

From data acquisition to ML model for predicting outcomes of chemical reactions

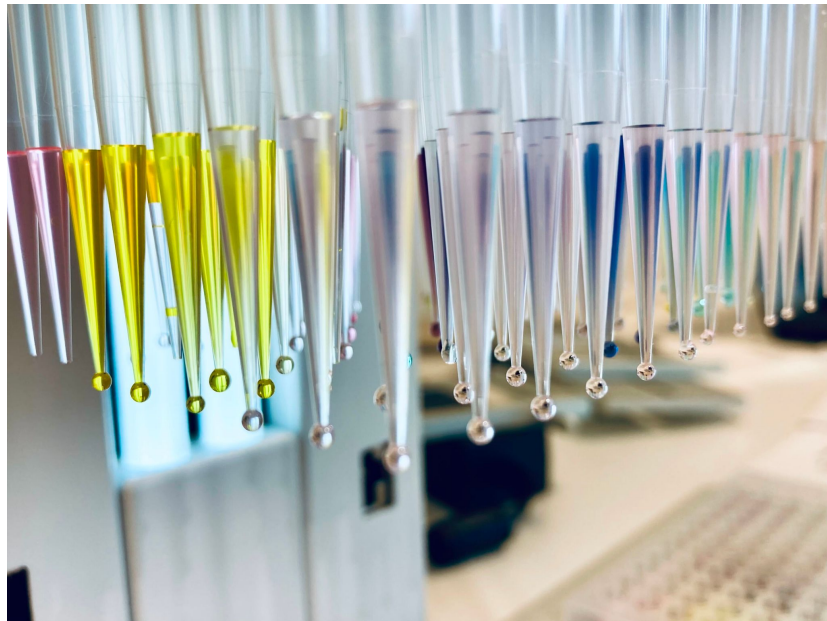


Michał Sadowski

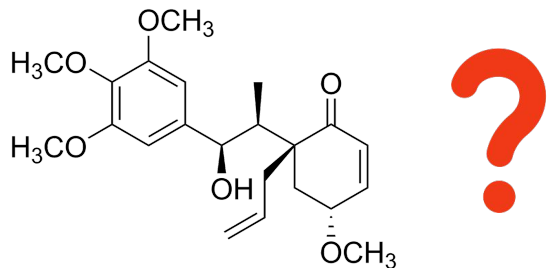
Molecule.one: solving unpredictability of chemistry

Predicts chemical reactions with unprecedented accuracy for faster drug discovery

- We created the first deep learning based commercial software for synthesis planning (2019-2022)
- Currently focused training deep learning models on data from our own laboratory (2022)



