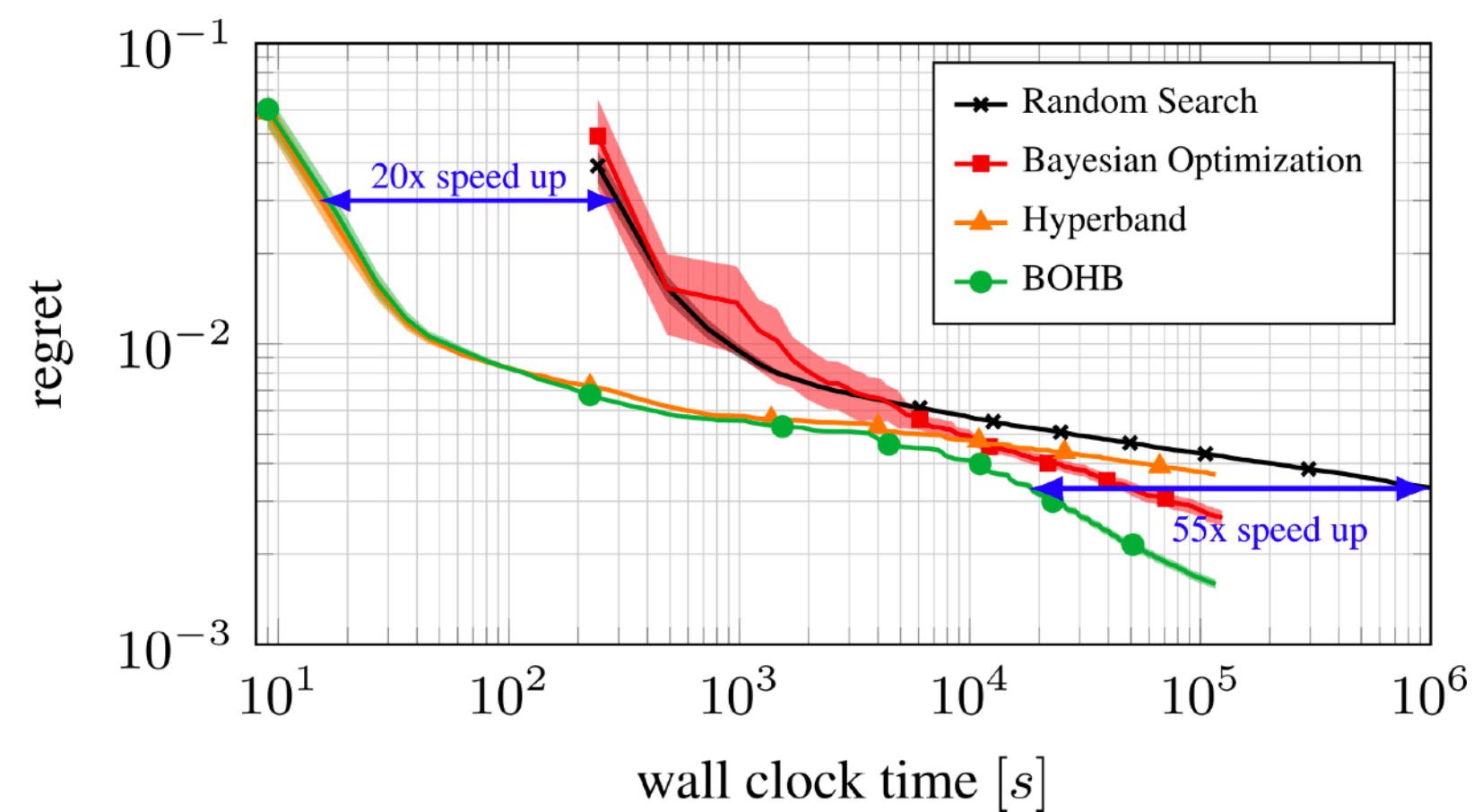


Practical Two-Step Lookahead Bayesian Optimization

Jian Wu, Peter Frazier

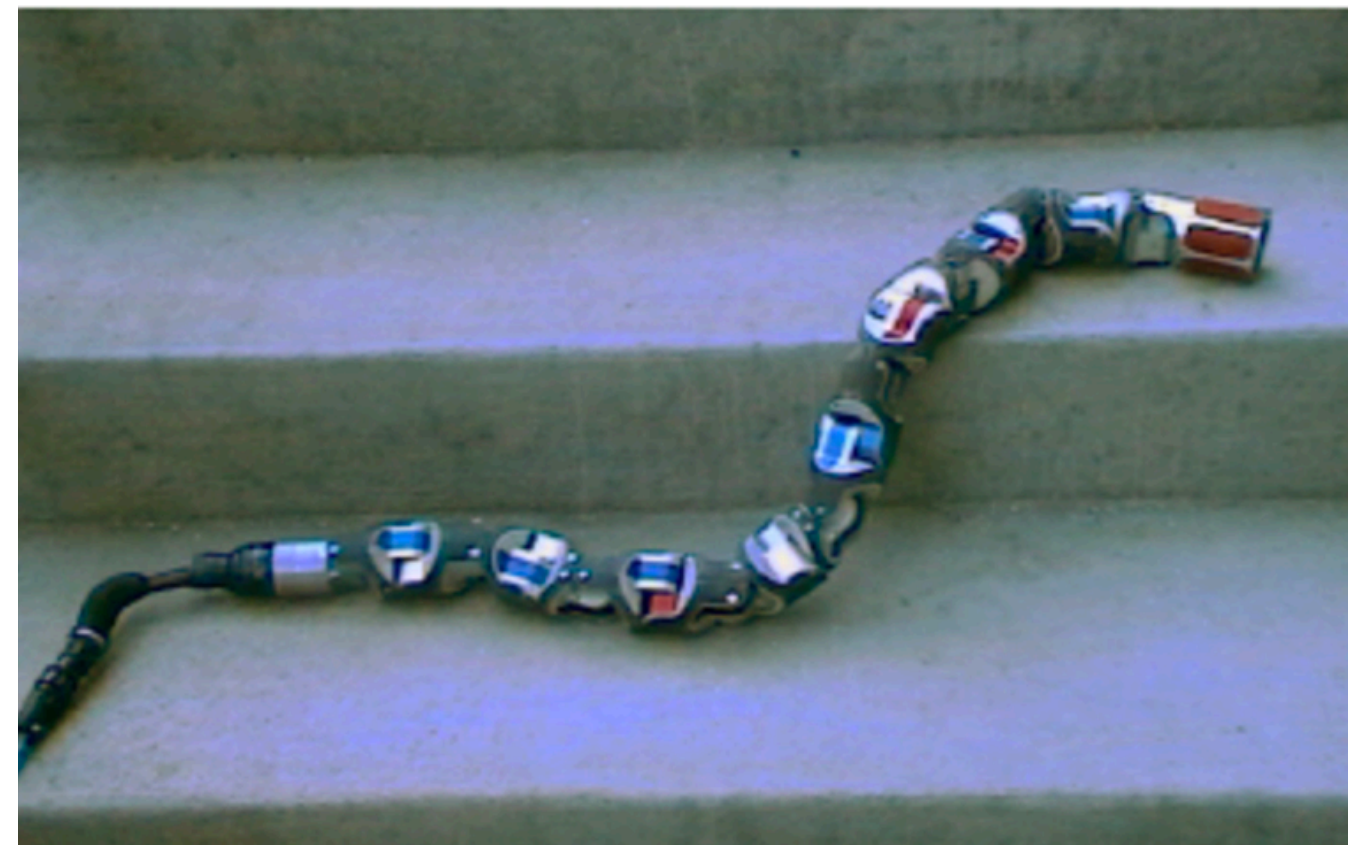
Poster #33, Th 10:45am

BayesOpt is a powerful black-box optimizer



AutoML

[automl.org]



Policy Search
in Robotics

[Tesch, Schneider, Choset 2011]



Product Improvement
with A/B testing

[botorch.org]

$$EI(x) = \underbrace{E[\max(f^*, f(x))]}_{\substack{\text{Value AFTER} \\ \text{observing } f(x)}} - \underbrace{f^*}_{\substack{\text{Value BEFORE} \\ \text{observing } f(x)}}$$

