

Learning in Stackelberg Markov Games

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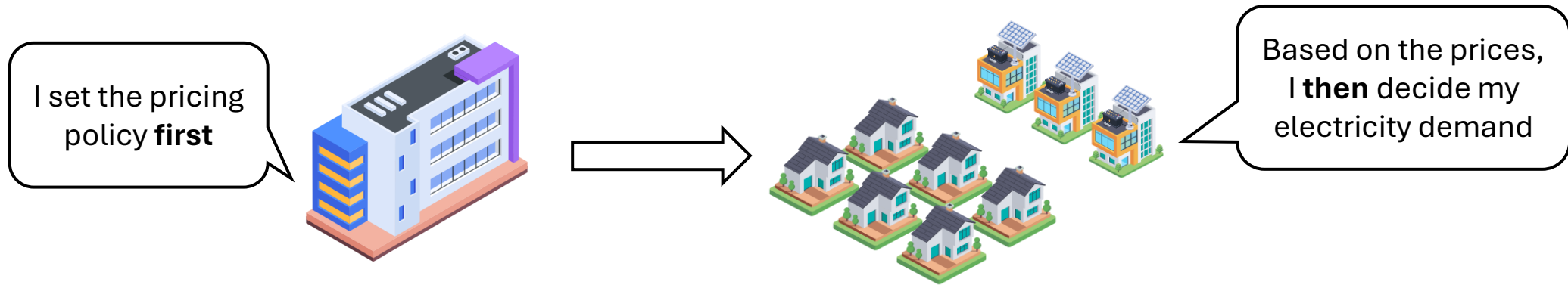
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Motivation

- Many real-world policy and market problems follow a leader-follower structure (Stackelberg Games).



- Dynamic, uncertain environments:
 - **Learning Algorithm:** for leader-follower best-response in Stackelberg Markov Games – dynamic leader-follower interactions with discounted rewards over infinite horizon.

