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AN ENERGY FRONTIER RESEARCH CENTER

NEXTorch: A Design and Bayesian Optimization Toolkit for Chemical Sciences and Engineering

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Office of Science

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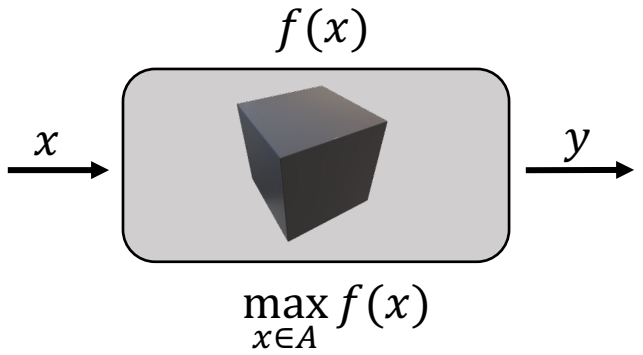
About Me

- Graduated from Vlachos Group Dec 2021
- ML Research Scientist at Meta
- Based in San Francisco
- Research interests: AI data annotation, active learning, generative AI, reinforcement learning from human feedback (RLHF)
- LinkedIn: https://www.linkedin.com/in/wangyifan411/

Global Optimization of Blackbox Functions

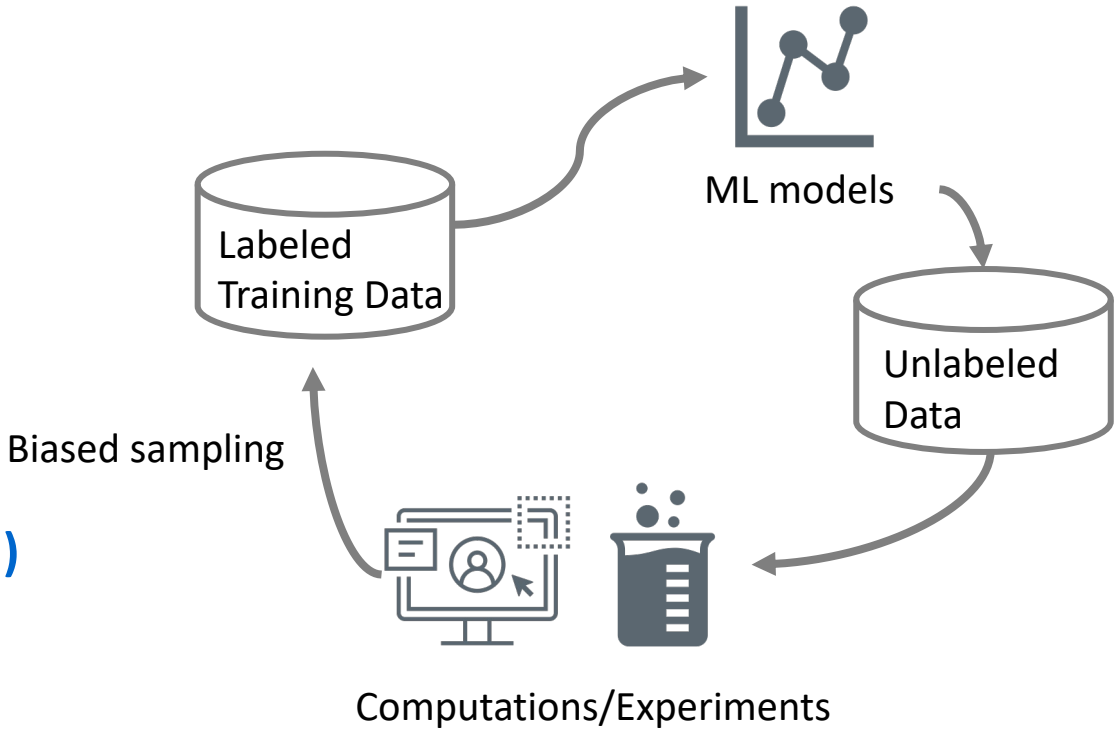
Blackbox functions [1]

- Expensive computer model or laboratory experiments
- Unknown explicit model form
- Multi-dimensional



Active learning [2]

- An algorithm "learning" from data, proposing next experiments, and improving prediction accuracy with fewer training data or lower cost



Use active learning to reduce experimental (computational) cost and improve accuracy of the surrogate model

[1] D.R. Jones, M. Schonlau, and W. J. Welch, J. Glob. Optim. **13**, 455 (1998).
[2] Settles, B. *Active Learning Literature Survey*. *Active Learning Literature Survey* (2009).

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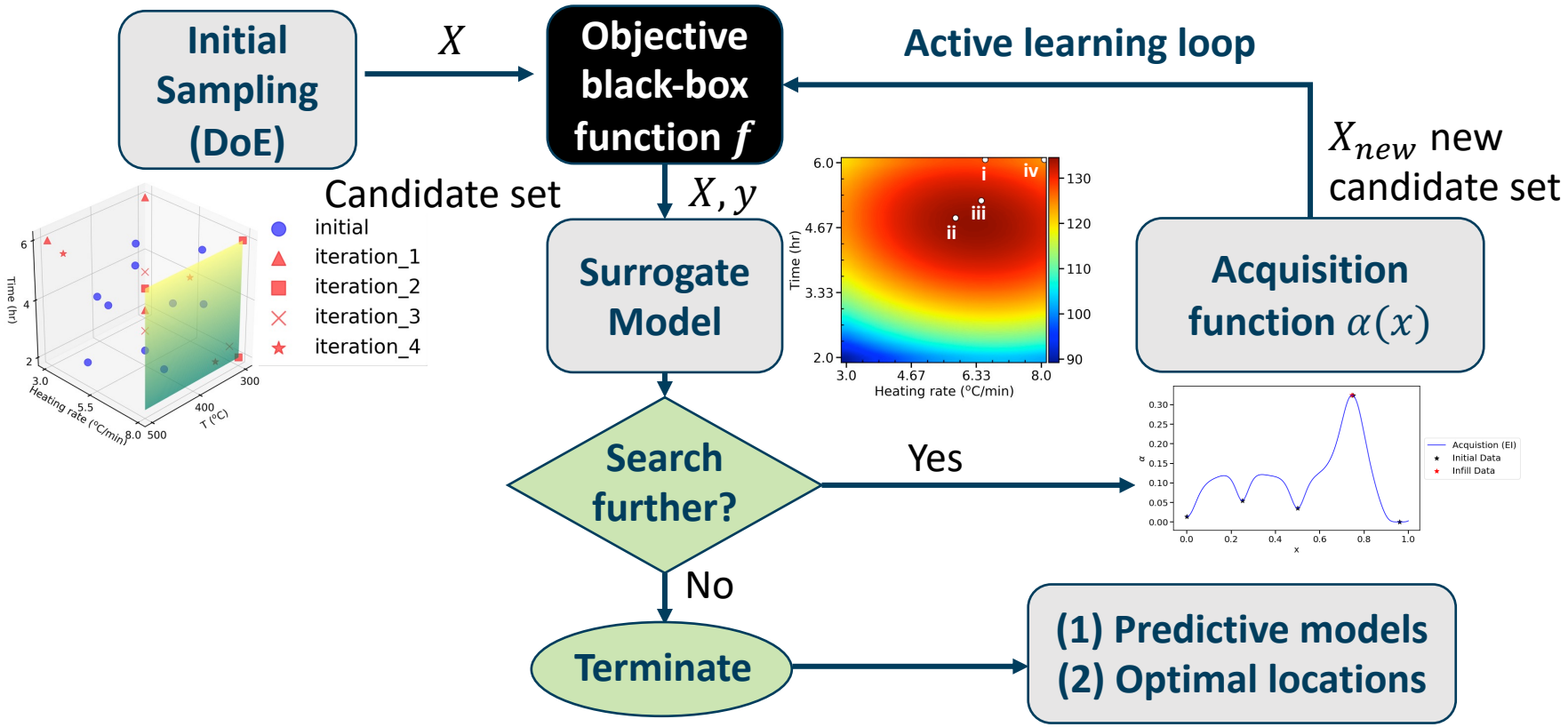
Bayesian Optimization (BO)