Classic
BayesOpt
looks only one
step ahead

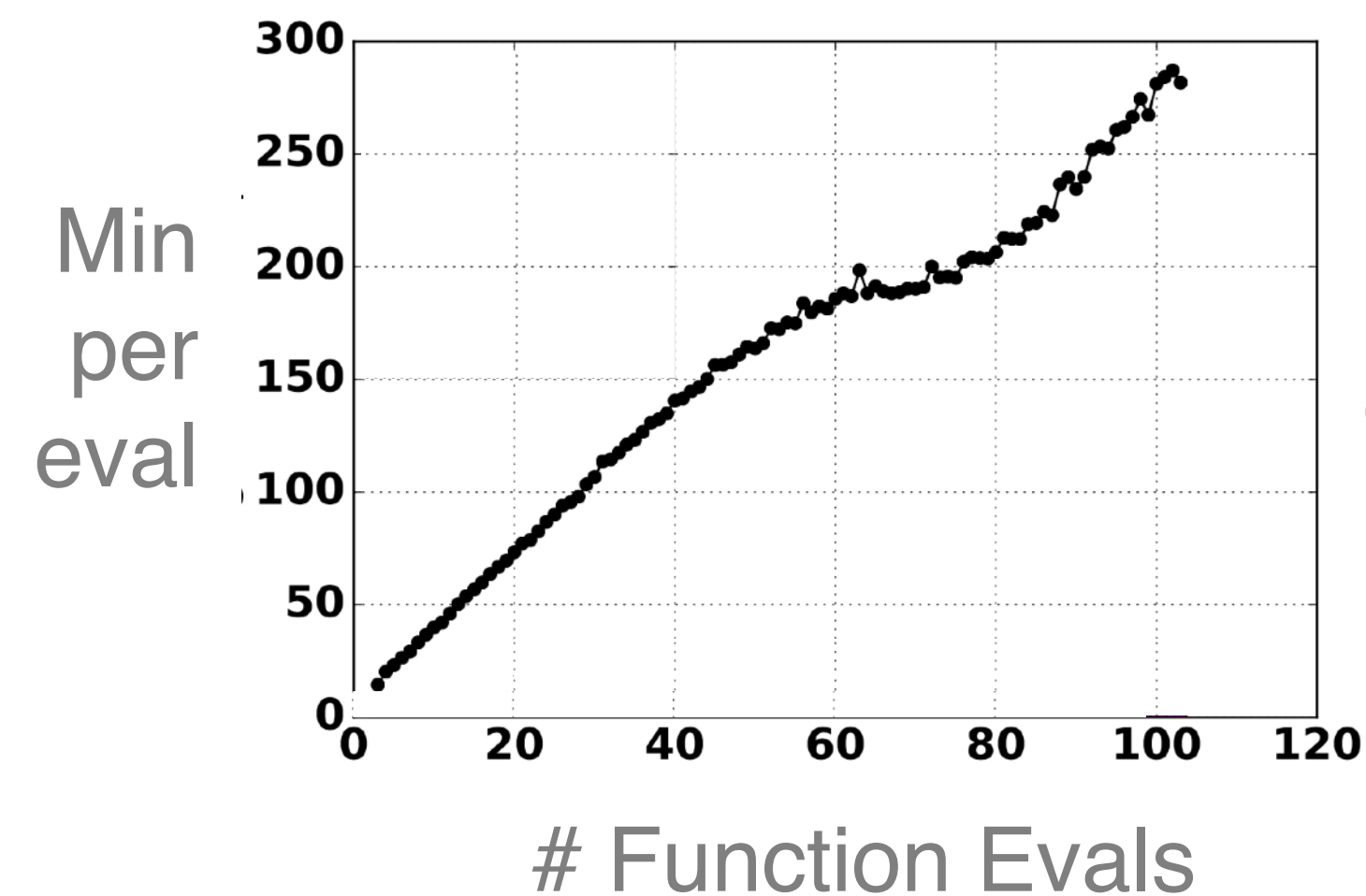
In THEORY:

- it's better to look multiple steps ahead



In PRACTICE:

