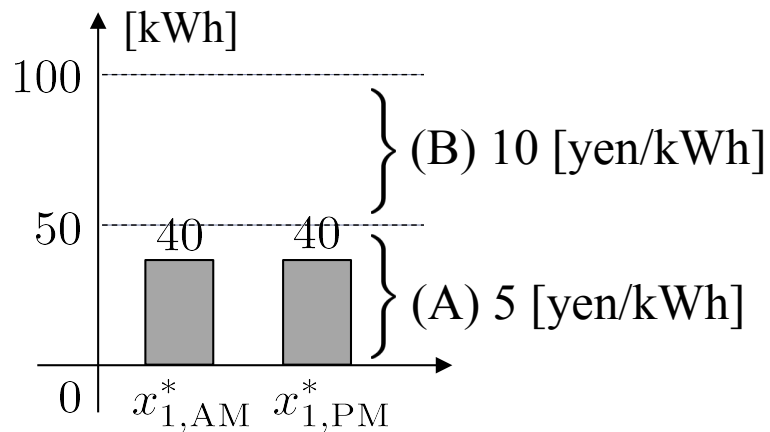
**Agg 1 (generators):**  $\begin{pmatrix} x_{1,AM} \\ x_{1,PM} \end{pmatrix} = \begin{pmatrix} g_{AM} \\ g_{PM} \end{pmatrix} \quad \begin{cases} (A) \ 0 \sim 50 \text{ [kWh]} \quad 5 \text{ [yen/kWh]} \\ (B) \ 0 \sim 50 \text{ [kWh]} \quad 10 \text{ [yen/kWh]} \end{cases}$

**Agg 2 (loads):**  $\begin{pmatrix} x_{2,AM} \\ x_{2,PM} \end{pmatrix} = \begin{pmatrix} -l_{AM} \\ -l_{PM} \end{pmatrix} = \begin{pmatrix} -20 \\ -60 \end{pmatrix}$

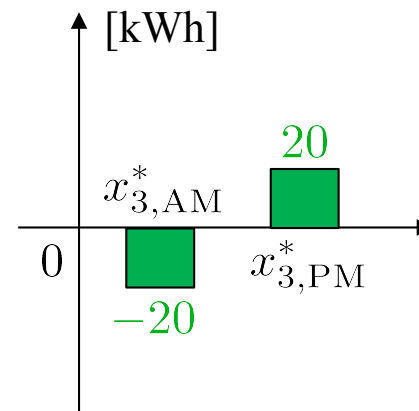
**Agg 3 (batteries):**  $\begin{pmatrix} x_{3,AM} \\ x_{3,PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$



Socially optimal market results (only god knows!!)



**Agg 1**



**Agg 3**

**Optimal clearing price:**

$$\begin{pmatrix} \lambda_{AM}^* \\ \lambda_{PM}^* \end{pmatrix} = \begin{pmatrix} 5 \\ 5 \end{pmatrix}$$

✓ optimal price levels off  
(i.e.  $\mu_{sft}^* = 0$ )

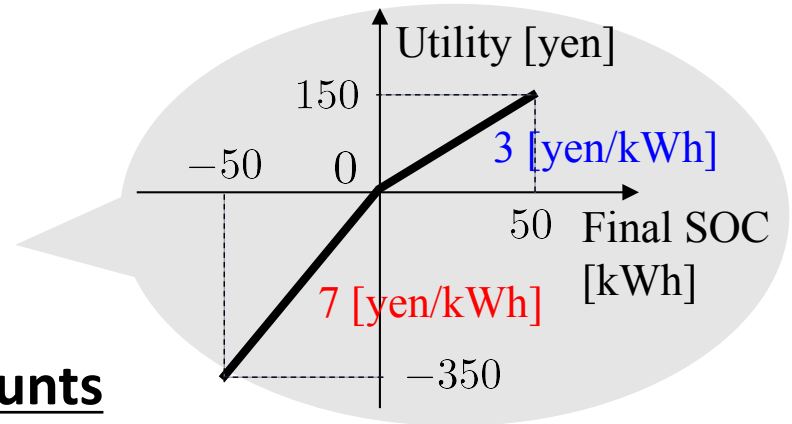


# Example: Sequential Market Clearing

**Agg 1 (generators):**  $\begin{pmatrix} x_{1,AM} \\ x_{1,PM} \end{pmatrix} = \begin{pmatrix} g_{AM} \\ g_{PM} \end{pmatrix} \quad \begin{cases} (A) \ 0 \sim 50 \text{ [kWh]} \ 5 \text{ [yen/kWh]} \\ (B) \ 0 \sim 50 \text{ [kWh]} \ 10 \text{ [yen/kWh]} \end{cases}$

