

# Computational Screening of Metal Organic Frameworks (MOFs) for Carbon Capture

Research &  
Innovation Center

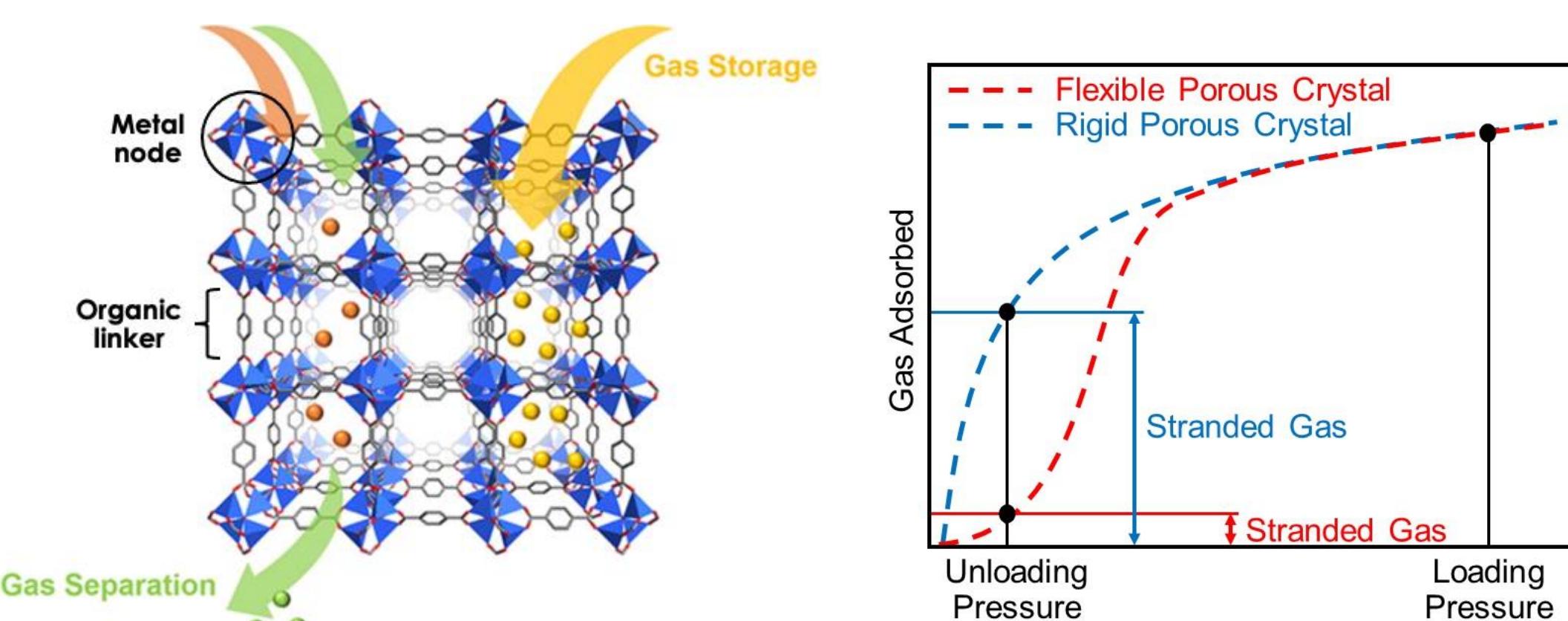


Samir Budhathoki<sup>1,2</sup>, Chris E. Wilmer<sup>3</sup>, Jan Steckel<sup>1</sup>

<sup>1</sup>National Energy Technology Laboratory, <sup>2</sup>NETL Support Contractor, <sup>3</sup>University of Pittsburgh