- It's slow



Runtime of GLASSES
[Gonzales et al. 2016]
on 10-d Rosenbrock

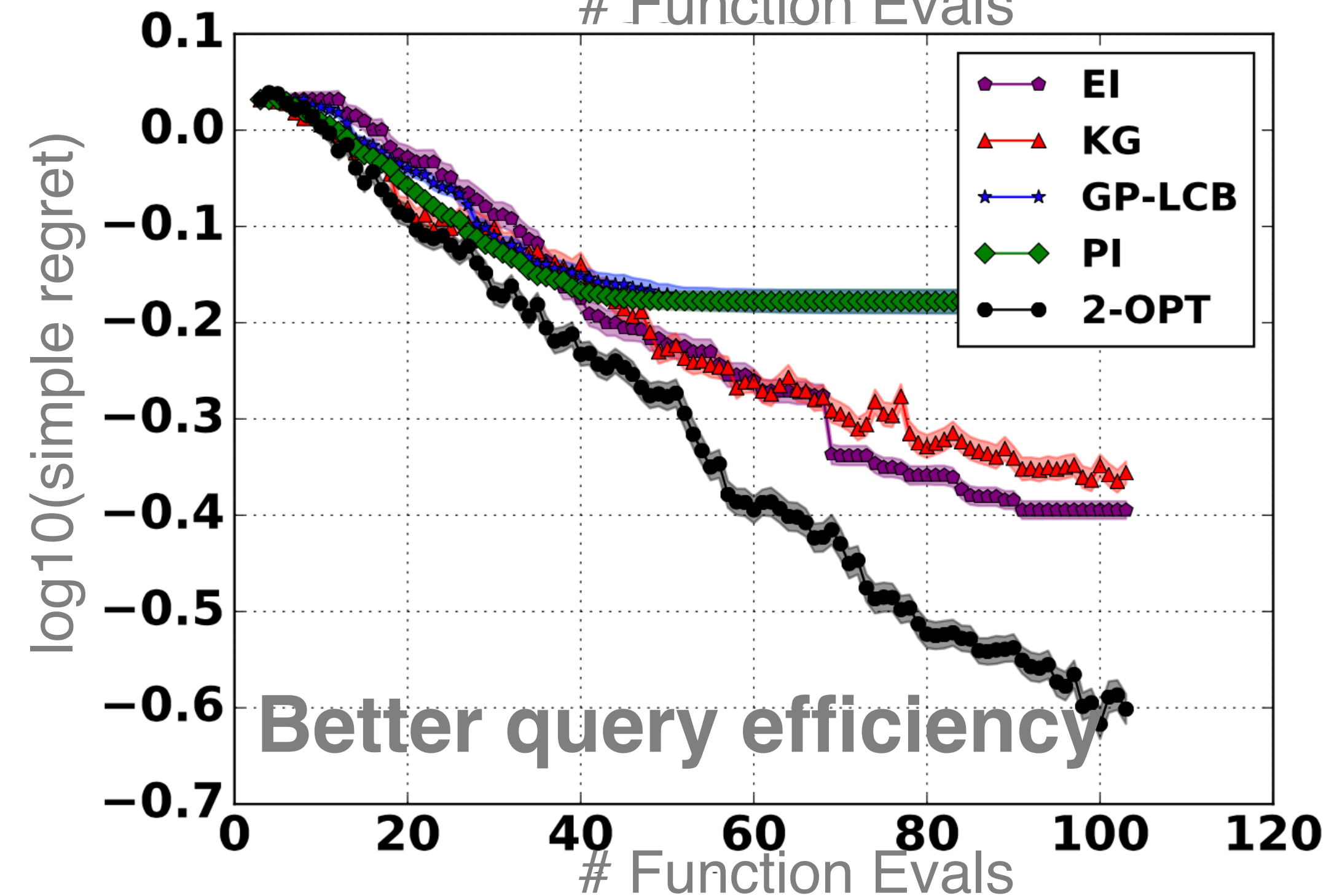
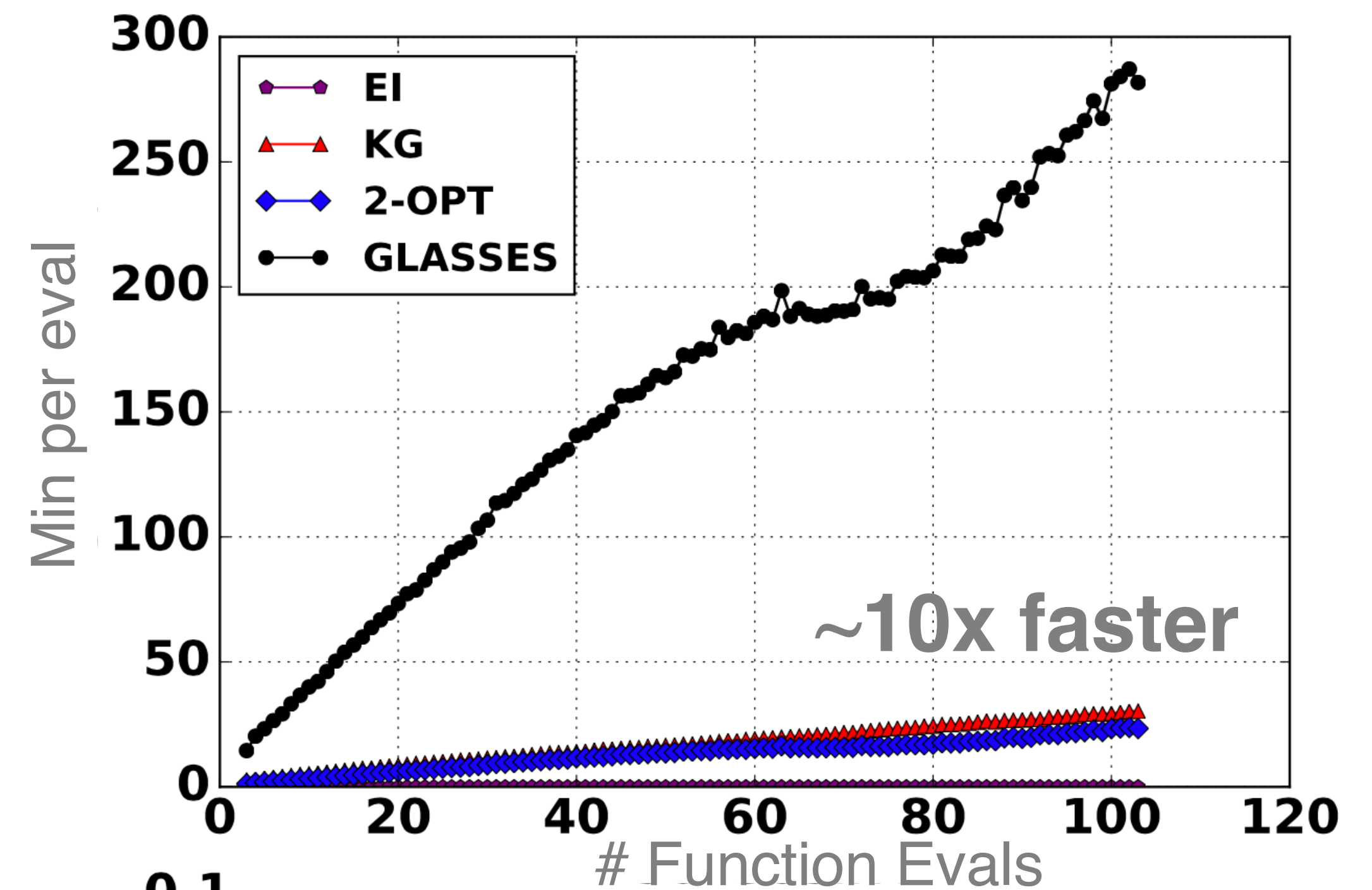
- Approximation errors erase most of the benefit

Our Contribution

A new algorithm that efficiently & accurately optimizes the **two-step lookahead** acquisition function.

It provides:

- **Better query efficiency** than 1-step and previous multi-step methods
- **Batch** evaluations
- **~10x faster** than previous multi-step methods (usually seconds to at most several minutes per batch)



Check out our poster!
#33, Th 10:45-12:45