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Optimal Power Dispatch of Power Networks with Potential Games



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FL11-11-3
15th, July, 2011

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Backgrounds




Smart Grid

- Two-way energy management system between supply and demand side
- Use of **renewable energy** (ex. solar, wind)
 - Environmental conservation
 - Alternative energy of conventional energy
 - Decentralization of energy

Outputs depend largely on weather

- ➔ It is **more difficult** to satisfy supply-demand balance
- ➔ Centralized and Distributed generations (renewable, battery) need to allocate electricity to customers in a coordinated way

No design policy of optimal power dispatch...

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Objective (Game-theoretic Control)

Game-theoretic approach

- Agents are self-interested
 - ➔ **Non-cooperative game**
- The solution to the problem = the equilibrium of the game

Applications : resource allocation, sensor coverage

Advantages


- robustness to failures and environmental disturbances
- guarantee global convergence
- improved scalability

➔ **Potential game** has

- Design policies of objective function
- Learning algorithms

Objective of this work

To apply potential game to optimal power dispatch problem of power network



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Definition of Game

- Player set $N = \{1, \dots, n\}$
- Collection of action sets $\mathcal{A} = \mathcal{A}_1 \times \dots \times \mathcal{A}_n$
 - Agent i 's action set : \mathcal{A}_i
 - Agent i 's action : $\mathbf{a}_i \in \mathcal{A}_i$
- Collection of objective function $U = \{U_1, \dots, U_n\}$
 - Agent i 's objective function $U_i : \mathcal{A} \rightarrow \mathbb{R}$
 - Every agent chooses \mathbf{a}_i to maximize U_i

➔ **Game $G = \langle N, \mathcal{A}, U \rangle$**

Nash equilibrium

A pure Nash equilibrium is an action $\mathbf{a}^* \in \mathcal{A}$ such that $\forall i \in N$

$$U_i(\mathbf{a}_i^*, \mathbf{a}_{-i}^*) = \max_{\mathbf{a}_i \in \mathcal{A}_i} U_i(\mathbf{a}_i, \mathbf{a}_{-i}^*)$$

$(\mathbf{a}_{-i} = (\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_{i-1}, \mathbf{a}_{i+1}, \dots, \mathbf{a}_n))$

(Ex.) Payoff Matrix

		Player 2	
		NE A	B
Player 1	A	(2,2)	(1,0)
	B	(0,1)	(4,4)
		NE	

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Potential Game

- **Potential function (global objective function)** $\phi : \mathcal{A} \rightarrow \mathbb{R}$
 - $\phi : \max$ ➔ Objective of a group is achieved

Potential game

A game $G = \langle N, \mathcal{A}, U \rangle$ is a potential game if there is a potential function ϕ such that $\forall i \in N, \forall \mathbf{a}_{-i} \in \mathcal{A}_{-i}$ and $\forall \mathbf{a}_i^*, \mathbf{a}_i'' \in \mathcal{A}_i$,

$$U_i(\mathbf{a}_i^*, \mathbf{a}_{-i}) - U_i(\mathbf{a}_i'', \mathbf{a}_{-i}) = \phi(\mathbf{a}_i^*, \mathbf{a}_{-i}) - \phi(\mathbf{a}_i'', \mathbf{a}_{-i})$$

Key property

- the guaranteed existence of a Nash equilibrium
- **Local maxima of ϕ are Nash equilibria**

➔ Application of appropriate learning algorithms maximize ϕ

(Ex.) Payoff Potential

		A		B	
		A	B	A	B
A		(2,2)	(1,0)	2	0
	B	(0,1)	(4,4)	0	3
				Global objective	

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Outline

- Introduction
- **Optimal Power Dispatch Problem**
- Optimal Power Dispatch Game
- Conclusions

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Power Networks

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◆ Node: $\mathcal{V} = \mathcal{B} \cup \mathcal{R} \cup \mathcal{D}$