Bayesian Statistics

Posterior **Data** **Prior**

$$P(f|D) \propto P(D|f)P(f)$$

Thomas Bayes

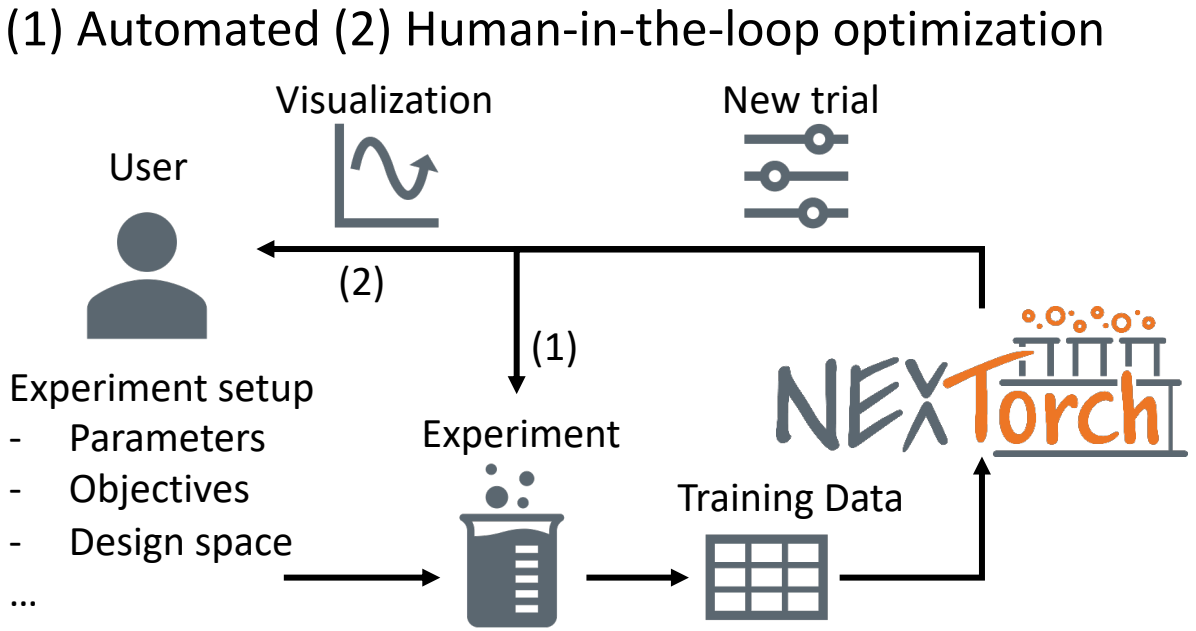
- Initial sampling can be generated through design of experiments (DoE)
- The surrogate model is typically a Gaussian Process (GP)
- Next experiment points are generated by acquisition functions (exploration vs. exploitation)

[1] P.I. Frazier, 1 (2018). http://arxiv.org/abs/1807.02811

NEXTorch

Toolkit for Design of Experiments + Bayesian Optimization

Workflow



- Install via pip `pip install nextorch`
- Key dependencies

python™ **pyDOE** PyTorch BoTorch

https://botorch.org/

[1] **Y. Wang**, T. Chen, and D.G. Vlachos, J. Chem. Inf. Model. 61, 5312–5319 (2021).
GitHub: https://github.com/VlachosGroup/nextorch
Documentation: https://nextorch.readthedocs.io/en/latest/index.html

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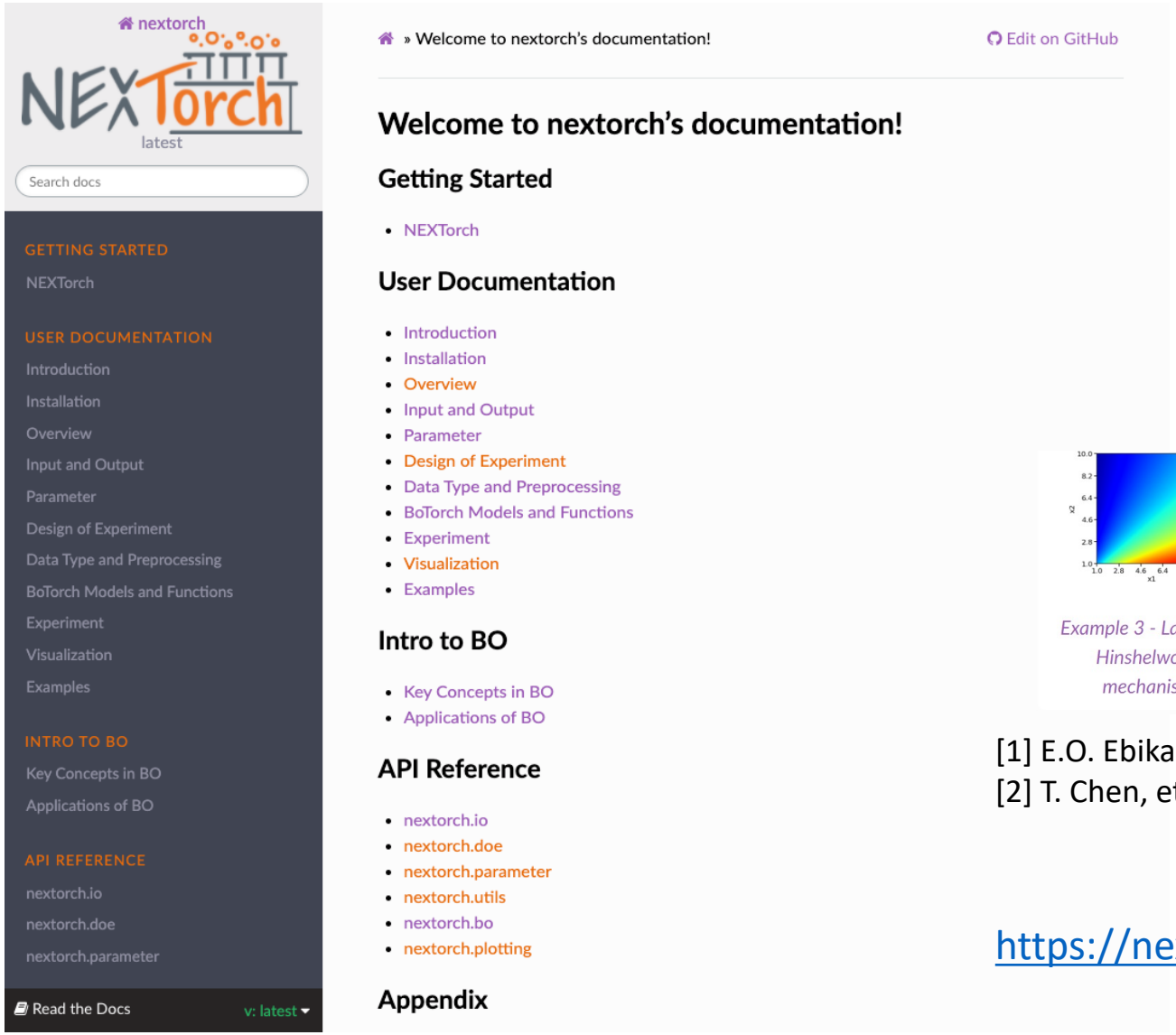
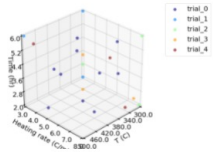
Online Documentation Page

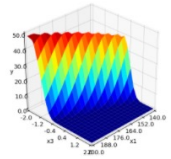


- NEXTorch modules and functions
- Tutorials with code examples
- Introduction to BO theory
- BO applications in literature

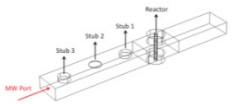
Example 3 - Langmuir Hinshelwood mechanism



Example 4 - Nitrogen-doped carbon catalysts



Example 5 - Plug flow reactor yield



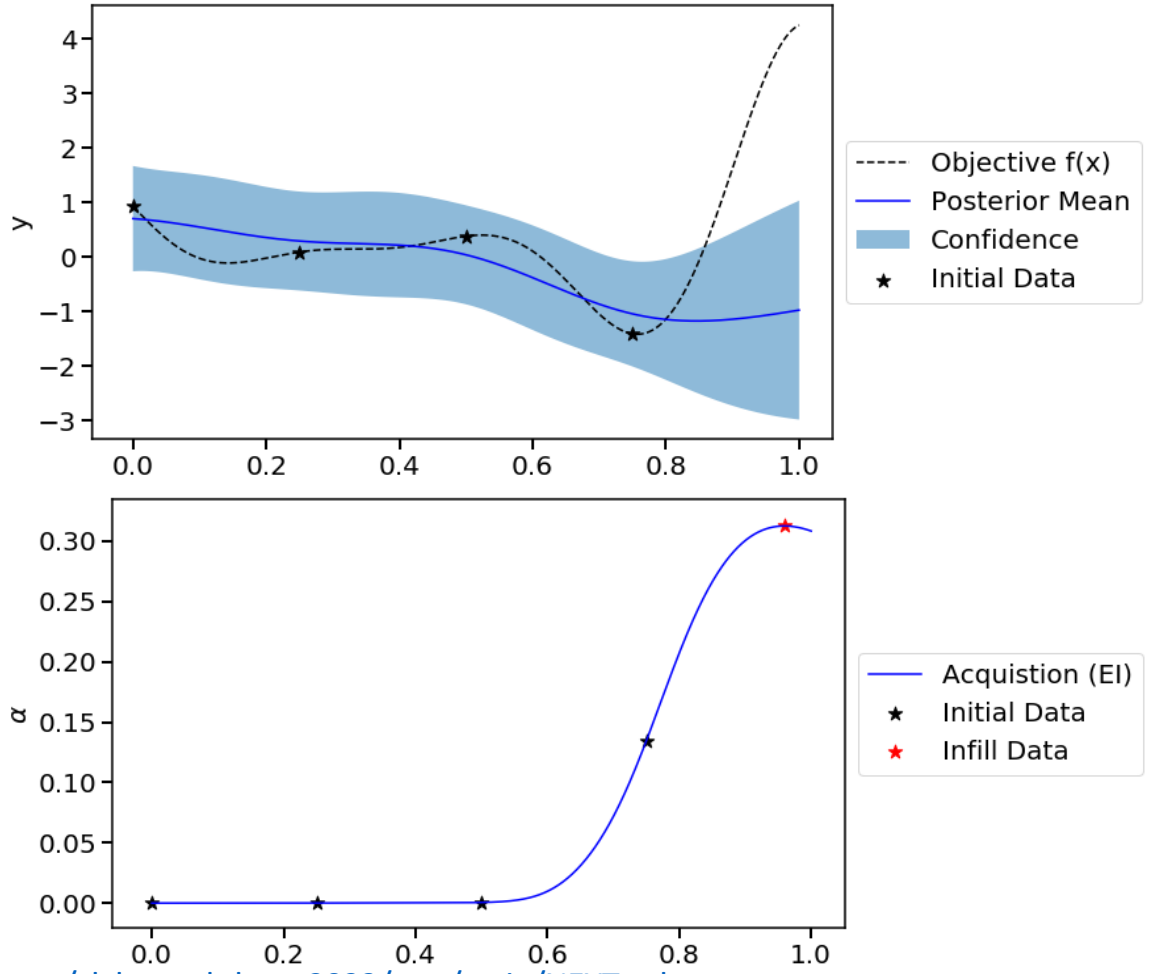
Example 8 - Stub tuner of the microwave cavity

