# What is the Right Notion of Distance between Predict-then-Optimize Tasks?

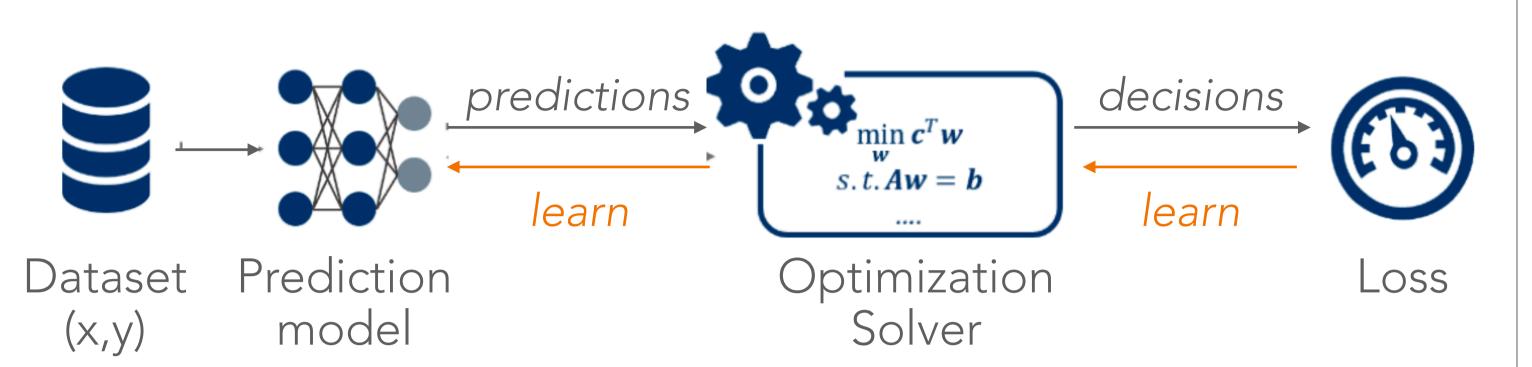


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## Predict-then-Optimize (PtO)

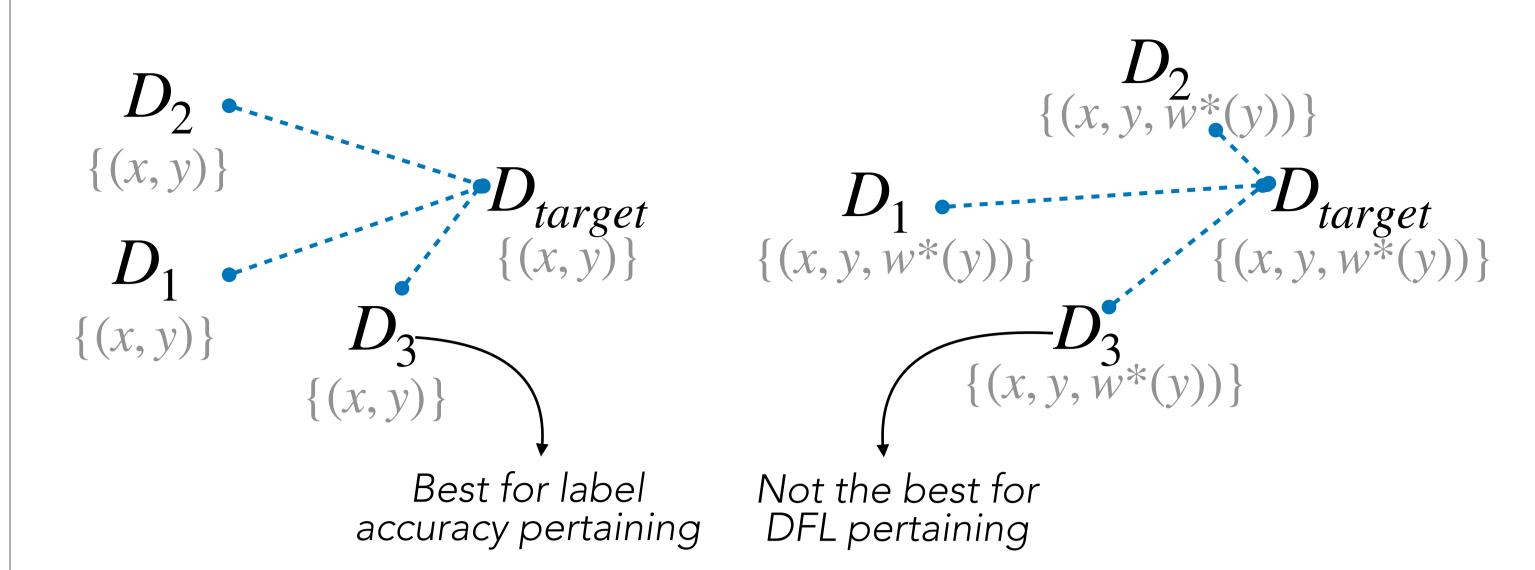
Predictions act as inputs to a downstream optimization, shifting the goal from minimizing prediction error to minimizing decision regret



In decision-focused learning (DFL), the loss is decision regret: the difference in objective value between the decision made using the prediction and the optimal decision using the true label.

$$q(\hat{y}, y) = |y \cdot w^*(\hat{y}) - y \cdot w^*(y)|$$

#### Distances between Datasets



- Enable principled reasoning about task similarity and generalization, especially when data or compute is limited
- Core for transfer learning, multitasks learning, data valuation, etc
- Model agnostic, pre-computed decisions

In decision-focused ML, two tasks that look similar in their feature-label space can lead to very different decisions. OTD<sup>3</sup> is a principled way to measure dataset similarity that accounts for the full feature-label-decision relationship, providing a more reliable signal for transferability when the goal is to learn good decisions—not just good predictions.

