

# Using Machine Learning to Predict the Sequences of Optimization Passes

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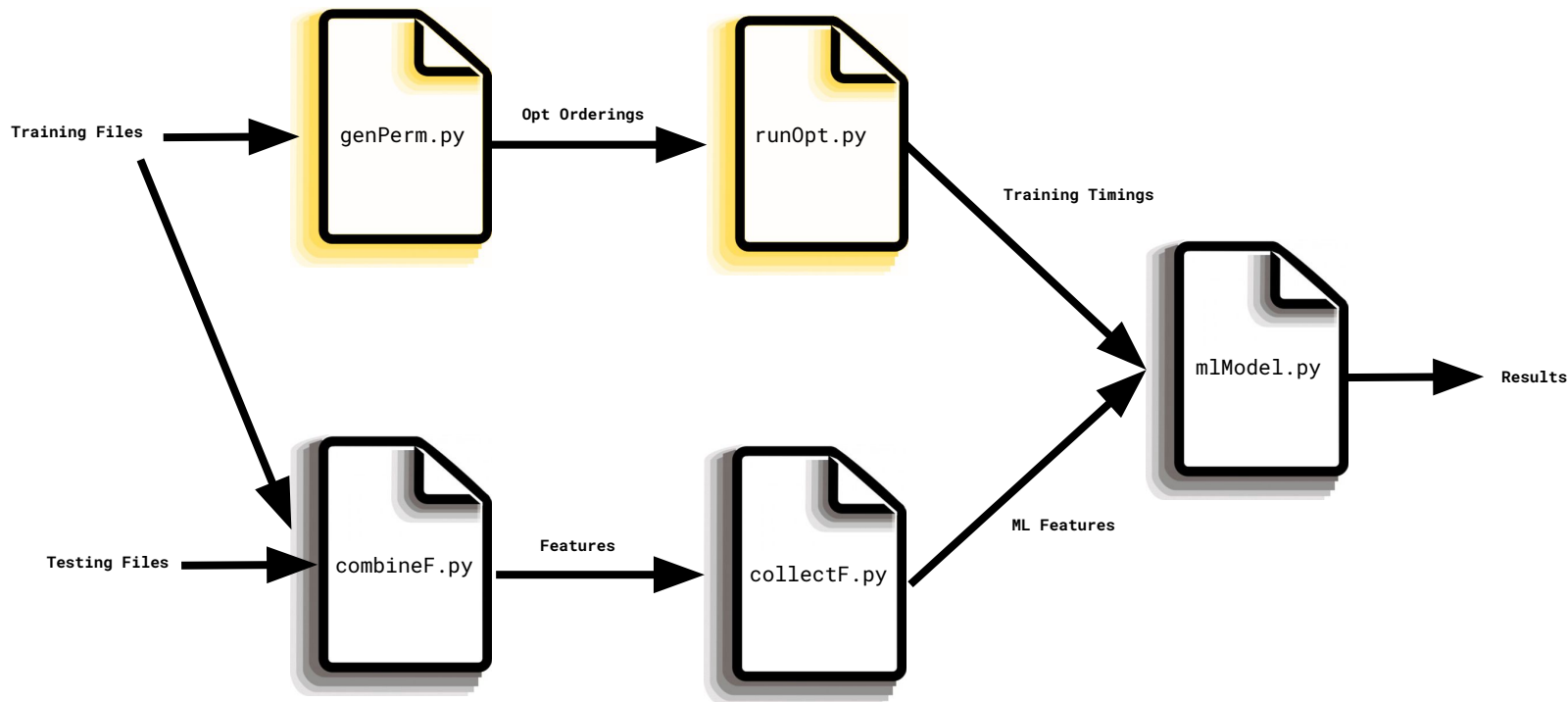
# Table of Contents

- I. Project Summary**
- II. Project Walkthrough**
  - A. Gathering training data**
    - 1. Timing Collection
    - 2. Feature Collection
  - B. Training ML Models**
    - 1. KNN
    - 2. Random Forest
    - 3. SVR
    - 4. Ensemble Methods
- III. Results**
  - A. O3 vs Our ML Models vs Original Paper
- IV. Demo**
- V. Q&A**