[1] E.O. Ebikade, et al, React. Chem. Eng. (2020).
[2] T. Chen, et al, Ind. Eng. Chem. Res. 59, 10418 (2020).

https://nextorch.readthedocs.io/en/latest/index.html

BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Starting from $x = 0, 0.25, 0.5, 0.75$

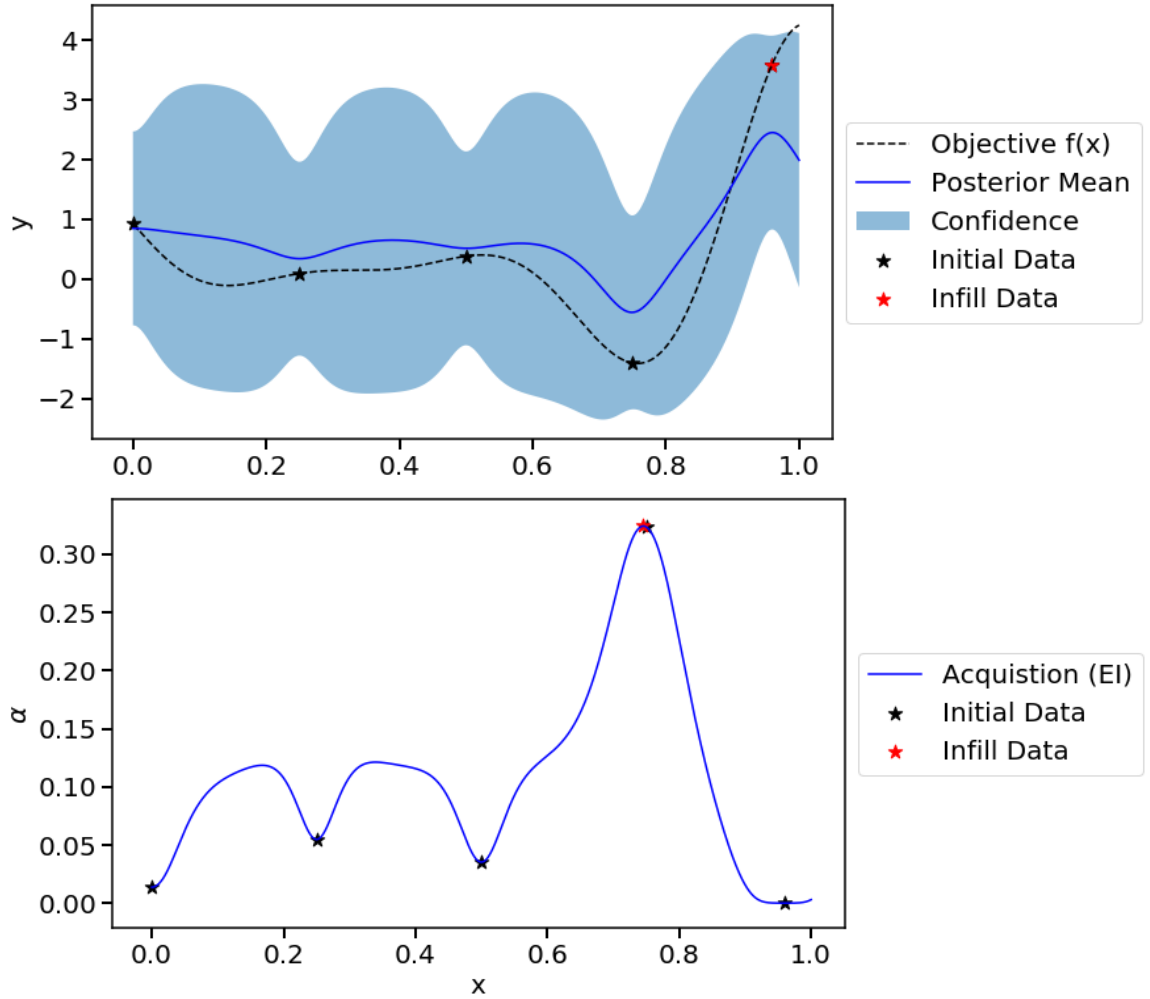


Demo instructions: https://github.com/VlachosGroup/vlab_workshop_2023/tree/main/NEXTorch

Notebook location: https://github.com/VlachosGroup/nextorch/blob/main/examples/notebooks/01_simple_1d.ipynb

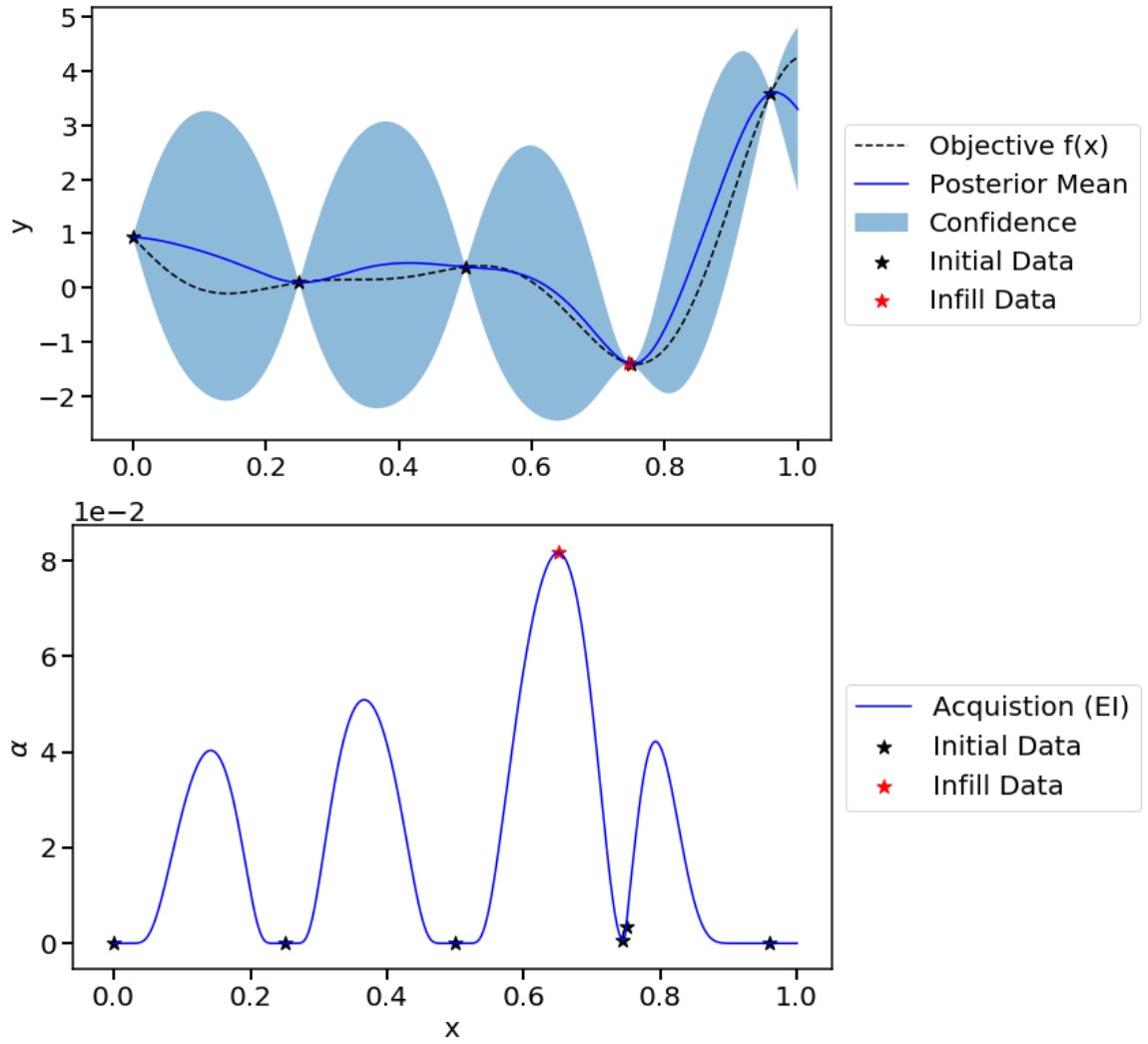
BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 1, acquisition function – expected improvement (EI)



