

Motivation for approach

- Current data repositories collect:
- ❖ Bacterial target or genetic data
 - ❖ Antibiotic compound libraries
 - ❖ Surveillance dataset



No effort to systematically collect preclinical data and leverage the data for guided drug discovery

Preclinical databases

- ❖ CO-ADD
- ❖ PubChEM
- ❖ ChEMBL
- ❖ BindingDB
- ❖ DrugCentral and more

Existing ML pipelines focus on

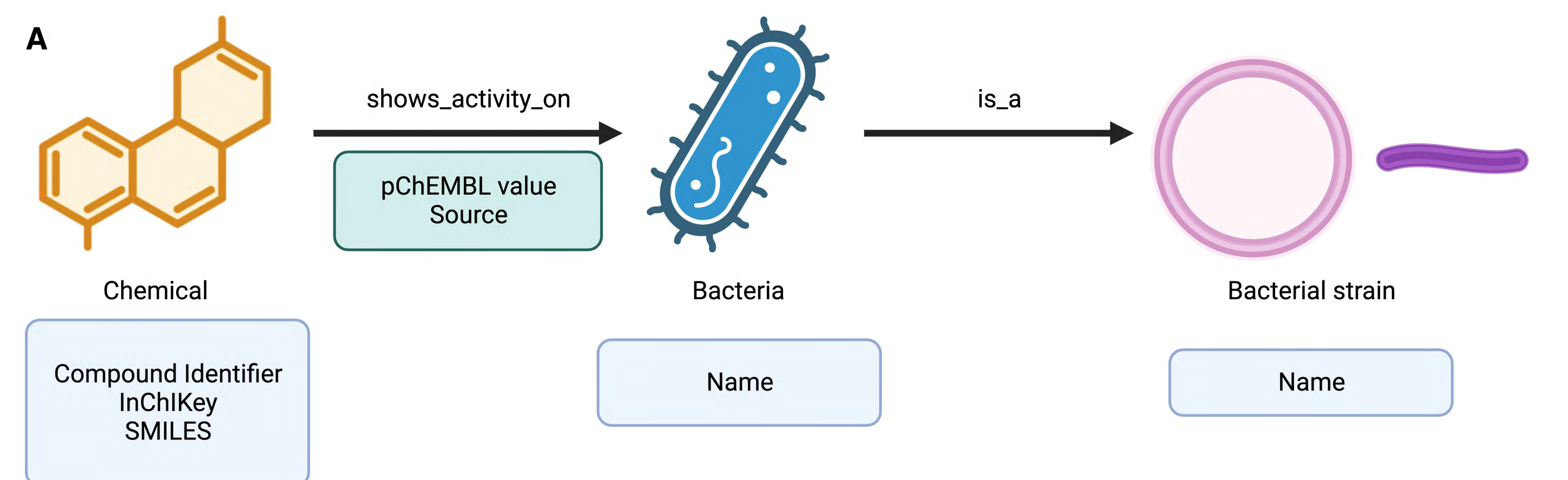
- ❖ Virtual library generation
- ❖ Resistance mechanisms
- ❖ Resistant strain prediction
- ❖ Drug optimization among others

AMR Knowledge Graph

Collected, harmonized and generated a **graph database** for **publicly available bioactivity data**

AMR-KG consist of:

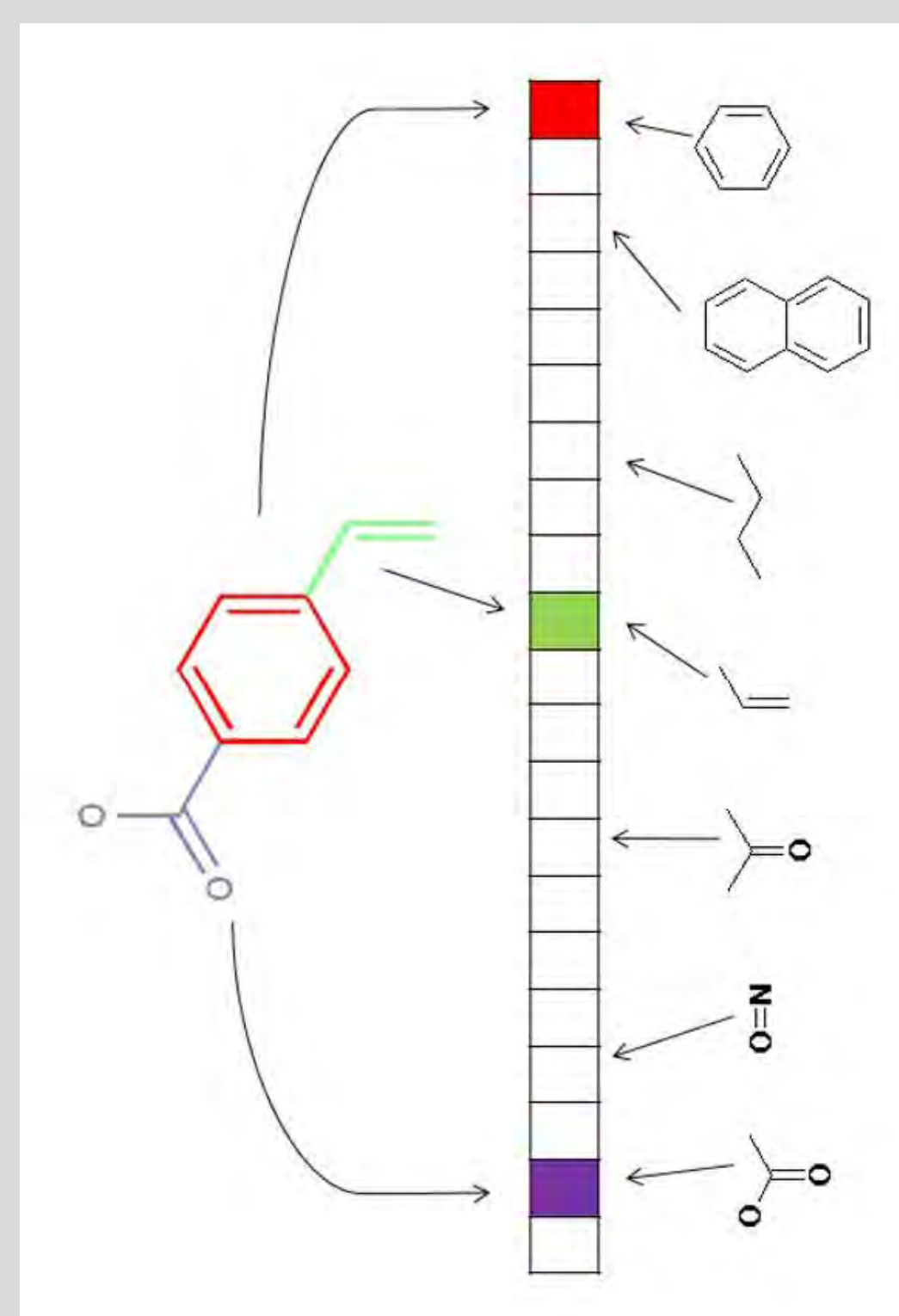
- ❖ ~54,000 compounds
- ❖ 382 bacterial strains across 3 classes (Gram+/- and acid-fast)



AMR-KG Schema

Machine learning model evaluation and predictions

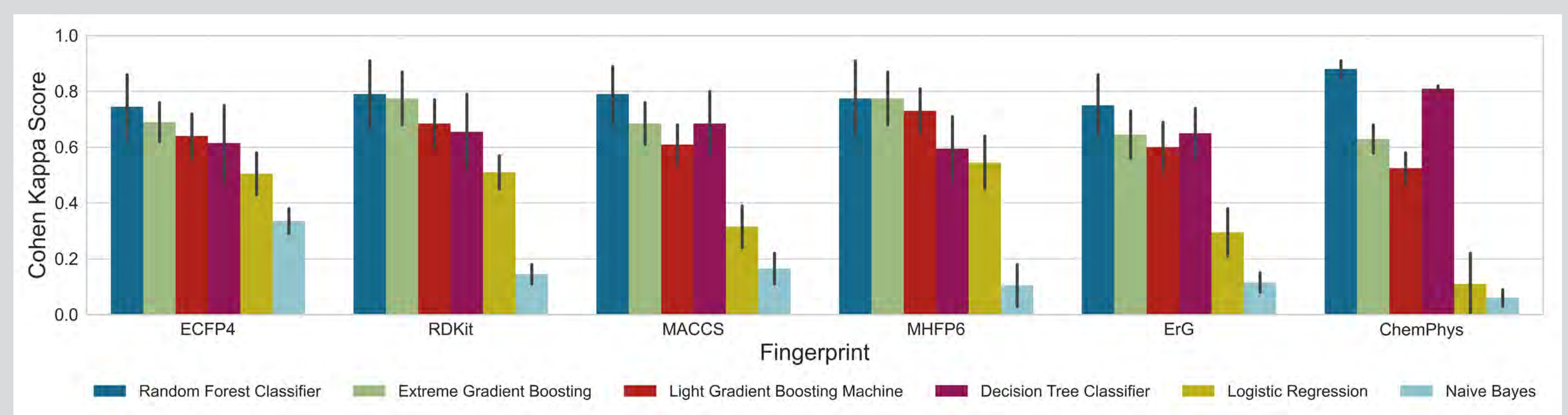
Model trained with five fingerprints: RDKit, MACCS, ECFP4, ErG, MHFP6 and ChemPhys