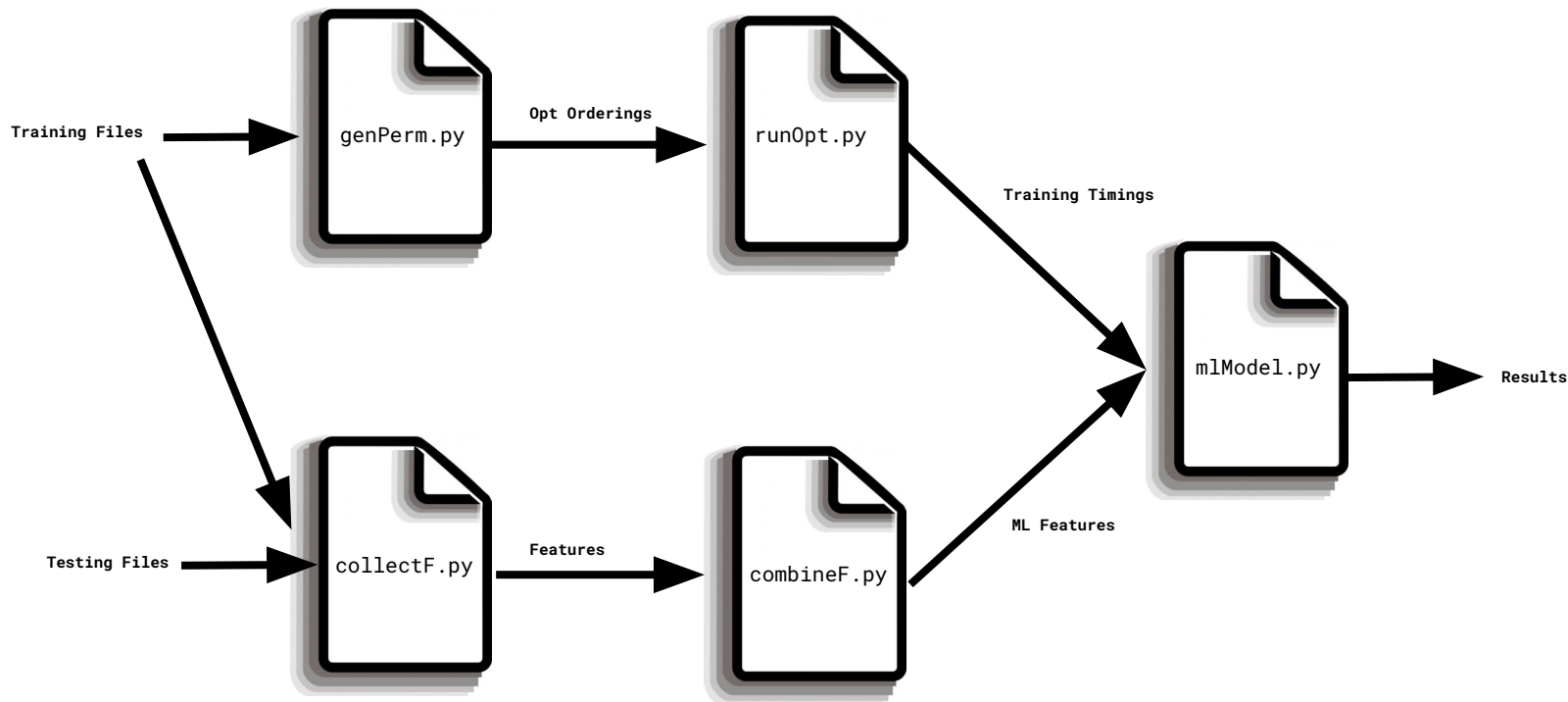
# Project Summary

- Using Machine Learning to Optimize Phase Ordering
- Gathering Training Data from Training Files
  - Find “Best” Phase Orderings
  - Collect Program Features
- Train ML Models on Gathered Data
  - KNN, SVR, Random Forest, Ada Boost, Gradient Boosting
- Used Trained Models on Test Files

