



# Python for early-stage design of sustainable aviation fuels

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A.M. MARTZ, A.E. COMESAÑA, V.H. RAPP, K.E. NIEMEYER

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# Acknowledgements



Lawrence Berkeley  
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**PI**  
Vi Rapp



**Applied Math**  
Ana Comesaña



**Software  
Development**  
Tyler Huntington



**Experimental  
Research**  
Sharon Chen



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Corinne Scown



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University



**PI**  
Kyle Niemeyer



**Mechanical  
Engineering**  
Ali Martz



Energy Efficiency &  
Renewable Energy

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# Design a cake!

- Make a recipe
- Goals:
  - Taste
  - Cost
  - Appearance



# Why is it difficult?

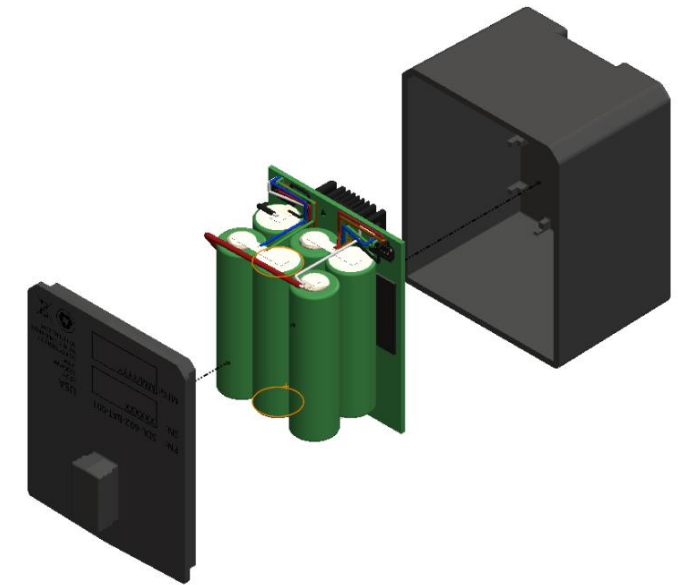
- Multi-objective
  - Taste
  - Cost
  - Appearance
- Multi-parameter
  - Flour
  - Sugar
  - Egg





# No-recipe Cakes in the Real World

- Manufacture of disintegrating tablets<sup>[1]</sup>
- Battery design<sup>[2]</sup>
- Transportation decarbonization



# Transportation Decarbonization

- Different modes have different pathways
- Gasoline vehicles → electric vehicles



- Aviation 2-3% of global CO<sub>2</sub> emissions
- Difficult to decarbonize



# Conventional Aviation Fuel

- Liquid, petroleum fuel
- Stringent requirements
  - Freezing point
  - Flash point
  - Viscosity
  - Density



# Sustainable Aviation Fuel

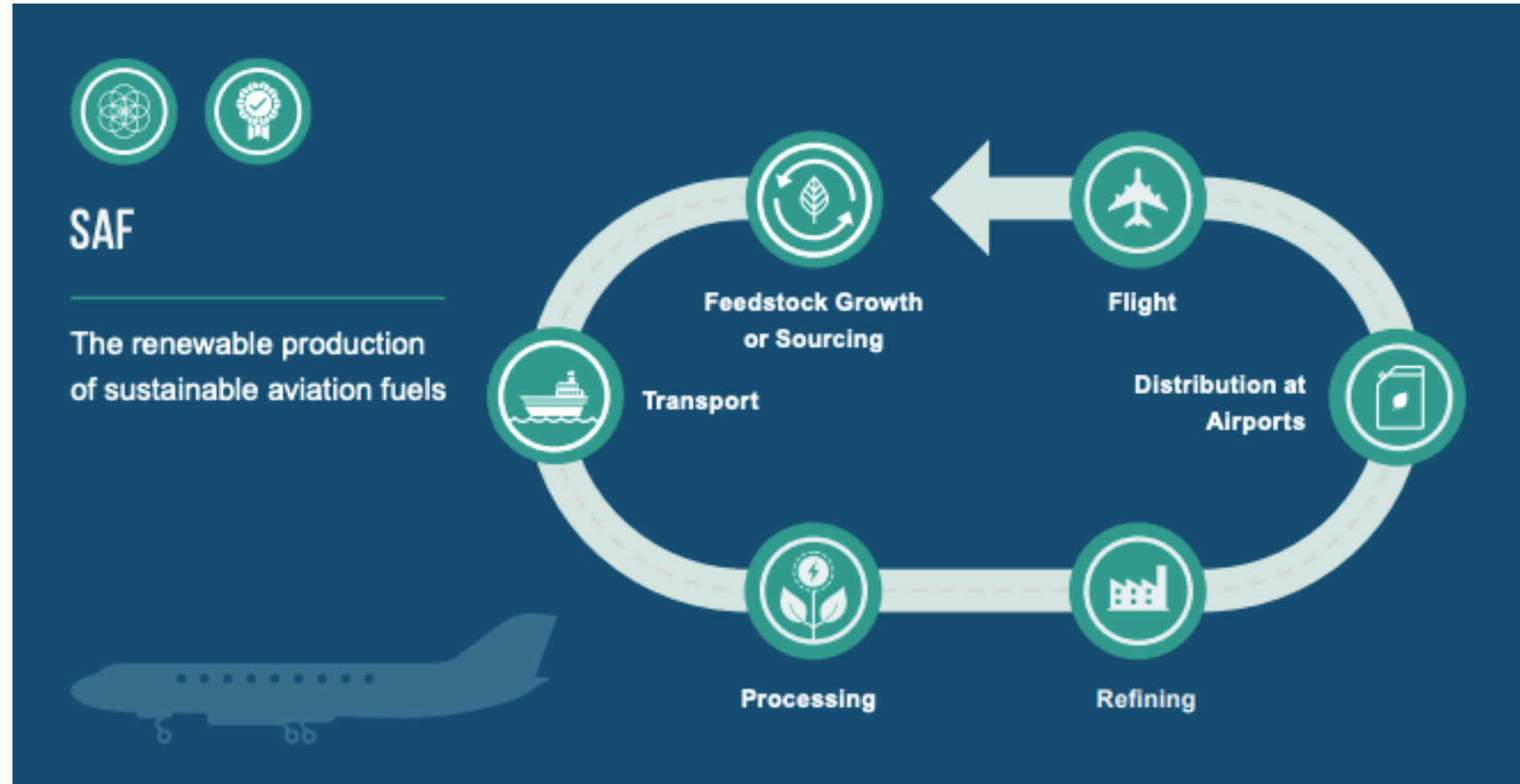
- Primary decarbonization opportunity
- Derived from biomass feedstocks
  - Cooking/plant oil
  - Agricultural residue
- Comparable performance to conventional fuel