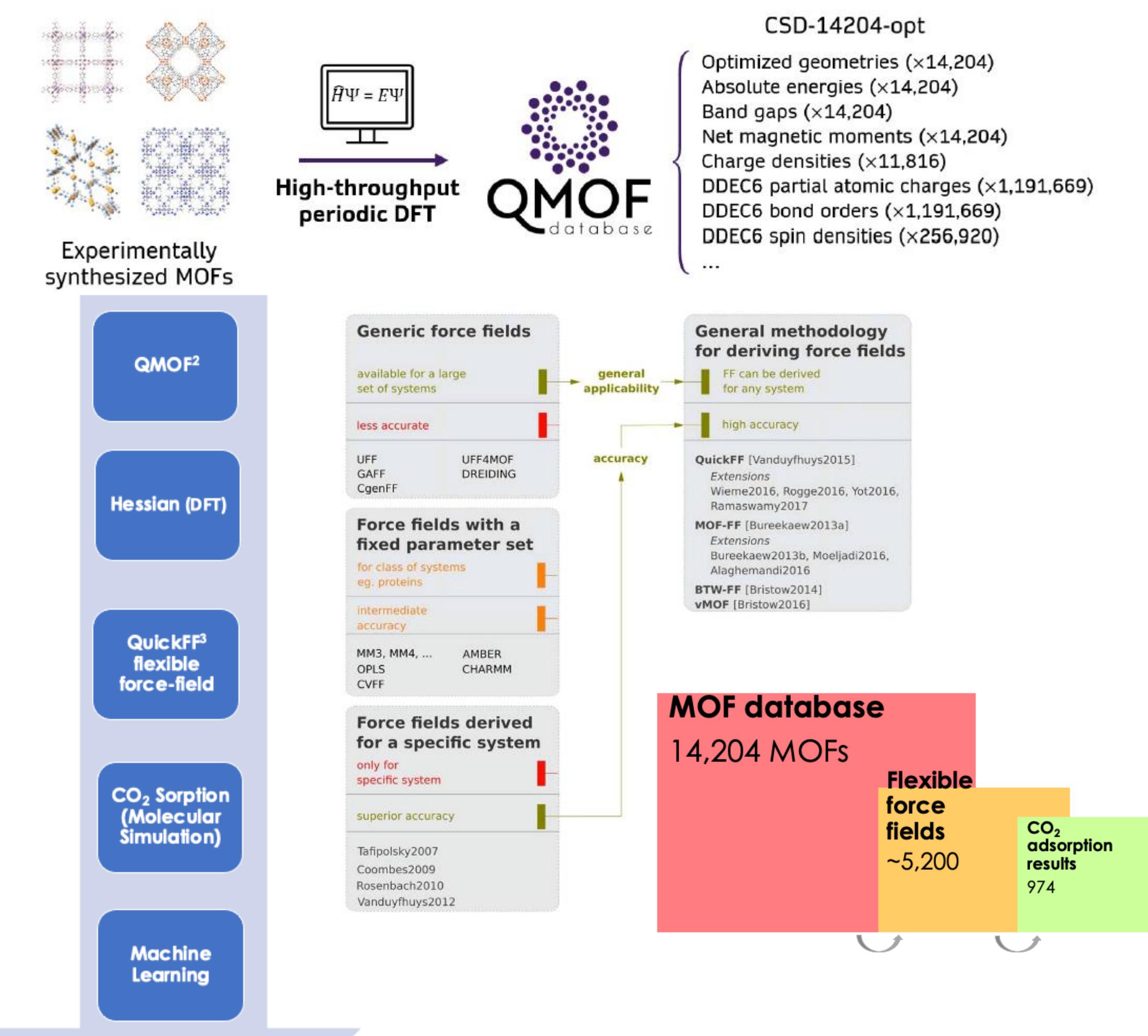
## Motivation

- MOFs are organic inorganic crystalline materials
- MOFs have excellent CO<sub>2</sub> sorption
- MOF design space is enormous due to a wide variety of constituents