# Project Structure

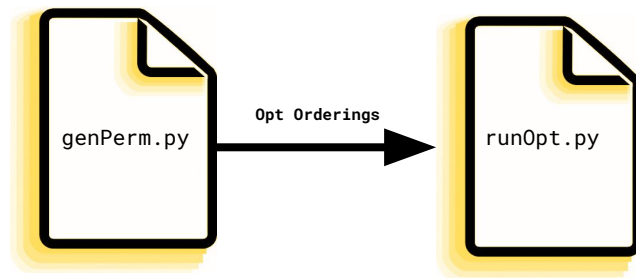


# Project Structure



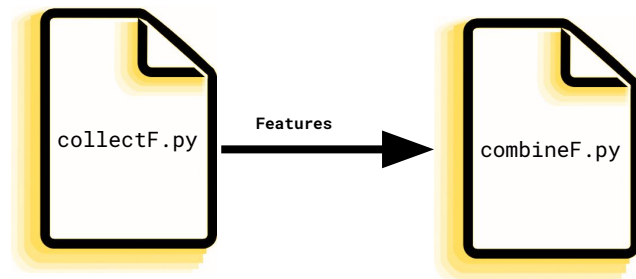
# Gathering Training Data: Timings

- generatePermutations.py
  - Works with a static set of optimizations
  - Generates permutations of optimization orderings on training files
- runOptimizations.py
  - Gathers the “best” orderings by profiling each permutation



# Gathering Training Data: Feature Collection

- `collectFeatures.py`
  - Finds and stores all features from training data
  - Finds and stores all features from test data
- `combineFeatures.py`
  - Transforms features to be ML Model Readable



# Gathering Training Data: LLVM Passes

1. Function Pass
  - a. High-level and control flow
2. Loop Pass
  - a. Depth and loop hierarchy
3. Module Pass
  - a. Static instruction counts
  - b. Compare our results to the features used by our research paper

***65 total features***



# Function Pass

- Total Basic Block Count
- Average Instruction Count Per BB
- Dynamic Instruction Count Categories
- Static Instruction Count Categories
- Dynamic-Static Instruction Count Ratios
- Biased/Unbiased Branch Counts
- Loop Count
- Basic Blocks Per Loop
- Average Static Instruction Count Categories Per Loop
- Recursive Call Count