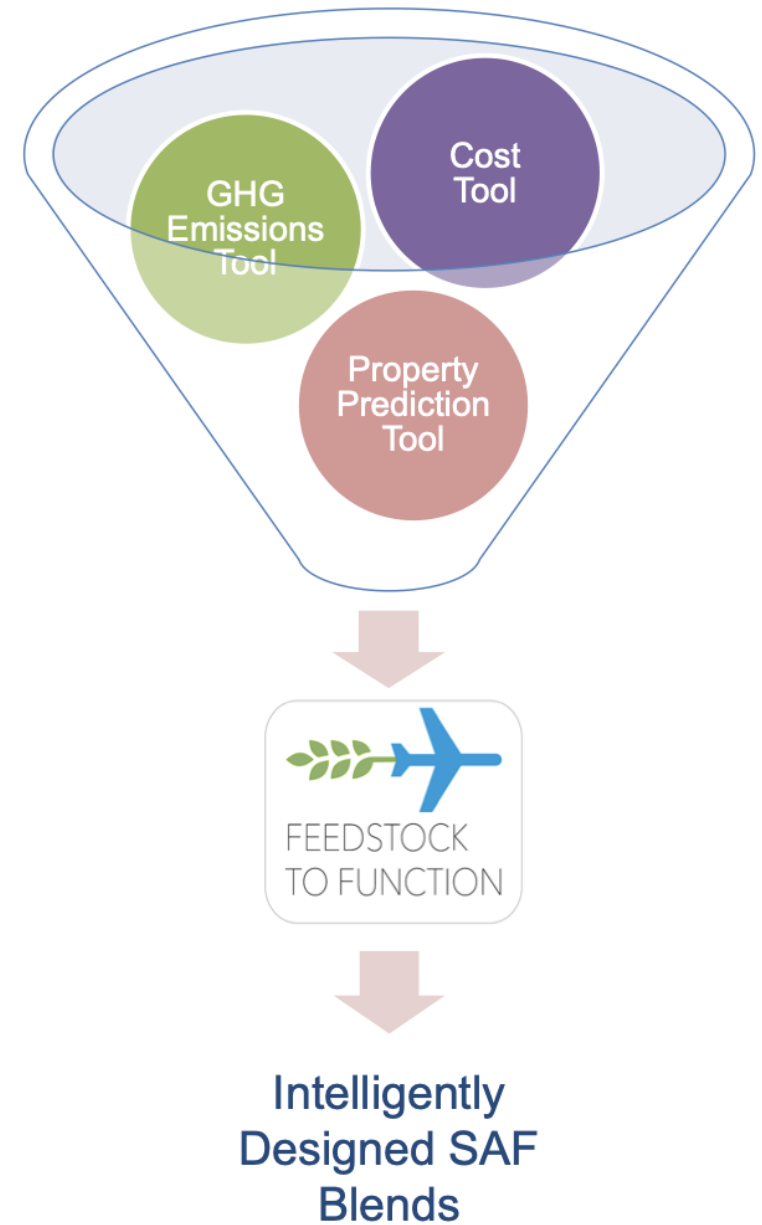


# Sustainable Fuel Implementation Challenges

- Cost
- Scale-up
- Slow development



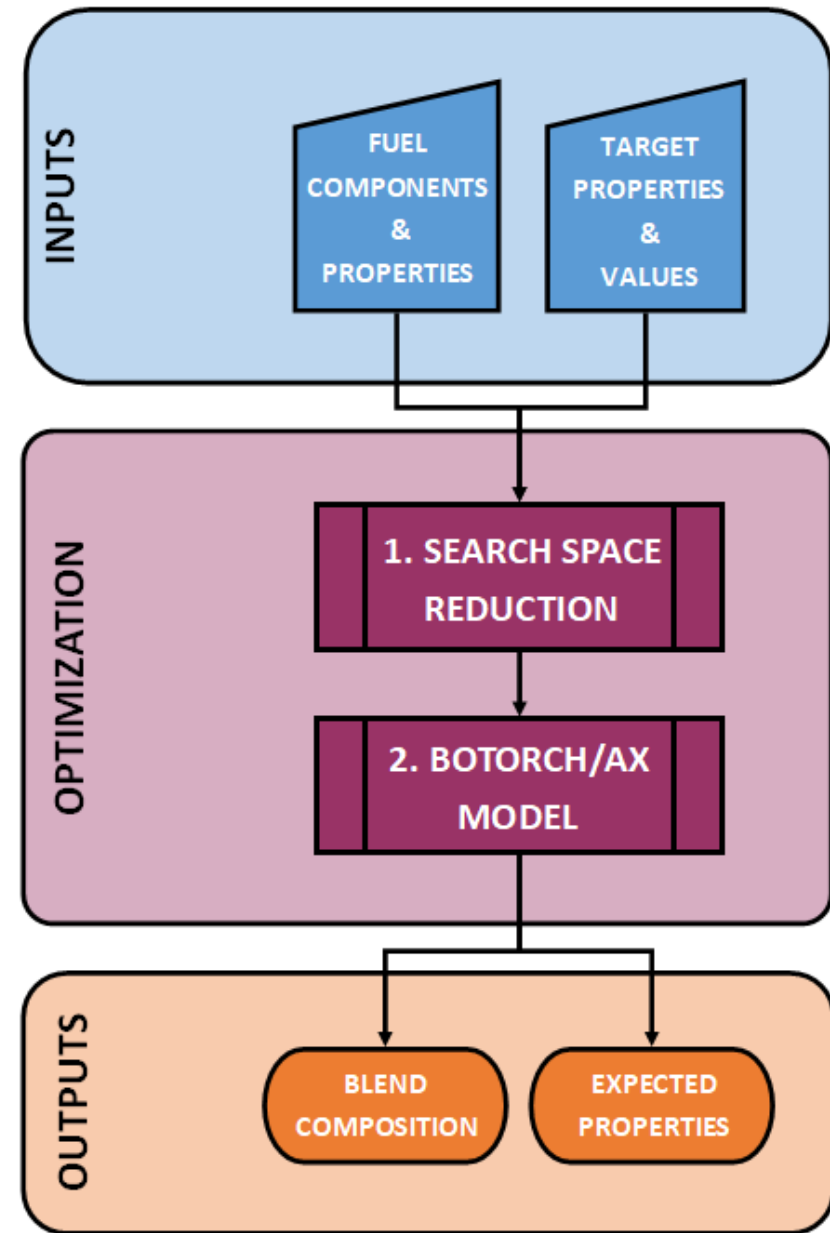
# Create an optimization tool for early-stage design of novel sustainable aviation fuels



