

# Brake pad **recipe creation** using Gen AI



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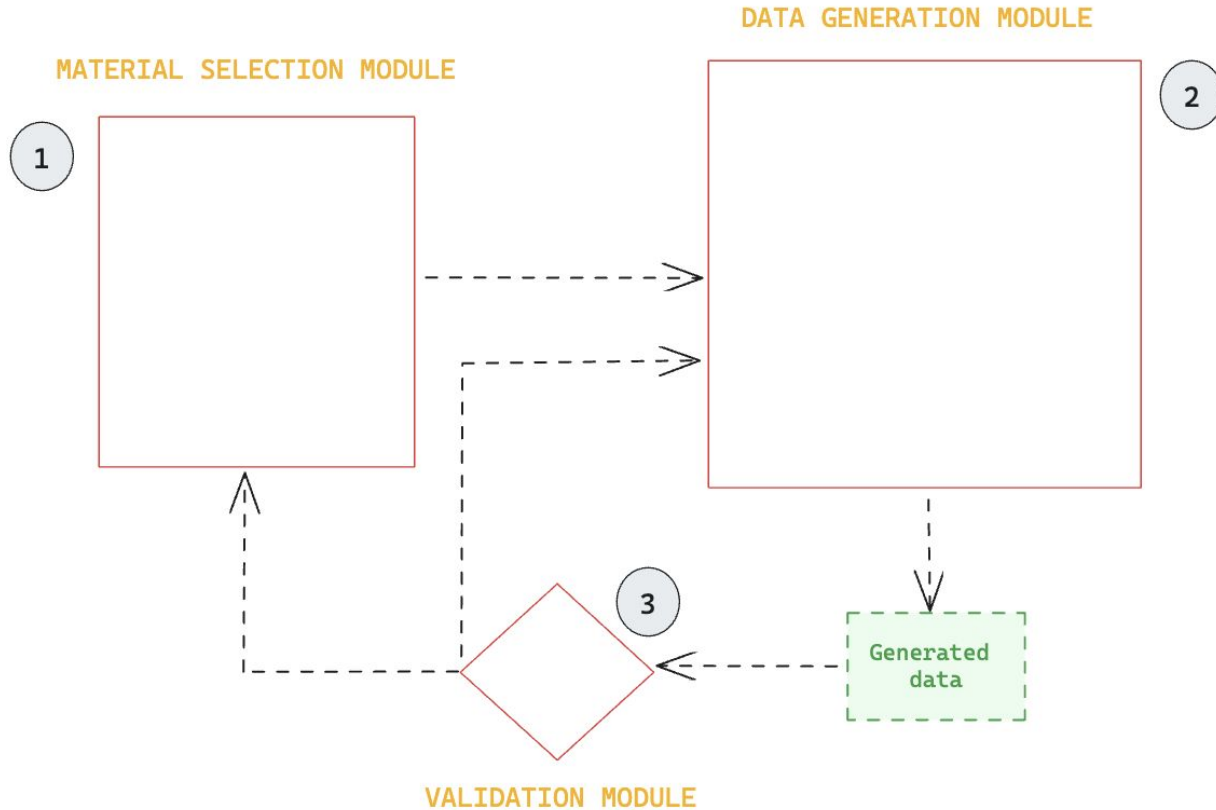
# Motivation

Potential for high Impact

Large Language Models (LLMs) perform good on time series

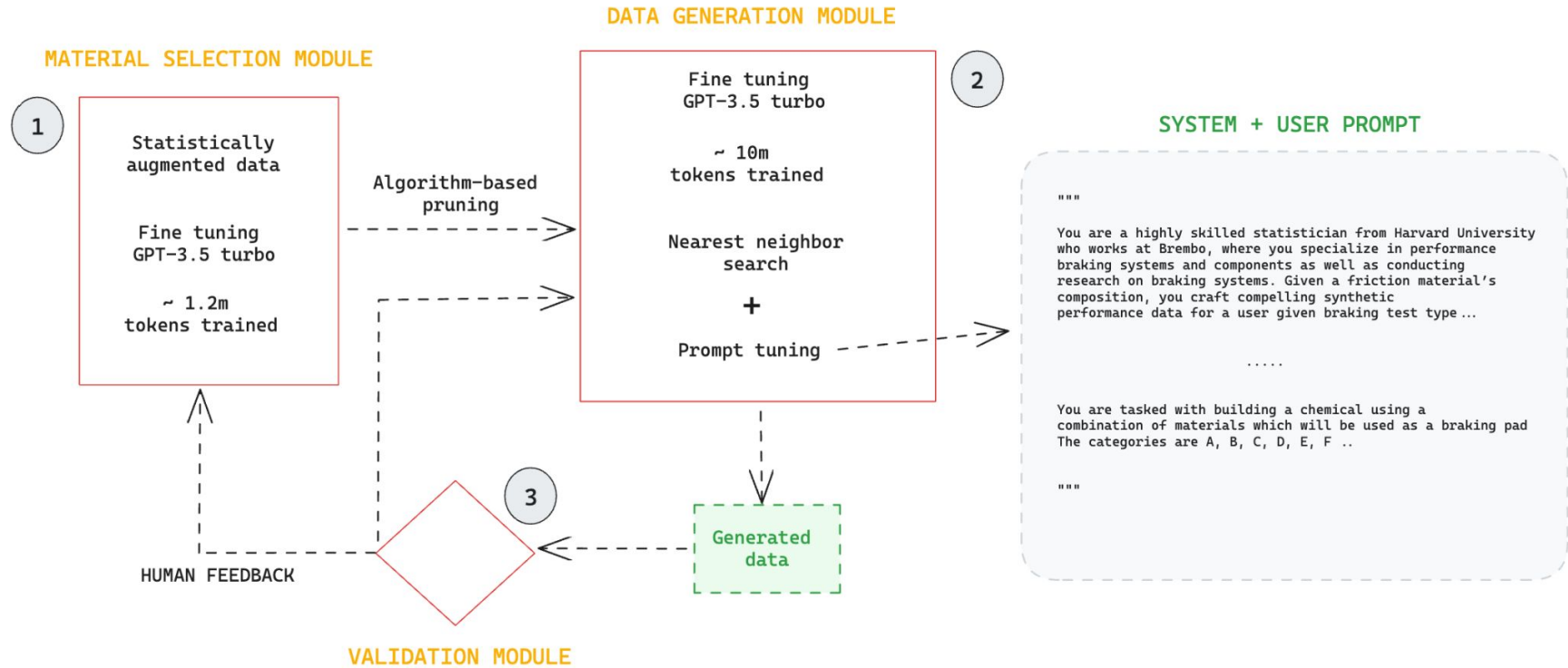
Challenging!