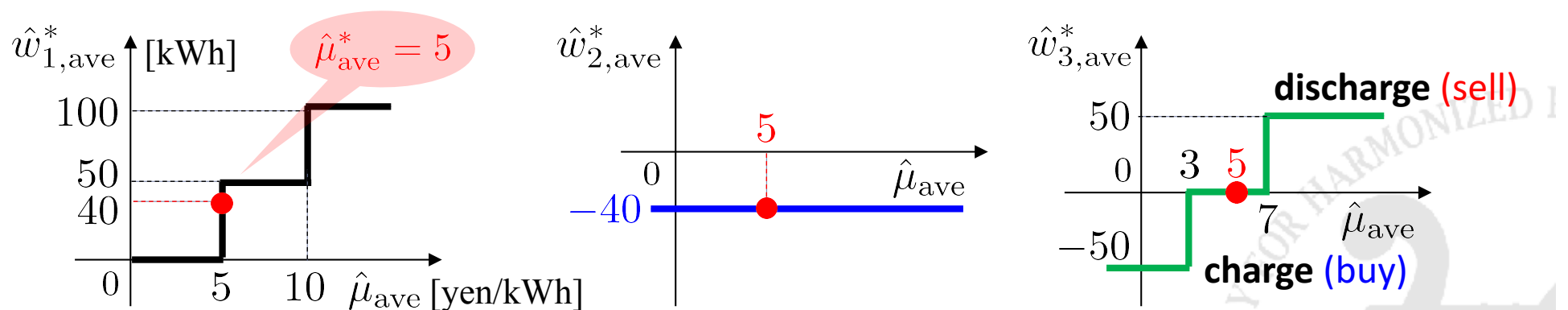
**Agg 2 (loads):**  $\begin{pmatrix} x_{2,AM} \\ x_{2,PM} \end{pmatrix} = \begin{pmatrix} -l_{AM} \\ -l_{PM} \end{pmatrix} = \begin{pmatrix} -20 \\ -60 \end{pmatrix}$

**Agg 3 (batteries):**  $\begin{pmatrix} x_{3,AM} \\ x_{3,PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$



## 1) Market clearing of average energy amounts

**Approximate bid function:**  $\hat{w}_{\alpha,ave}^*(\hat{\mu}_{ave}) = w_{\alpha,ave}^*(\hat{\mu}_{ave}, \mu_{sft})|_{\mu_{sft}=0}$



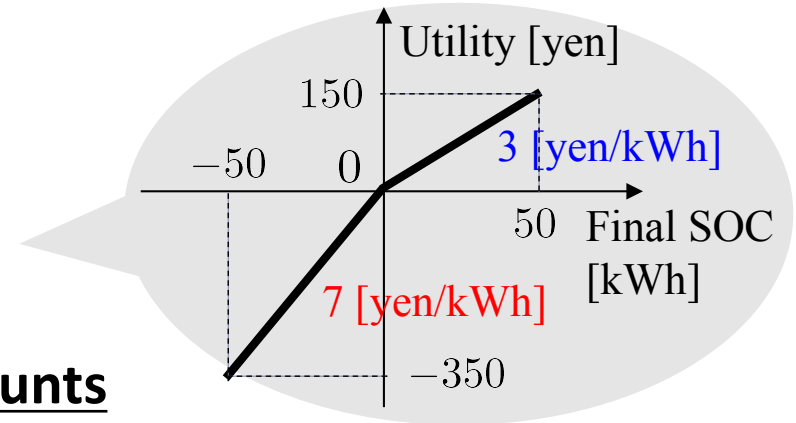


# Example: Sequential Market Clearing

**Agg 1 (generators):**  $\begin{pmatrix} x_{1,AM} \\ x_{1,PM} \end{pmatrix} = \begin{pmatrix} g_{AM} \\ g_{PM} \end{pmatrix} \quad \begin{cases} (A) & 0 \sim 50 \text{ [kWh]} & 5 \text{ [yen/kWh]} \\ (B) & 0 \sim 50 \text{ [kWh]} & 10 \text{ [yen/kWh]} \end{cases}$

**Agg 2 (loads):**  $\begin{pmatrix} x_{2,AM} \\ x_{2,PM} \end{pmatrix} = \begin{pmatrix} -l_{AM} \\ -l_{PM} \end{pmatrix} = \begin{pmatrix} -20 \\ -60 \end{pmatrix}$