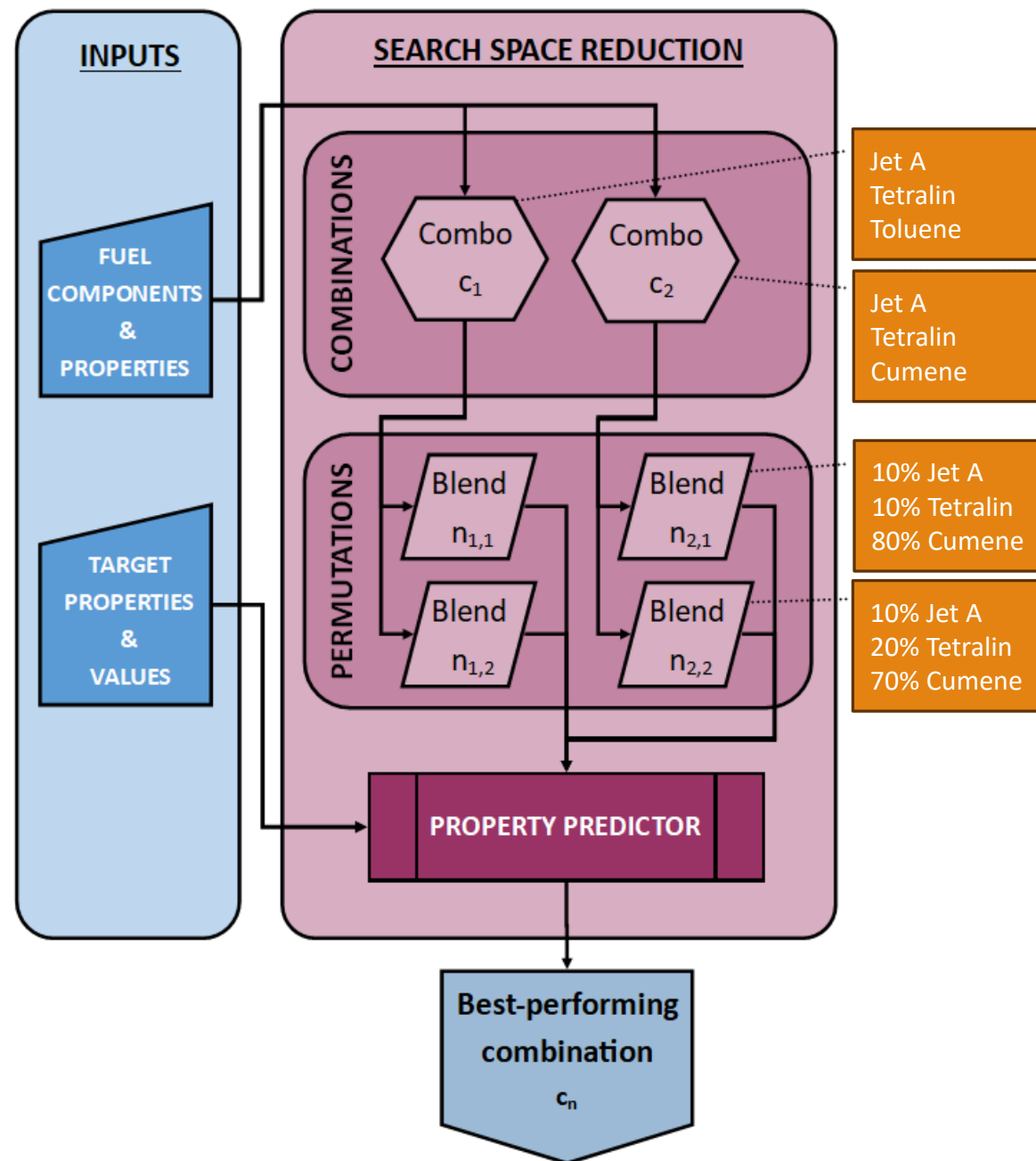
# Methods

Sequential optimization:

1. Search space reduction
2. Multi-objective Bayesian optimization

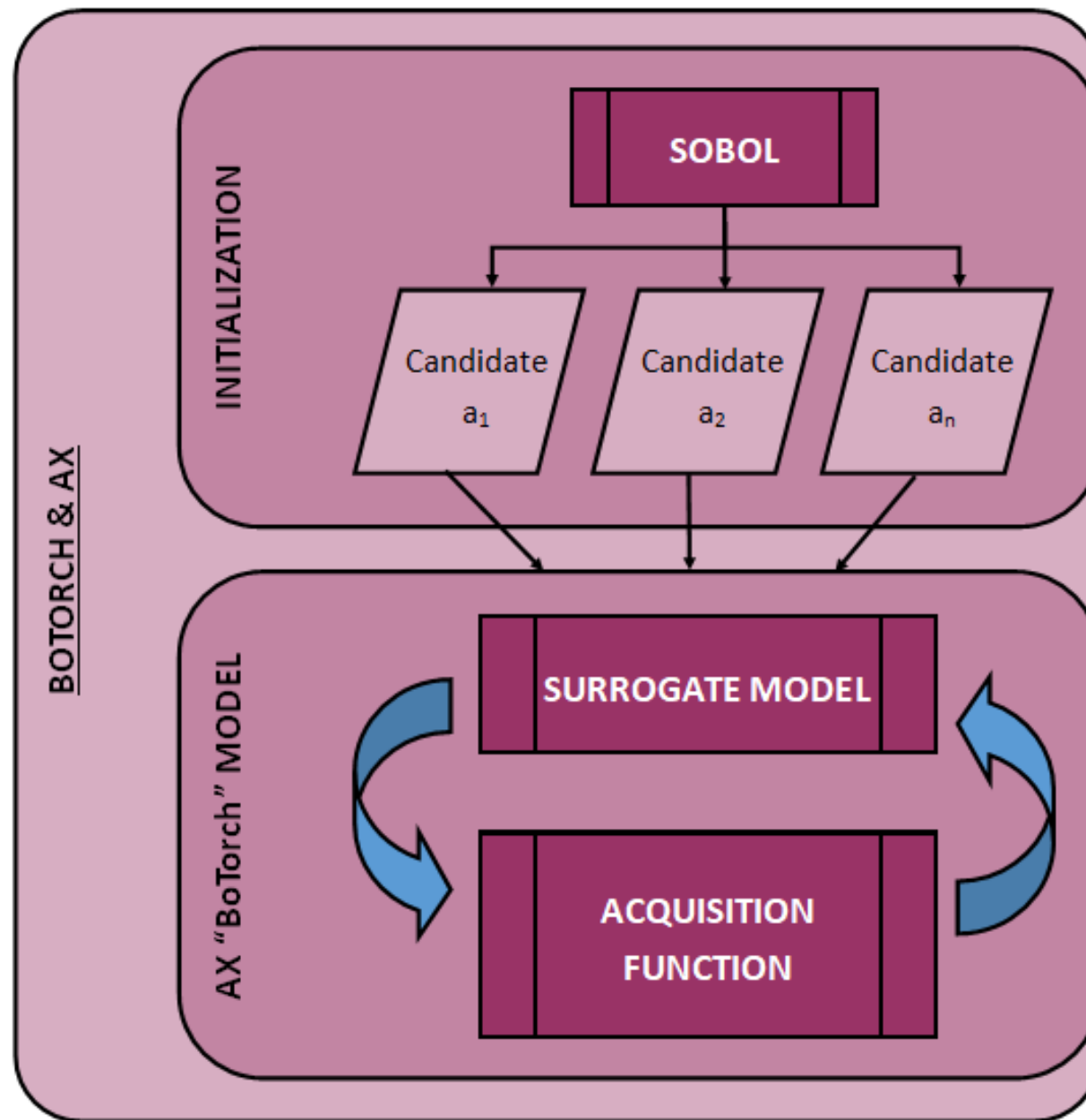


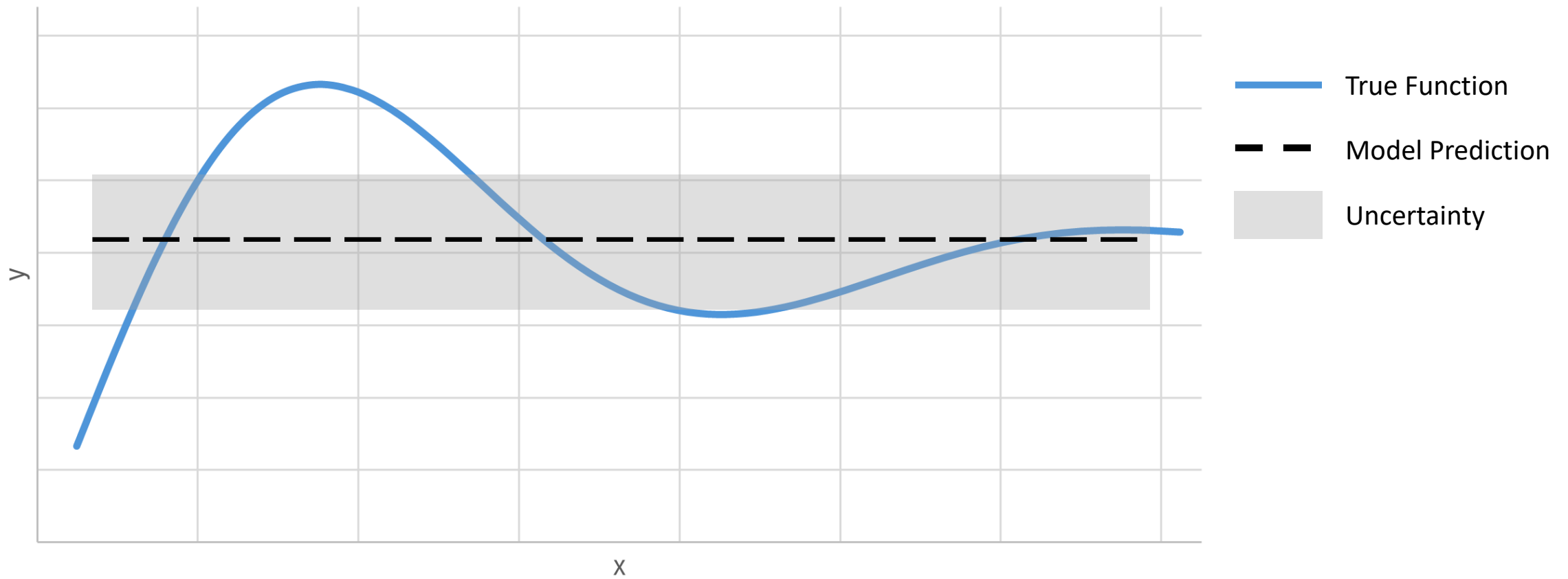
# Search Space Reduction

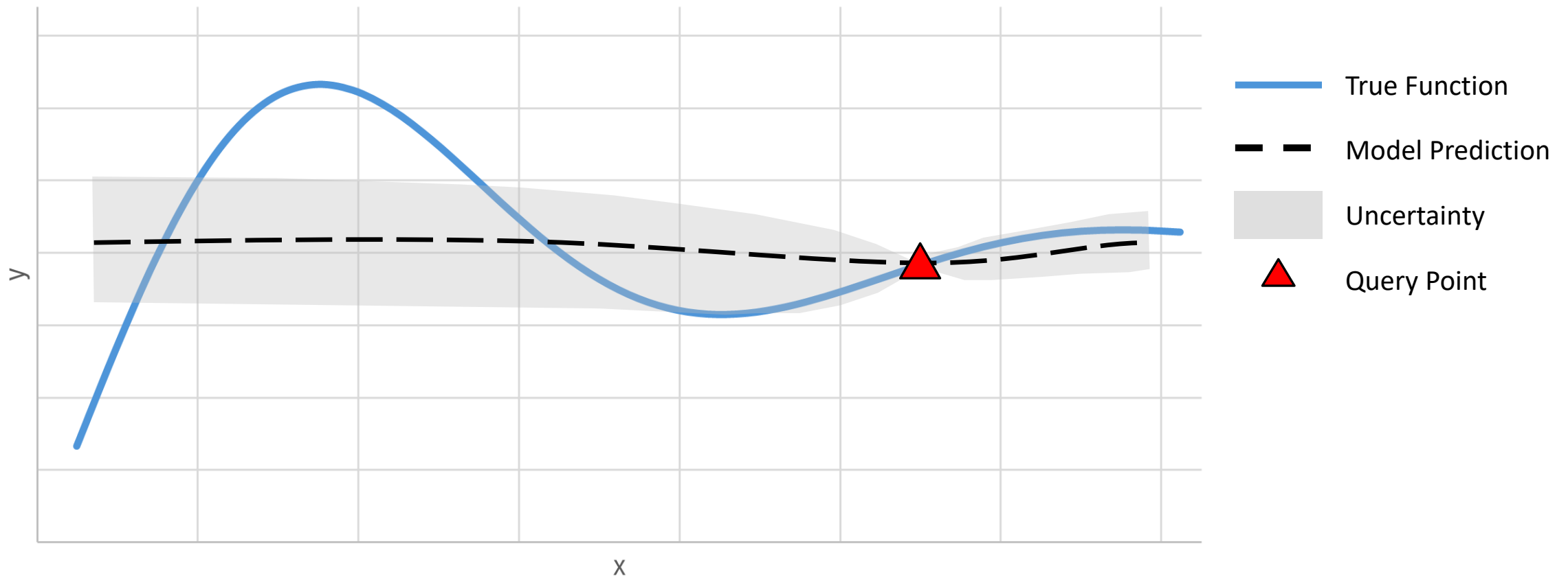


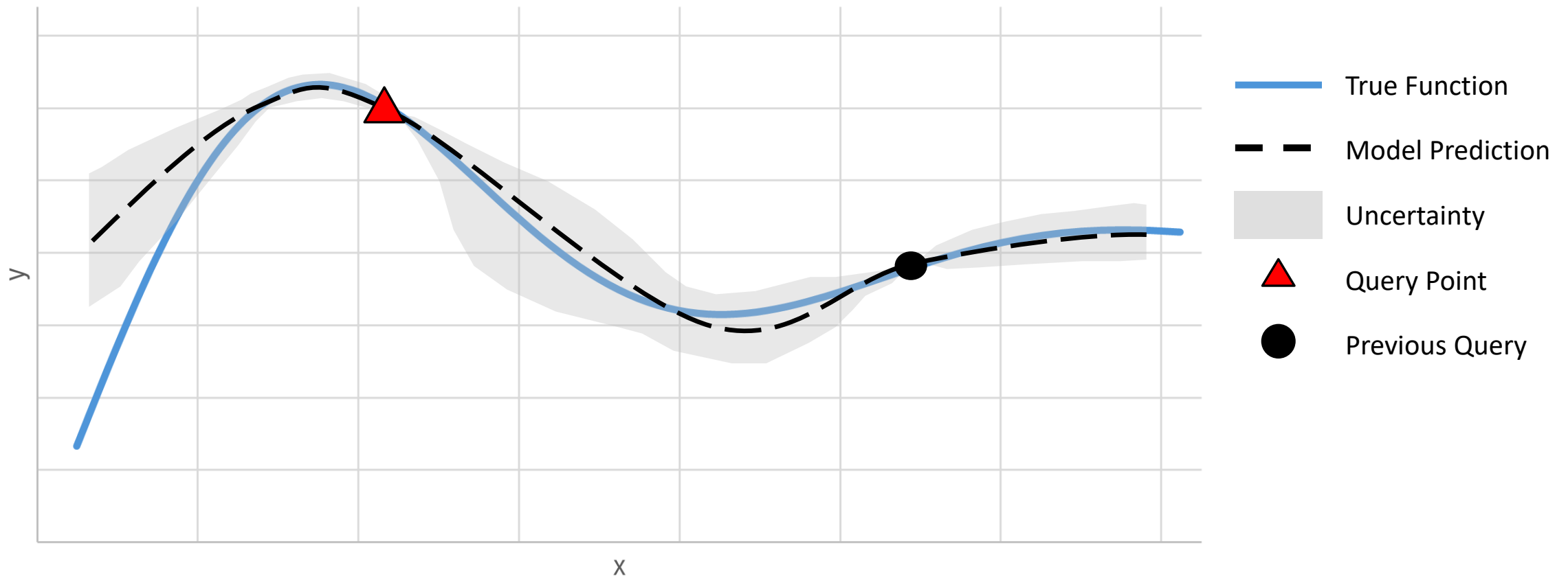


