



Toward Integrating Mathematical Programming and Machine Learning: Boosting Power System Operation Economics via Closed-Loop Predict-then-Optimize

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Closed-Loop Predict-then-Optimize (C-PO)

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Preliminaries
and Motivations

Closed-Loop Predict-then-Optimize Idea

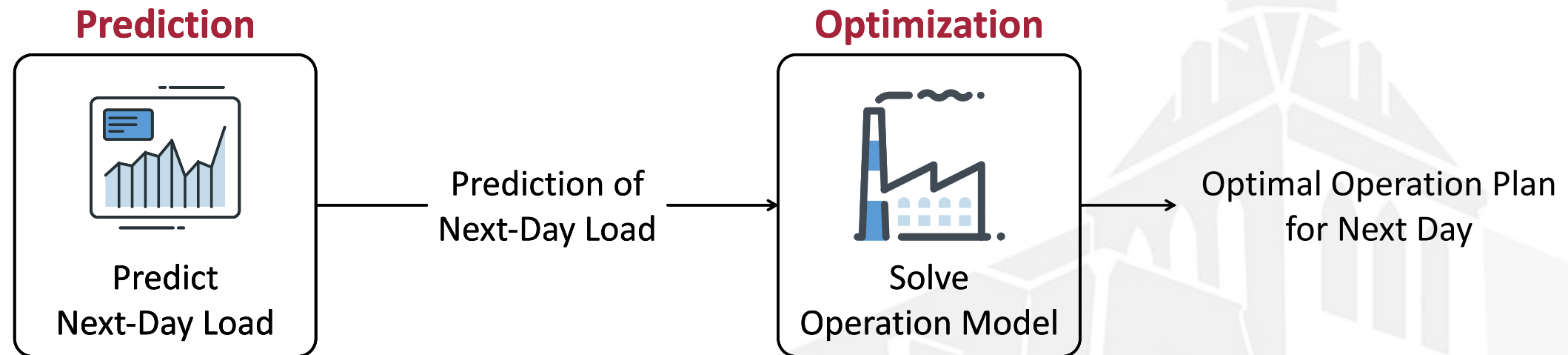
Case Studies

Summary



Preliminaries

- **Operations in Open-Loop Predict-then-Optimize (e.g., PSE&G around Hoboken):**



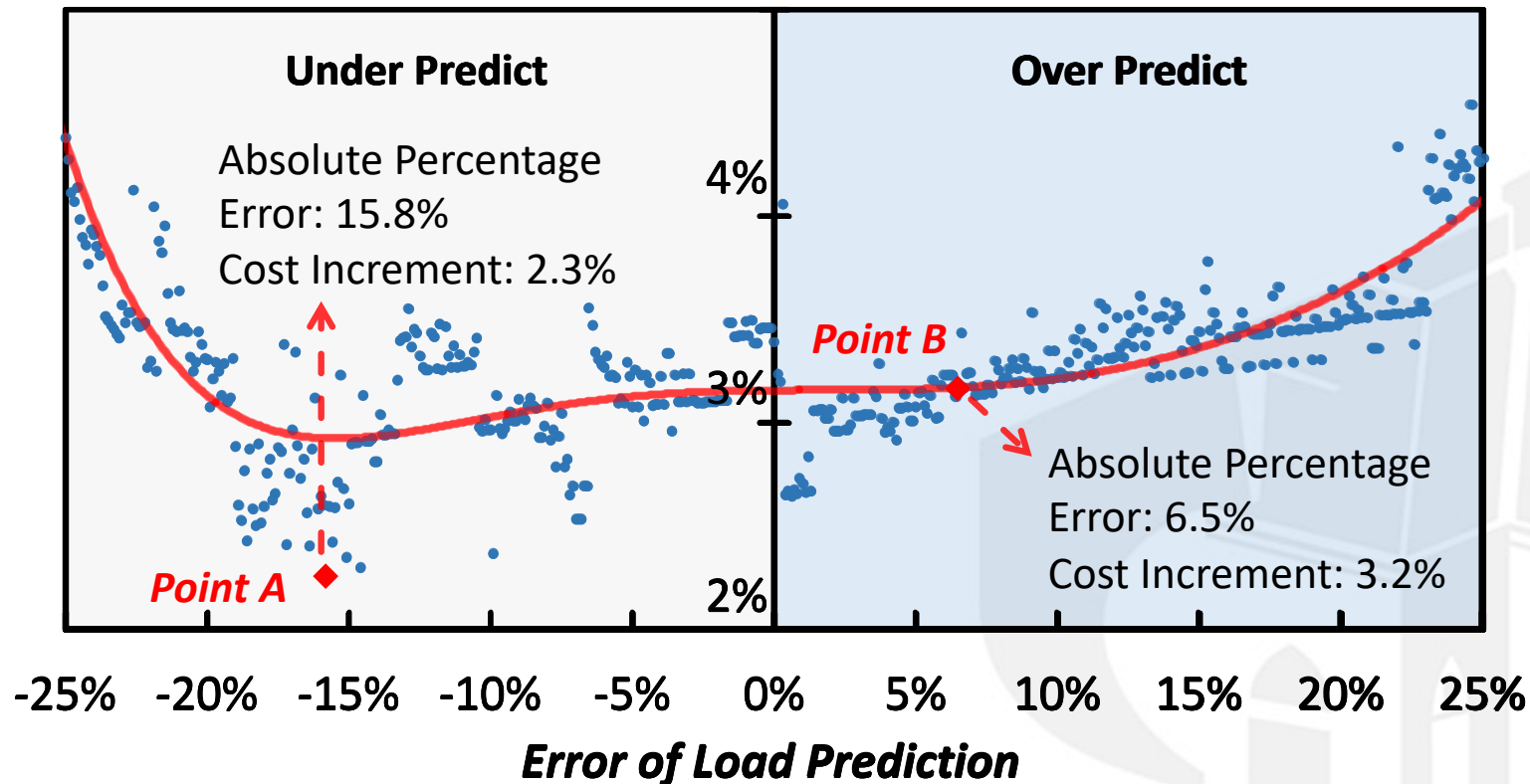
- **Step 1:** Machine learning *predicts* load demand as *accurately* as possible
- **Step 2:** Operation plans are *optimized* by solving mathematical models
- **Operator's Goal:** Minimum Operation Cost (*Lower Cost Means Better Economics*)