Bottlenecks in organic chemistry

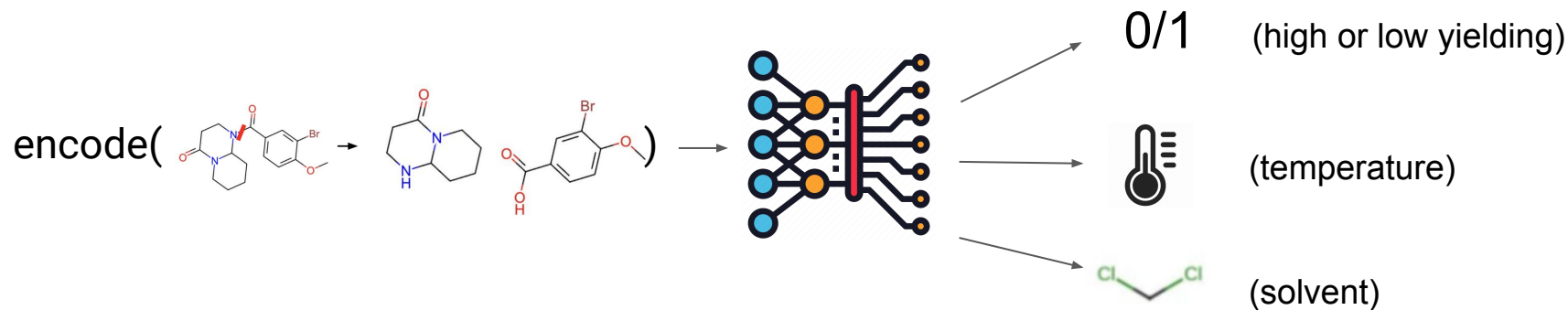


Finding cheap synthesis plan is difficult

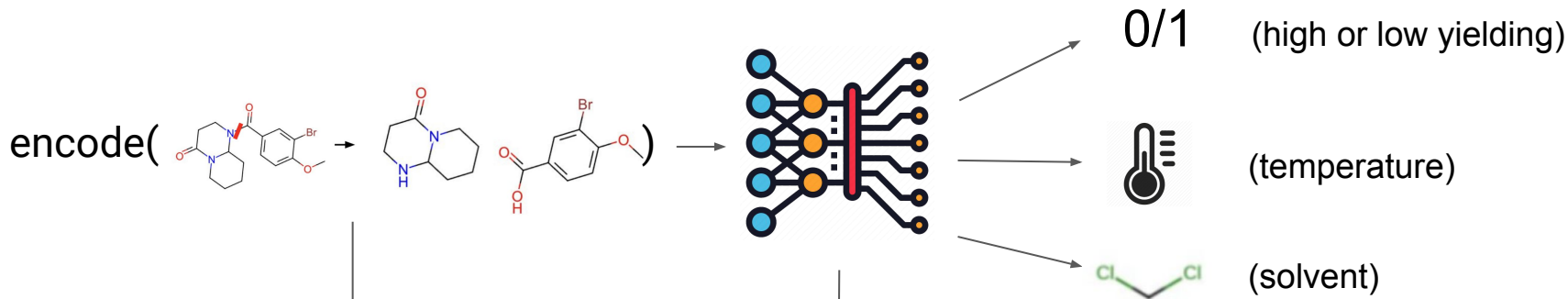


~50% of reactions fail

Deep learning for chemical reactions

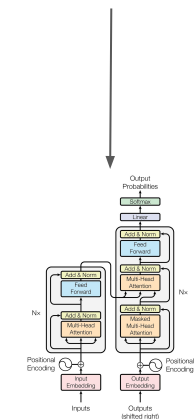


Deep learning for chemical reactions



For example:

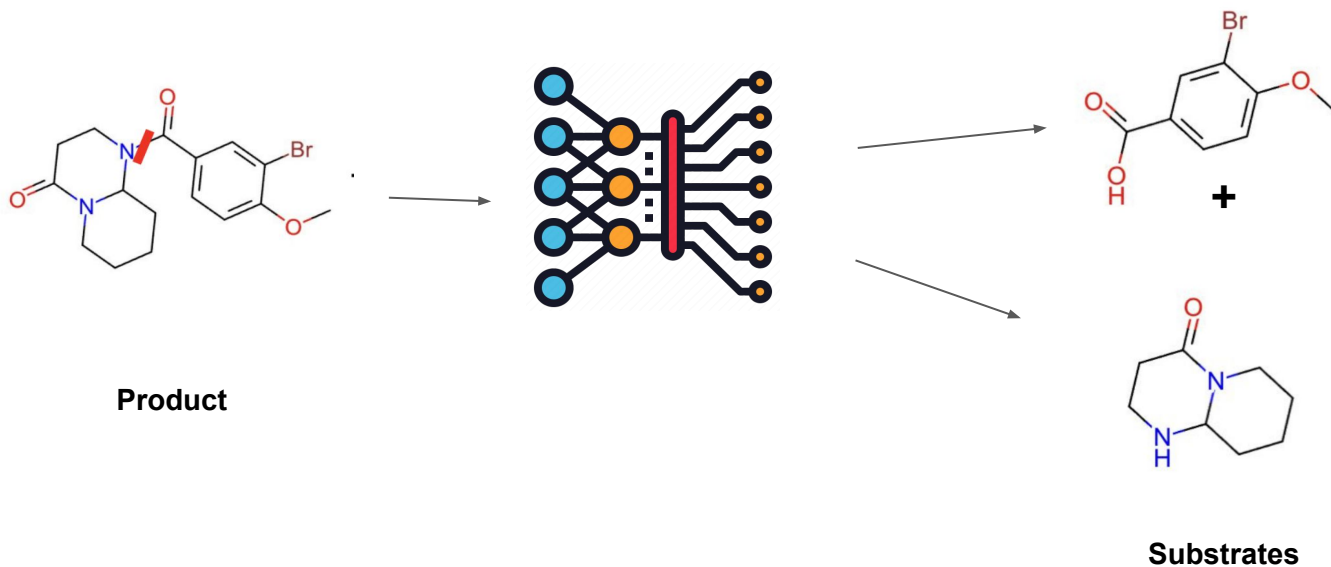
CC(=O)[OH].[OH]C>>CC(=O)OCC
SMILES notation