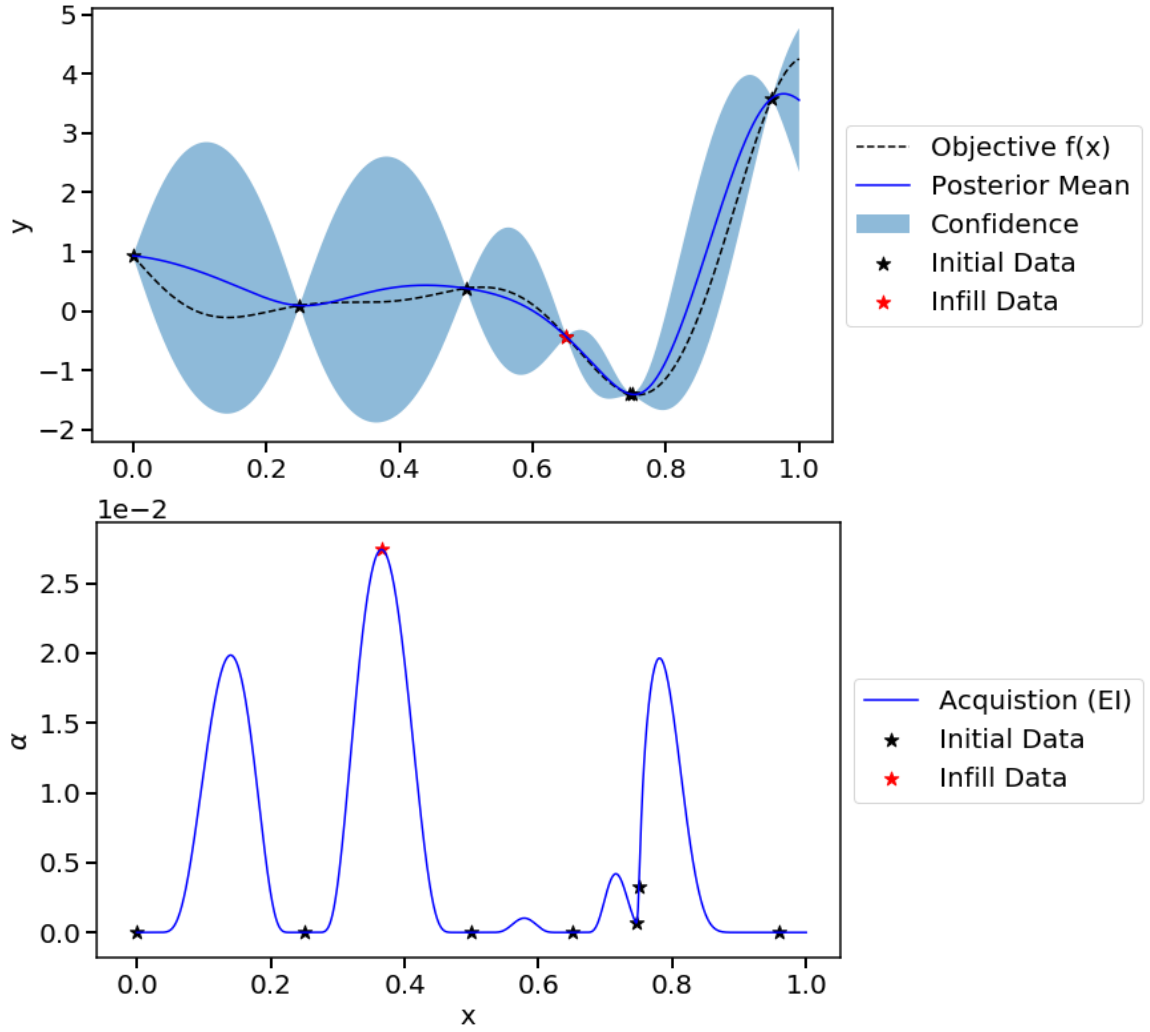
BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 2



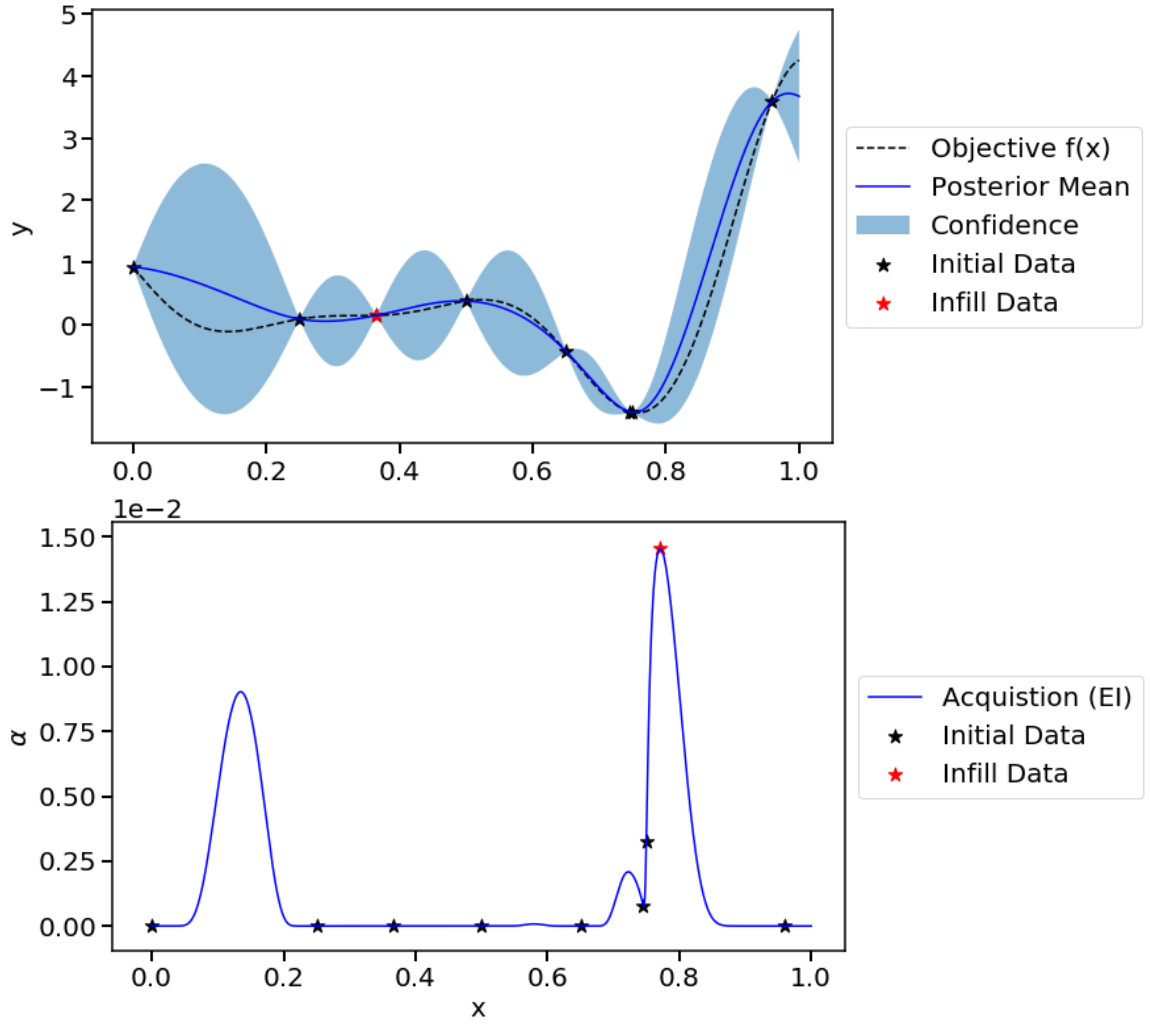
BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 3



BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 4



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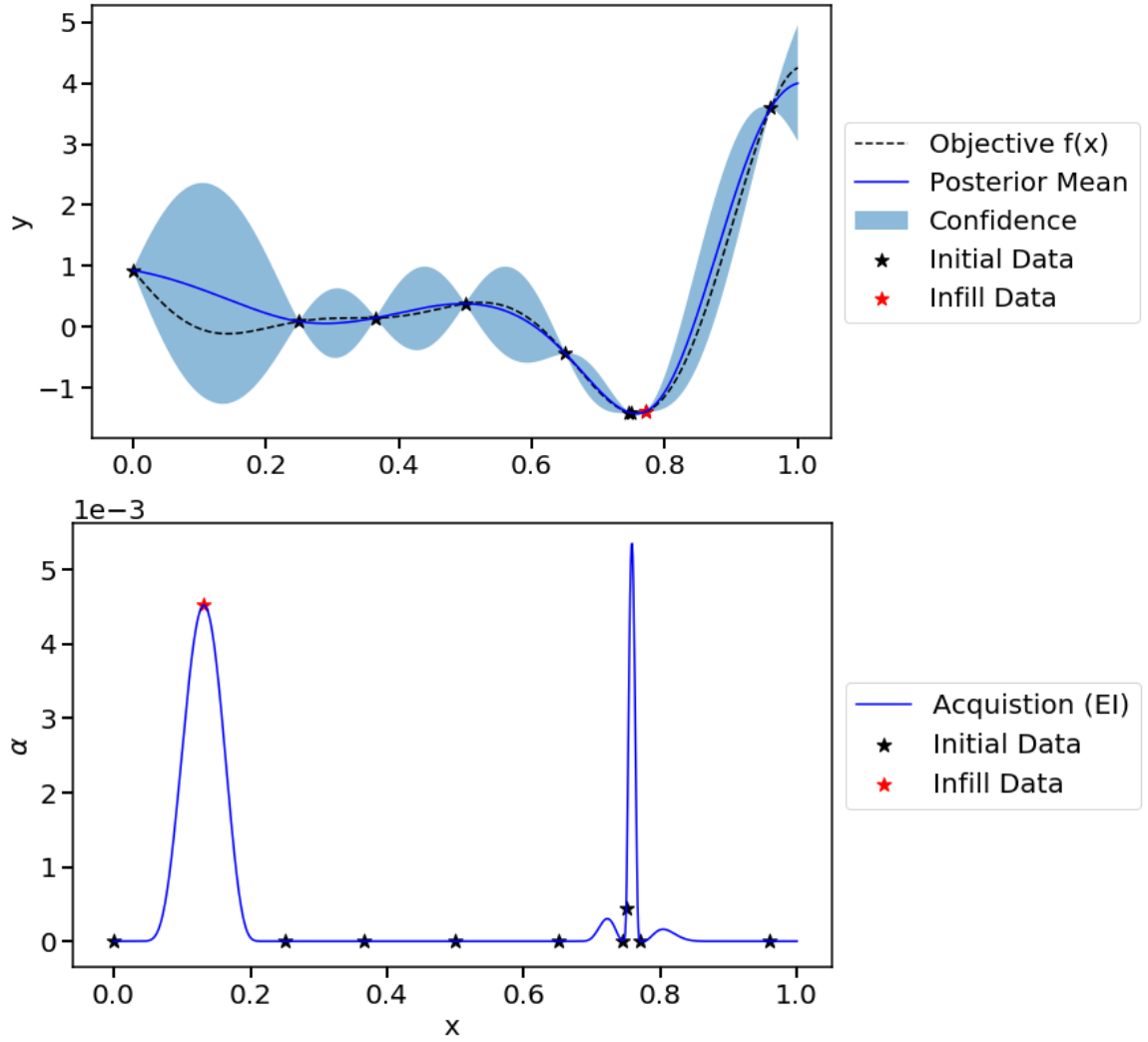
BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 5



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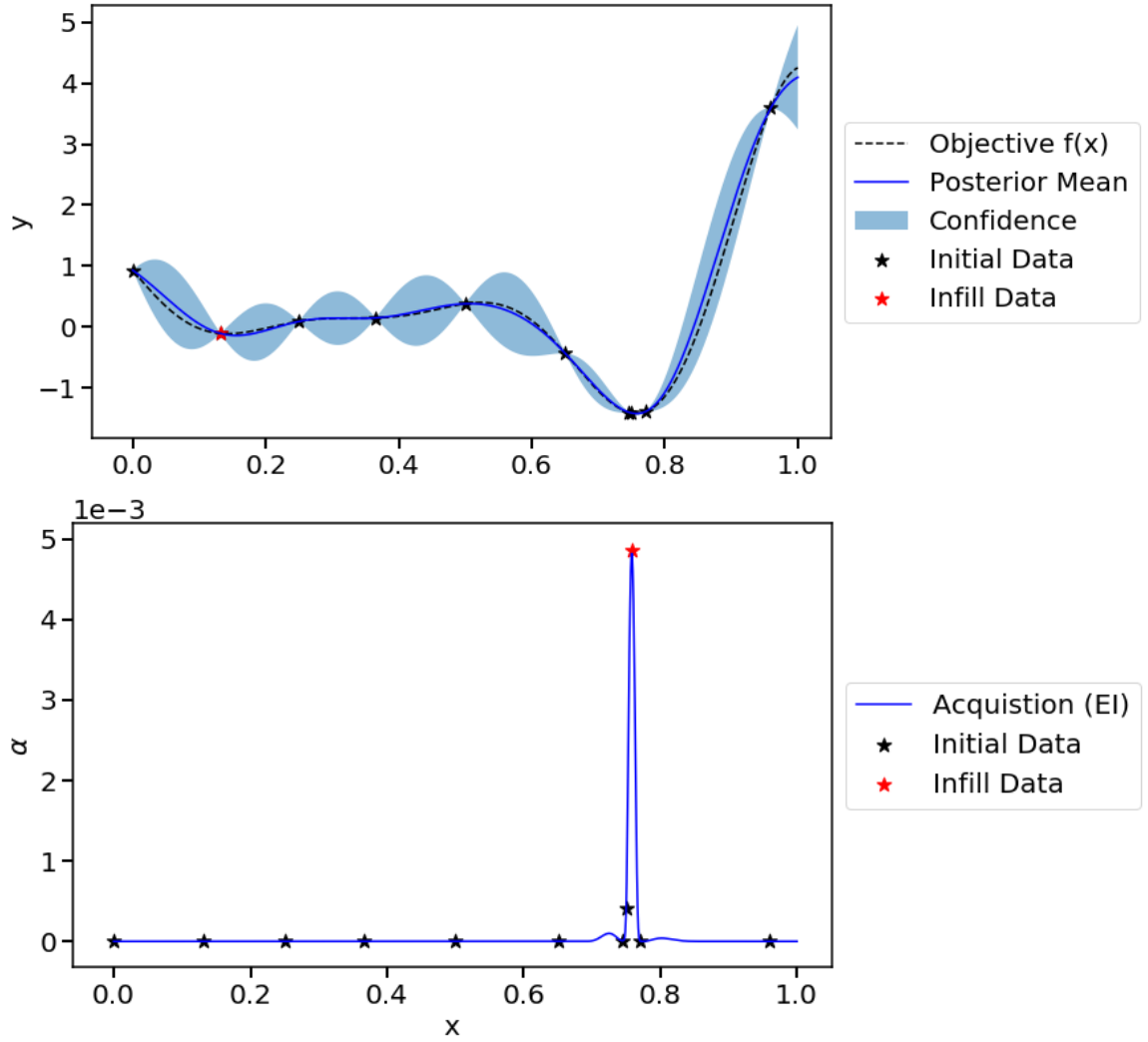
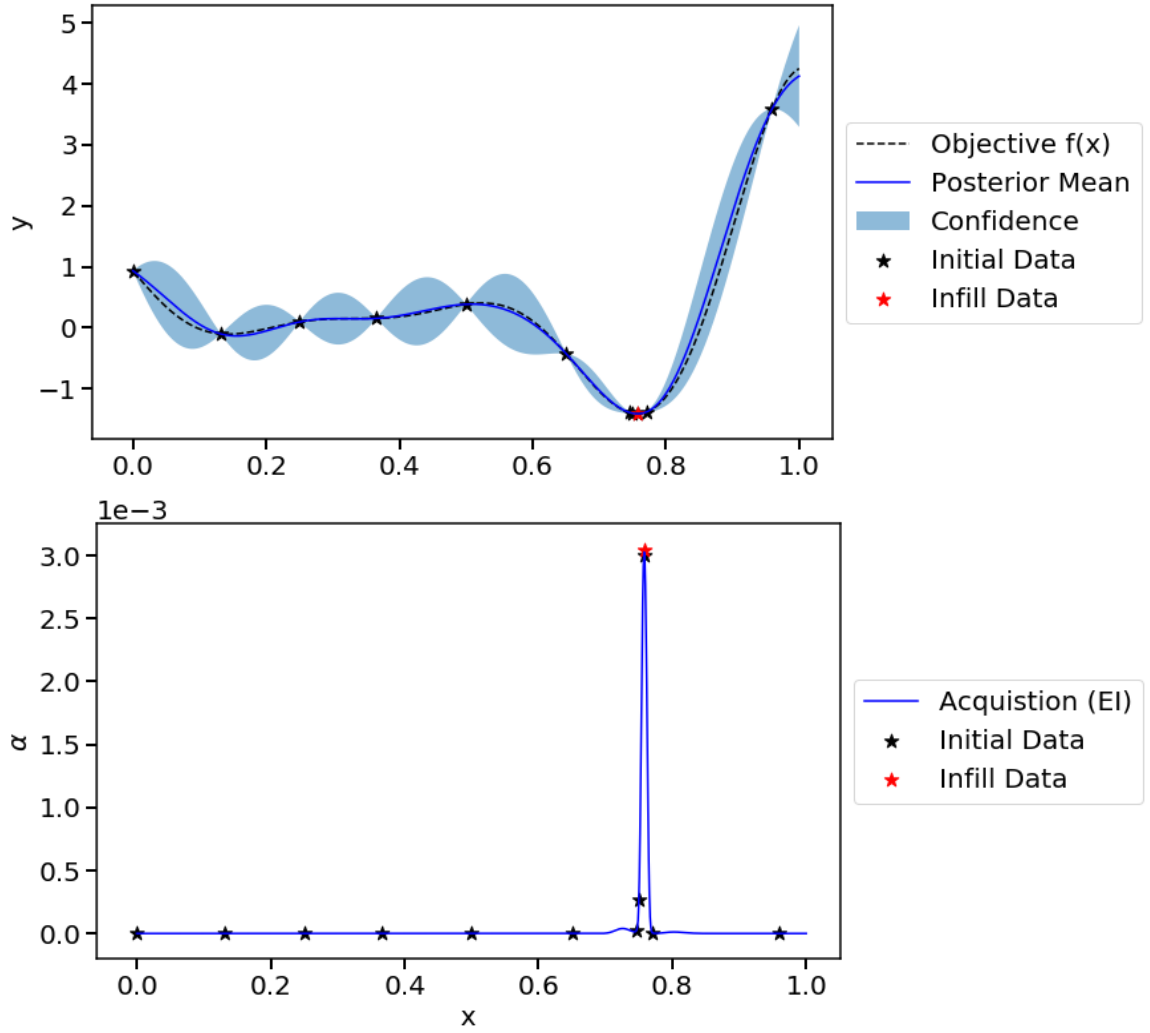
BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 6