**Agg 3 (batteries):**  $\begin{pmatrix} x_{3,AM} \\ x_{3,PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$

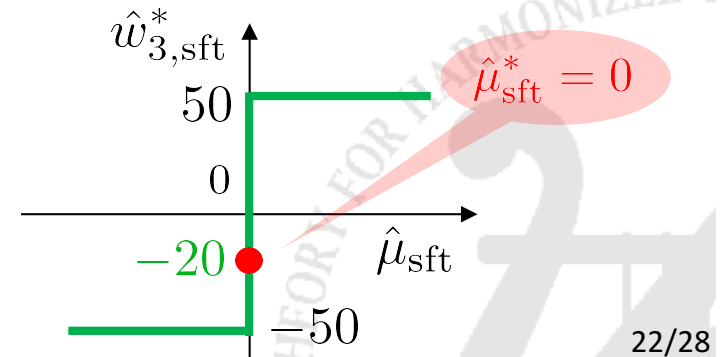
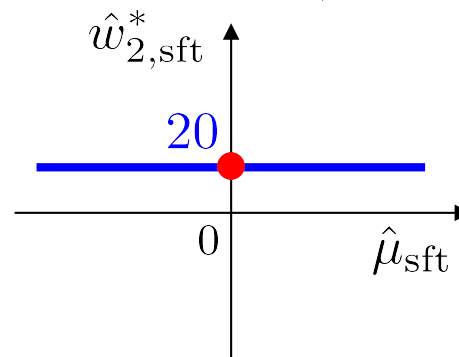
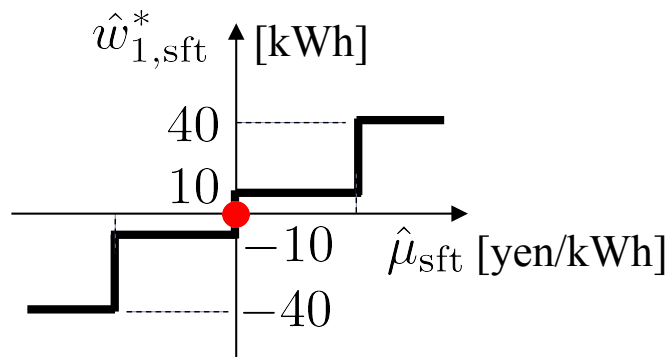


## 1) Market clearing of average energy amounts

Balancing amounts:  $\hat{w}_{1,ave}^* = 40$   $\hat{w}_{2,ave}^* = -40$   $\hat{w}_{3,ave}^* = 0$  **Average price:**  $\hat{\mu}_{ave}^* = 5$

## 2) Market clearing of shift energy amounts

**Approximate bid function:**  $\hat{w}_{\alpha,sft}^*(\hat{\mu}_{sft}) = \arg \max_{w_{\alpha,sft}} \{ \hat{\mu}_{sft} w_{\alpha,sft} - H_{\alpha}(\hat{w}_{\alpha,ave}^*, w_{\alpha,sft}) \}$



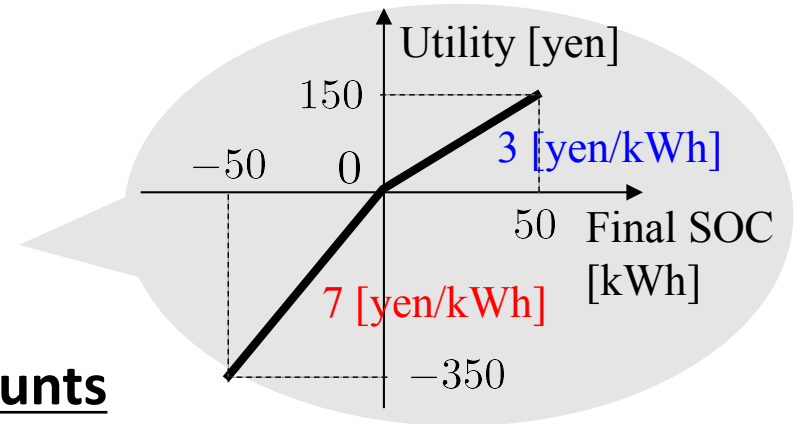


# Example: Sequential Market Clearing

**Agg 1 (generators):**  $\begin{pmatrix} x_{1,AM} \\ x_{1,PM} \end{pmatrix} = \begin{pmatrix} g_{AM} \\ g_{PM} \end{pmatrix} \quad \begin{cases} (A) \ 0 \sim 50 \text{ [kWh]} \ 5 \text{ [yen/kWh]} \\ (B) \ 0 \sim 50 \text{ [kWh]} \ 10 \text{ [yen/kWh]} \end{cases}$

