

Model Predictive Control of Industrial Loads and Energy Storage for Demand Response

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Introduction

- Demand response demonstrates potentials to enhance the power system's operational flexibility in a cost-effective way.
- Advantages of Industrial Loads in DR
 - **Infrastructure:** Most industrial loads have already installed the infrastructures for control, measurement, and communications which are necessary for demand response.
 - **Response:** The adjustments of the power consumption from many industrial loads can be very large, fast, and accurate.
 - **Economic Incentive:** The industrial loads are also strongly motivated to participate in demand response programs even at the cost of increasing their operation complexity, as making profits is their primary concern.
- Examples
 - smelting pots, furnaces, fans, freezers, pumps, mills, crushers, etc.
- Challenges for Industrial Demand Response
 - reliability, complexity, granularity
- Granularity Restriction
 - Most of these industrial loads can only provide power changes in a discrete manner, e.g. the power change is several MWs at a time.
 - This coarse granularity hinders those industrial loads from providing the most valuable ancillary services, as regulation and load following in the current electricity markets require a continuous change of power.
- **Research Objective**
 - Fully utilize the DR potentials from the industrial loads.
 - Overcome the granularity restriction.

Prediction

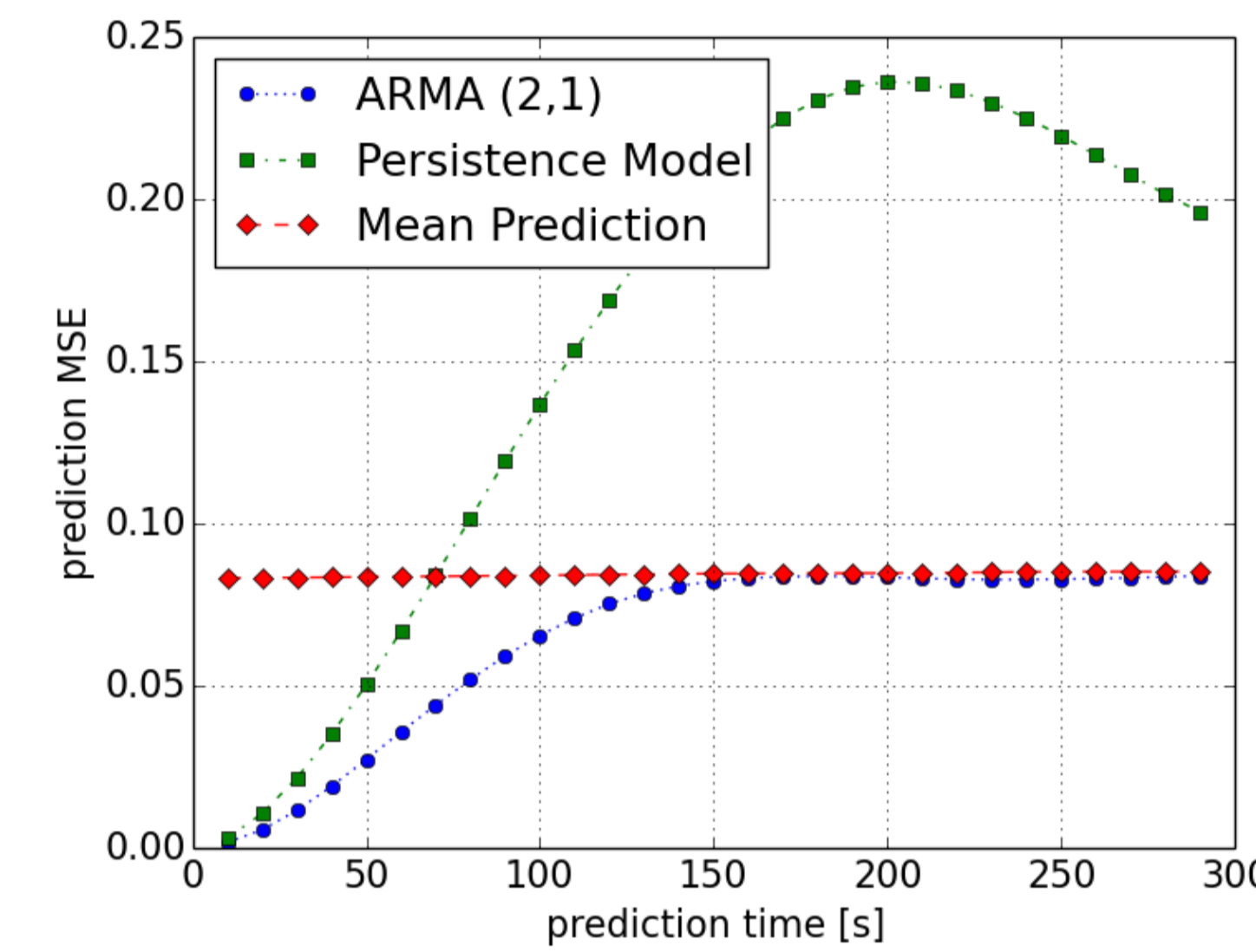


Figure 2: Prediction mean squared errors.