# Our system in a nutshell

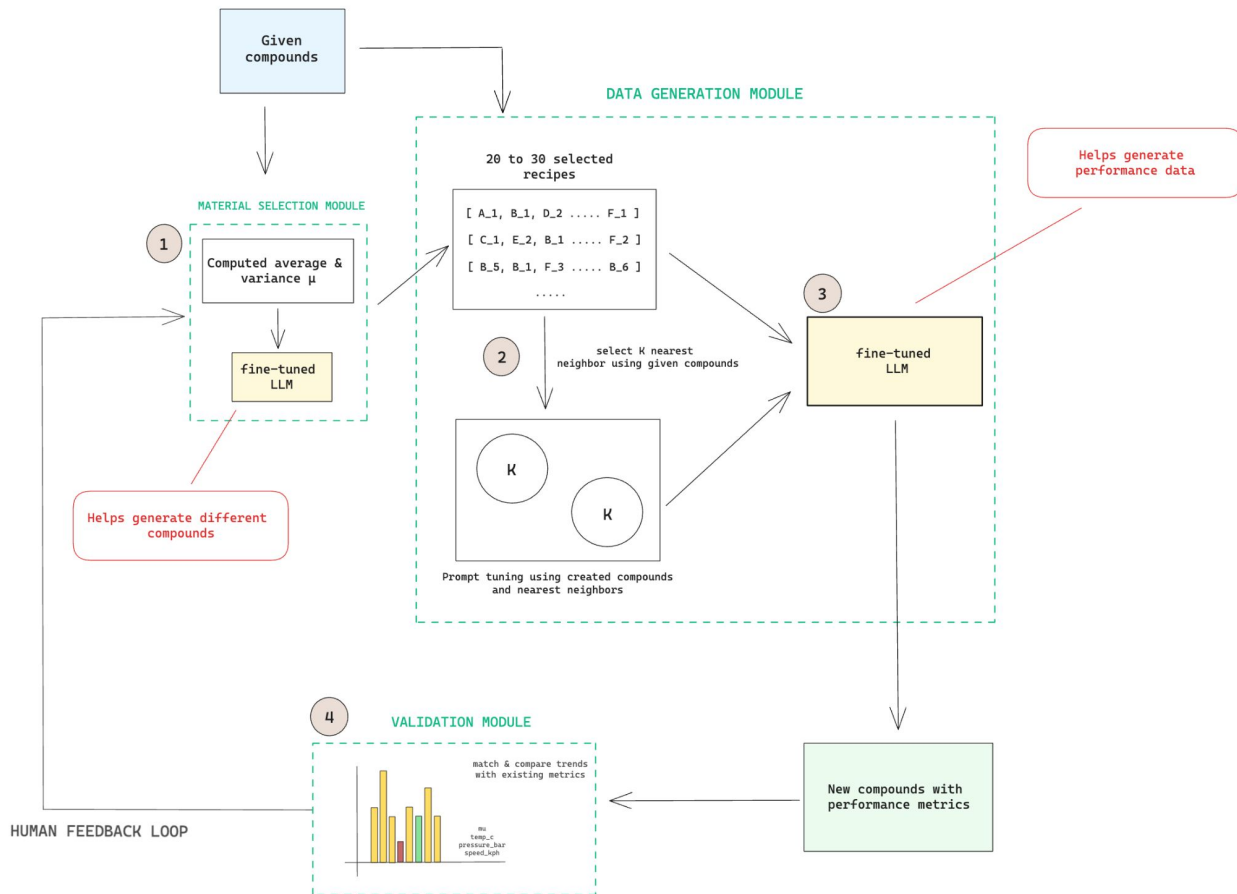


**Simple Much?**  
**Let's Zoom In**

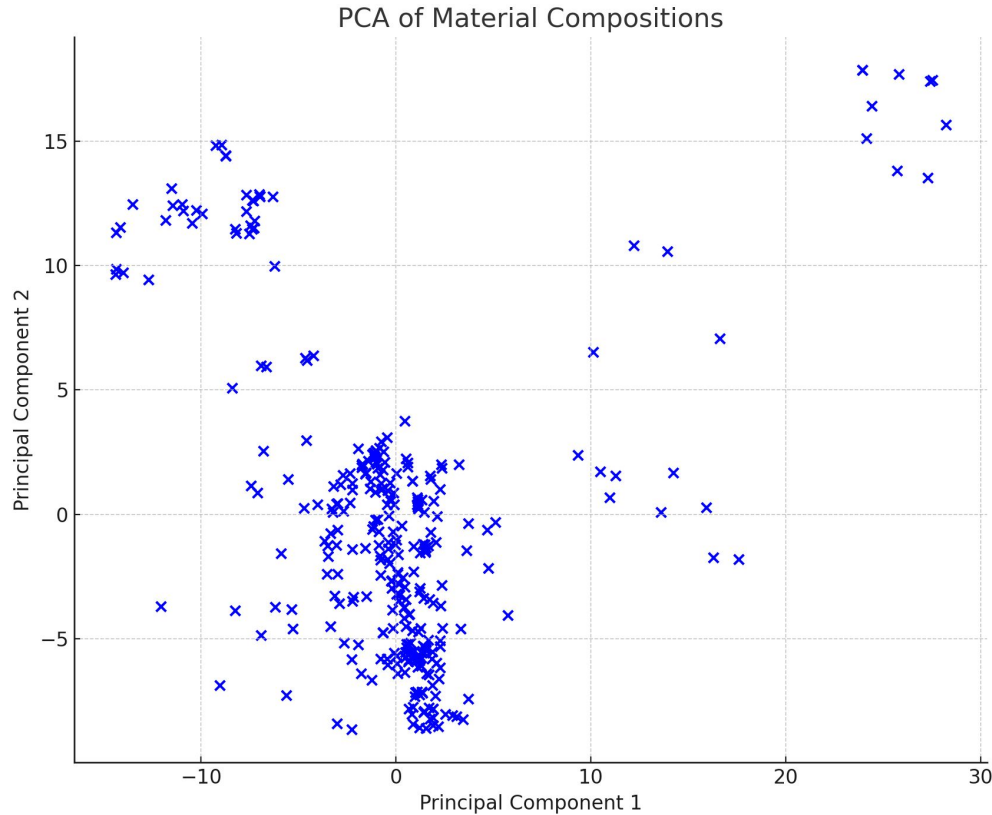
# Techniques used



# Diving