# BoTorch/Ax Bayesian Optimization

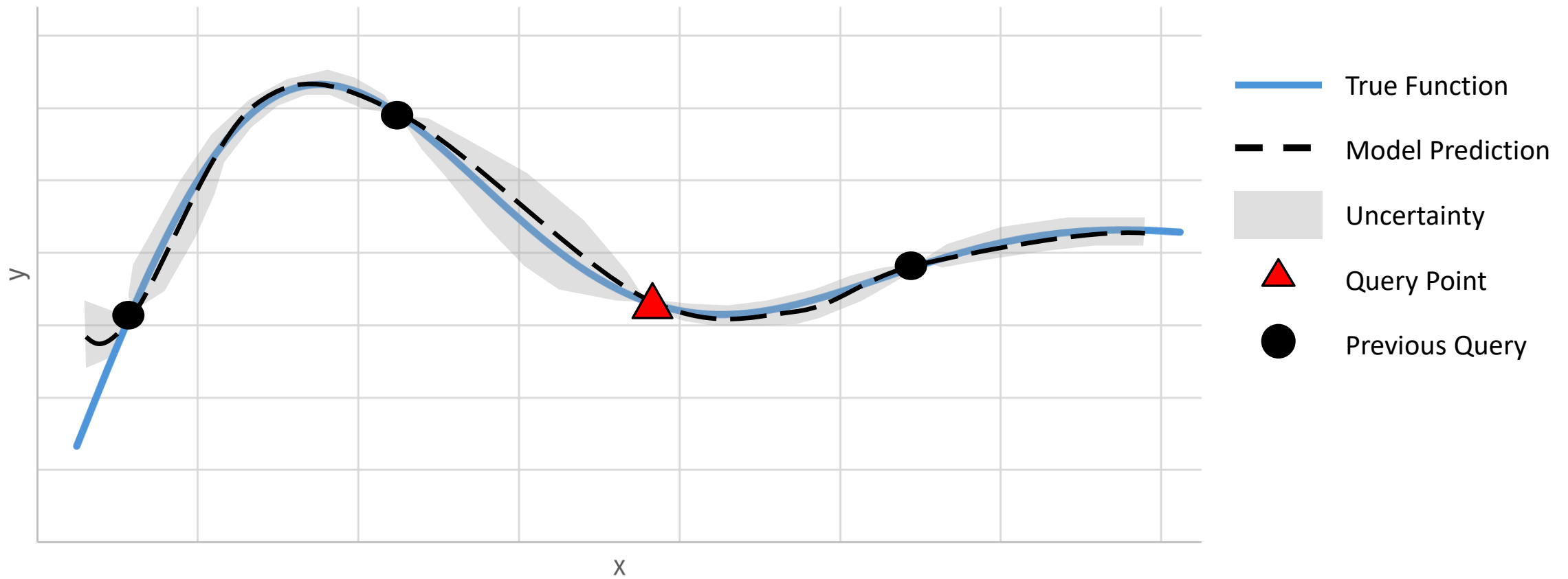


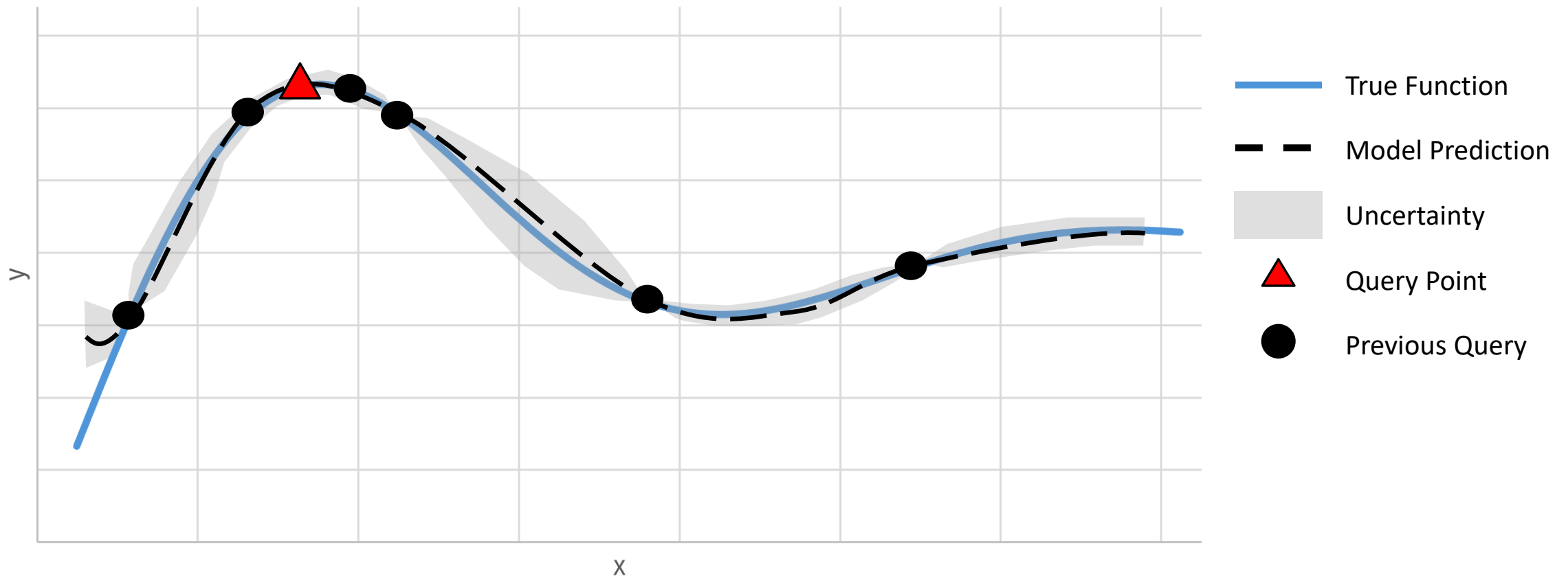






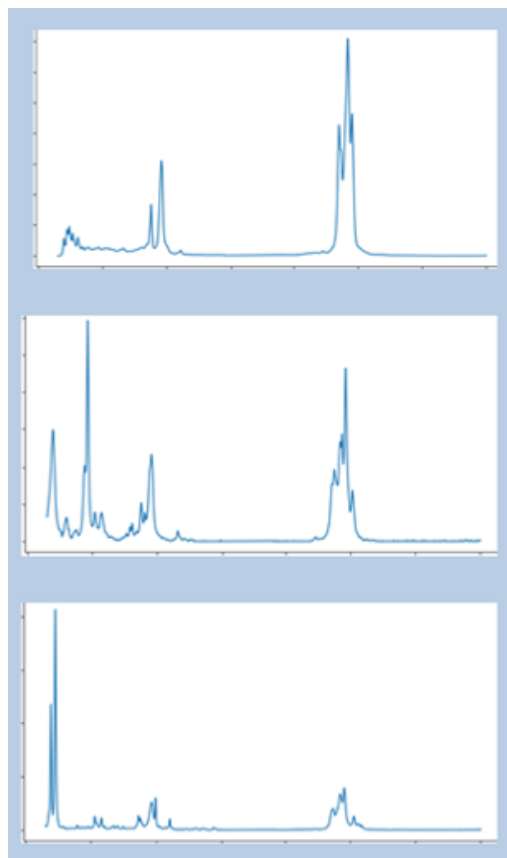






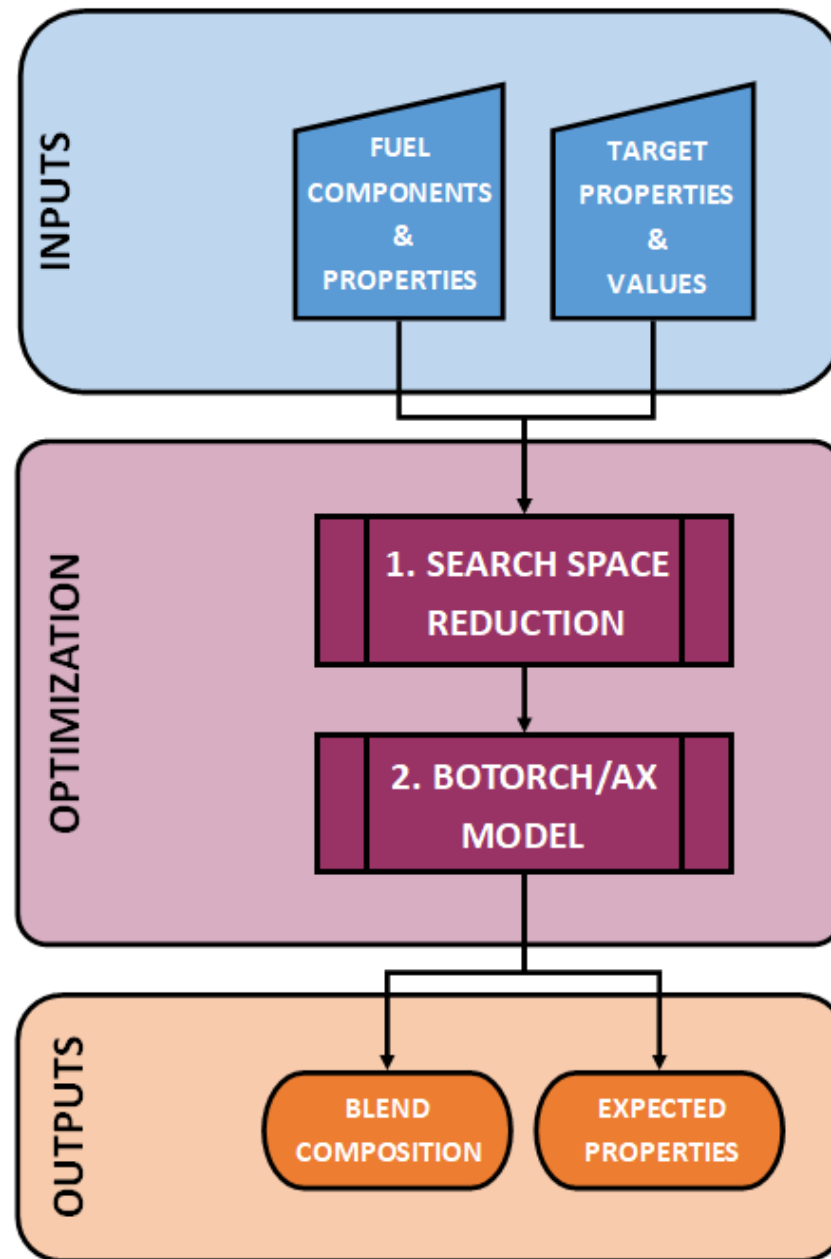
# Property Predictors

- “Black box” prediction
- Modular substitution
