

# 1 Optimization as a layer

## 1.1 Some Concepts

(.) is the alias:

- **(e2e)** End-to-end: raw-input to ultimate outputs of interests.
  - without tuned features/feature engineering
- **(optlayer)** Optimization as a Layer
  - inputs  $\rightarrow$  outputs:  $x \rightarrow y$ , includes an optimization problem  $y = \arg \min f(x)$
- Differentiation, forward-and-backward pass.
  - for differentiable (convex) problem, Jacobian can be calculated via KKT.
  - for LP, can be calculated via interior point HSD formulation, [Ye et al. \(1994\)](#)
  - other specified solvers, for QP, conic, ...

## 1.2 Differentiations

[optimal condition] + solver

- KKT + QP solver, [Amos and Kolter \(2017\)](#)
- CVX  $\rightarrow$  Conic (HSD embedding) [optimal condition]  $\rightarrow$  conic solver (SCS, ...), [Amos and Kolter \(2017\)](#)
- LP  $\rightarrow$  HSD, [Mandi and Guns \(2020\)](#)

## 1.3 Application

- e2e, stochastic programming (single stage.)
  - sp  $\rightarrow$  deterministic.
- sensitivity analysis
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## 1.4 Reference

10 Amos B, Kolter JZ (2017) Optnet: Differentiable optimization as a layer in neural networks. *International Conference on Machine Learning*. (PMLR), 136–145.

Mandi J, Guns T (2020) Interior Point Solving for LP-based prediction+ optimisation. *arXiv preprint arXiv:2010.13943*.

Ye Y, Todd MJ, Mizuno S (1994) An  $O(\sqrt{n}L)$ -iteration homogeneous and self-dual linear programming algorithm. *Mathematics of operations research* 19(1):53–67.