0 4 2

Output tokens

Finding synthesis plans with deep learning



Dataset sizes in chemistry are tiny

**Estimated number
drug-like molecules**
(10^{20} - 10^{63})

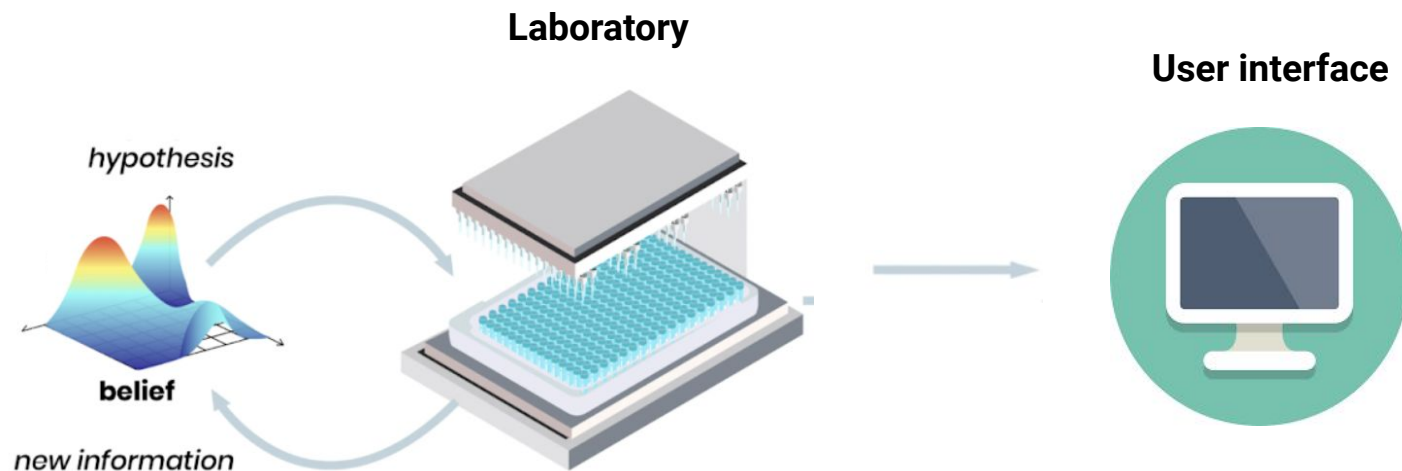
Size of publicly available
reactions datasets (10^6)



