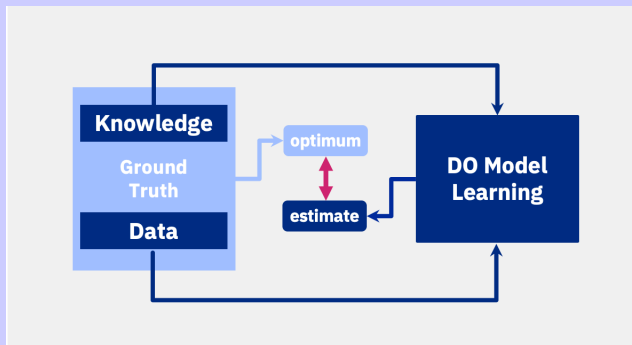


DOFramework: a testing framework for DO model learners.



Knowledge
 Ω

Bounded convex polytope $\Omega \subseteq \text{dom}(f) \subseteq \mathbb{R}^d$, $d = n + m + k \geq 2$.

Ground
Truth f

Continuous PWL f with (combinatorially) known $\mathbf{x}^* \in \arg \min_{\mathbf{x} \in \Omega} f(\mathbf{x})$.

Data D

Gaussian mix model in $\text{dom}(f)$.

DO Problem Instance: (f, Ω, D, x^*)

Estimate $\hat{\mathbf{x}}^*$ score:

$$\text{score}(\hat{\mathbf{x}}^*) = \frac{f(\hat{\mathbf{x}}^*) - f(\mathbf{x}^*)}{f_{\max} - f_{\min}}$$

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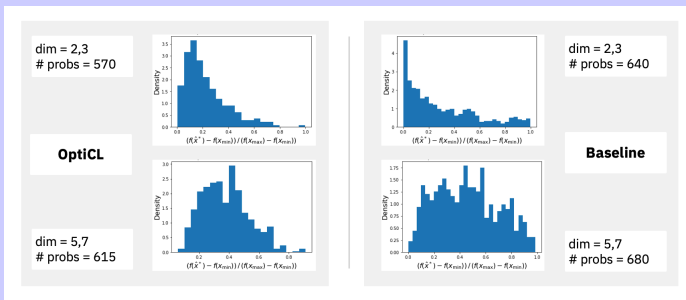
Estimate $\hat{\mathbf{x}}^*$ score:

$$\text{score}(\hat{\mathbf{x}}^*) = \frac{f(\hat{\mathbf{x}}^*) - f(\mathbf{x}^*)}{f_{\max} - f_{\min}} \implies \text{score density}$$

Solution Quality Probability:

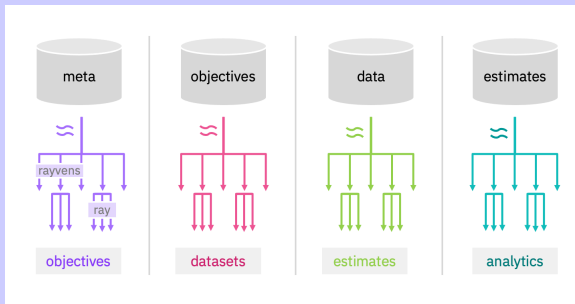
$$Pr[f(\hat{\mathbf{x}}^*) - f(\mathbf{x}^*) < \epsilon(f_{\max} - f_{\min})]$$

Profiling



- Maragno, Wiberg. *OptiCL: Mixed-integer optimization with constraint learning* (2021).
<https://github.com/hwiberg/OptiCL/>.
- Mitchell, OSullivan, Dunning. *PuLP: A Linear Programming Toolkit for Python* (2011).
<https://coin-or.github.io/pulp/>.

Design



- Moritz et. al.. *Ray: A Distributed Framework for Emerging AI Applications* (OSDI 2018).
<https://github.com/ray-project/ray>.
- Bercea, Tardieu. *Rayvens: Event sources and sinks on Ray*, <https://github.com/project-codeflare/rayvens>

Deployment

Requirements:

- Local / Cloud Object Storage (configs.yaml)
- Local / k8s Cluster (AWS / IBM Cloud)

Installation:

```
$ pip install doframework
$ cd <project_folder>
$ doframework-setup.sh --configs configs.yaml
```

Running:

```
$ python user_module.py --configs configs.yaml
$ ray submit doframework.yaml user_module.py \\  
--configs configs.yaml
```