Motivations: Flaw in Open-Loop Predict-then-Optimize (O-PO)

- **More Accurate Prediction \nRightarrow Lower Operation Cost**

Operation Cost Increment Caused by Prediction Error (Lower is Better)



Point A vs Point B

Worse prediction error enables lower operation cost

Why?

Real-world operation problems are nonlinear

Accuracy-cost relationship is nonlinear

O-PO ignores this



Motivations: Flaws in Open-Loop Predict-then-Optimize (O-PO)

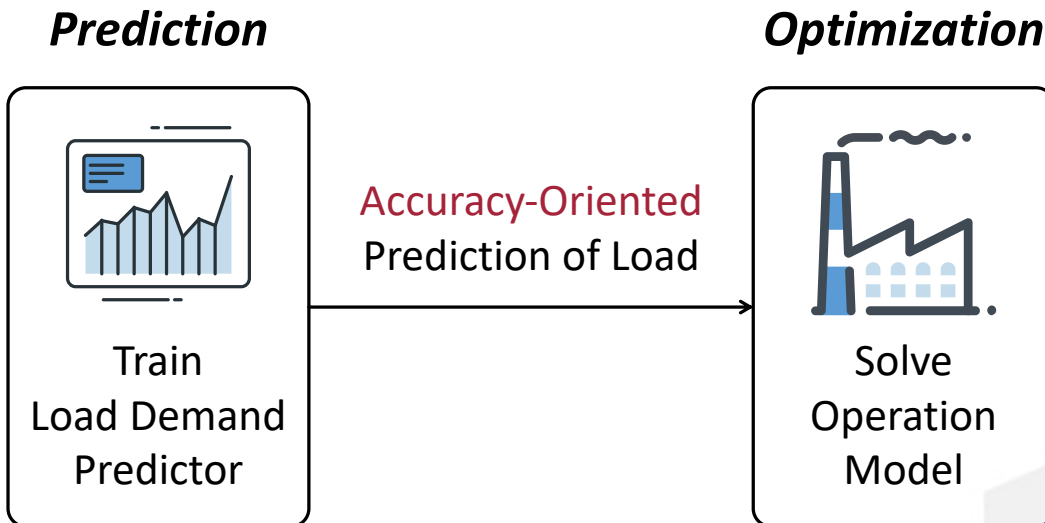
*“In many real-world applications, the **ultimate goal** is not to make good predictions, but rather to use the often noisy predictions to **make good decisions.**”*

----- Yoshua Bengio
in Using a Financial Training Criterion Rather than a Prediction Criterion,
1997