deeper into the system

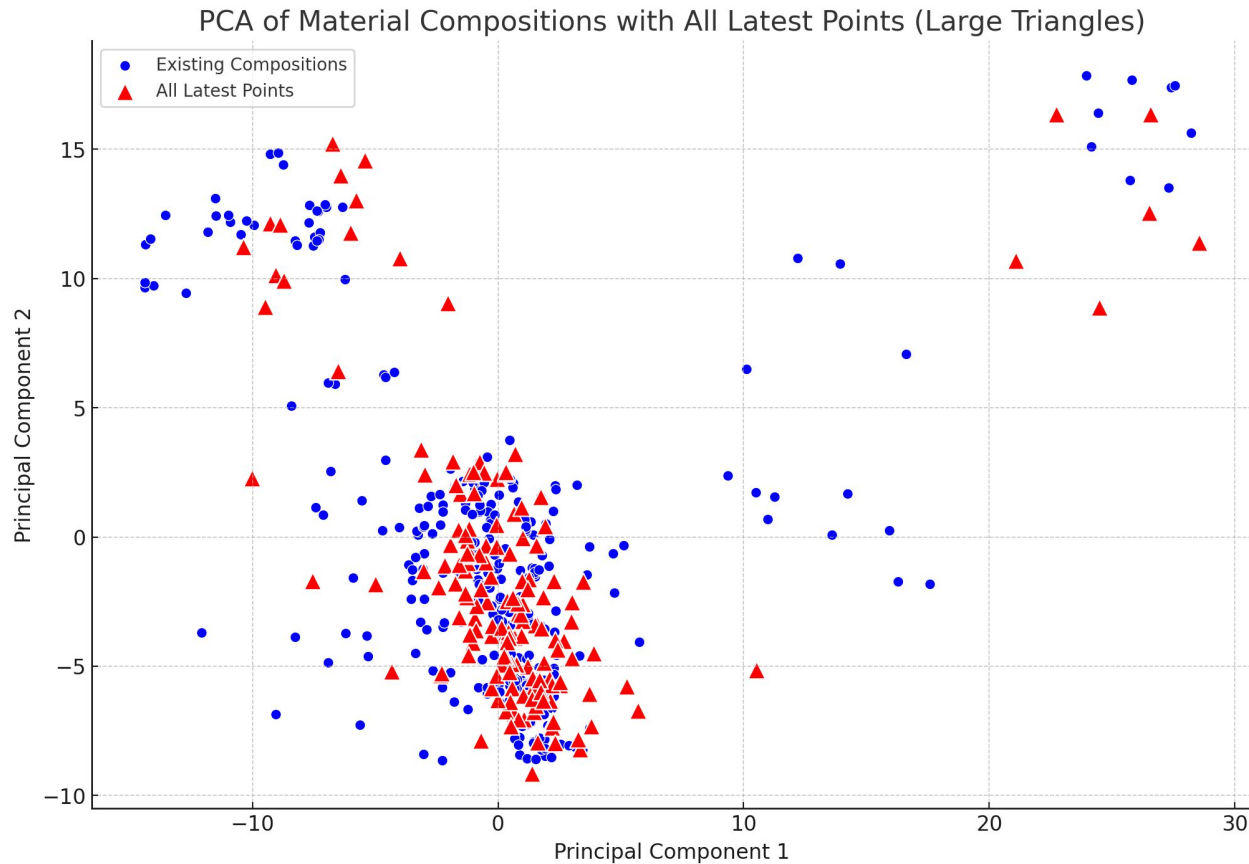


# Initial data analysis



This is a visual representation of the existing material compositions in the dataset.

