

Speedup Techniques for Hyperparameter Optimization

Overview

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Beyond Black-box Optimization

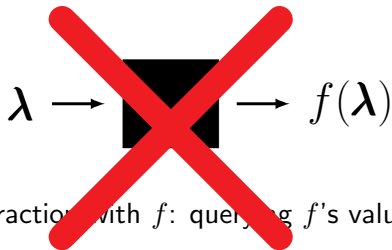
Recall general blackbox optimization:



Only mode of interaction with f : querying f 's value at a given λ

Beyond Black-box Optimization

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Only mode of interaction with f : querying f 's value at a given λ

Too slow for tuning expensive models

Methods for Going Beyond Blackbox Bayesian Optimization

- Sum of little black boxes

- ▶ Each little black box is fast but only yields a noisy estimate
- ▶ SMAC [Hutter et al. 2011] directly solves $\arg \min_{\lambda \in \Lambda} \sum_{i=1}^N c(\lambda, i)$
- ▶ Auto-WEKA [Thornton et al, 2013] used this to optimize 10-fold cross-validation performance

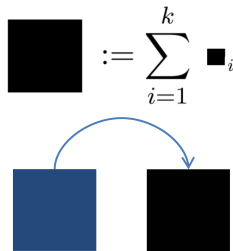
$$\blacksquare := \sum_{i=1}^k \blacksquare_i$$

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- Meta-learning across problems / datasets



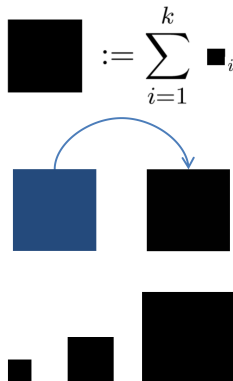
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- Multi-fidelity optimization



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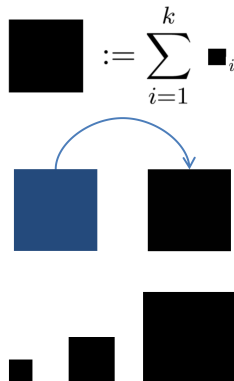
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- Meta-learning across problems / datasets

- Multi-fidelity optimization

- Graybox optimization / learning curve prediction



Learning Goals of this Lecture

After this lecture, students can ...

- Describe many different ways of using [meta-learning](#) to speed up HPO
- Explain the concept of [multi-fidelity](#) optimization to speed up HPO
- Explain the [Successive Halving](#) and [Hyperband](#) algorithms
- Explain how to combine Bayesian optimization and Hyperband in [BOHB](#)
- Describe how to [exploit multiple fidelities in Bayesian optimization](#)
- Discuss several ways of predicting [learning curves](#)
- Discuss [success stories](#) of speeding up Bayesian optimization