Compound fingerprint vector

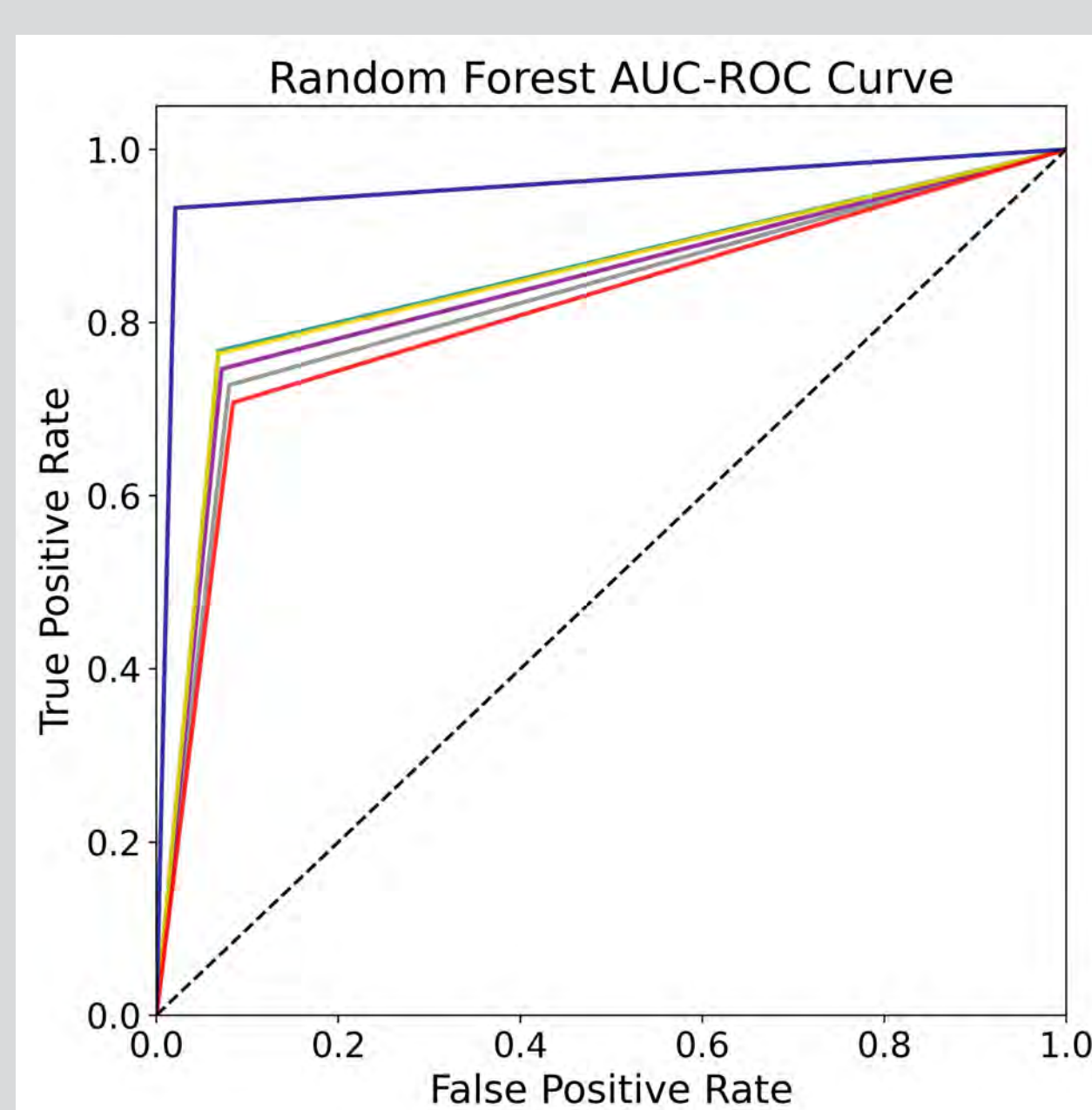