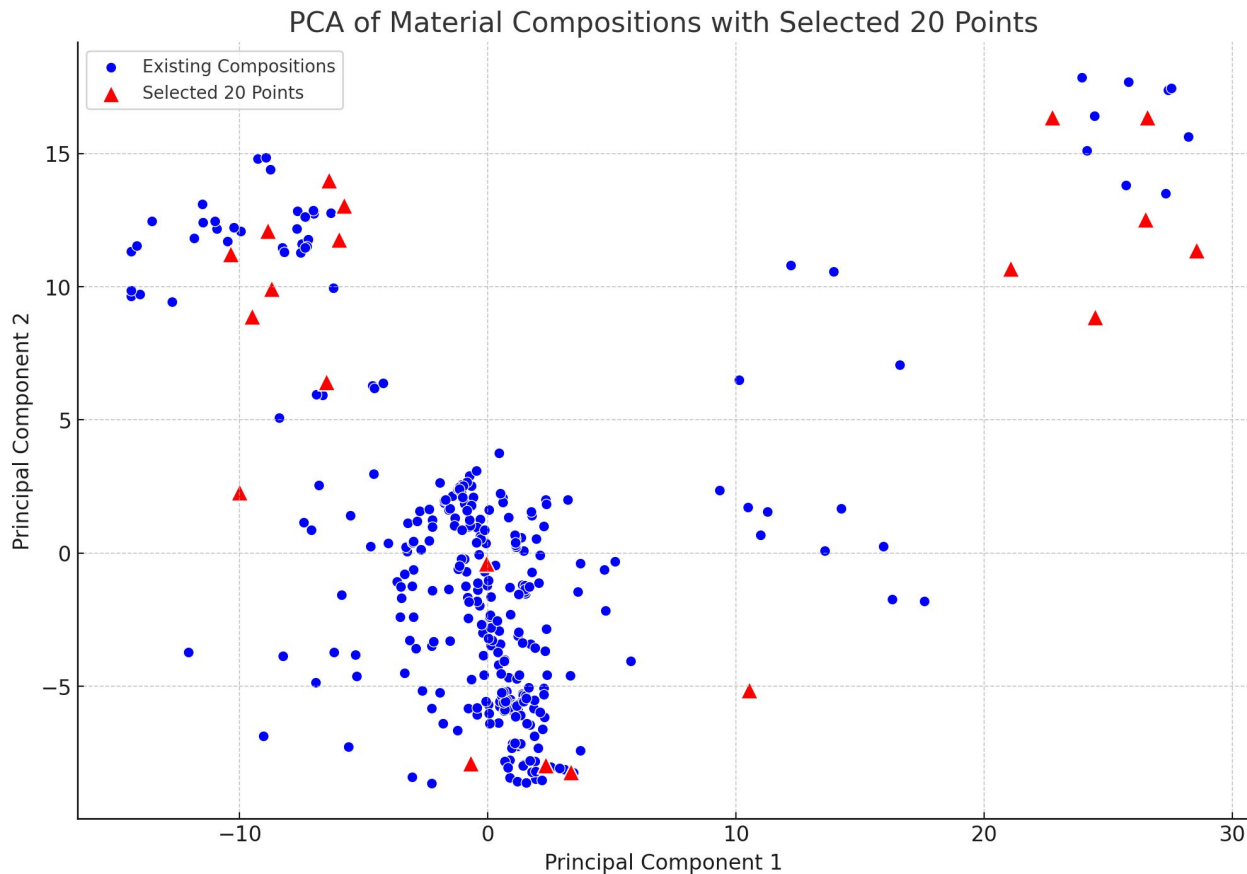
# Creating new compounds



Red points are new materials generated by our AI model

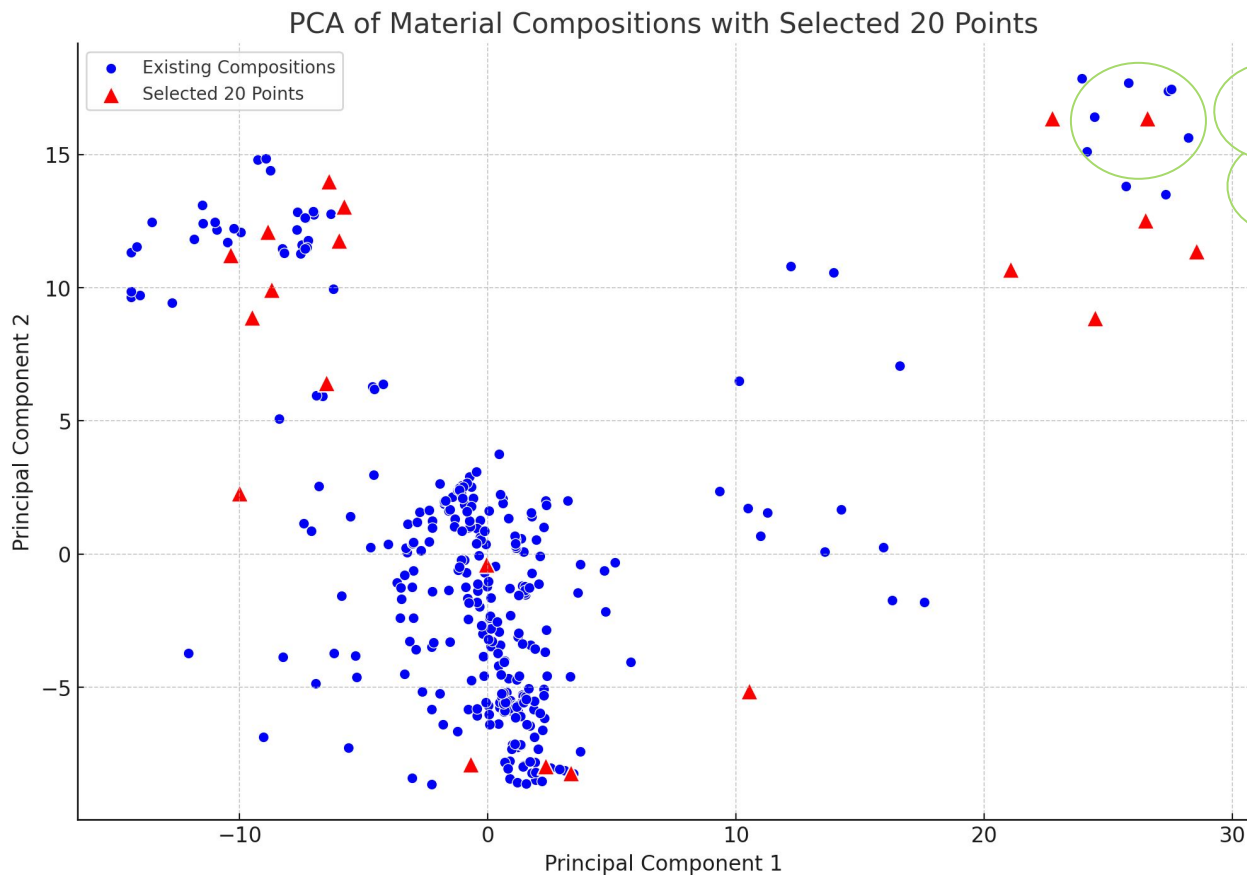