Battery: $\mathcal{B} = \{B_i | i = 1, \dots, n_B\}$

Renewable energy: $\mathcal{R} = \{R_i | i = 1, \dots, n_R\}$

Demand: $\mathcal{D} = \{D_i | i = 1, \dots, n_D\}$

◆ Link: $\mathcal{E} = \mathcal{V} \times \mathcal{V}$

Energy flow: $\mathcal{E}_P \rightarrow$ Graph: $G = (\mathcal{V}, \mathcal{E})$

Information flow: $\mathcal{E}_I \dashrightarrow$

Battery dynamics:

$$\Sigma_{\mathcal{B}}: b_i(k) = b_i(k-1) - \sum_{j \in \mathcal{N}_{DB,i}} u_{bij}(k), i \in \mathcal{B}$$

battery level $u_{bij} < 0$: charge

$\mathcal{N}_{DB,i}$: set of D_j can be supplied by B_i

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Power Networks

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Generator

It is difficult to satisfy supply-demand balance by only renewable energy and battery.

Generator adjust supply-demand balance:

$$\Rightarrow d_j(k) = u_{gj}(k) + \sum_{i \in \mathcal{N}_{RD,j}} u_{rij}(k) + \sum_{i \in \mathcal{N}_{BD,j}} u_{bij}(k)$$

Supply of generator $\mathcal{N}_{RD,j}$: set of D_j can be supplied by R_i

Supply of renewable energy and battery $\mathcal{N}_{BD,j}$: set of D_j can be supplied by B_i

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Optimal Power Dispatch Problem

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Assessment functions

- Maximize supply from renewable energy:
$$J_R = \sum_{i \in \mathcal{R}} w_{R,i} \left(r_i(k) - \sum_{j \in \mathcal{N}_{DR,i}} u_{rij}(k) \right)^2$$
- Remaining battery level = desired level:
$$J_B = \sum_{i \in \mathcal{B}} w_{B,i} (b_{i,rref} - b_i(k))^2$$
- Minimize electricity prices (fuel cost):
$$J_D = \sum_{j \in \mathcal{D}} w_{D,j} w_{D,j}^2(k)$$

$$= \sum_{j \in \mathcal{D}} w_{D,j} \left(d_j(k) - \left(\sum_{i \in \mathcal{N}_{RD,j}} u_{rij}(k) + \sum_{i \in \mathcal{N}_{BD,j}} u_{bij}(k) \right) \right)^2$$

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Optimal Power Dispatch Problem

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Constraint conditions

- Supply of renewable energy: $u_{rij}(k) \geq 0 \quad \sum_{j \in \mathcal{N}_{DR,i}} u_{rij} \leq r_i$
- Supply of renewable energy: $u_{gj}(k) > 0 \leftarrow$ running cost

$$d_j(k) = u_{gj}(k) + \sum_{i \in \mathcal{N}_{RD,j}} u_{rij}(k) + \sum_{i \in \mathcal{N}_{BD,j}} u_{bij}(k)$$

$$\Rightarrow \sum_{i \in \mathcal{N}_{RD,j}} u_{rij}(k) + \sum_{i \in \mathcal{N}_{BD,j}} u_{bij}(k) < d_j(k)$$

- Battery capacity: $b_i(k) \in [0, B_{i,max}]$
- Battery's charging and discharge rate:
$$\left| \sum_{j \in \mathcal{N}_{DB,i}} u_{bij}(k) \right| \leq \delta b_i$$

Objective: minimize each assessment function under constraint conditions

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Outline

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- Introduction
- Optimal Power Dispatch Problem
- Optimal Power Dispatch Game**
- Conclusions