Number of negative reactions
in datasets

~ 0

Solution: Closing the loop in organic chemistry

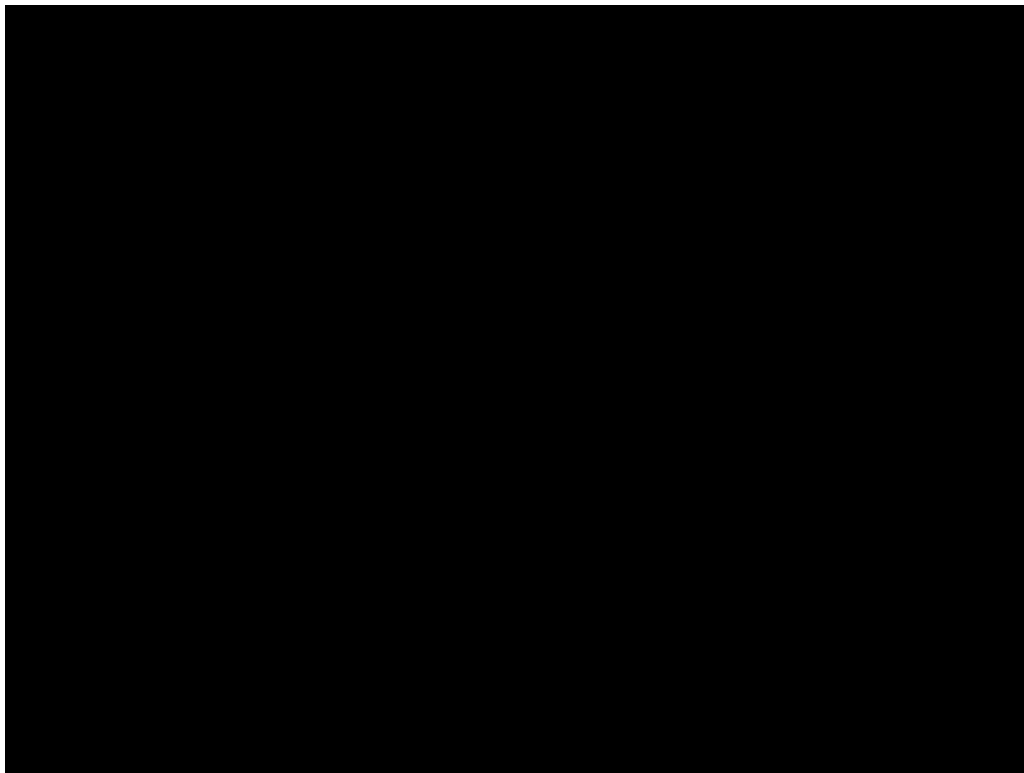


HT laboratory @ Molecule.one



- We set up our own laboratory
- Designed to perform diverse reactions based on automatic recommendations

HT laboratory @ Molecule.one



- It would not be possible without automation
- We are capable of performing a few thousand reactions per week
- It is more than a typical chemist performs in a year

Link to the [video](#)

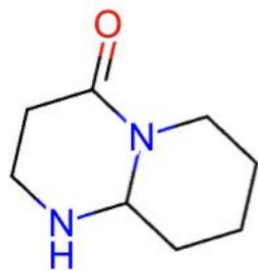
What reactions should we focus on?

Still it is impossible to generate enough data for model to predict every reaction

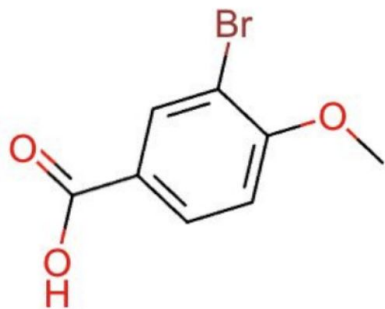
We've chosen type of reactions that are:

- Within the most commonly used reactions in MedChem
- Can be performed in a high-throughput laboratory
- Difficult to predict outcomes of reaction