- Linear by Volume
- Data-driven predictor



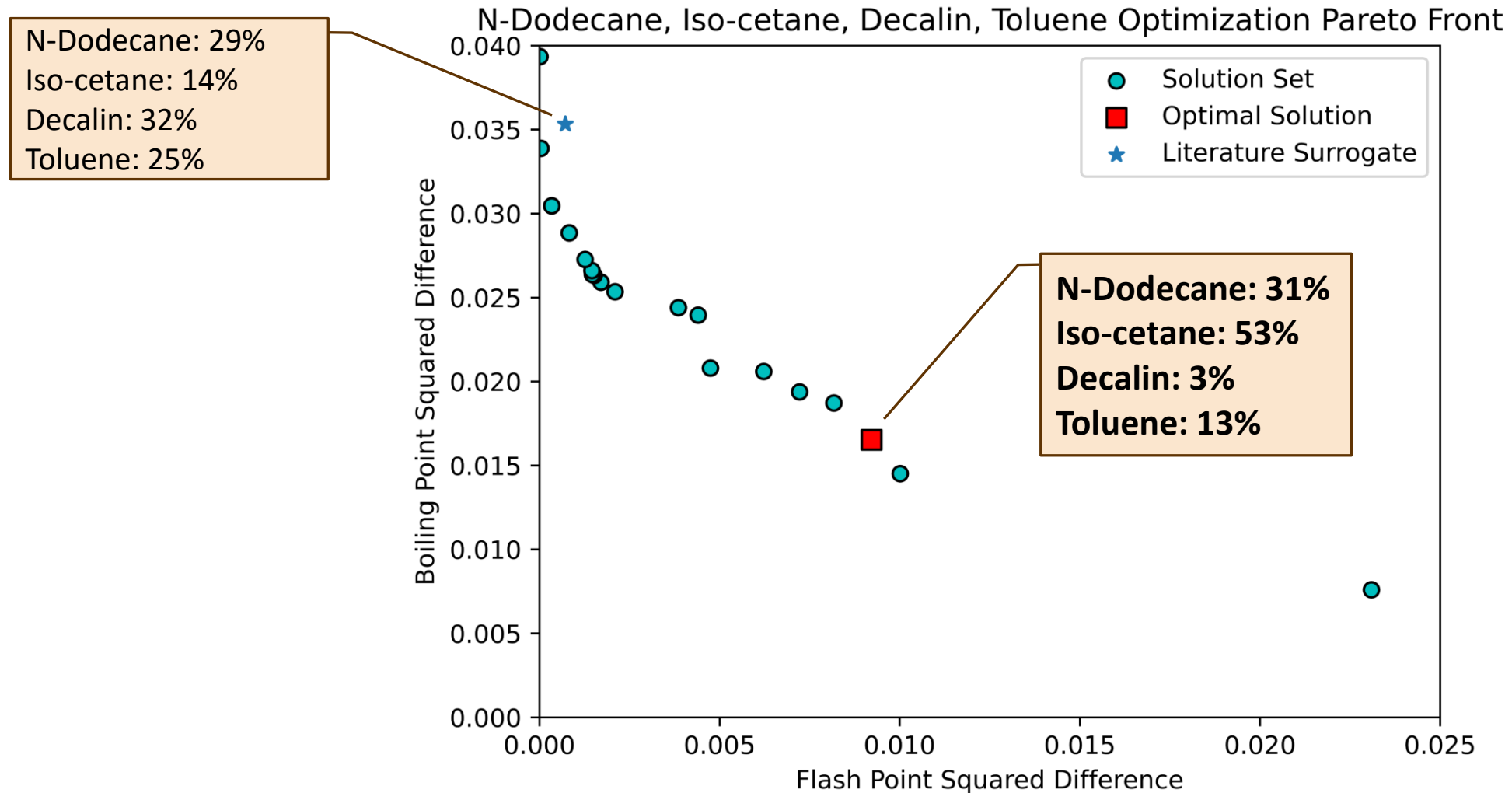
Machine Learning  
Prediction Model

# Review





# Surrogate Comparison [3]



# Design Challenge #1

Possible components:

cumene, methylcyclohexane, toluene, cyclopentane, hexene, isooctane

	Cumene	Methylcyclohexane		Hexene	
Blend Ratio	0.18	0.21		0.61	
	Boiling Point (K)	Flash Point (K)	Melting Point (K)	Density (kg/m <sup>3</sup> )	Viscosity (mm <sup>2</sup> /s)
Target Values	400	280	150	750	1.3
Blended Estimate	376	292	164	742	1.319
+/-	-24	+12	+14	-8	+0.019