***30 total features***

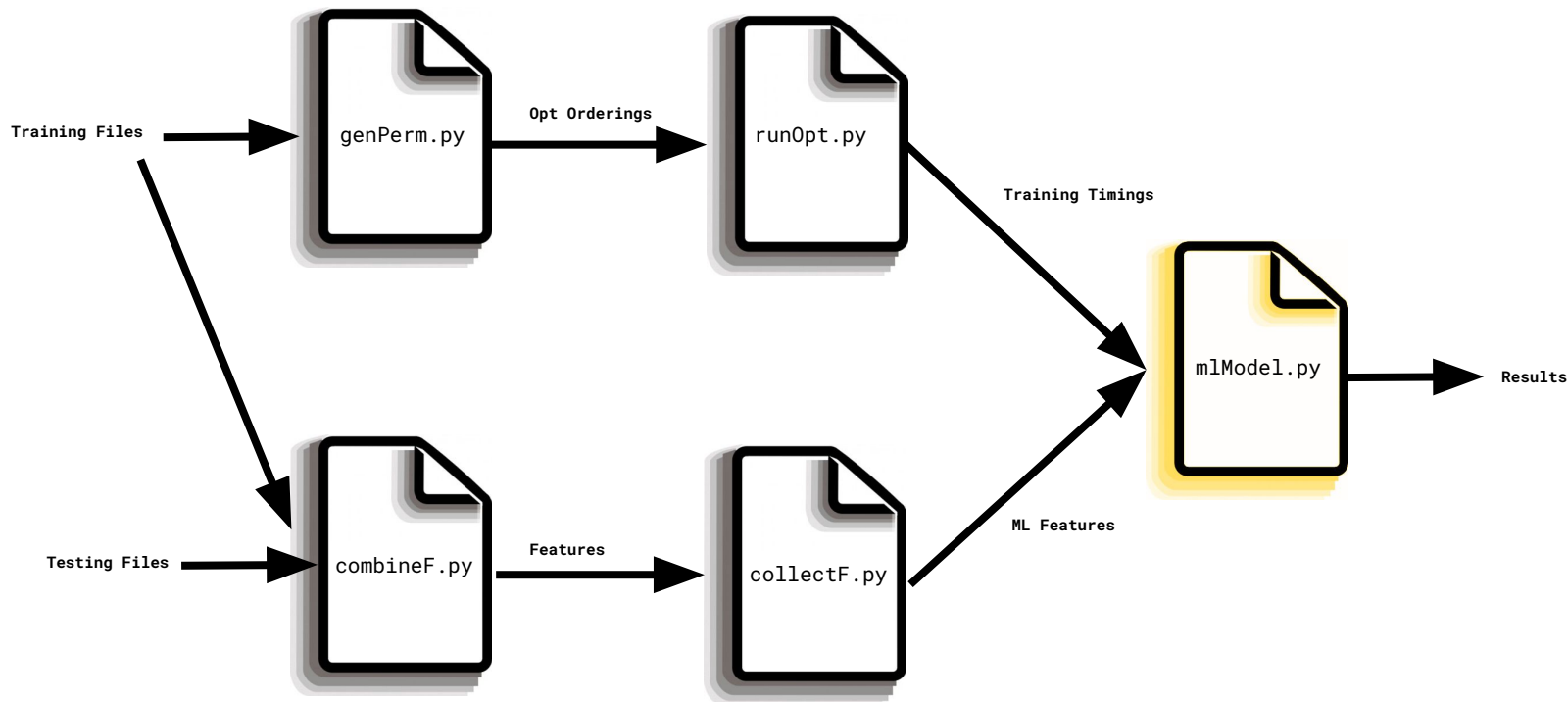
# Loop Pass

- Number Of Loops With Nesting
- Number Of Outermost Loops
- Average Loop Nesting Depth
- Average Outermost Loop Depth
- Deepest Loop Nesting Depth

***5 total features***

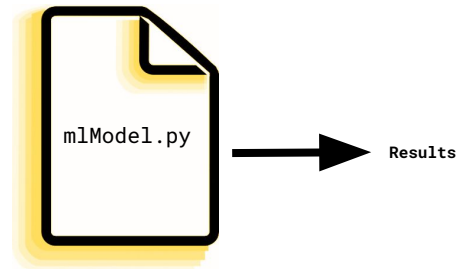
***35 total features combined with function pass***

# Project Structure



# Machine Learning Models

- mlModel.py
  - Contains:
    - K Nearest Neighbors
    - Support Vector Regression (SVR)
      - Kernels: Linear, Poly, RBF
    - Random Forest
  - Ensemble Methods (combine multiple models)
    - Gradient Boost
    - AdaBoost