**Agg 2 (loads):**  $\begin{pmatrix} x_{2,AM} \\ x_{2,PM} \end{pmatrix} = \begin{pmatrix} -l_{AM} \\ -l_{PM} \end{pmatrix} = \begin{pmatrix} -20 \\ -60 \end{pmatrix}$

**Agg 3 (batteries):**  $\begin{pmatrix} x_{3,AM} \\ x_{3,PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$



## 1) Market clearing of average energy amounts

Balancing amounts:  $\hat{w}_{1,ave}^* = 40$   $\hat{w}_{2,ave}^* = -40$   $\hat{w}_{3,ave}^* = 0$  **Average price:**  $\hat{\mu}_{ave}^* = 5$

## 2) Market clearing of shift energy amounts

Balancing amounts:  $\hat{w}_{1,sft}^* = 0$   $\hat{w}_{2,sft}^* = 20$   $\hat{w}_{3,sft}^* = -20$  **Shift energy price:**  $\hat{\mu}_{sft}^* = 0$

**【Theorem】** Socially optimal market clearing iff **optimal price levels off**

i.e.  $(T^{-1}\hat{w}_{\alpha}^*)_{\alpha \in \mathcal{A}} = (x_{\alpha}^*)_{\alpha \in \mathcal{A}}, \quad T^{-1}\hat{\mu}^* = \lambda^* \iff \lambda_1^* = \dots = \lambda_n^*.$



# Example: Sequential Market Clearing

**Agg 1 (generators):**  $\begin{pmatrix} x_{1,AM} \\ x_{1,PM} \end{pmatrix} = \begin{pmatrix} g_{AM} \\ g_{PM} \end{pmatrix} \quad \begin{cases} (A) & 0 \sim 50 \text{ [kWh]} & 5 \text{ [yen/kWh]} \\ (B) & 0 \sim 50 \text{ [kWh]} & 10 \text{ [yen/kWh]} \end{cases}$

**Agg 2 (loads):**  $\begin{pmatrix} x_{2,AM} \\ x_{2,PM} \end{pmatrix} = \begin{pmatrix} -l_{AM} \\ -l_{PM} \end{pmatrix} = \begin{pmatrix} -20 \\ -60 \end{pmatrix}$

**Agg 3 (batteries):**  $\begin{pmatrix} x_{3,AM} \\ x_{3,PM} \end{pmatrix} = \begin{pmatrix} \delta_{AM}^{out} - \delta_{AM}^{in} \\ \delta_{PM}^{out} - \delta_{PM}^{in} \end{pmatrix}$  (without battery aggregator)

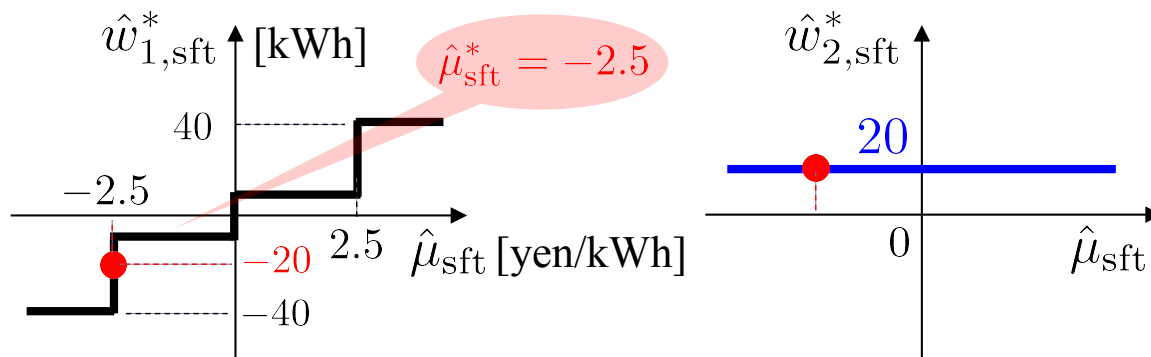
**Optimal clearing price:**

$$\begin{pmatrix} \lambda_{AM}^* \\ \lambda_{PM}^* \end{pmatrix} = \begin{pmatrix} 5 \\ 10 \end{pmatrix}$$

## 1) Market clearing of average energy amounts

Balancing amounts:  $\hat{w}_{1,ave}^* = 40$   $\hat{w}_{2,ave}^* = -40$   $\hat{w}_{3,ave}^* = 0$  **Average price:**  $\hat{\mu}_{ave}^* = 5$