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Optimal Power Dispatch Game

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- Player set $N = \{1, \dots, n\}$
Renewable energy, Battery
- Collection of action sets $\mathcal{A} = \mathcal{A}_1 \times \dots \times \mathcal{A}_n$
Supply of renewable energy and battery: u_r, u_b
- Collection of objective function $U = \{U_1, \dots, U_n\}$
- Feasible action sets $R(a) = R_1(a) \times \dots \times R_n(a), R_i(a) \subseteq \mathcal{A}_i$
Constraint conditions
- State $X = \{X_1, \dots, X_n\}$
Battery level b
- Transition of states $P: \mathcal{A} \times X \rightarrow X$
Battery dynamics: $b_i(k) = b_i(k-1) - \sum_{j \in \mathcal{N}_{DB,i}} u_{bij}(k)$

$$\Rightarrow \text{Strategic game } \Gamma = \{N, \mathcal{A}, U, R, X, P\}$$

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Optimal Power Dispatch Game

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Design of each agent's objective function

Potential game at state \mathbf{x}

A game is a potential game at state \mathbf{x} if there is a potential function ϕ such that $\forall i \in N, \forall a_{-i} \in \mathcal{A}_{-i}$ and $\forall a_i' \in \mathcal{A}_i$,

$$U_i(a_i', a_{-i}, \mathbf{x}) - U_i(a_i, \mathbf{x}) = \phi(a_i', a_{-i}, \mathbf{x}) - \phi(a_i, \mathbf{x})$$

$(a_{-i} = (a_1, a_2, \dots, a_{i-1}, a_{i+1}, \dots, a_n))$

➔ At each state \mathbf{x} , optimal power dispatch game has Nash equilibriums

Nash equilibrium at state \mathbf{x}

A Nash equilibrium is an action $a^* \in \mathcal{A}$ such that $\exists \mathbf{x}$,

$$U_i(a_i^*, a_{-i}^*, \mathbf{x}) = \max_{a_i \in \mathcal{A}_i} U_i(a_i, a_{-i}^*, \mathbf{x}), \forall i \in N$$

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Optimal Power Dispatch Game

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Agent : Renewable energy, Battery Action: $\mathbf{u}_r, \mathbf{u}_b$

Potential function (global objective)

$$\phi(\mathbf{u}_r, \mathbf{u}_b, \mathbf{b}) = -q_D J_D - q_R J_R - q_B J_B$$

Each agent's objective function: q_D, q_R, q_B : weight

Renewable energy $\mathcal{R}_i \in \mathcal{R}$:

$$U_{R_i} = - \sum_{j \in N_{DR,i}} w_{D,j} \left(d_j(k) - \left(\sum_{l \in N_{RD,j}} a_{r,lj}(k) + \sum_{l \in N_{BD,j}} u_{bl}(k) \right) \right)^2 - w_{B,i} \left(r_i(k) - \sum_{j \in N_{DR,i}} a_{r,lj}(k) \right)$$

Battery $\mathcal{B}_i \in \mathcal{B}$:

$$U_{B_i} = - \sum_{j \in N_{DB,i}} w_{D,j} \left(d_j(k) - \left(\sum_{l \in N_{RD,j}} a_{r,lj}(k) + \sum_{l \in N_{BD,j}} u_{bl}(k) \right) \right)^2 - w_{B,i} (b_{i,rnf} - b_i(k))^2$$

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Optimal Power Dispatch Game: Example

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Assessment functions

 $J_D = (3 - u_r - u_b)^2 \quad (d = 3)$
 $J_R = (2 - u_r)^2 \quad (r = 2)$
 $J_B = (1 - (b - u_b))^2 \quad (b_{ref} = 1)$

Payoff matrix Renewable energy

		0	1	2
Battery	$b = 0$			
	0	(-91,-94)	(-41,-41)	(-11,-10)
	-1	(-160,-164)	(-90,-91)	(-40,-40)
		0	1	2
1	(-41,-44)	(-11,-11)	(-1,0)	
0	(-90,-94)	(-40,-41)	(-10,-10)	
-1	(-161,-164)	(-91,-91)	(-41,-40)	
		0	1	2
1	(-40,-44)	(-10,-11)	(0,0)	
0	(-91,-94)	(-41,-41)	(-11,-10)	
-1				