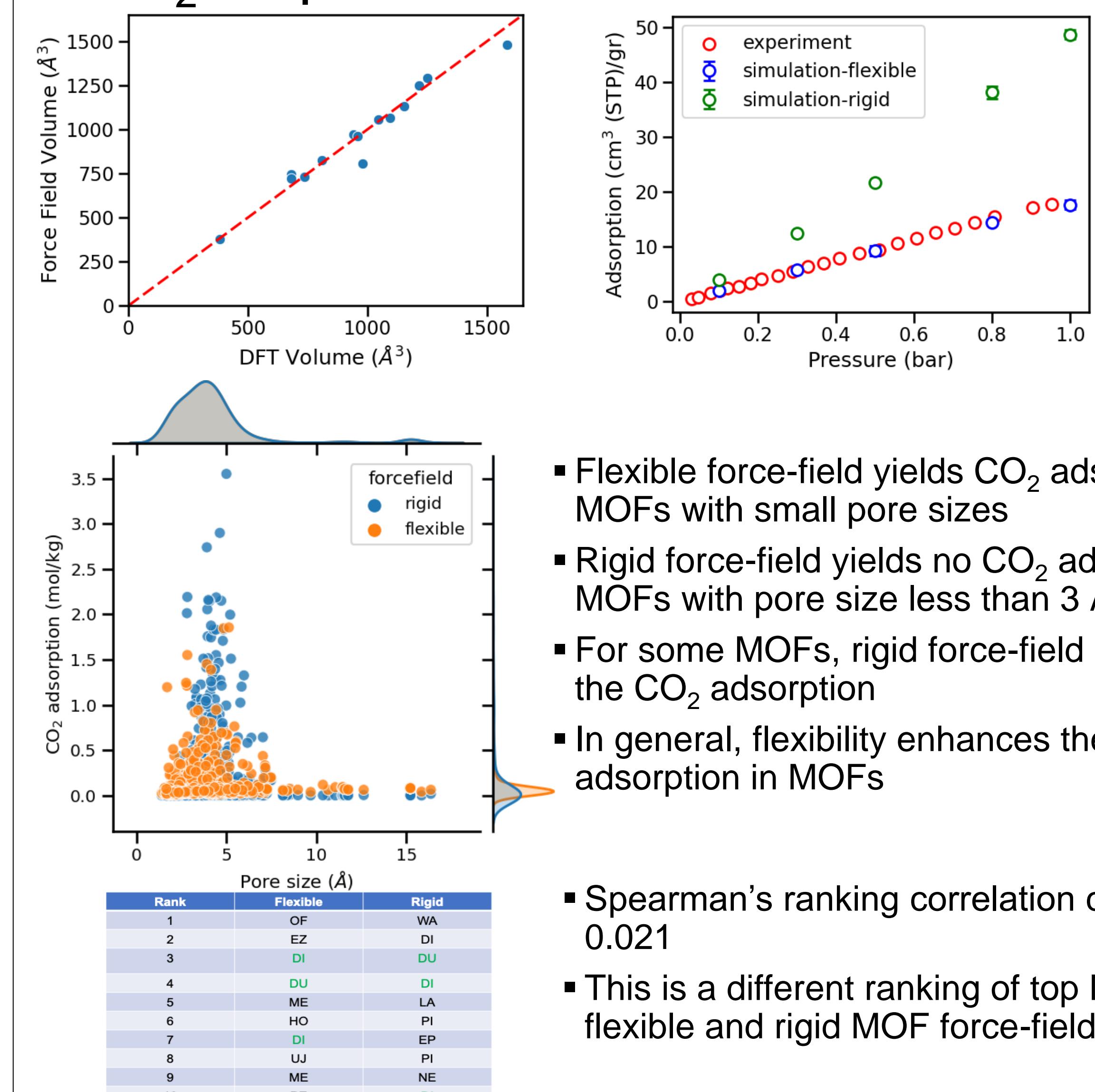
- Subtle flexible modes and large framework volume changes can influence MOF performance regarding adsorption-based applications<sup>1</sup>
- Most of the studies have focused on using fixed-atom force-fields (MOF atoms do not move), which are less accurate
- Flexible force-field describes movement of the MOFs environment, which is a more realistic model