## 2) Market clearing of shift energy amounts



**Clearing price:**

$$\begin{pmatrix} 2.5 \\ 7.5 \end{pmatrix} = \hat{\mu}_{ave}^* \begin{pmatrix} 1 \\ 1 \end{pmatrix} + \hat{\mu}_{sft}^* \begin{pmatrix} 1 \\ -1 \end{pmatrix} \neq \begin{pmatrix} \lambda_{AM}^* \\ \lambda_{PM}^* \end{pmatrix}$$

**At least approximate clearing even if optimal price does not level off**



# Numerical Example

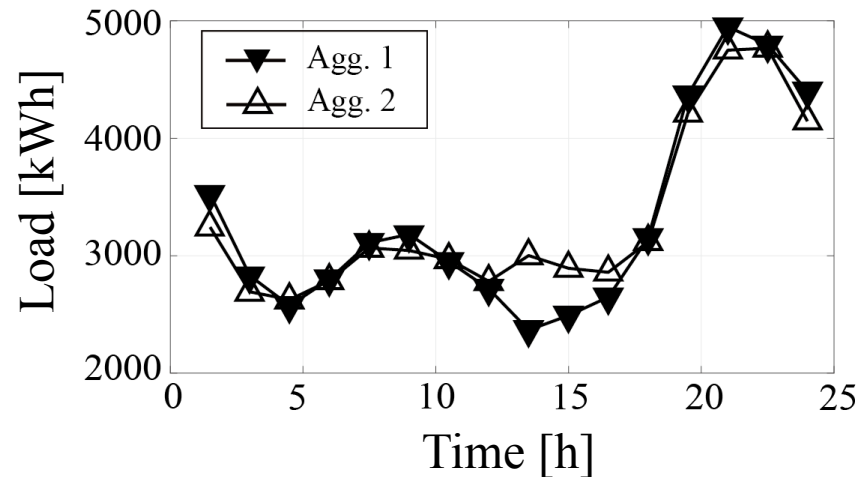
## Agg 1, Agg 2 (loads & batteries)

Charge/discharge efficiencies:  $\begin{cases} 95\% \text{ (Agg 1)} \\ 94\% \text{ (Agg 2)} \end{cases}$

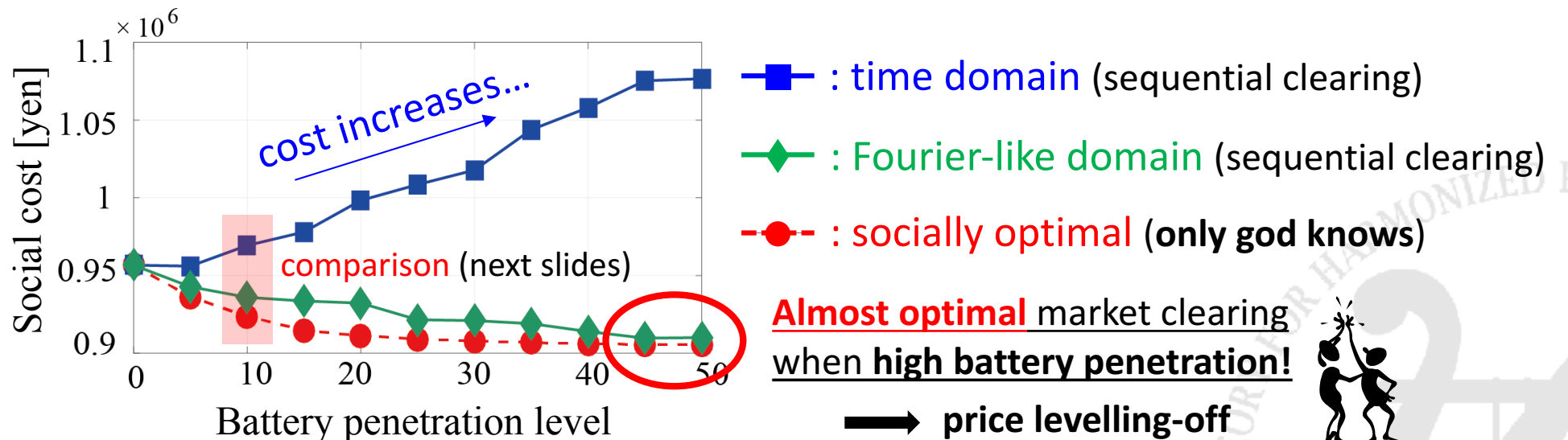
SOC/inverter constraints: **parameters**