Potential function

 $\phi = -10J_D - J_R - J_B$

Utility

R : $U_R = -10J_D - J_R$

B : $U_B = -10J_D - J_B$

Actions

R : $u_r = \{0, 1, 2\}$

B : $u_b = \{-1, 0, 1\}$

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Optimal Power Dispatch Game: Example

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Best response

$a_i = \arg \max_{a_i \in \mathcal{A}_i} U_i(a_i, a_{-i}, b)$ Renewable energy

		0	1	2
Battery	$b = 0$	1	2	3
	0	①(-91,-94)	②(-41,-41)	③(-11,-10)
	-1	(-160,-164)	(-90,-91)	(-40,-40)
		0	1	2
1	(-41,-44)	(-11,-11)	(-1,0)	
0	④(-90,-94)	⑤(-40,-41)	⑥(-10,-10)	
-1	(-161,-164)	(-91,-91)	(-41,-40)	
		0	1	2
1	(-40,-44)	(-10,-11)	(0,0)	
0	(-91,-94)	⑧(-41,-41)	⑨(-11,-10)	
-1				

Equilibrium

$[u_r^*, u_b^*, b^*] \Rightarrow [u_r^*, 0, b^*] ?$

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Convergence Analysis: Renewable energy's action

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$$U_{R_i} = - \sum_{j \in N_{DR,i}} w_{D,j} \left(d_j(k) - \left(\sum_{l \in N_{RD,j}} a_{r,lj}(k) + \sum_{l \in N_{BD,j}} u_{bl}(k) \right) \right)^2 - w_{B,i} \left(r_i(k) - \sum_{j \in N_{DR,i}} a_{r,lj}(k) \right)$$

- Renewable energy's action is **not affected by battery state**.

$$\frac{\partial U_{R_i}}{\partial a_{r,lj}} = 2 \sum_{j \in N_{DR,i}} w_{D,j} \left(d_j - \left(\sum_{l \in N_{RD,j}} a_{r,lj} + \sum_{l \in N_{BD,j}} u_{bl} \right) \right) + 2w_{B,i} \left(r_i - \sum_{j \in N_{DR,i}} a_{r,lj} \right)$$

$$\frac{\partial U_{R_i}}{\partial a_{r,lj}} > 0 \quad (* \text{ constraint conditions})$$

- Renewable energy **takes action to increase his supply**.
- Renewable energy's action has **upper limit**: $\sum_{j \in N_{DR,i}} a_{r,lj} \leq r_i$

➔ Renewable energy \mathcal{R} 's action change is finite.

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Convergence Analysis: Battery's action

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Assumption

Battery \mathcal{B}_i 's action set: $\mathcal{A}_{B_i} = \{-\delta b_i, 0, \delta b_i\}$, $\left| \sum_{j \in N_{DB,i}} u_{bj} \right| = \delta b_i$

$$U_{B_i} = - \sum_{j \in N_{DB,i}} w_{D,j} \left(d_j(k) - \left(\sum_{l \in N_{RD,j}} a_{r,lj}(k) + \sum_{l \in N_{BD,j}} u_{bl}(k) \right) \right)^2 - w_{B,i} (b_{i,rnf} - b_i(k))^2$$

- Battery's action is **not affected by other battery's state**.
- Battery **doesn't take action to restore battery's state**.

proof

Assumption:

$$U_{B_i}(0, b_i) < U_{B_i}(u_{bij}, b_i - \delta b_i) \quad \dots \textcircled{1}$$

$$U_{B_i}(0, b_i - \delta b_i) < U_{B_i}(-u_{bij}, b_i) \quad \dots \textcircled{2}$$

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Convergence Analysis: Battery's action