Stackelberg Markov Games

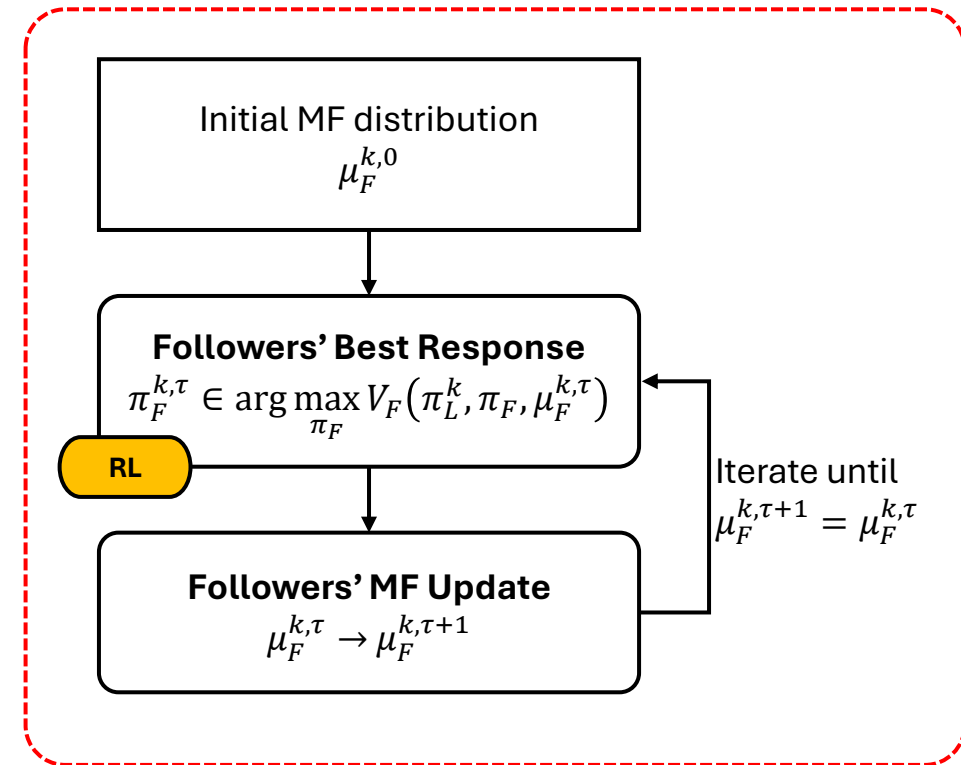
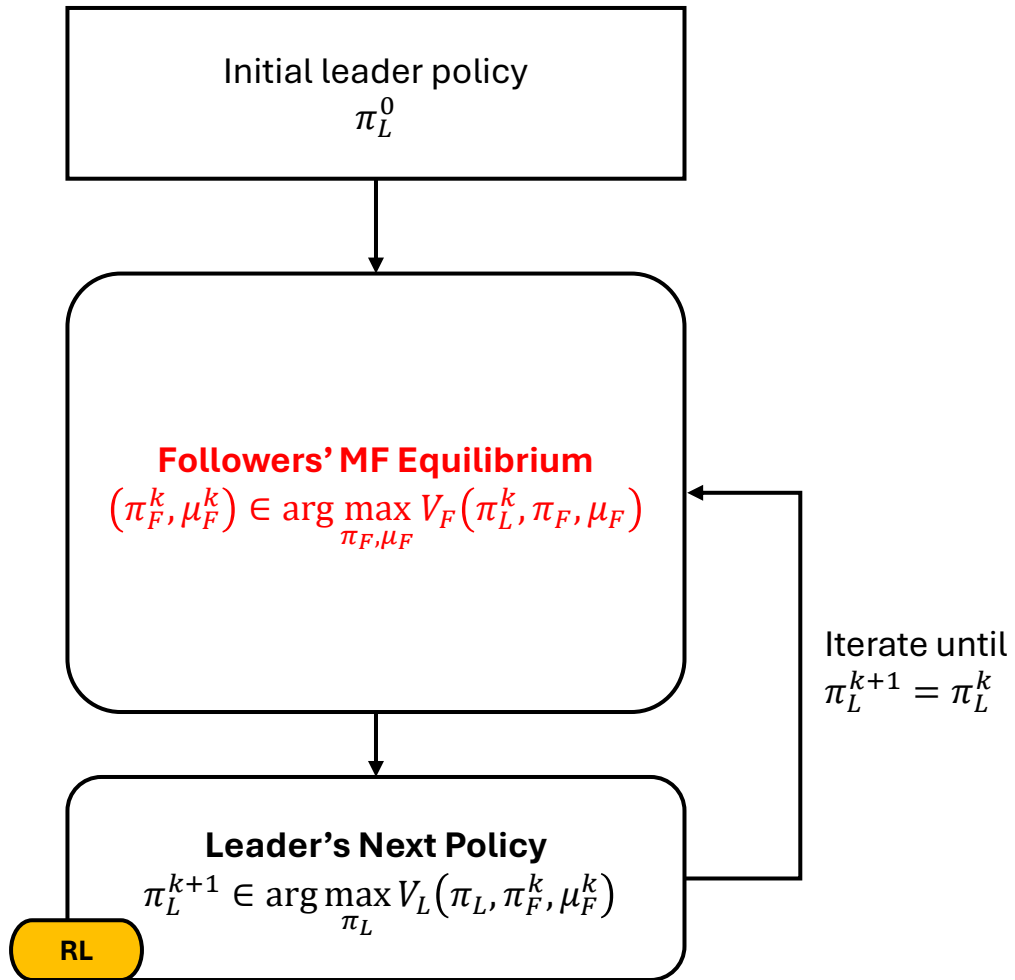
- Single Leader Single Follower:
 - Leader commits to a policy first: π_L
 - Follower then chooses a policy: π_F
- **Follower:** best response to any leader's policy:
 - Best response: $\text{BR}_F(\pi_L) := \arg \max_{\pi_F} V_F(\pi_L, \pi_F)$
- **Leader:** anticipates follower will take its best response to π_L
 - Leader's optimal policy $\pi_L^{\text{SSE}} \in \arg \max_{\pi_L} V_L(\pi_L, \text{BR}_F(\pi_L))$
 - As a result: $\pi_F^{\text{SSE}} \in \text{BR}_F(\pi_L^{\text{SSE}})$

**Stationary Stackelberg
Equilibrium (SSE)**
 $(\pi_L^{\text{SSE}}, \pi_F^{\text{SSE}})$

Extension to Infinite Followers

- Single Leader Infinite Follower
 - **Mean field (MF):** followers' states and actions have a distribution
 - **Followers:** reduce to a single representative agent $\leftrightarrow$ MF distribution μ_F
 - **Followers' MF equilibrium:** for any π_L , they stabilize at $(\pi_F, \mu_F)(\pi_L)$
 - **Stationary Stackelberg MF equilibrium (SS-MFE):** $(\pi_L^*, \pi_F^*, \mu_F^*)$

Learning Framework

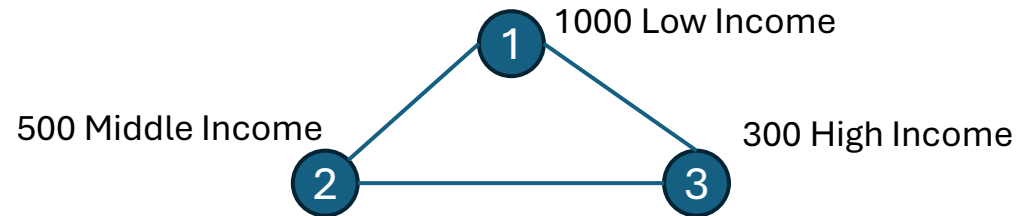


Numerical Experiment (3-Node Power Network)

- **Utility company (leader):** charge electricity rates
 - To minimize difference of **energy expenditure incidence (EEI)** between different income groups.