# Pruning for variability in compounds



A subset of the red points were selected to optimise for variance

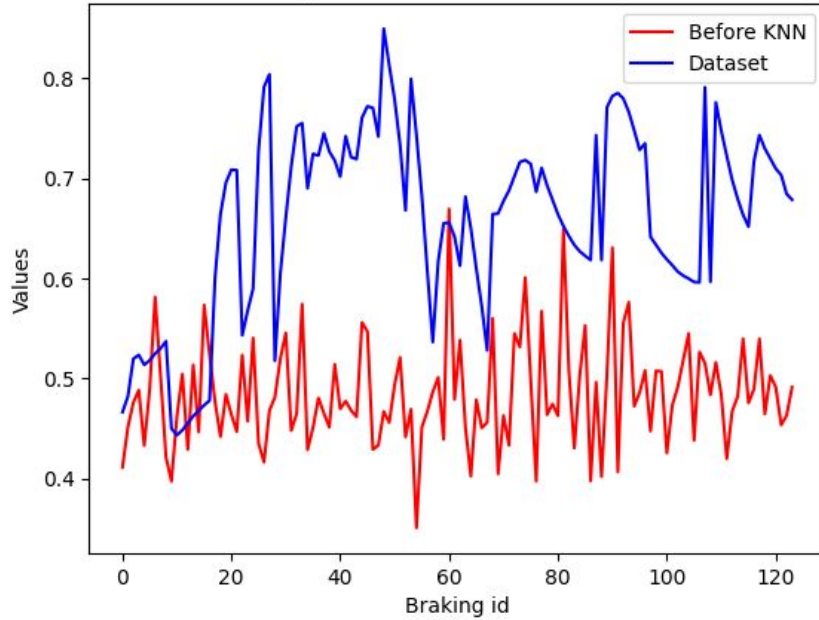
# How did we improve performance?