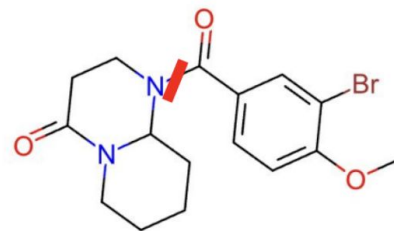
First experiments - Amide couplings



amine

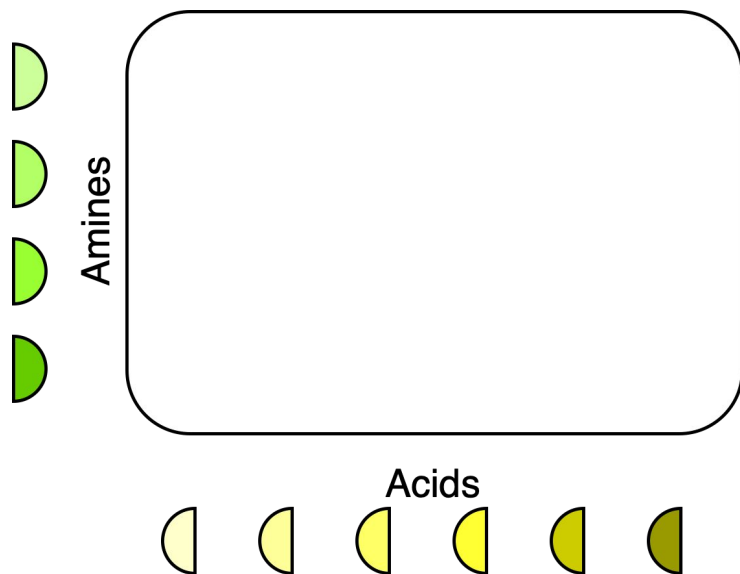


carboxylic acid

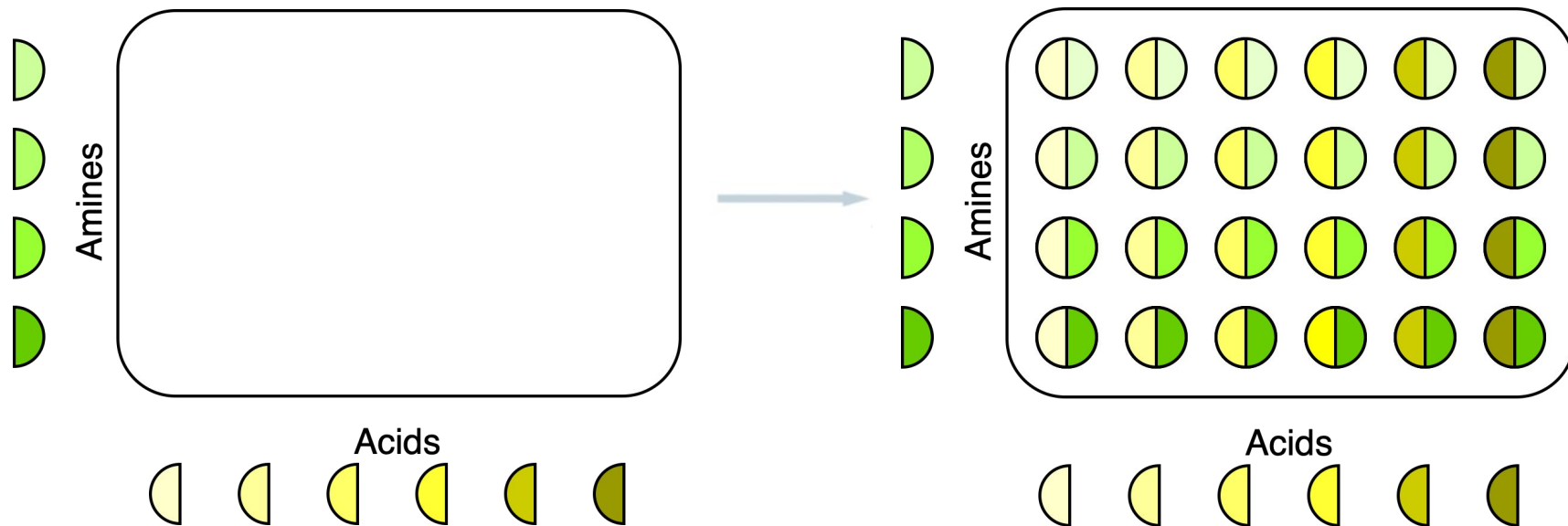


amide

Grid dataset design

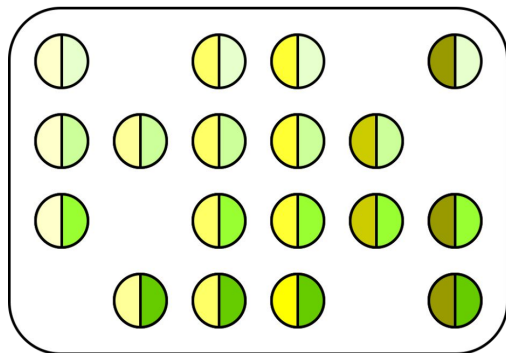


Grid dataset design