$\Delta = d_j - \sum_{i \in \mathcal{N}_{RD,j}} w_{i,j} \quad \Theta = b_{i,r,j} - b_i$

$\Rightarrow U_{Bj} = - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j}} w_{i,j} \right)^2 - w_{B,j} \Theta^2$

①: $- \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j}} w_{i,j} \right)^2 - w_{B,j} (\Theta + \delta b_i)^2$

$> - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} \right)^2 - w_{B,j} \Theta^2 \dots \textcircled{3}$

②: $- \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} + w_{i,j} \right)^2 - w_{B,j} \Theta^2$

$> - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} \right)^2 - w_{B,j} (\Theta + \delta b_i)^2 \dots \textcircled{4}$

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Convergence Analysis: Battery's action

③, ④: $2 \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} \right)^2$

$> \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j}} w_{i,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} + w_{i,j} \right)^2$

$\Rightarrow \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} \right)^2$

$> \frac{\sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j}} w_{i,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{RD,j} \setminus \{i\}} w_{i,j} + w_{i,j} \right)^2}{2}$

Contradiction! (*: property of convex function)

\therefore Battery **doesn't** take action to restore battery's state.

Battery has constraint: $b_i(k) \in [0, B_{i,max}]$
 $\mathcal{A}_{Bj} = \{-\delta b_i, 0, \delta b_i\}$

\Rightarrow Battery **B's** action change is finite.

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Convergence Analysis: Optimal Power Dispatch Game

- ϕ, U_{Ri}, U_{Bi}
- Best response: $a_i = \arg \max_{a_i \in \mathcal{A}_i} U_i(a_i, a_{-i}, b)$
- Renewable energy \mathcal{R} 's action change is finite.
- Battery \mathcal{B}_i 's action set: $\mathcal{A}_{Bj} = \{-\delta b_i, 0, \delta b_i\}$
- Battery \mathcal{B} 's action change is finite.

\Rightarrow Optimal power dispatch game $\Gamma = \{N, \mathcal{A}, U, R, X, P\}$ has a stationary state (equilibrium).

Stationary state (equilibrium)

$\left[\begin{array}{l} [u_r^*, u_b^*, b^*] \Rightarrow [u_r^*, 0, b^*] \\ \frac{\partial \phi}{\partial u_b} (u_r^*, 0, b^*) = 0 \end{array} \right.$

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Conclusion

Summary

- \triangleright Definition of optimal power dispatch game
- \triangleright Analysis of optimal power dispatch game
Convergence to equilibrium

Future Works

- \triangleright Analysis of different configuration of power networks
- \triangleright Simulations

Check $\left[\begin{array}{l} \mathcal{B}'s \text{ and } \mathcal{R}'s \text{ actions} \\ \text{equilibrium} \end{array} \right.$

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Appendix

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Convergence Analysis: Battery's action

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$\Delta = d_j - \sum_{i \in \mathcal{N}_{DB,j}} w_{i,j} \quad \Theta = b_{i,r,j} - b_i$

$\Rightarrow U_{Bj} = - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 - w_{B,j} \Theta^2$

①: $- \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 - w_{B,j} \left(\Theta + \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2$
 $> - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2 - w_{B,j} \Theta^2 \dots \textcircled{3}$

②: $- \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} + w_{M,j} \right)^2 - w_{B,j} \Theta^2$
 $> - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2 - w_{B,j} \left(\Theta + \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 \dots \textcircled{4}$

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Convergence Analysis: Battery's action

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③: $- \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2$
 $> - w_{B,j} \Theta^2 + w_{B,j} \left(\Theta + \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 \dots \textcircled{5}$

④: $- w_{B,j} \Theta^2 + w_{B,j} \left(\Theta + \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2$
 $> - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} + w_{M,j} \right)^2$
 $\dots \textcircled{6}$

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Convergence Analysis: Battery's action