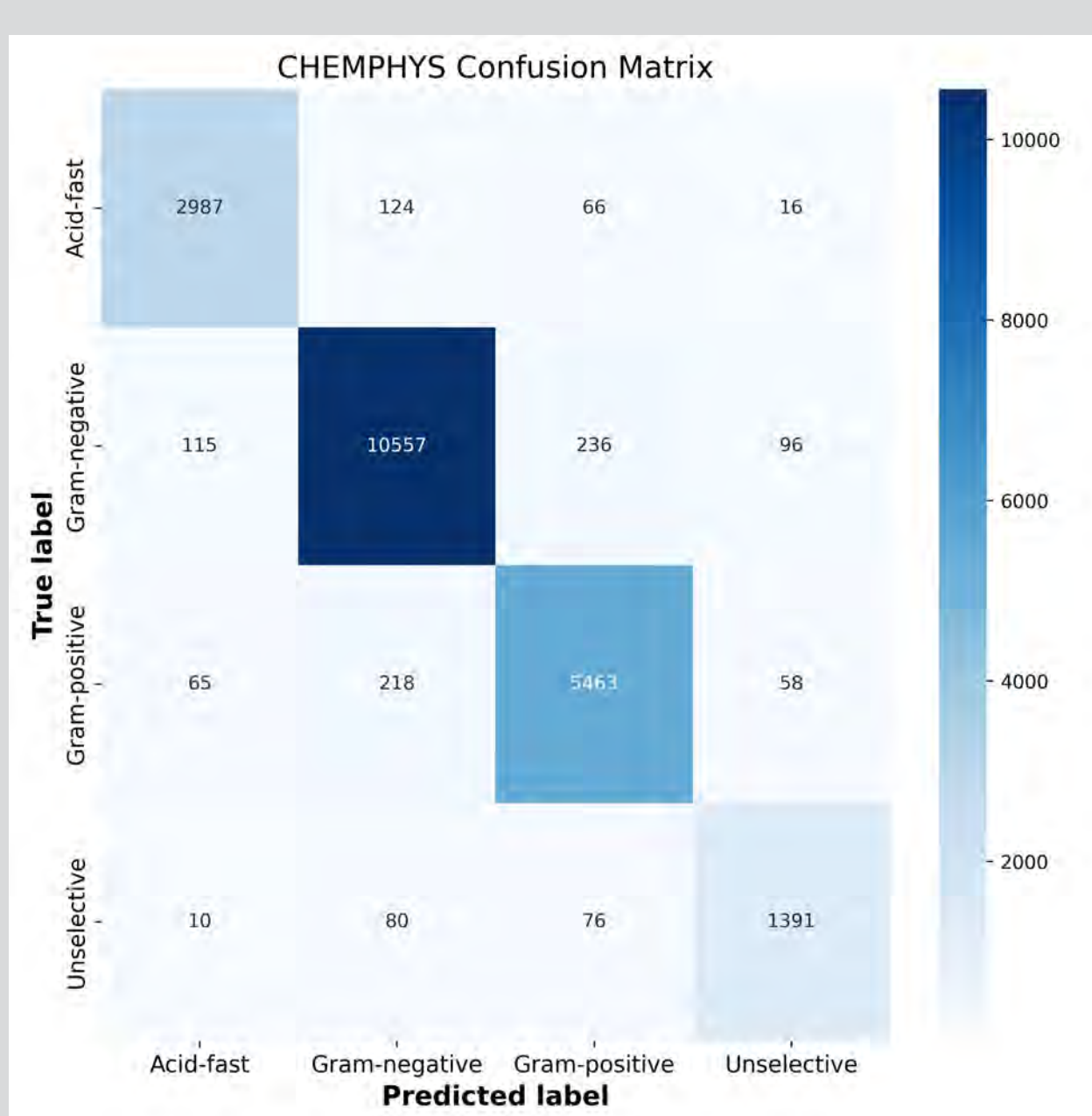
Find the best model