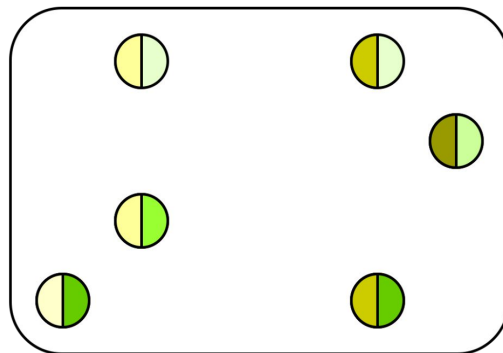


Model evaluation

Train set

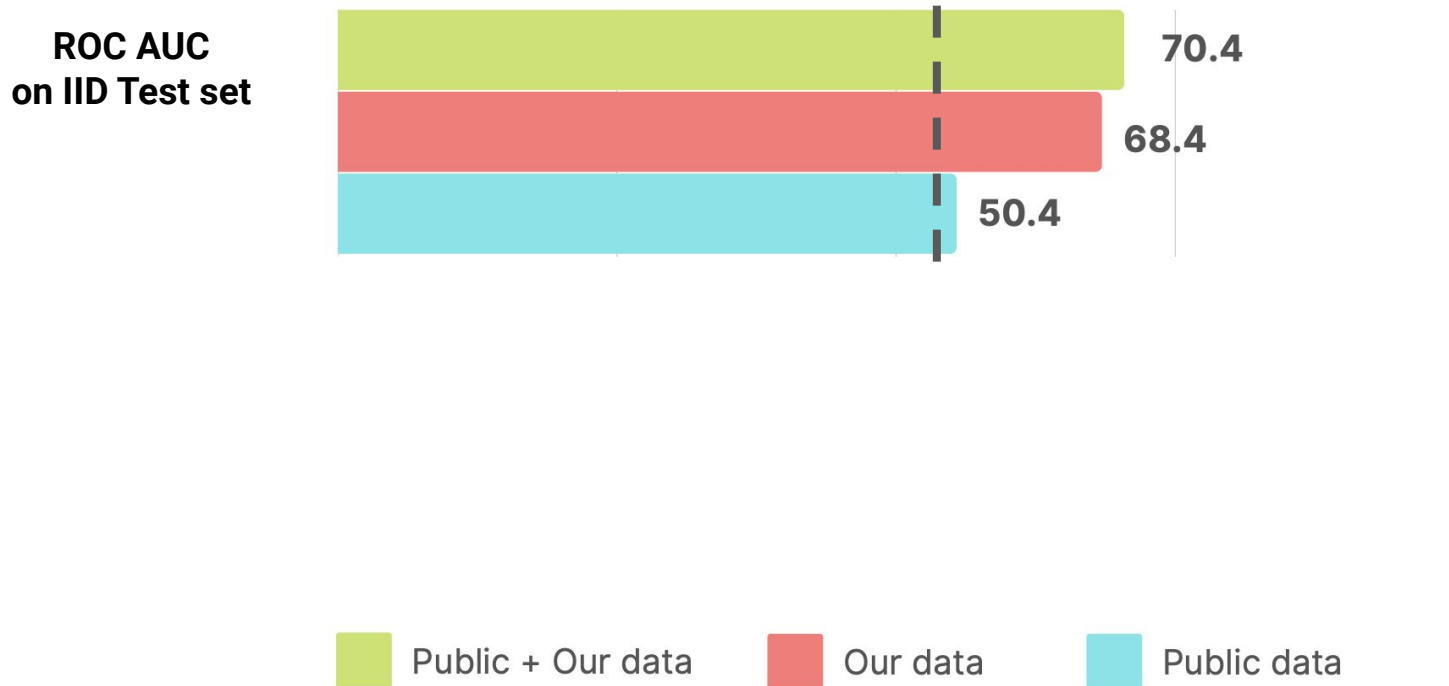


IID Test set



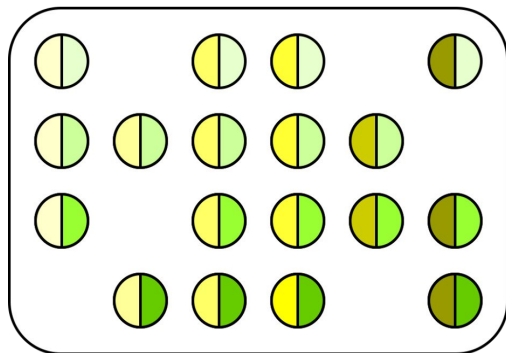
Evaluation on IID Test set

Using lab data enables achieving significantly better results than using public data