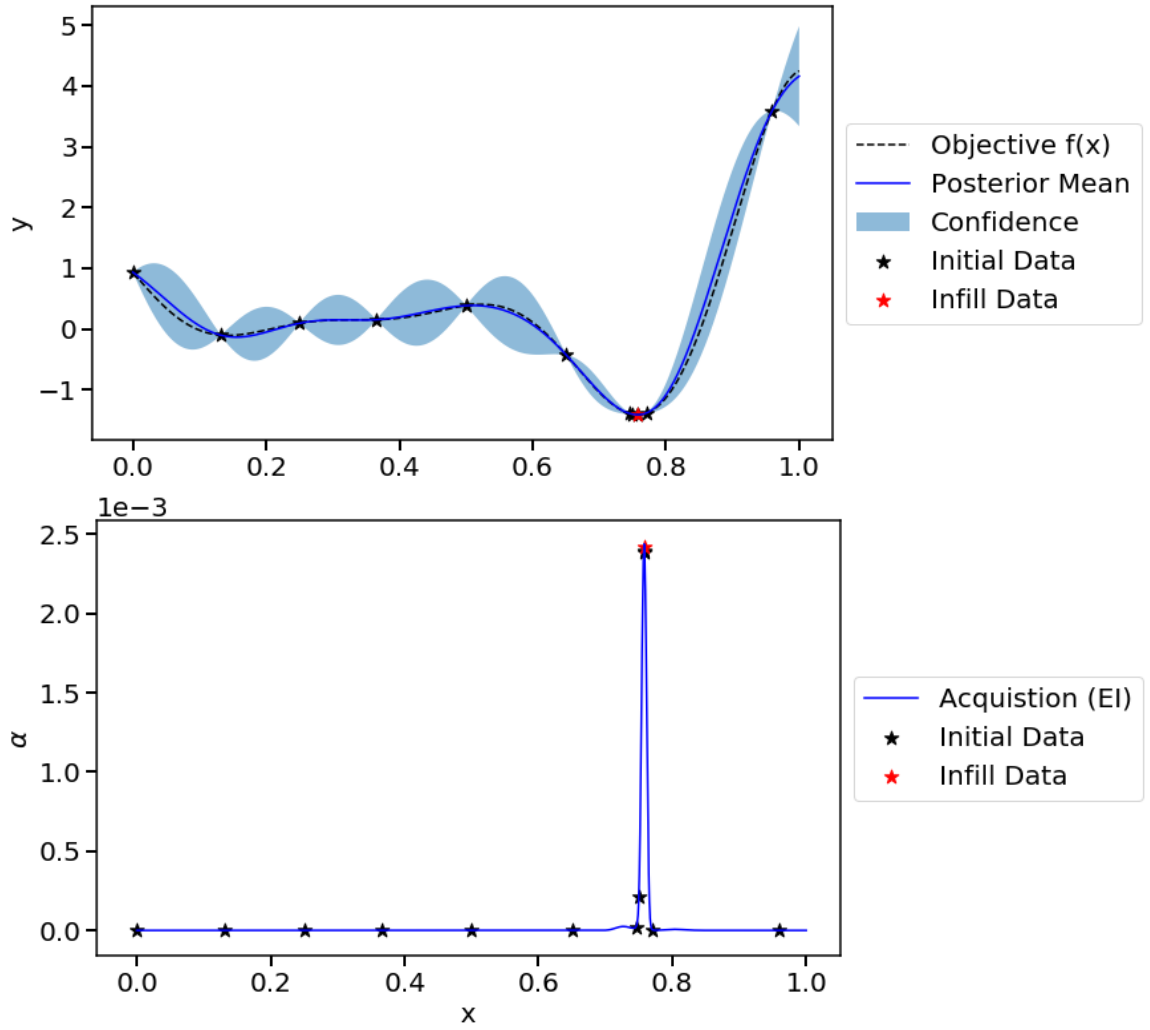
BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 7



BO Examples in 1D

- Find the minima of $f(x) = (6x-2)^2 sin(12x-4); x \in [0,1]$
- Iteration 8



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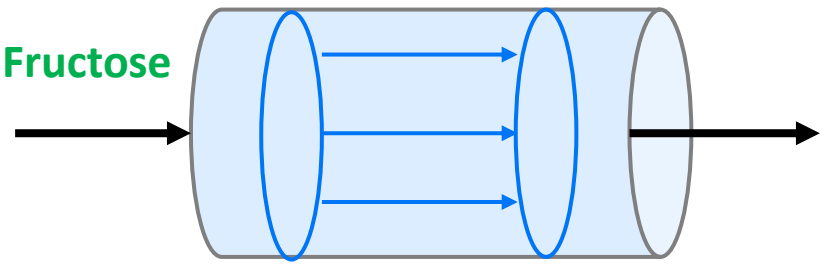
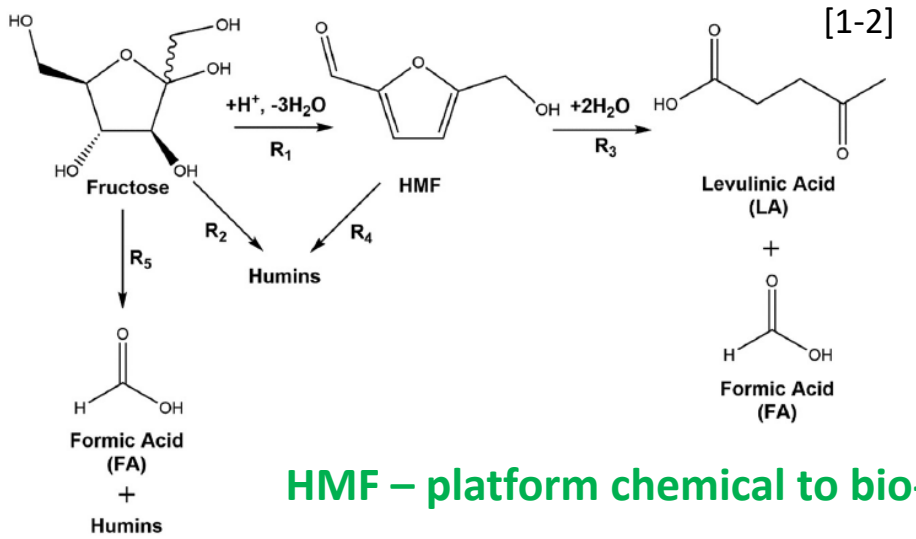
Case Study – HMF Yield Optimization



$$u\frac{dC_i}{dx} = R_i$$

HMF
LA,
FA,
Fructose,
Humins
…



[1-2]