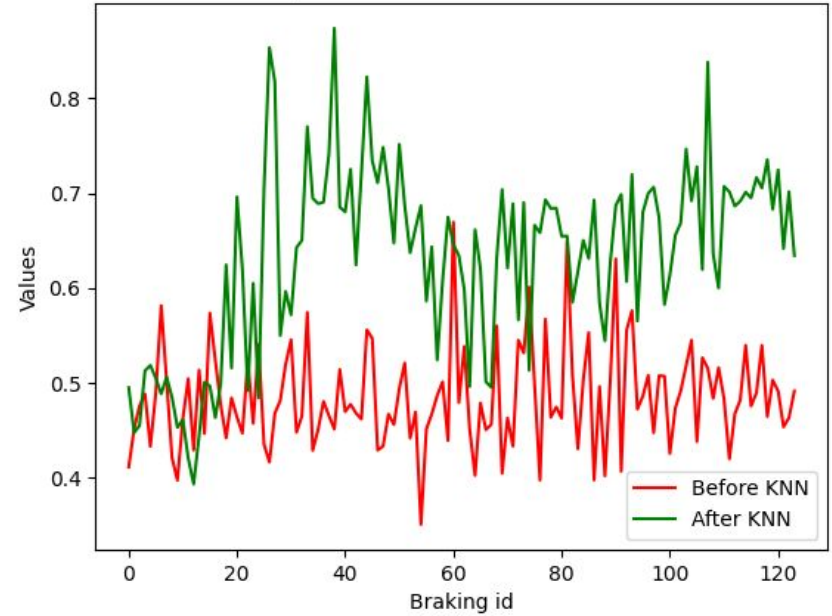
To get time series data for the red compound, we prompt our model the data of the blue ones

# How did we improve performance?

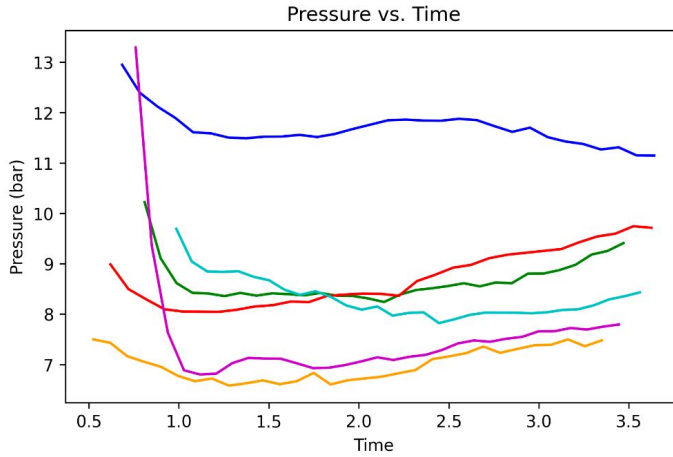
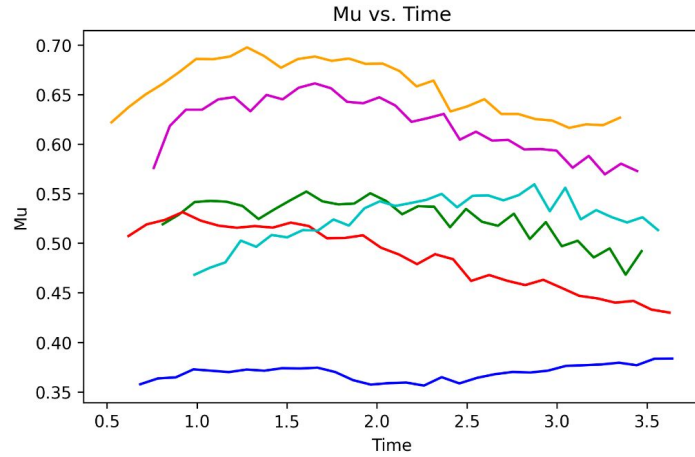
Line Graph of Mu before KNN and dataset