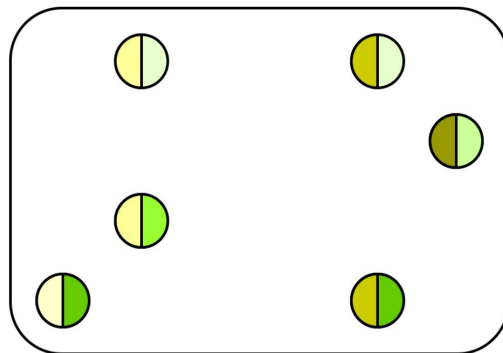


Model evaluation

Train set

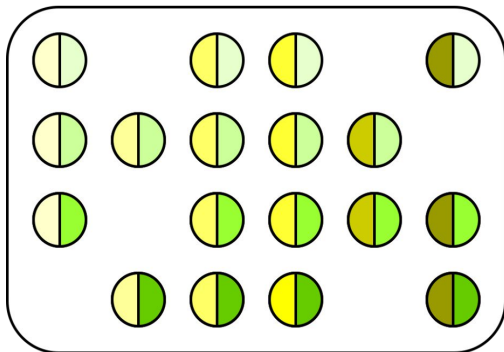


IID Test set

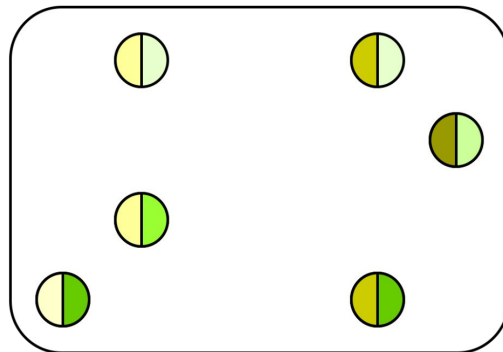


Model evaluation - out of distribution

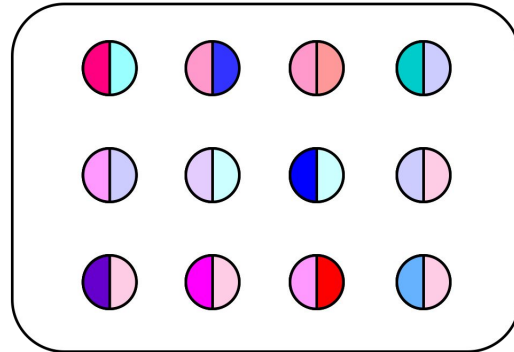
Train set



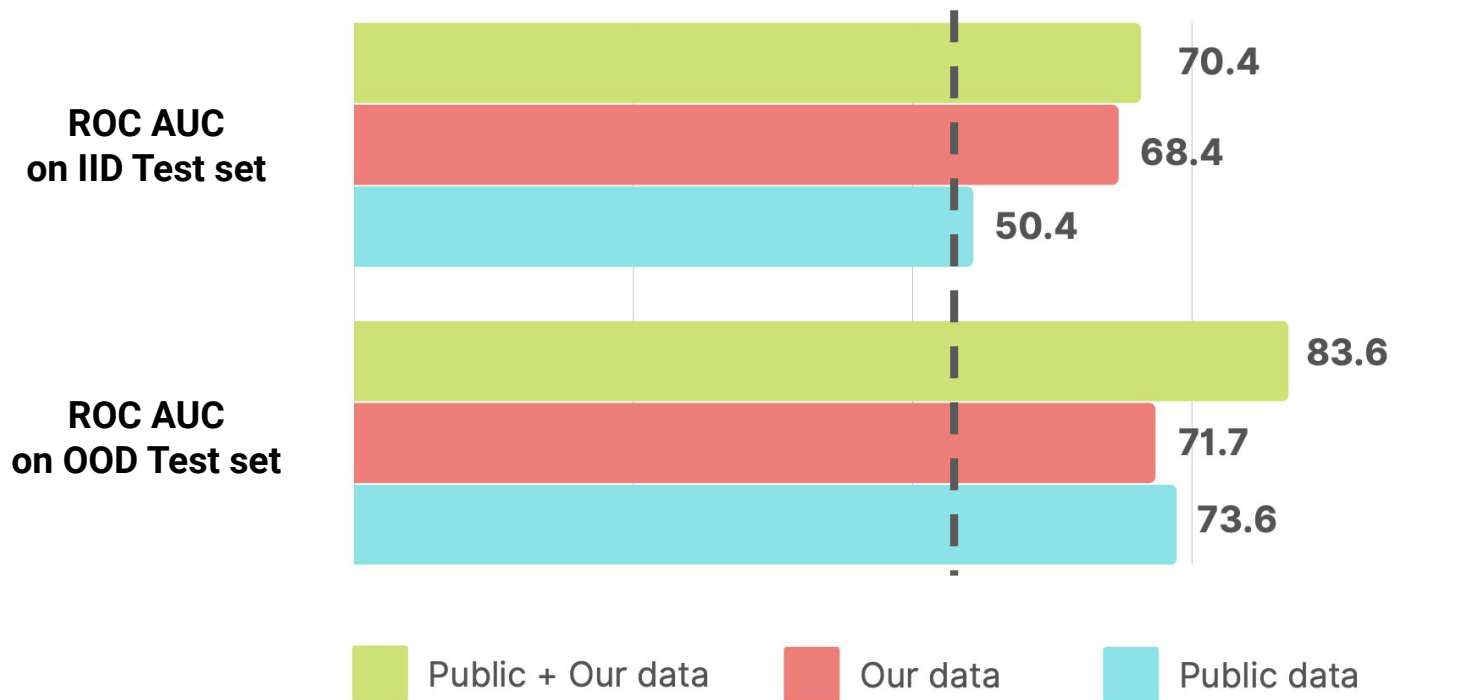
IID Test set



OOD Test set

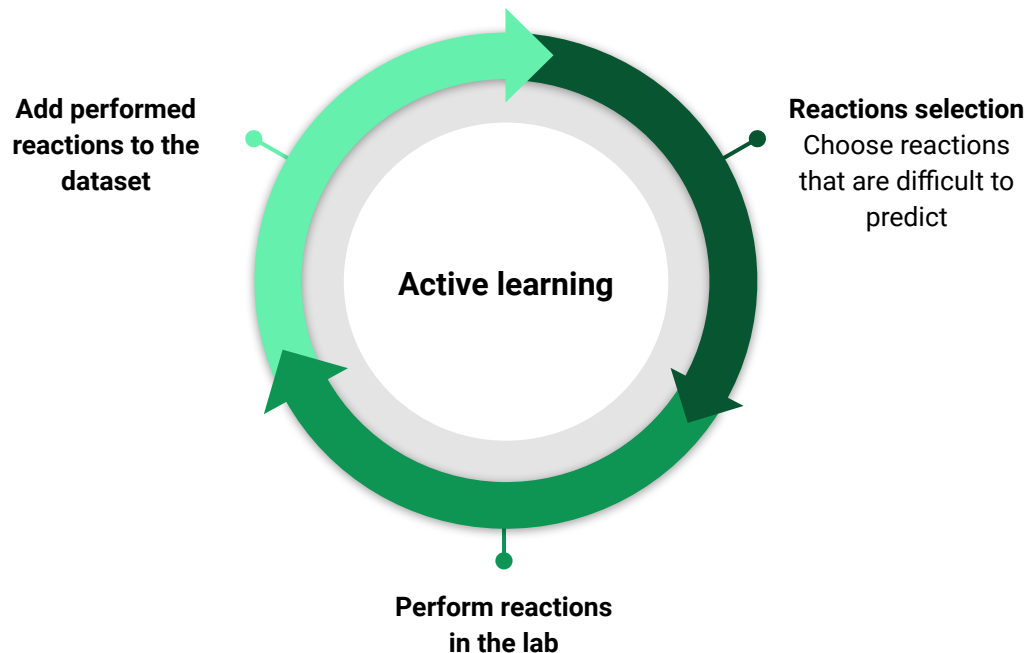


Evaluation on IID and OOD Test set



Open challenges

- How to estimate the model uncertainty?
- How to preserve the variety of performed reactions?
- How to better predict reaction conditions?
- How to implement data pipelines and ensure data correctness?



Thank you

- High failure rate of organic reactions is a key bottleneck in drug discovery
- Molecule.one solution: a closed loop with in-house wet lab focused on high throughput (performs in 1 week more reactions that a chemist usually performs in a year)
- First experiments show strong improvement of deep learning models; public data alone doesn't allow accurate predictions
- Many challenges to solve, especially related to active learning/continual learning!



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Molecule.one

We are hiring!