HMF – platform chemical to bio-products

- Goal: maximize the HMF yield (Y) to improve productivity and reduce downstream costs.
- Three key input parameters ($\boldsymbol{X}$) and bounds:
  - X_1, **Temperature** – (140 – 200 °C)
  - X_2, **pH** – (0 – 1)
  - X_3, **Residence time** – (0.01 – 100 min), sampled in log space (-2,2)

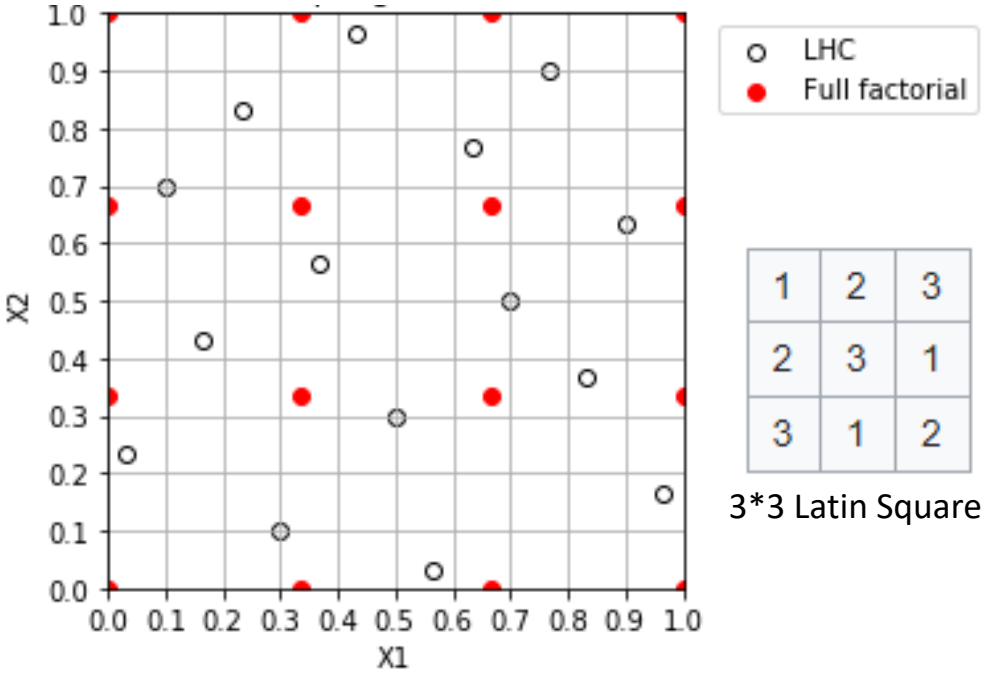
[1] T.D. Swift et al., *ACS Catal.* 4, **2014**, 259.
[2] Desir, P.; Saha, B.; Vlachos, D. G. Energy Environ. Sci. 2019.
Model available in Python and MATLAB: https://github.com/VlachosGroup/Fructose-HMF-Model

Optimized Latin Hypercube design (LHS)

Initial Sampling plan:

- Monte Carlo sampling
- Maximize the distance between points
- Maximize the information gain in the sampling space

2D example (level = 4, 16 samples)

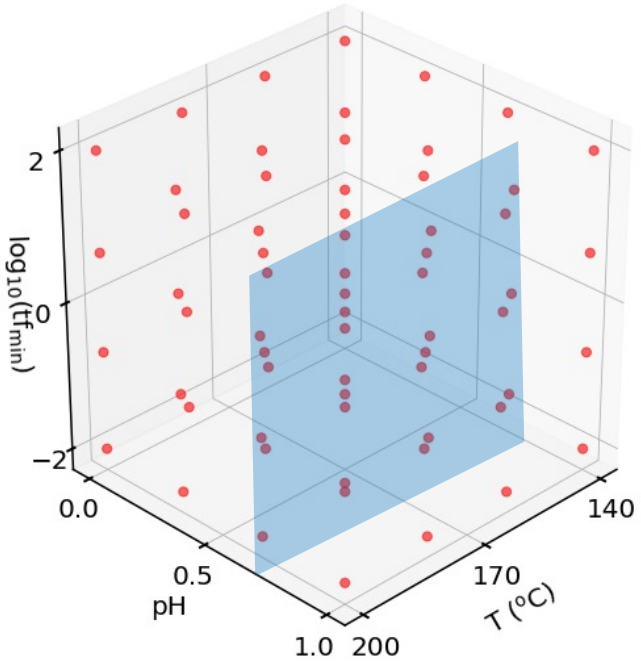


1	2	3
2	3	1
3	1	2

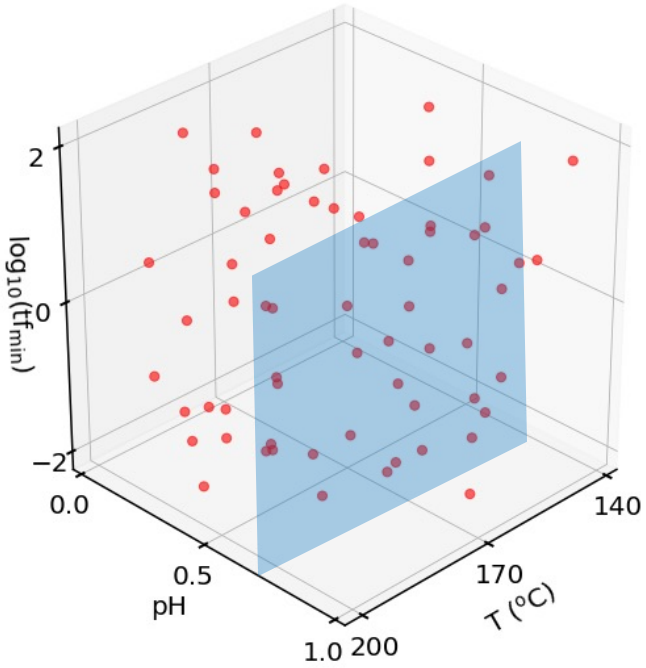
3*3 Latin Square

Sampling Plans

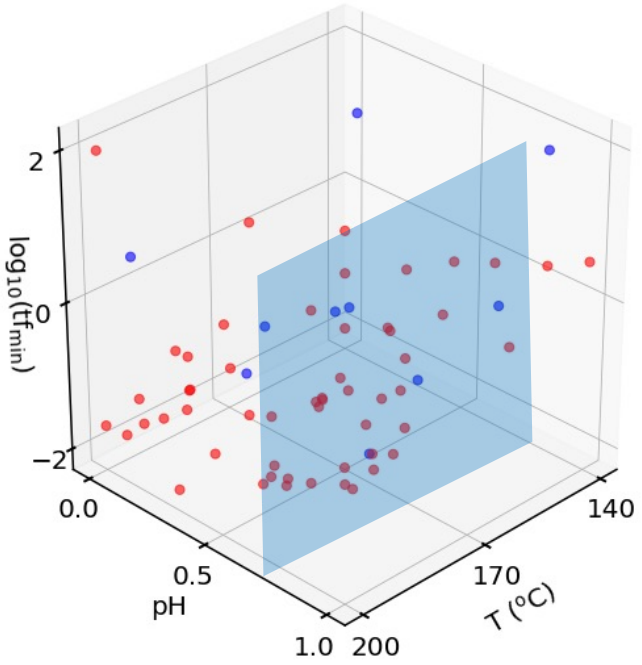
Design 1
Full factorial (level = 4)
64 samples