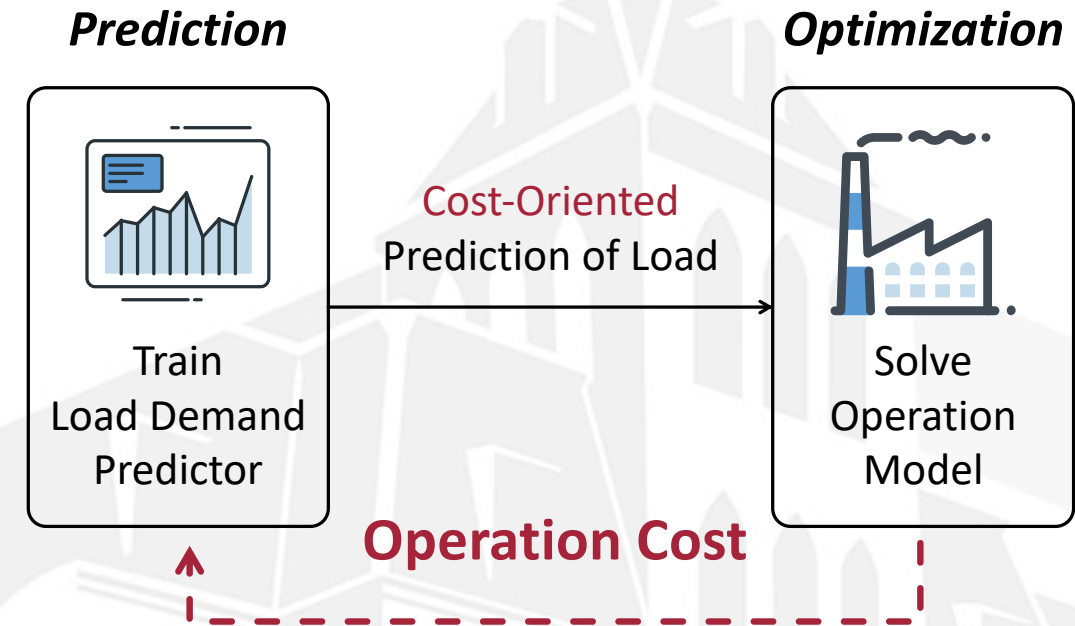
Closed-Loop Predict-then-Optimize Idea: What is It?

Open-Loop Predict-then-Optimize (O-PO)



- Train predictor using **accuracy criterion**
- **Open-loop** and **accuracy-oriented**

Closed-Loop Predict-then-Optimize (C-PO)



- Train predictor using **cost criterion**
- **Closed-loop** and **cost-oriented**



C-PO Idea: How to Close the Opened Loop?

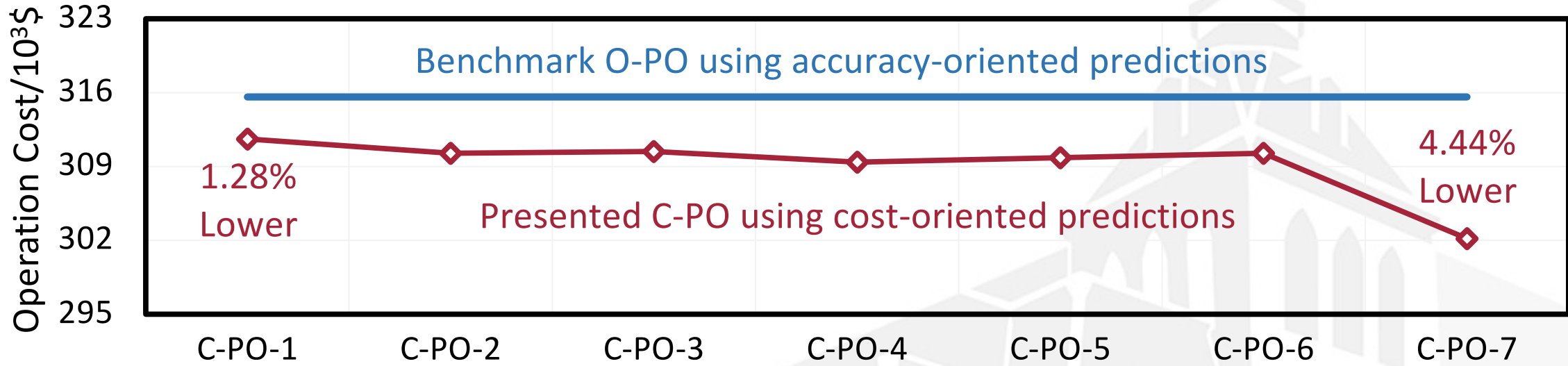
- ***Train Machine Learning-based Predictor \mathcal{P} via Cost-Oriented Loss Function (Don't Simply Pursue Prediction Accuracy!)***

$$\min_{\mathcal{P}} \frac{1}{S} \sum_{s=1}^S |\text{Operation Cost}(\mathcal{P})_s - \text{Operation Cost}_s^{\text{Perfect}}|$$