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⑤, ⑥: $- \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2$
 $> - \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} + w_{M,j} \right)^2$

$2 \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2$

$> \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} + w_{M,j} \right)^2$

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Convergence Analysis: Battery's action

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$2 \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2$
 $> \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} + w_{M,j} \right)^2$

$\Rightarrow \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} \right)^2$
 $> \frac{\sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j}} w_{M,j} \right)^2 + \sum_{j \in \mathcal{N}_{DB,j}} w_{D,j} \left(\Delta - \sum_{i \in \mathcal{N}_{DB,j} \setminus \{j\}} w_{M,j} + w_{M,j} \right)^2}{2}$

Contradiction! (*, property of convex function)

\therefore Battery doesn't take action to restore battery's state.

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Architecture for Potential Games and Utility Designs

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Utility Design \rightarrow Cost sharing methodologies

Wonderful Life
Shapley Value
Weighted Shapley Value

Potential Games

Gradient play
Log-linear learning
Learning Design

Player set $S \subseteq N, i \in S$

- Wonderful Life Utility (WLU)
 $f_r(i, S) = W_r(S) - W_r(S \setminus \{i\})$
- The Shapley Value (SV)
 $f_r(i, S) = \sum_{T \subseteq S \setminus \{i\}} \omega_T (W_r(S) - W_r(S \setminus \{i\}))$
 $\omega_T = \frac{|T|!(|S| - |T|)!}{|S|!}$
- The Weighted Shapley Value (WSV)

Distribution Rule	Existence of Equilibrium	Potential Game	Budget Balanced	Tractable	Informational Requirement
WLU	yes	yes	no	yes	Medium
SV, WSV	yes	yes	yes	no	High

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State-Based Non-Cooperative Design

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State-based games state $x(t) \in X$

- Utility function $U_i(a_i, a_{-i}, x)$
- State transition function $P: A \times X \rightarrow \Delta(X)$: the set of probability distribution over X
- Each player selects an action to maximize his expected utility
 $\therefore a_i(t) \in \arg \max_{a_i(t) \in A_i} \mathbf{E}[U_i(a(t), x(t))]$
- $x(t+1)$ is chosen randomly according to $P(a(t), x(t)) \in \Delta(X)$

State-based Nash Equilibrium $[a^*, x^*]$
 for every x^j in the support of $P(a^*, x^*)$,
 $U_i(a_i^*, a_{-i}^*, x^j) = \max_{a_i \in A_i} U_i(a_i, a_{-i}^*, x^j)$

State-based Potential games
 $\exists \phi: A \rightarrow \mathbb{R}$ such that $a_i^j \in A_i$,
 $U_i(a_i^j, a_{-i}, x) - U_i(a_i, x) > 0 \Rightarrow \phi(a_i^j, a_{-i}) - \phi(a) > 0$

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Priority-Based Distribution Rule

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state = priority \Rightarrow **Priority-based**

x_i^r : the priority of player i at resource r $\left\{ \begin{array}{l} x_i^r < x_j^r : i \text{ has higher priority than } j \\ x_i^r = 1 : \text{top priority} \end{array} \right.$

State dynamics : first in first out (FIFO)

- Multiple players seek to join a resource simultaneously
 \Rightarrow The order of the entering players is randomly chosen
- $a(t) = a(t-1) \Rightarrow x(t+1) = x(t)$

Priority-based utility $U_i(a', x) = \mathbb{E}_{P(a', x)} V_i(a', x')$

Expectation which x' is chosen : $\mathbb{E}_{P(a', x)} \quad x'_i := \{j \in N : x_j^r \leq x_i^r\}$

Marginal contribution : $V_i(a, x) = \sum_{r \in a_i} (W^r(x_i^r) - W^r(x_i^r \setminus i))$



Distribution Rule	Budget Balanced	Tractable	PoS	PoA
Priority-Based	yes	yes	1	1/2

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