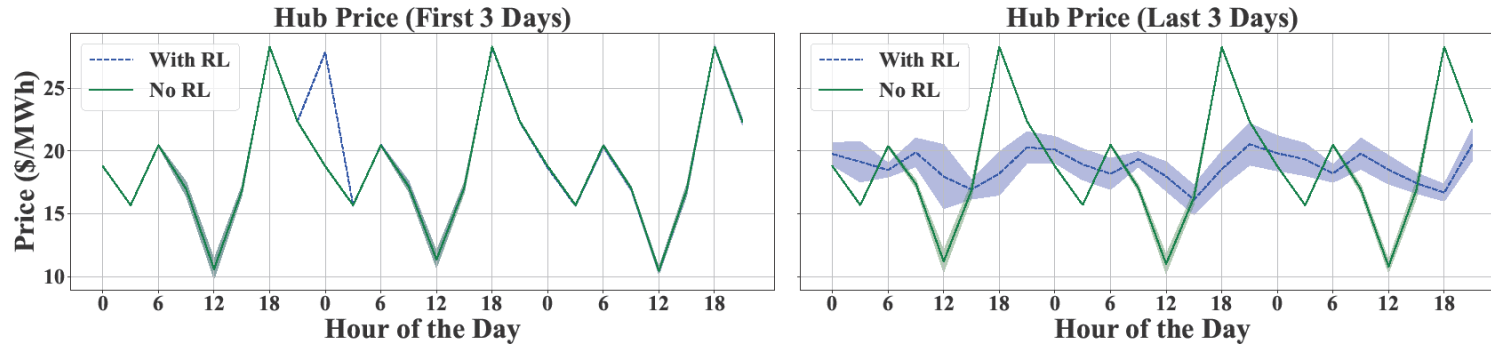
$$EEI = \text{Energy Spending} / \text{Income}$$

- **Prosumers (followers):** learns battery charge/discharge strategies based on price signals

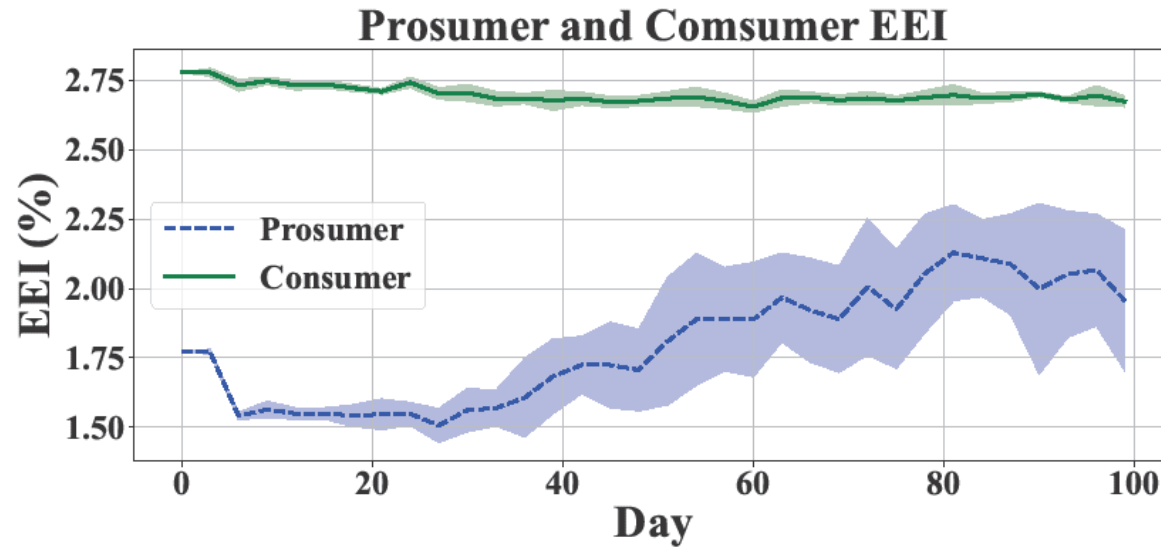


- **Consumers:** 3000 VERY Low Income per node
- RL Algorithm: PPO; Simulation: 100 Days

Results



Having renewables makes price changes less extreme



Ensures **equity** across income groups while **promoting renewable adoption**.

Thank you!