

Mathematical Decision-Optimization (DO) Model $M(\mathbf{w})$ (*):

$$\begin{aligned} \mathbf{x}^*(\mathbf{w}) \in \arg \min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{R}^m} \quad & f(\mathbf{x}, \mathbf{y}, \mathbf{w}) \\ \text{s.t.} \quad & \mathbf{g}(\mathbf{x}, \mathbf{y}, \mathbf{w}) \leq \mathbf{0} \\ & \mathbf{y} = \mathbf{h}(\mathbf{x}, \mathbf{w}) \\ & \mathbf{x} \in \Omega(\mathbf{w}) \end{aligned}$$

$\mathbf{w} \in \mathbb{R}^k$ – fixed uncontrollable.

$\Omega(\mathbf{w})$ – polytope (possibly unbounded).

(*) Maragno*, D., Wiberg*, H., Bertsimas, D., Birbil, S. I., Hertog, D. d., and Fajemisin, A. (2021). *Mixed-Integer Optimization with Constraint Learning*.

Learned DO Model $\hat{M}(\mathbf{w})$:

$$\begin{aligned} \hat{\mathbf{x}}^*(\mathbf{w}) \in \arg \min_{\mathbf{x} \in \mathbb{R}^n, \mathbf{y} \in \mathbb{R}^m} \quad & \hat{f}(\mathbf{x}, \mathbf{y}, \mathbf{w}) && \Leftarrow \text{learn} \\ \text{s.t.} \quad & \mathbf{g}(\mathbf{x}, \mathbf{y}, \mathbf{w}) \leq \mathbf{0} \\ & \mathbf{y} = \hat{\mathbf{h}}(\mathbf{x}, \mathbf{w}) && \Leftarrow \text{learn} \\ & \mathbf{x} \in \Omega(\mathbf{w}) \end{aligned}$$

$\mathbf{w} \in \mathbb{R}^k$ – fixed uncontrollable.

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- UN World Food Programme (INFORMS Edelman Award 2021): Food palatability prediction in food basket cost minimization.
- Louisville Metropolitan Sewer District and Tetra Tech (INFORMS Edelman Award 2019 Finalist): Rainfall prediction in wastewater storage maximization.

What We **Want**:

$$\begin{aligned}\hat{\mathbf{x}}^* &= \hat{\mathbf{x}}^*(\mathbf{w}), & \hat{\mathbf{y}}^* &= \mathbf{h}(\hat{\mathbf{x}}^*, \mathbf{w}) \\ \mathbf{x}^* &= \mathbf{x}^*(\mathbf{w}), & \mathbf{y}^* &= \mathbf{h}(\mathbf{x}^*, \mathbf{w})\end{aligned}$$

(a) close to optimum

$$f(\hat{\mathbf{x}}^*, \hat{\mathbf{y}}^*, \mathbf{w}) - f(\mathbf{x}^*, \mathbf{y}^*, \mathbf{w}) \leq \epsilon$$

(b) feasible

$$\mathbf{g}(\hat{\mathbf{x}}^*, \hat{\mathbf{y}}^*, \mathbf{w}) \leq \mathbf{0}$$

What We **Can**:

known policy: $\mathbf{x}_0 = \mathbf{x}_0(\mathbf{w}), \mathbf{y}_0 = \mathbf{h}(\mathbf{x}_0, \mathbf{w})$

(a) improve upon policy

$$\Pr[f(\mathbf{x}_0, \mathbf{y}_0, \mathbf{w}) - f(\hat{\mathbf{x}}^*, \hat{\mathbf{h}}(\hat{\mathbf{x}}^*, \mathbf{w}), \mathbf{w}) \geq \epsilon] \geq (1 - \delta_1)$$

(b) likely feasible

$$\Pr[\mathbf{g}(\hat{\mathbf{x}}^*, \hat{\mathbf{h}}(\hat{\mathbf{x}}^*, \mathbf{w}), \mathbf{w}) \leq \mathbf{0}] \geq (1 - \delta_2)$$

What We **Can**:

known policy: $\mathbf{x}_0 = \mathbf{x}_0(\mathbf{w}), \mathbf{y}_0 = \mathbf{h}(\mathbf{x}_0, \mathbf{w})$

(a) improve upon policy: **Probability of Improvement (Pol)**

$$\Pr[f(\mathbf{x}_0, \mathbf{y}_0, \mathbf{w}) - f(\hat{\mathbf{x}}^*, \hat{\mathbf{h}}(\hat{\mathbf{x}}^*, \mathbf{w}), \mathbf{w}) \geq \epsilon] \geq (1 - \delta_1)$$

(b) likely feasible: **Probability of Constraint Satisfaction (PoCS)**

$$\Pr[\mathbf{g}(\hat{\mathbf{x}}^*, \hat{\mathbf{h}}(\hat{\mathbf{x}}^*, \mathbf{w}), \mathbf{w}) \leq \mathbf{0}] \geq (1 - \delta_2)$$

Gaussian Process: f as a “random variable”.

$$f|D \sim GP$$

\Rightarrow Value @ point $(\mathbf{x}, \mathbf{y}, \mathbf{w})$

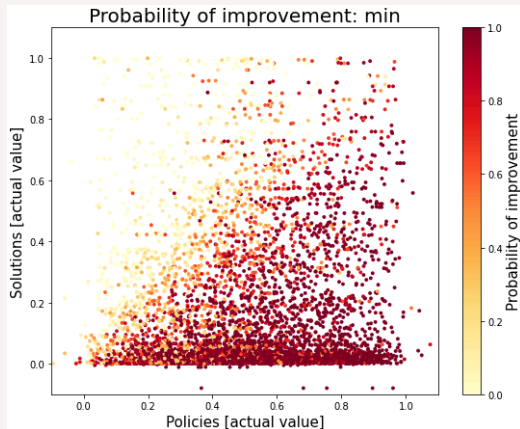
$$f(\mathbf{x}, \mathbf{y}, \mathbf{w}) \sim \mathcal{N}(\mu_{GP}(\mathbf{x}, \mathbf{y}, \mathbf{w}), \sigma_{GP}^2(\mathbf{x}, \mathbf{y}, \mathbf{w}))$$

\Rightarrow Value @ 2 points $(\mathbf{x}', \mathbf{y}', \mathbf{w}'), (\mathbf{x}'', \mathbf{y}'', \mathbf{w}'')$

$$(f(\mathbf{x}', \mathbf{y}', \mathbf{w}'), f(\mathbf{x}'', \mathbf{y}'', \mathbf{w}'')) \sim \mathcal{N}(\mu_{GP}, \Sigma_{GP})$$

\Rightarrow Pol and PoCS estimation

Pol and PoCS



Pol of 5593 pairs
 $(\hat{\mathbf{x}}^*, \mathbf{x}_0)$ associated
with 1718 randomly
generated $(\mathbf{f}, \Omega, \mathbf{D}, \mathbf{x}^*)$
in dimensions 2, 3 using
DOFramework.

FP	2.6%
FN	9.2%

FP: $\text{Pol} \geq 0.5, f(\hat{\mathbf{x}}^*) \geq f(\mathbf{x}_0)$

FN: $\text{Pol} \leq 0.5, f(\hat{\mathbf{x}}^*) \leq f(\mathbf{x}_0)$