## Agg 3 (9 types of generators)

Generation costs: 3, 6, ..., 27 [yen/kWh]



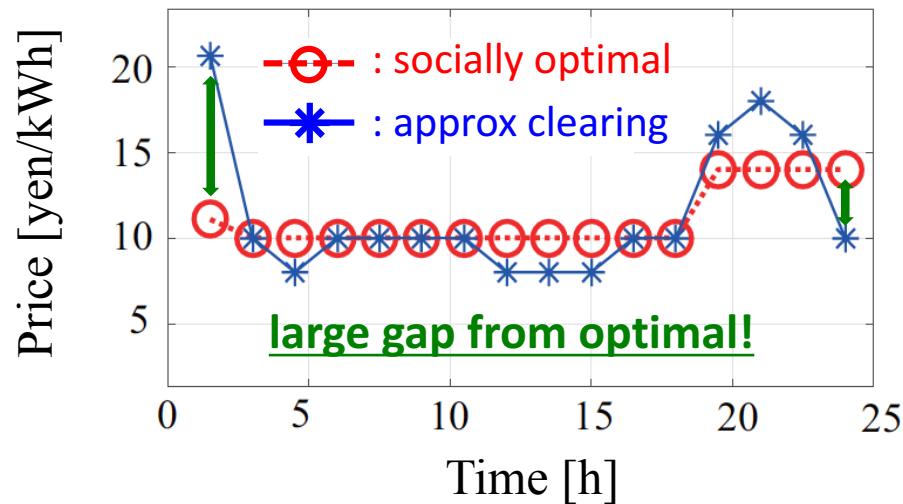
## Resultant social costs when varying battery penetration levels (16 time spots)