Choosing best model from a cohort based on Kappa score



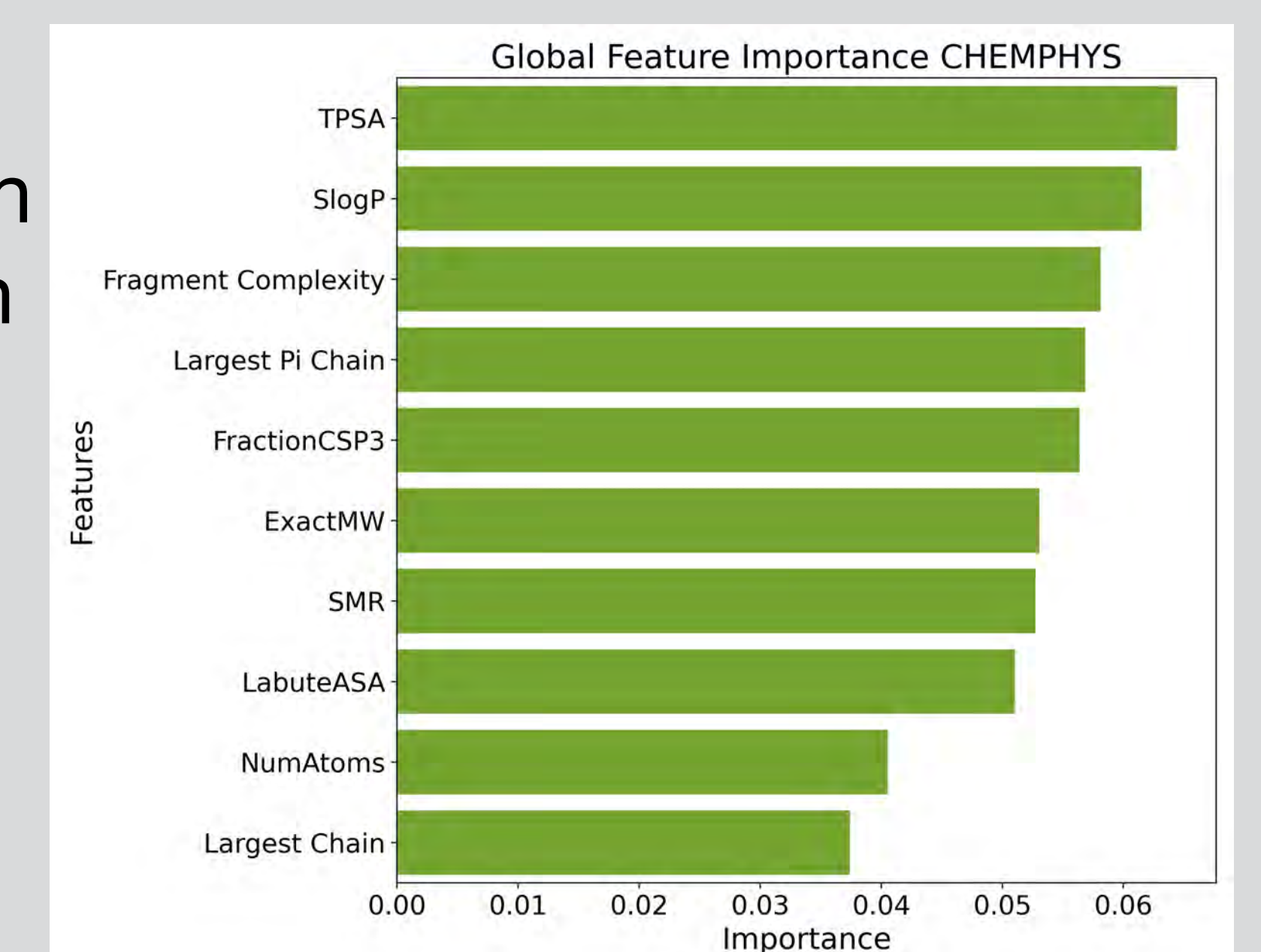
Fingerprint	Accuracy	Kappa	Macro F1
ErG	0.80	0.67	0.72
ECFP4	0.82	0.70	0.76
RDKit	0.82	0.72	0.76
MACCS	0.82	0.72	0.77
MHFP6	0.84	0.74	0.79
ChemPhys	0.95	0.92	0.94

Optimizing the best model



Finalizing the best model with hyperparameter optimization using Optuna

Metrics shown here are on the test dataset (21,000 compounds)



Applicability of model

Use the model to select antibacterial compounds in:

- Prioritizing compounds in screening
- Filtering virtual screening for potential active compounds
- Drug repositioning
- In complement with antibacterial active vs inactive model

Where can you find us?

Contact: Yojana.Gadiya@itmp.fraunhofer.de

GitHub: https://github.com/IMI-COMBINE/broad_spectrum_prediction

Model will be deposited on BioModel with manuscript submission.