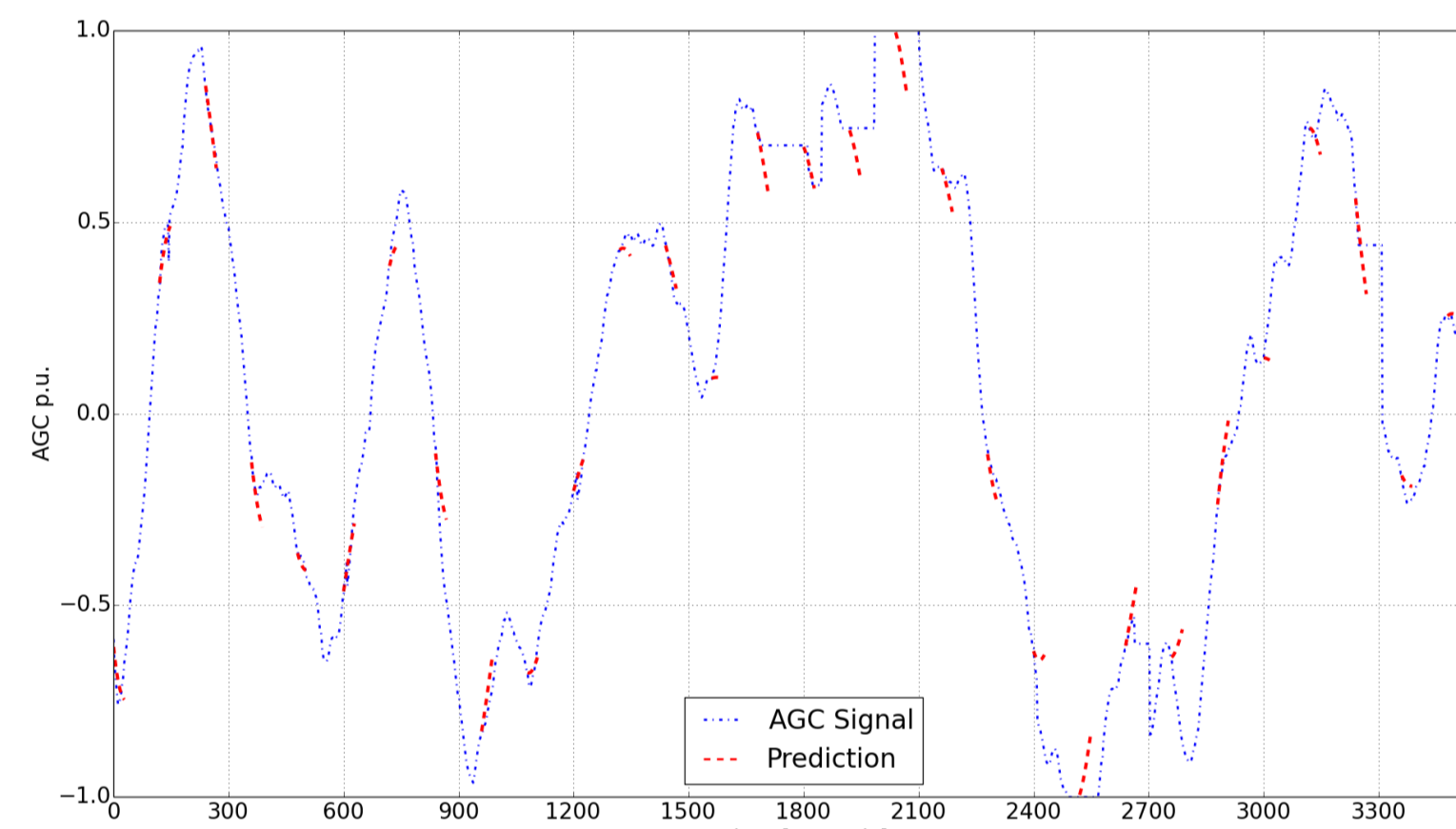


Figure 3: Regulation signal (AGC) over one hour and its prediction.