Line Graph of Mu before KNN and after KNN



# Results: Guess what's real and what's generated.

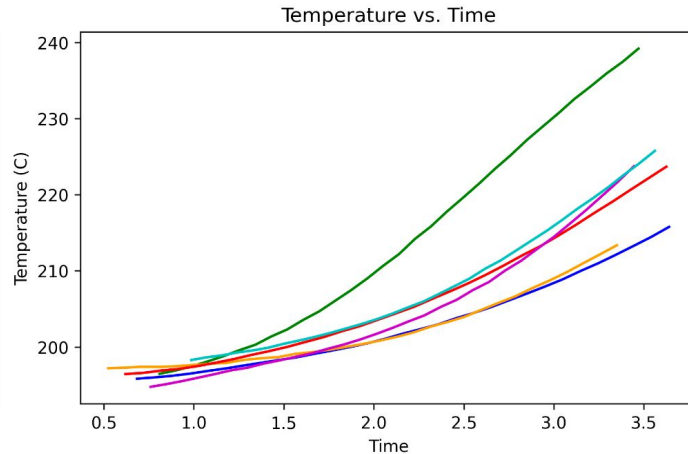
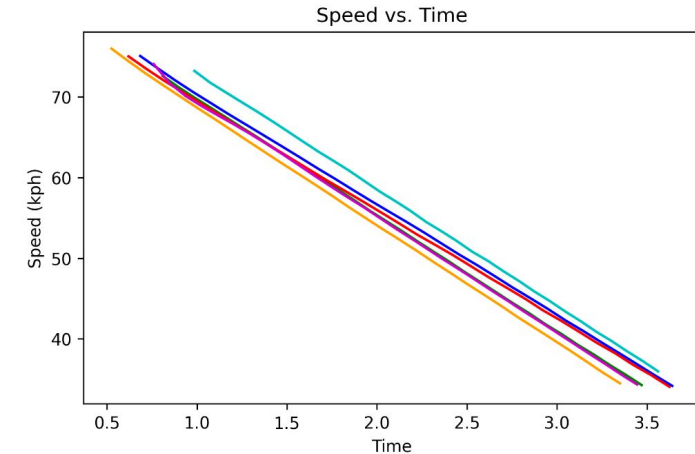


5 of these graphs are from existing data.

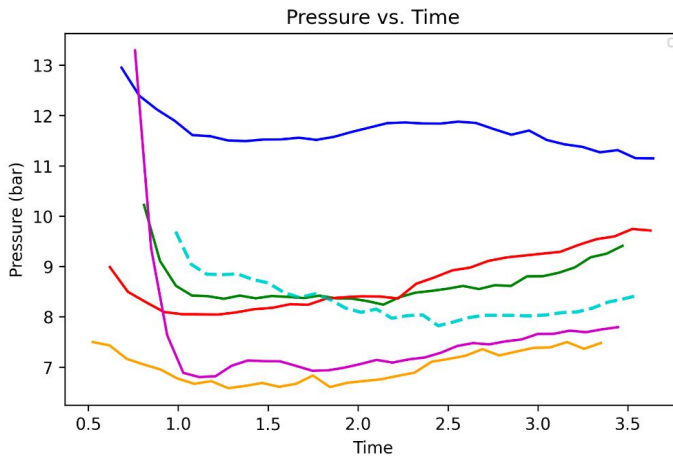
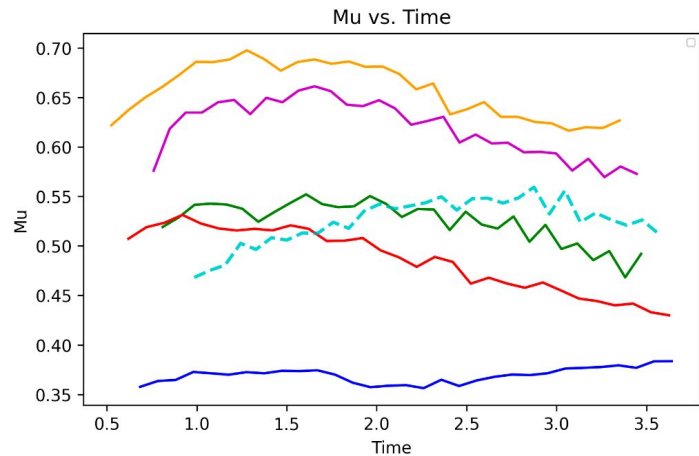
One has been generated by our model.

Can you guess which one?

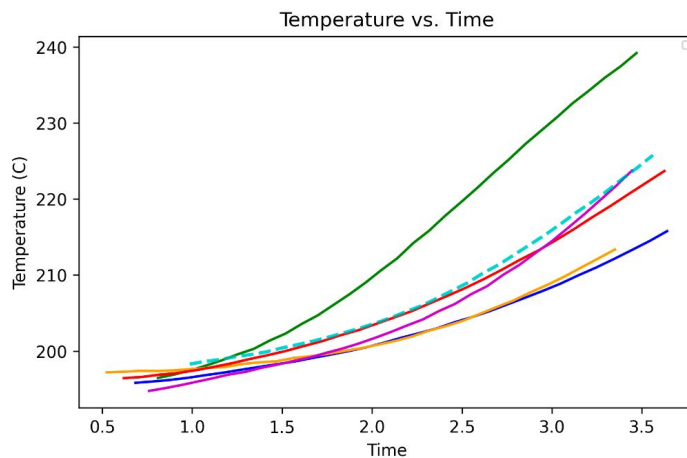
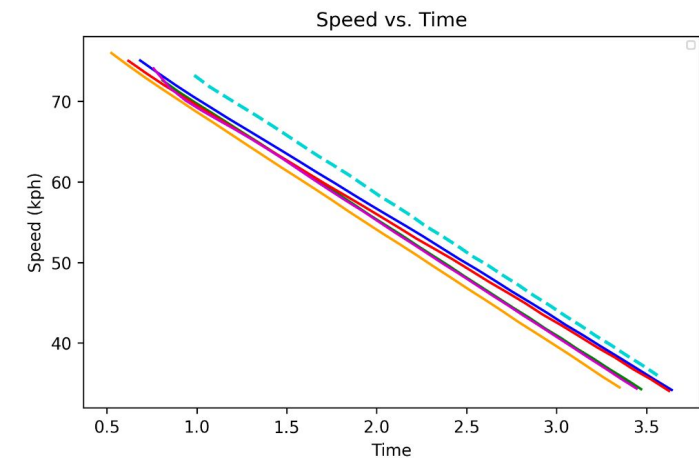
Look closely!