How should we compare decisions? We propose decision quality disparity: the difference in objective value achieved by two decisions when evaluated on the same true labels.

$$l_q(y, z; y^*, z^*) := |q(y, y^*) - q(z, z^*)|$$

#### What is the distance between feature-label-decision pairs?

We take a convex combination of the feature-label distance and the disparity in decision quality —used as the cost in the optimal transport formulation.

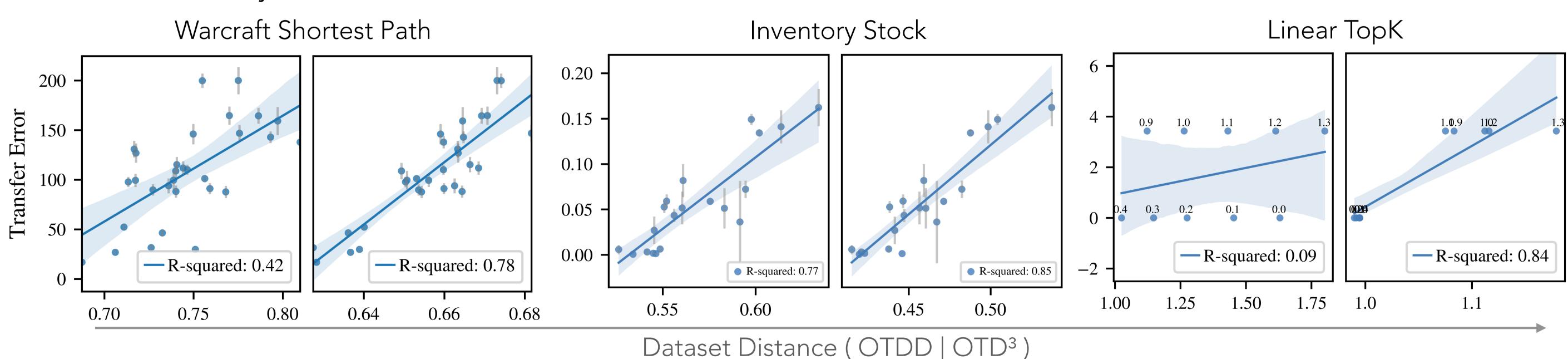
$$c_{PtO}((x, y, z), (x', y', z')) \triangleq \alpha_X \cdot d_X(x, x') + \alpha_Y \cdot d_Y(y, y') + \alpha_W \cdot l_q(z, z'; y', y')$$

$$\mathsf{OTD}^3(D, D'; \alpha) \triangleq \min_{\pi \in \Pi(\gamma, \beta)} \int_{W \times W'} c_{PtO}^{\alpha}(w, w') \, \mathrm{d}\pi$$

- 1. A **true metric** in the space of measures of feature-label-decision pairs
- 2. Provides a bound on the target regret

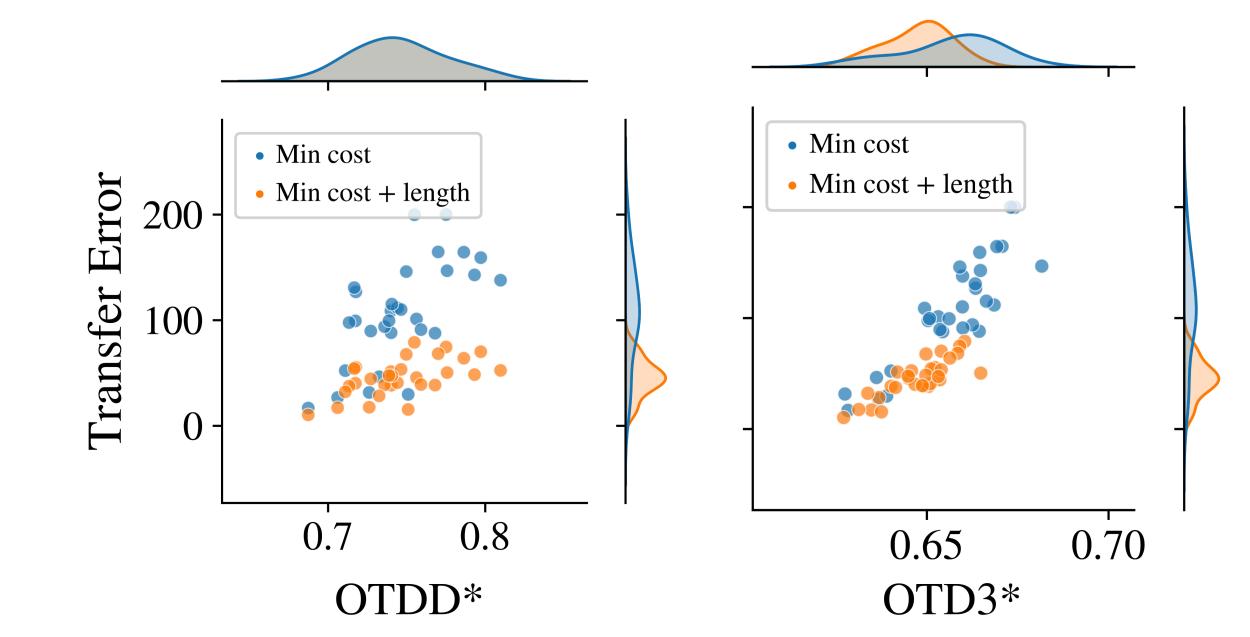
#### Applications and Experimental Results

Predict transferability of decision-focused learned models on shifted domains



Distinguish transferability success

### across different downstream tasks



# Signal DFL transfer performance under limited data availability

