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A bilevel framework for decision-making under uncertainty with contextual information

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Cooking a data-driven decision

$$\min_{z \in Z} \mathbb{E}[f_0(z; y) | X = x]$$

x contextual info
 y uncertain parameter
 z decision



A receipt:

1. Select an uncertain **parameter model**
2. Train the model with a **data set**
3. Issue a **point forecast** for the unseen period \tilde{t}
4. Solve the deterministic optimization problem

$$\hat{y} = g(x; w)$$

$$w^* = \phi(g(x_{\tilde{t}}; w), y_{\tilde{t}})$$

$$\hat{y}_{\tilde{t}} = g(x_{\tilde{t}}; w^*)$$

$$z_{\tilde{t}} = \arg \min_{z \in Z} f_0(z; \hat{y}_{\tilde{t}})$$

How to train the model?

Predictive approach

$$w^{\text{FO}} = \arg \min_{w \in \mathbb{R}^q} \sum_{t \in \mathcal{T}} l^{\text{FO}}(g^{\text{FO}}(x_t; w), y_t)$$

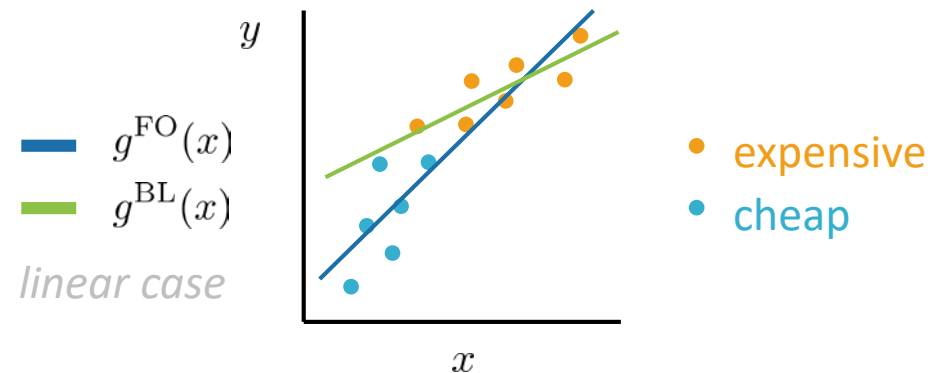
- ✓ Perform well in most tasks
- ✓ Simpler
- ✗ Suboptimal for decision-making

Bilevel prescriptive approach

$$w^{\text{BL}} = \arg \min_{w \in \mathbb{R}^q} \sum_{t \in \mathcal{T}} f_0(\hat{z}_t; y_t)$$

s.t. $\hat{z}_t = \arg \min_{z \in Z} f_0(z; g^{\text{BL}}(x_t; w)) \quad \forall t \in \mathcal{T}$

- ✓ Yields better decisions
- ✗ More complex
- ✗ Issues for linear f_0



Reformulation of the lower level (I)

Lower Level

$$w^{\text{BL}} = \arg \min_{w \in \mathbb{R}^q} \sum_{t \in \mathcal{T}} f_0(\hat{z}_t; y_t)$$

$$\text{s.t. } \hat{z}_t = \arg \min_z f_0(z; g^{\text{BL}}(x_t; w)) \quad \forall t \in \mathcal{T}$$

$$\left. \begin{array}{l} f_i(z_t) \leq 0, \quad \forall i \\ h_j(z_t) = 0, \quad \forall j \end{array} \right\} \quad \forall t \in \mathcal{T}$$

} Feasible set Z



KKT
optimality
conditions

$$w^{\text{BL}} = \arg \min_{w, \hat{z}_t, \lambda_{it}, v_{jt}} \sum_{t \in \mathcal{T}} f_0(\hat{z}_t; y_t)$$

$$\text{s.t. } \nabla f_0(\hat{z}_t, g^{\text{BL}}(x_t, w)) + \sum_{i=1}^I \lambda_{it} \nabla f_i(\hat{z}_t) + \sum_{j=1}^J v_{jt} \nabla h_j(\hat{z}_t) = 0, \quad \forall t \in \mathcal{T}$$

(stationarity)

$$\left. \begin{array}{l} f_i(\hat{z}_t) \leq 0, \quad \forall i \\ h_j(\hat{z}_t) = 0, \quad \forall j \\ \lambda_{it} \geq 0, \quad \forall i \\ \lambda_{it} f_i(\hat{z}_t) = 0, \quad \forall i \end{array} \right\} \quad \forall t \in \mathcal{T}$$

(primal feasibility)

(dual feasibility)

(complementary slackness)

Reformulation of the lower level (II)

$$\lambda_{it} f_i(\hat{z}_t) = 0, \quad \forall i, \forall t \in \mathcal{T}$$

(complementary slackness)

Approach 1: **Regularization**

$$-\sum_{\forall it} \lambda_{it} f_i(\hat{z}_t) \leq \epsilon \quad \epsilon \rightarrow 0$$

Solver: CONOPT ↑ fast, ↓ non-optimal

Approach 2: **Big-M**

$$\left. \begin{array}{l} \lambda_{it} \leq u_{it} M^D, \\ f_i(\hat{z}_t) \geq (u_{it} - 1) M^P, \\ u_{it} \in \{0, 1\}, \end{array} \right\} \quad \forall i, \forall t \in \mathcal{T}$$

Under some conditions of f_0 and $Z \rightarrow$ **MIQP**

↳ Solvers: CPLEX, Gurobi

↑ optimality, ↓ slow, big-M tuning



THANKS!

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Checkout **more at:**

M. A. Muñoz, S. Pineda, J. M. Morales, [A bilevel framework for decision-making under uncertainty with contextual information](#). *arXiv preprint arXiv:2008.01500*, 2020.