# Were you able to guess correctly?



The light blue one (dotted) is ours!



# Future Work

Fine tune our models using more data (Increases physical relevance)

Better selection of input materials for fine tuning (Increases physical relevance)

RLHF – Automatically incorporate feedback from validator through prompt tuning

Inject some non-determinism into material generator to get varied set of materials (Increases variance)

**Cheers!**



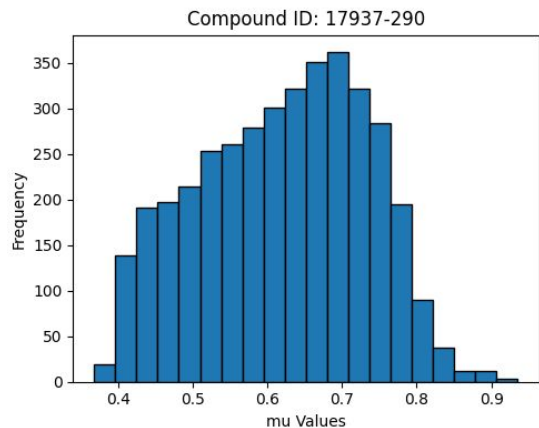
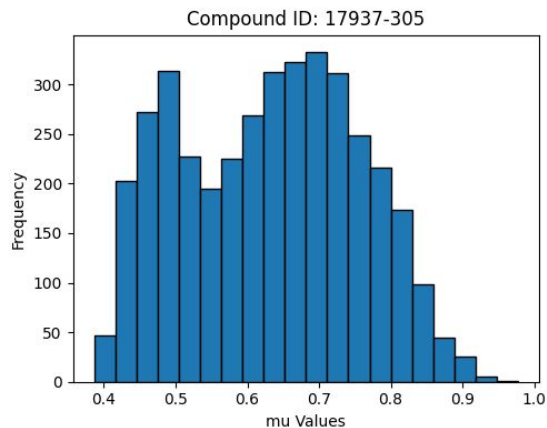
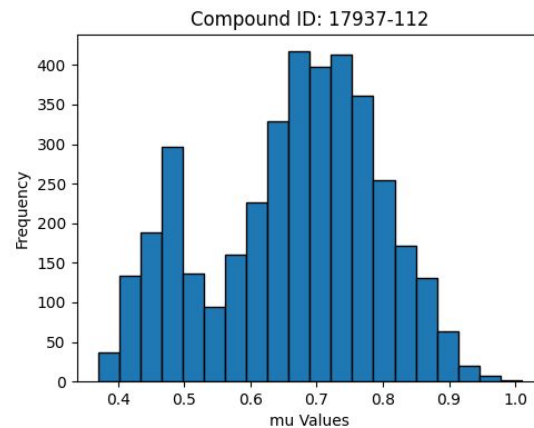
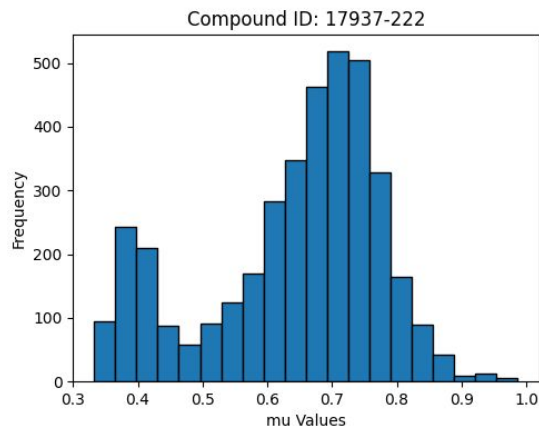
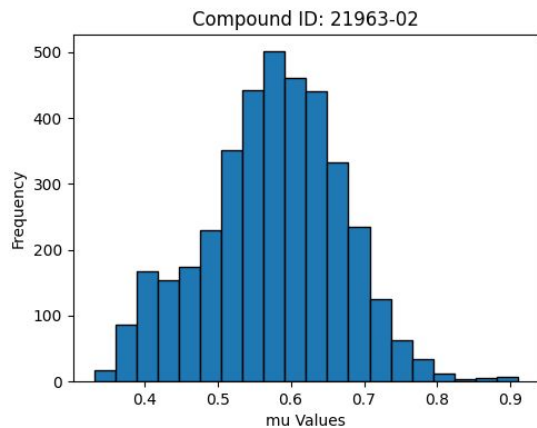
# Appendix



# Why use GenAI?

- GenAI is known to perform better on time series forecasting than traditional ML techniques ([research paper](#)).
- After analysing the given data, we also incorporated traditional ML techniques (like KNN, K means clustering) into our data generation pipelines to achieve great results.

# Initial data analysis



Plotting out the range of  $\mu$  values across different compounds