# Design Challenge #2

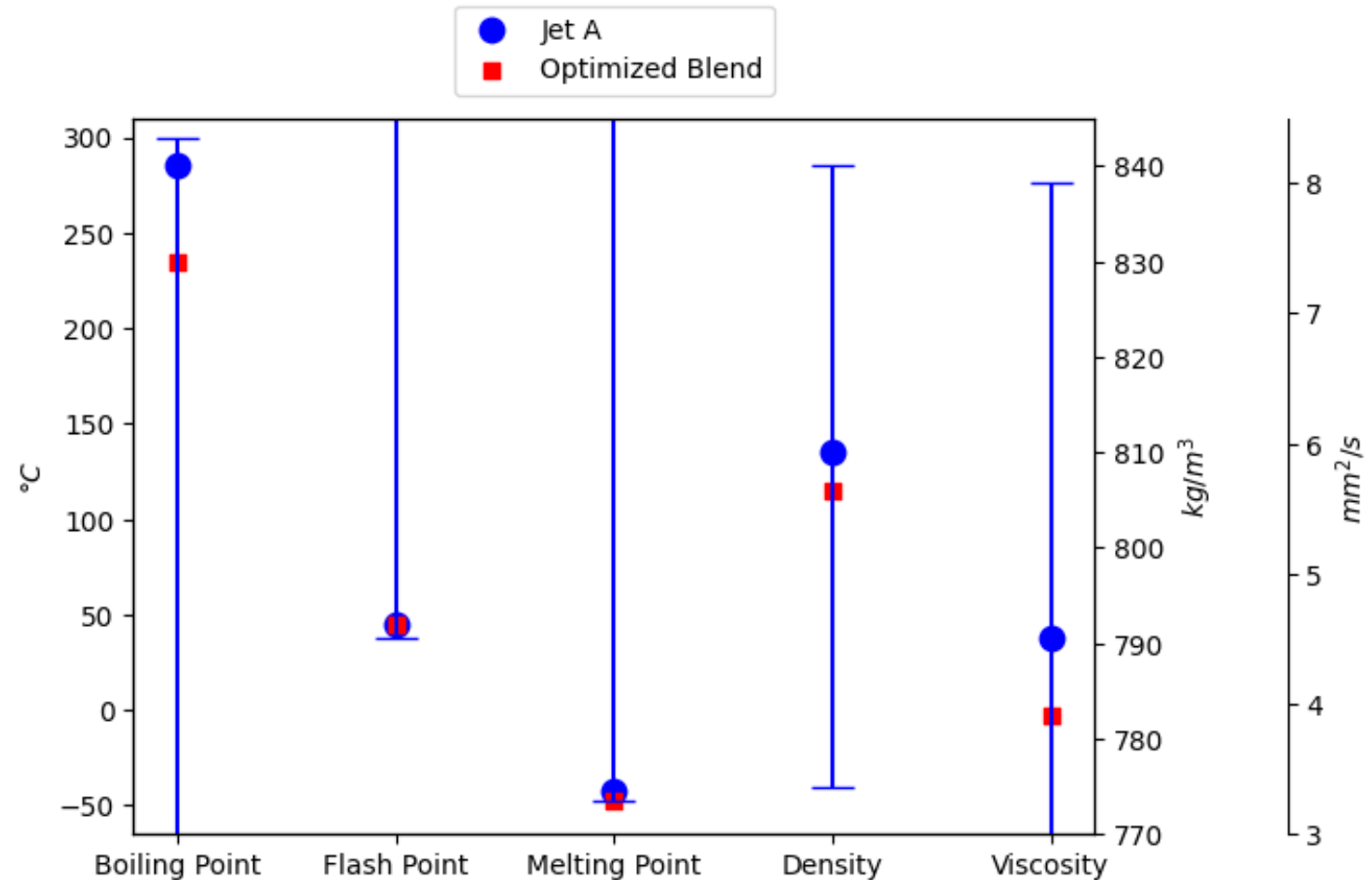
Requirement: 0-75% Jet A

Possible components:

60% heptane/40% isooctane,  
tetralin, cumene, toluene,  
methylpentane, isooctane

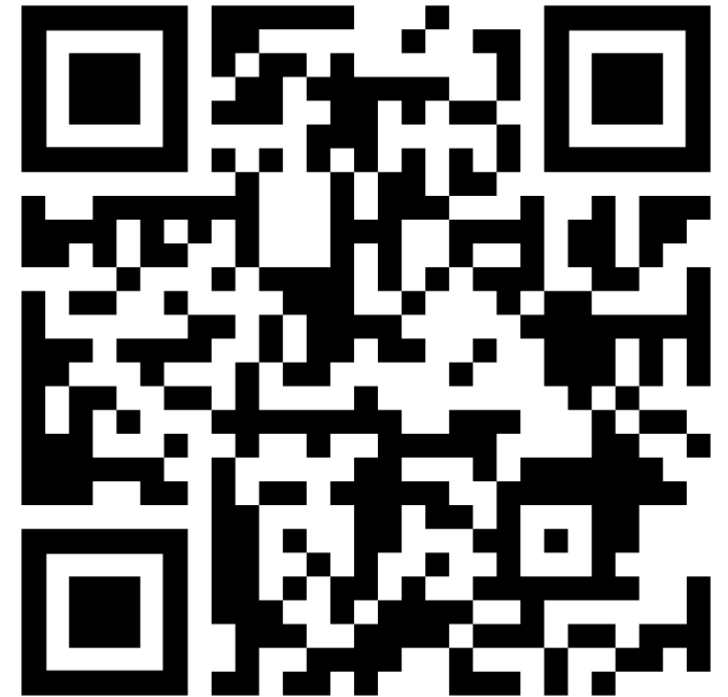
Blend Ratio:

74% Jet A  
08% 60/40 heptane/isooctane  
08% Tetralin  
10% Cumene



# Summary

- Developed multi-objective, multi-parameter optimization approach
- Tool designed for variability
  - Number of components
  - Required components
  - Predictors
- Expanded Feedstock to Function capabilities



<https://feedstock-to-function.lbl.gov/>



# Credits

- [1] Sano, Syusuke, et al. “Application of Bayesian Optimization for Pharmaceutical Product Development.” *Journal of Pharmaceutical Innovation*, vol. 15, no. 3, Sept. 2020, pp. 333–43. *Springer Link*, <https://doi.org/10.1007/s12247-019-09382-8>.
- [2] Thelen, Adam, et al. “Sequential Bayesian Optimization for Accelerating the Design of Sodium Metal Battery Nucleation Layers.” *Journal of Power Sources*, vol. 581, Oct. 2023, p. 233508. *ScienceDirect*, <https://doi.org/10.1016/j.ipowsour.2023.233508>.
- [3] Kim, Doohyun, Jason Martz, and Angela Violi. “A Surrogate for Emulating the Physical and Chemical Properties of Conventional Jet Fuel.” *Combustion and Flame*, vol. 161, no. 6, June 2014, pp. 1489–98. *ScienceDirect*, <https://doi.org/10.1016/j.combustflame.2013.12.015>