Case Study

- Simulation Setup
 - industrial machines: 4*2MW
 - on-site storage: $E = 1$ MWh, $P_s = 3$ MW
 - provide regulation $R = 6$ MW at baseline $B = 4$ MW
- Result
 - accurately provide a large range of regulation
 - over the hour: only 12 switches and 0.12 MWh violation integral

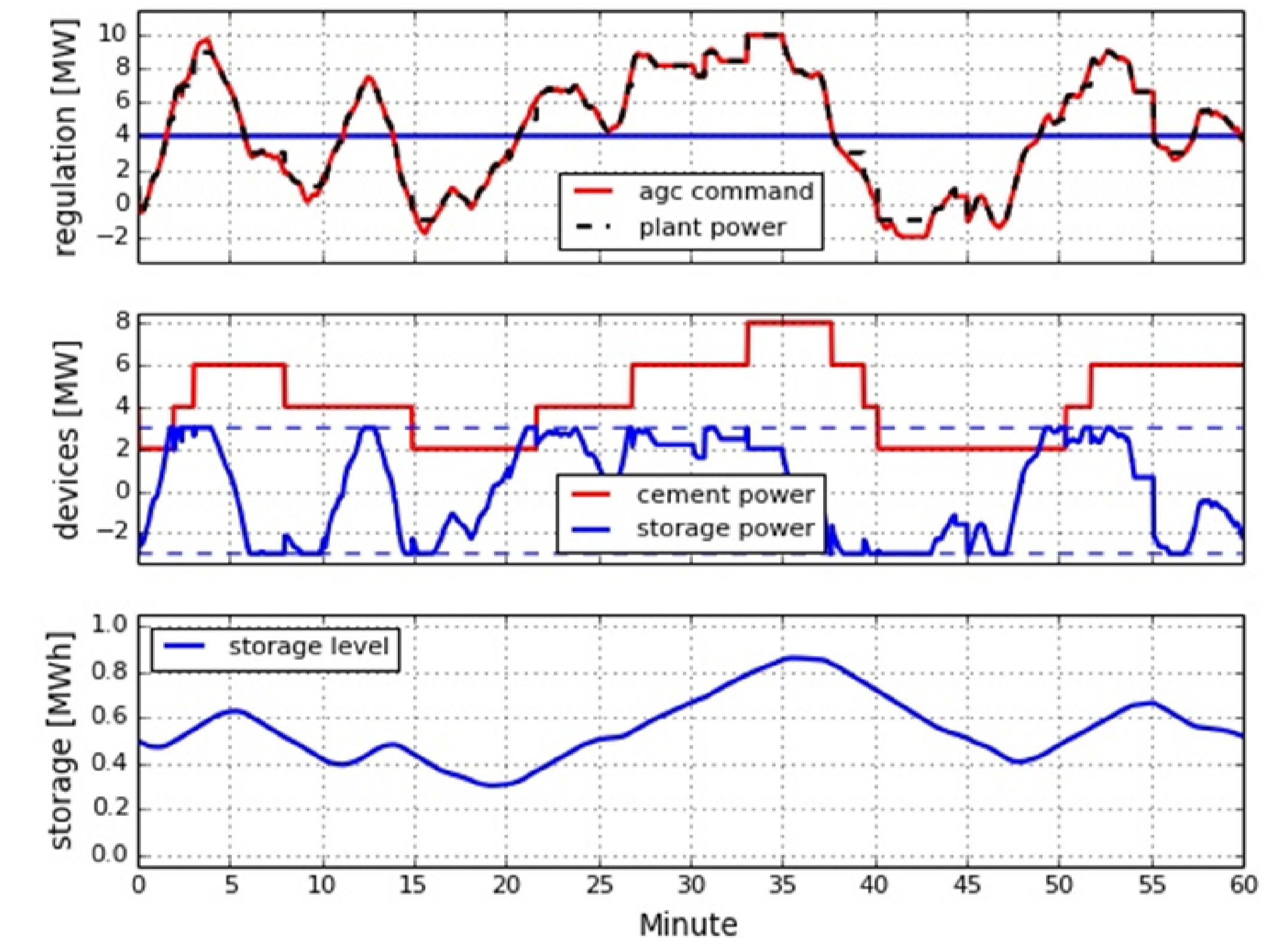


Figure 4: Simulation results with increased penalty on switch actions.