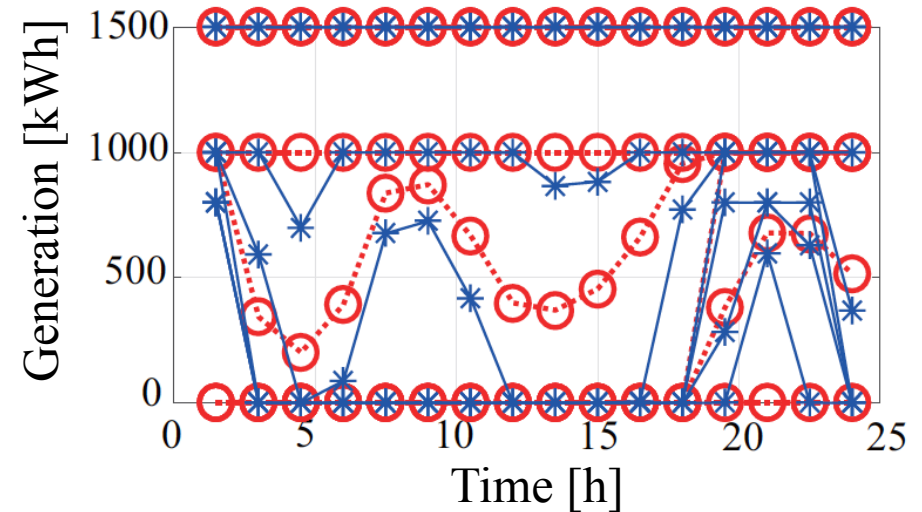


# Sequential Clearing in Time Domain

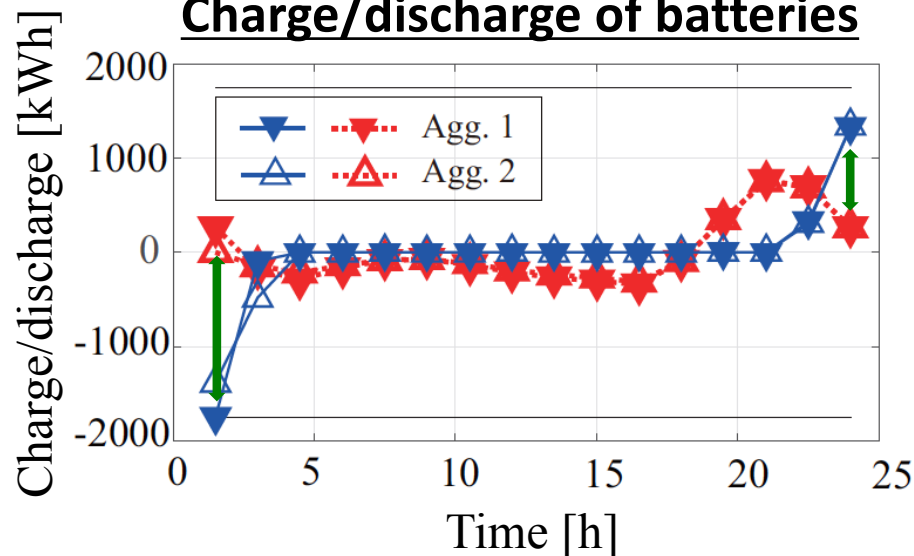
## Clearing price



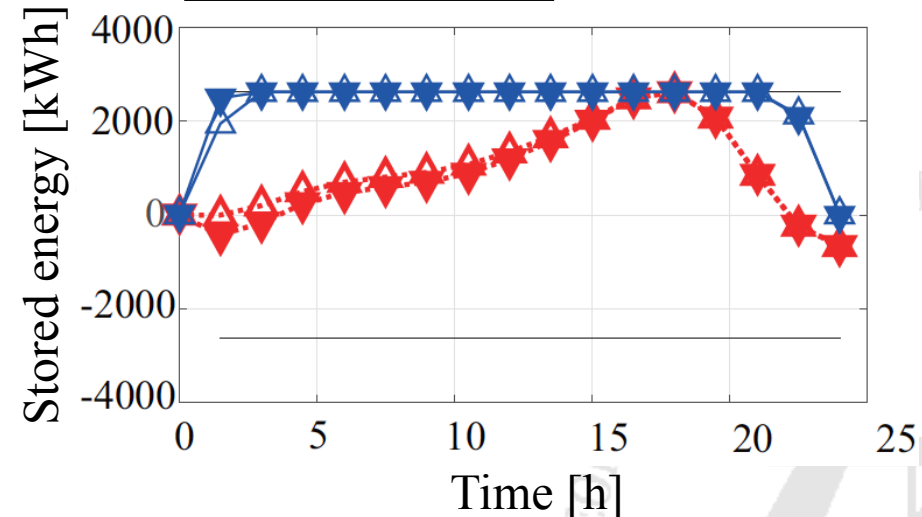
## Generators (9 types)