- Operation Cost(\mathcal{P})_s is the **cost induced by predictor \mathcal{P}** in scenario s
Operation Cost_s^{Perfect} is the **perfectly low cost** in scenario s
- Measure **operation cost increment** induced by predictor \mathcal{P}
- Predictor \mathcal{P} learns to generate cost-oriented predictions that can make the operation cost closer to its perfection



Case Studies: C-PO vs O-PO on Real-World Dataset



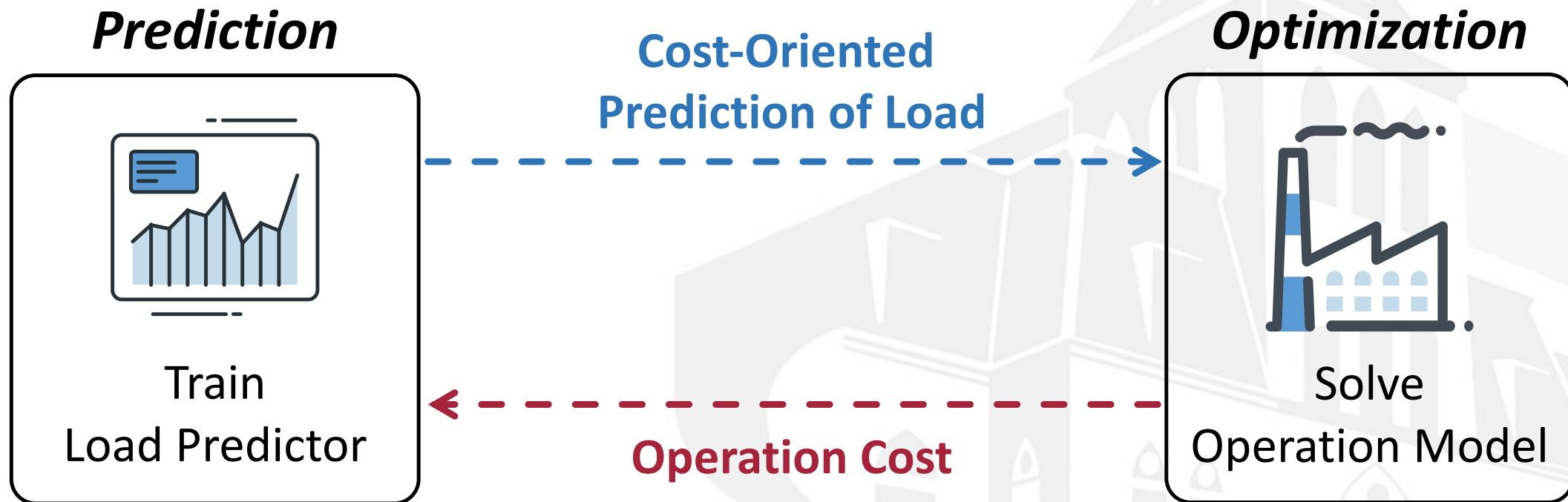
Type of Prediction	Mean Absolute Percentage Error (MAPE)	Root Mean Square Error (RMSE)
Accuracy-Oriented	39%	130MW
Cost-Oriented	34% (Better)	149MW (Worse)



Summary

- **Back to Essence: What is Closed-Loop Predict-then-Optimize?**

Idea that feeds the optimization back to the prediction for improving the optimization performance. **Don't simply pursue the accuracy!!! Pursue the ultimate goal!!!**





Summary

- ***Tools to Realize This Idea***

Deep learning, reinforcement learning, and so on.

- ***Industry Applications***

Power system operation, supply chain scheduling, and so on.

- ***Works based on This Idea***

1. Xianbang Chen, Yafei Yang, Yikui Liu, Lei Wu, “Feature-Driven Economic Improvement for Network-Constrained UC: A Closed-Loop Predict-and-Optimize Framework,” *IEEE Transactions on Power Systems*.
2. Xianbang Chen, Yikui Liu, Lei Wu, “Towards Improving UC Economics: An Add-On Tailor for Renewable Energy and Reserve Predictions,” *IEEE Transactions on Sustainable Energy (Under Second-Round Review)*.

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and Motivations

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