Design 2
64 random samples

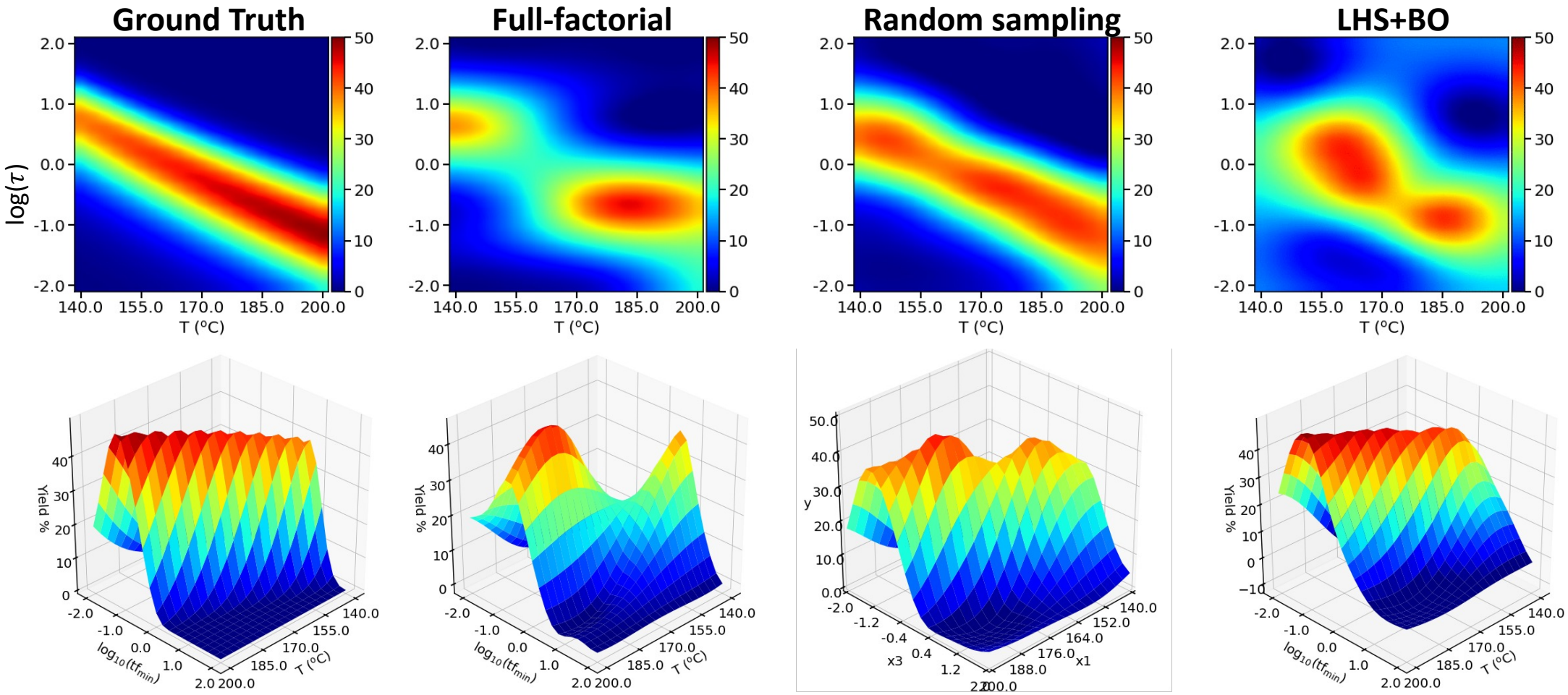


Design 3
10 samples from Latin hypercube (LHS)
54 samples from 54 BO loops,
Acquisition function – EI



- LHS is an efficient space-filling, Monte Carlo sampling method
- We compare response surfaces of HMF yield at **pH=0.7** with varying temperature and residence time

Surrogate Model Performance



- LHS+BO produces more accurate surrogate models

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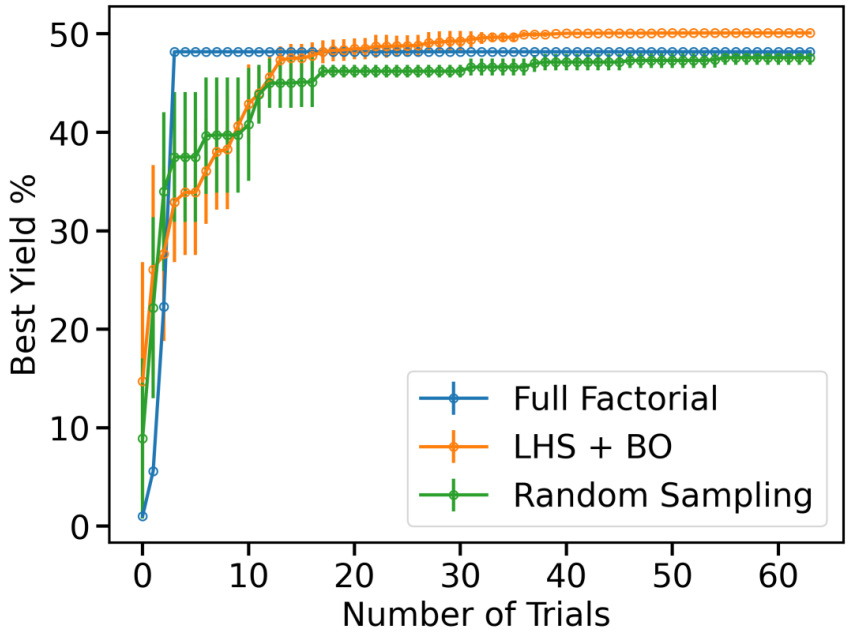
Single-, Multi-Objective Optimization

Maximize HMF Yield

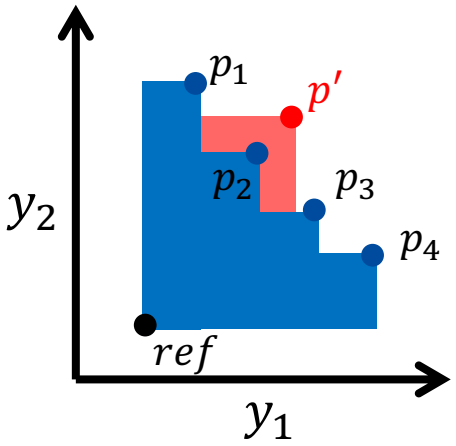
- Optimal condition: Temperature – 200 ° C
  pH – 0.705
  Residence time – 0.076 min (4.56 s)

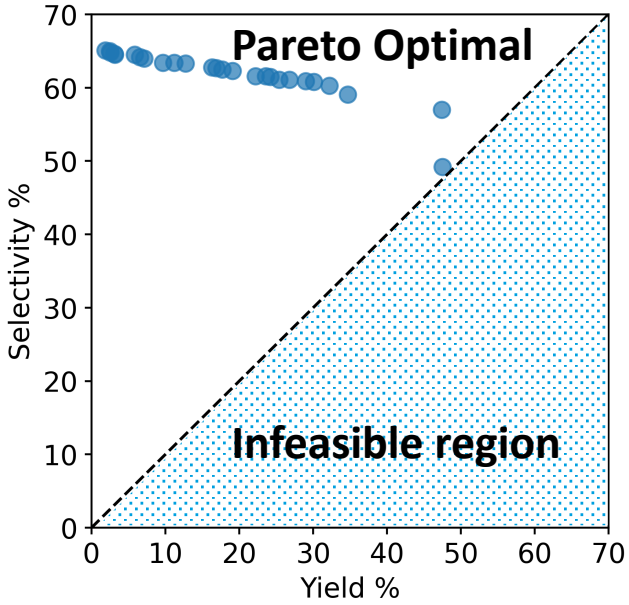


Co-maximize HMF Yield and Selectivity

- HMF Yield = Fructose Conversion × HMF Selectivity
- Fructose Conversion ≤ 100 %

Expected Hypervolume Improvement (EHVI)

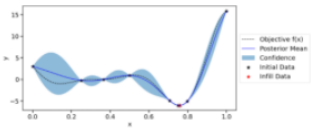




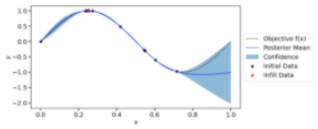
- LHS+BO locates a higher optimal value compared to others
- The runtime of core BO functions completes in seconds per iteration on a laptop CPU
- NEXTorch requires little code and reduces the time or materials for **computations** or **lab experiments**

More Examples

Basic API Usage

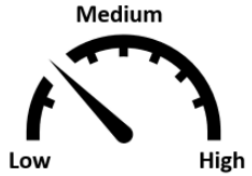


Example 1 - Simple 1d nonlinear function



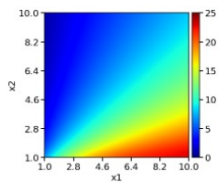
Example 2 - Sin(x) 1d function

Mixed Type Parameters

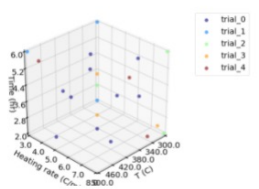


Example 10 - Plug flow reactor yield with mixed type inputs

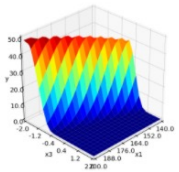
Applications in Reaction Engineering



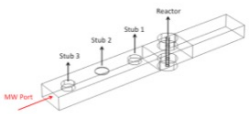
Example 3 - Langmuir Hinshelwood mechanism



Example 4 - Nitrogen-doped carbon catalysts

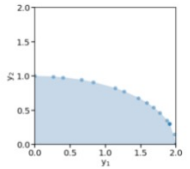


Example 5 - Plug flow reactor yield

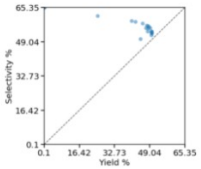


Example 8 - Stub tuner of the microwave cavity

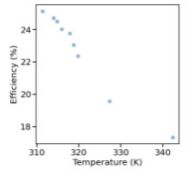
Multi-Objective Optimization(MOO)



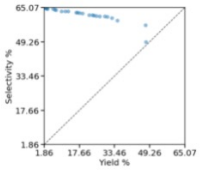
Example 6 - Multi-objective optimization for an ellipse function



Example 7 - Multi-objective optimization for plug flow reactor



Example 9 - Multi-objective optimization for Microwave operating conditions



Example 11 - Multi-objective optimization for plug flow reactor

Documentation: https://nextorch.readthedocs.io/en/latest/index.html