Coordination by Model Predictive Control

- Main Idea: support by on-site energy storage
 - industrial machines: **large/discrete** power change, **main body**
 - on-site storage: **fine/continuous** power change, **handle mismatch**

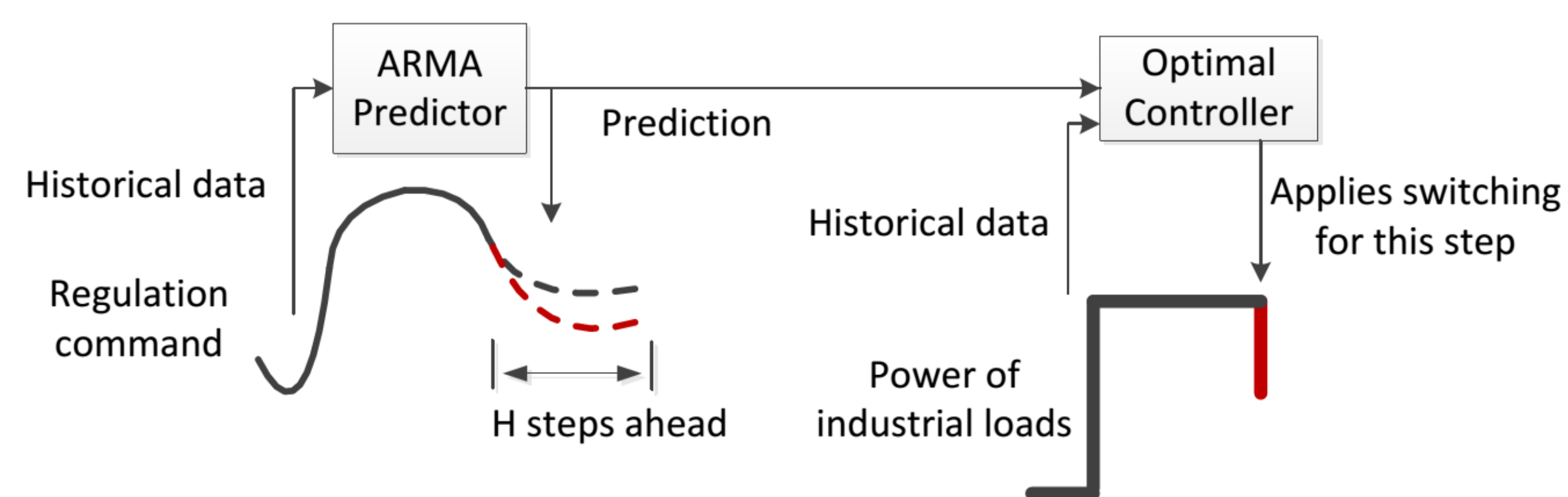


Figure 1: MPC coordination framework.

Optimal Control

- Objective

$$\text{minimize } \sum_{i \in H} (\alpha v_i + \beta s_i) + \gamma d \quad (1)$$
- Regulation Violation

$$v_{t+i} \geq |B + R\hat{w}_{t+i} - P_m x_{t+i} - y_{t+i}| \quad \forall i \in H \quad (2)$$
- Machine Switching

$$s_i \geq |x_{t+i} - x_{t+i-1}| \quad \forall i \in H \quad (3)$$
- Storage Level Deviation

$$d \geq |e_{t+H} - \bar{e}| \quad (4)$$
- Storage Energy Balance

$$e_{t+i} - e_{t+i-1} = y_{t+i} \delta \quad \forall h \in H \quad (5)$$
- Switching Limitation

$$\sum_{j=t+i-L}^{t-1} \tilde{s}_j + \sum_{j=t}^{t+i} s_j \leq \bar{s} \quad \forall i \in H \quad (6)$$
- Variable Ranges

$$x_{t+i} \in \{0, 1, \dots, n\} \text{ and } -P_s \leq y_{t+i} \leq P_s \quad \forall i \in H \quad (7)$$

Conclusion

- Fully Utilize Industrial Loads' DR Potentials
 - add more balancing resources to power system
 - encourage loads to be more active in DR
- MPC Coordination Framework
 - "the whole is greater than the sum of its parts"
 - many potential applications
 - e.g. the coordination among fast/slow generators, buildings, storage, ...

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- Relevant work available at <http://www.xiaozhang.work/>