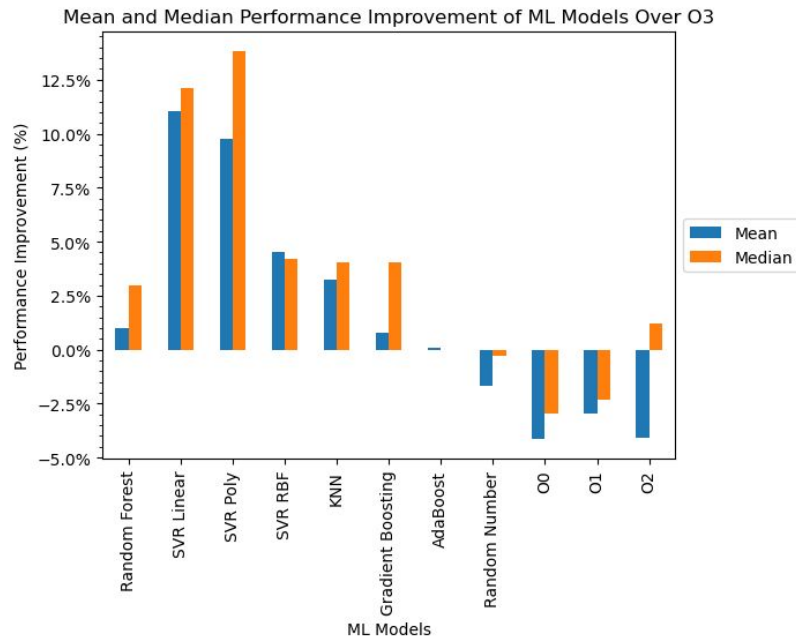


# Demo

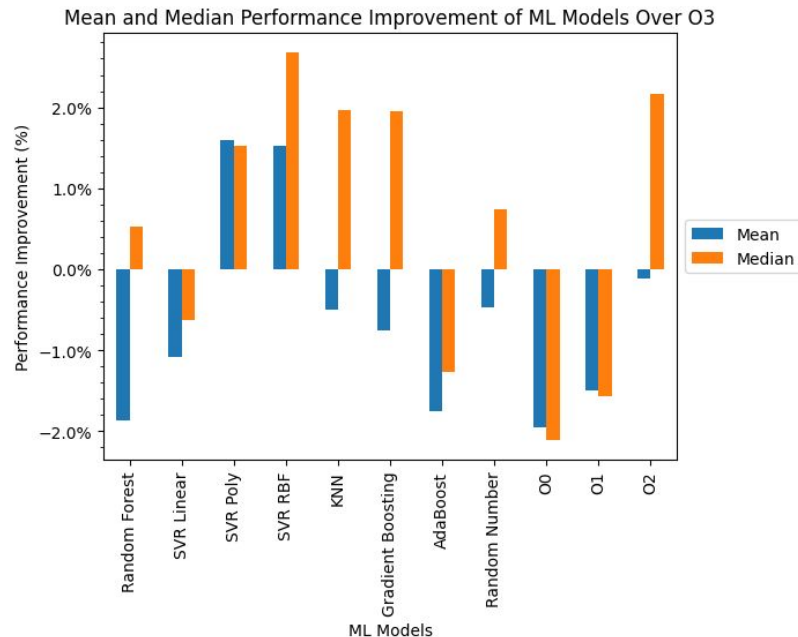
<https://github.com/EECS-583-Group-24/ML-LOOP/tree/main/demo>

# Results

- Our ML Results with our Features

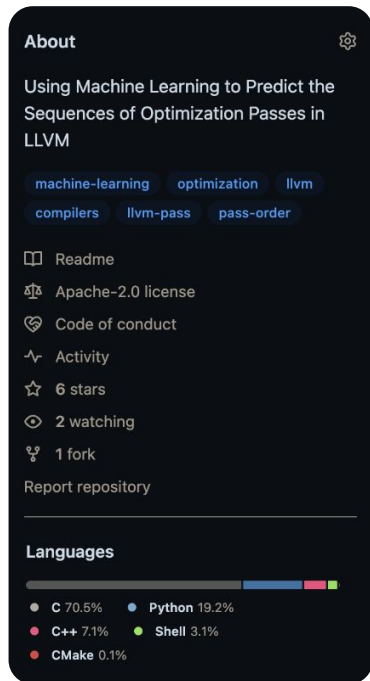


- Original Paper Features



# Future Works

- Github: [Link](#)
  - More Testing / Training data
    - More Diverse Data Set
    - More Complex Data Set
  - Better ML Models
    - Better Features
    - Better Reinforcement Learning
    - Better Training Heuristic
  - Get Open Source Contributors
    - More Forks / Stars



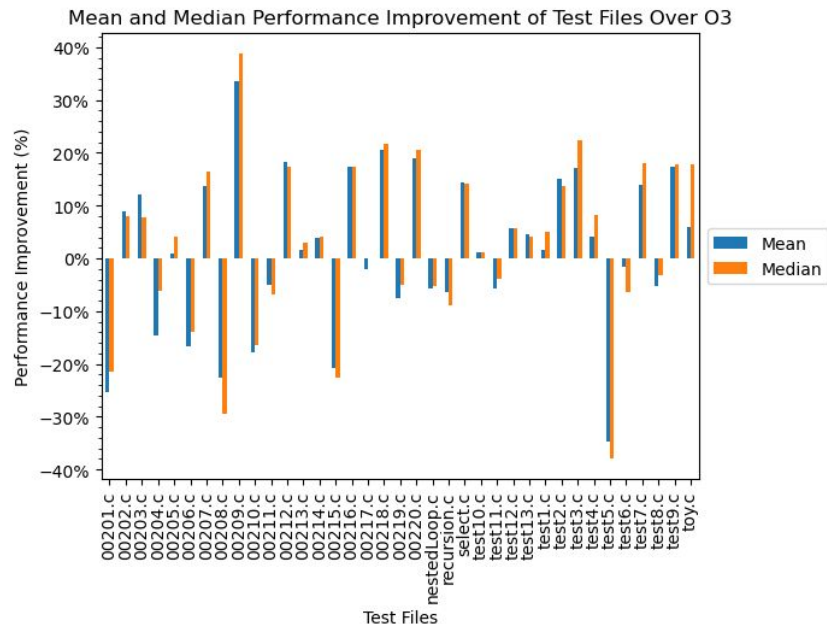
# Thank You!

Any Questions?





# Results Appendix



Many more graphs/tables: <https://github.com/EECS-583-Group-24/ML-LOOP/tree/main/figures>

Graphs for original paper: <https://github.com/EECS-583-Group-24/ML-LOOP/tree/paper/figures/originalPaper>