## Charge/discharge of batteries



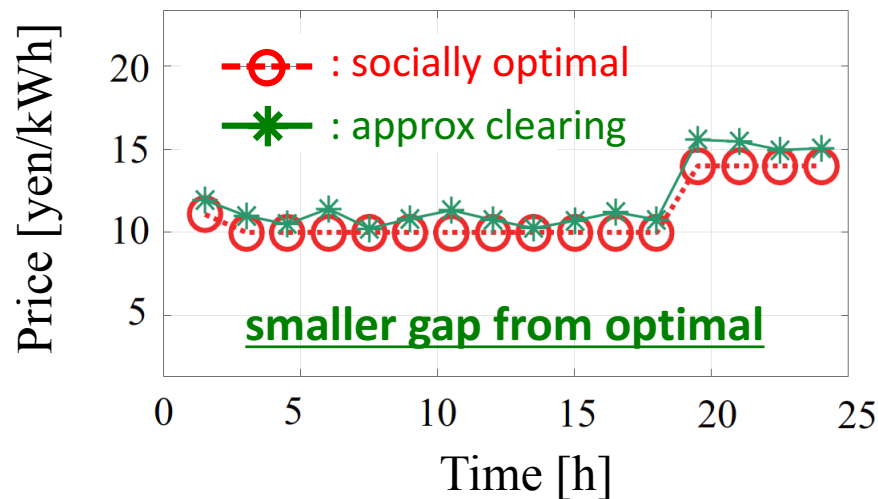
## SOC of batteries



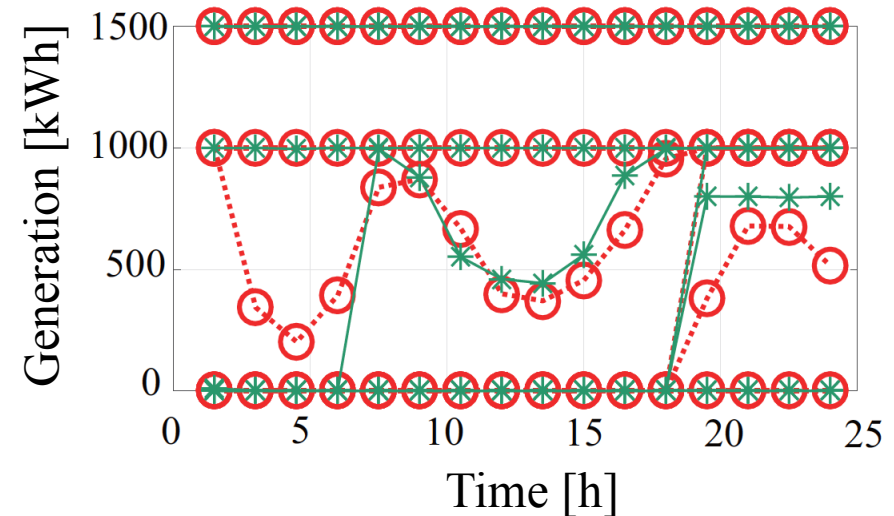


# Sequential Clearing in Fourier-Like Domain

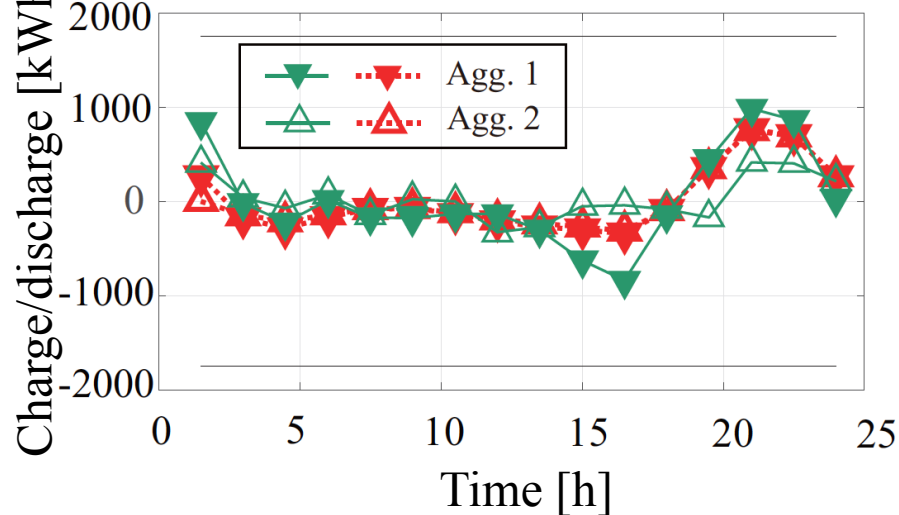
## Clearing price



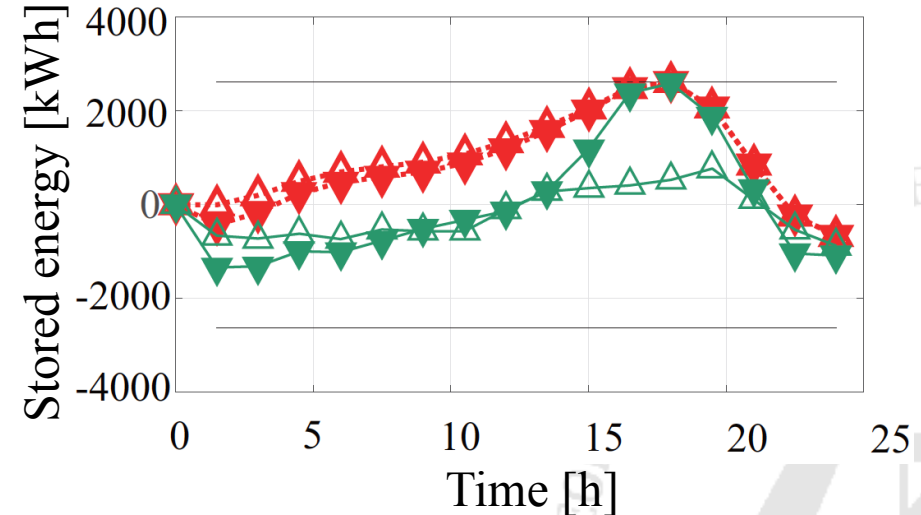
## Generators (9 types)



## Charge/discharge of batteries



## SOC of batteries





# Concluding Remarks

- ▶ **Bidding system design for multiperiod electricity markets**
  - ▶ **Distributed algorithm design for convex optimization**
    - ▶ Each aggregator submits bidding curves to ISO
    - ▶ ISO finds clearing price and balancing amounts by bidding curves
- ▶ **Proposed approach to bidding system design**
  - ▶ Basis transformation compatible with energy shift markets
  - ▶ Sequential clearing scheme based on approximate bidding curves

A Distributed Scheme for Power Profile Market Clearing under High Battery Penetration, IFAC WC 2017

Bidding System Design for Multiperiod Electricity Markets: Pricing of Stored Energy Shiftability, CDC 2017 (to appear)

## Thank you for your attention!