## Approach



Software Packages:  
Molecular Simulation: LAMMPS<sup>4</sup>, RASPA<sup>5</sup>  
Machine Learning: PyCaret<sup>6</sup>

## Results: Flexible force-field validation and CO<sub>2</sub> sorption at ambient condition

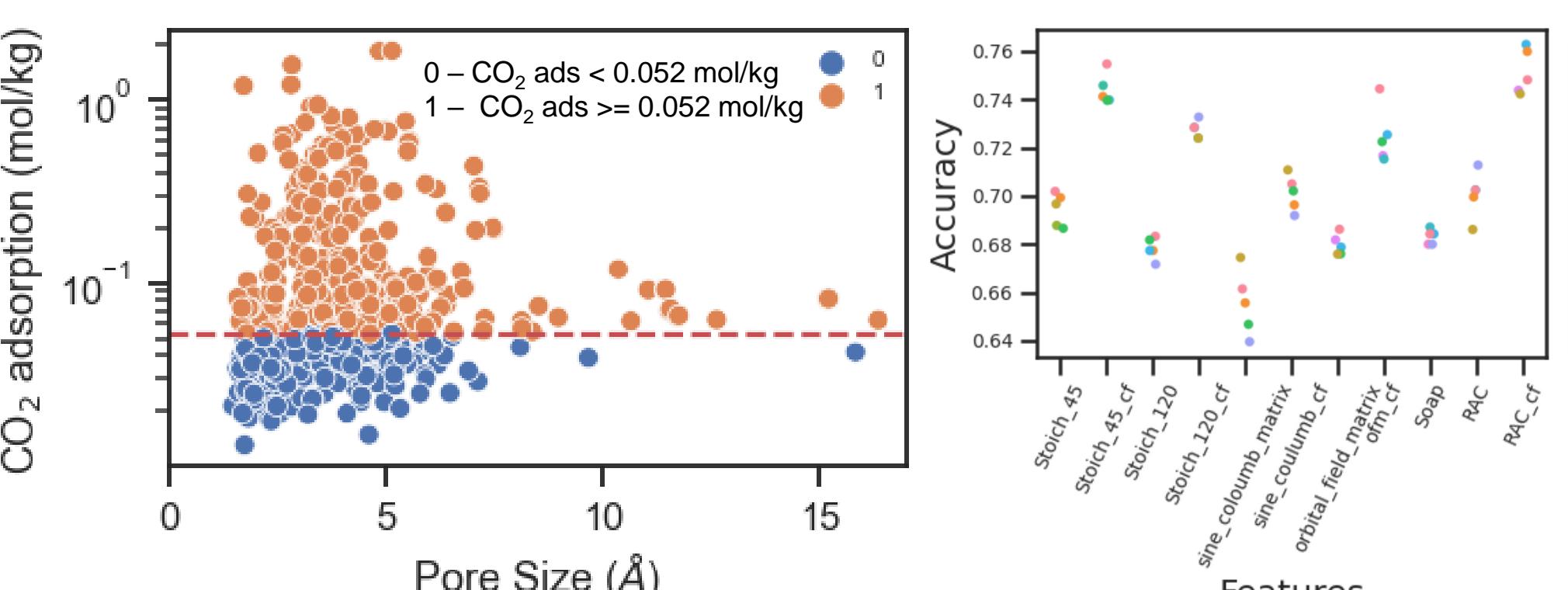


- Flexible force-field yields CO<sub>2</sub> adsorption in MOFs with small pore sizes
- Rigid force-field yields no CO<sub>2</sub> adsorption for MOFs with pore size less than 3 Å
- For some MOFs, rigid force-field overestimates the CO<sub>2</sub> adsorption
- In general, flexibility enhances the CO<sub>2</sub> adsorption in MOFs

▪ Spearman's ranking correlation coefficient is 0.021

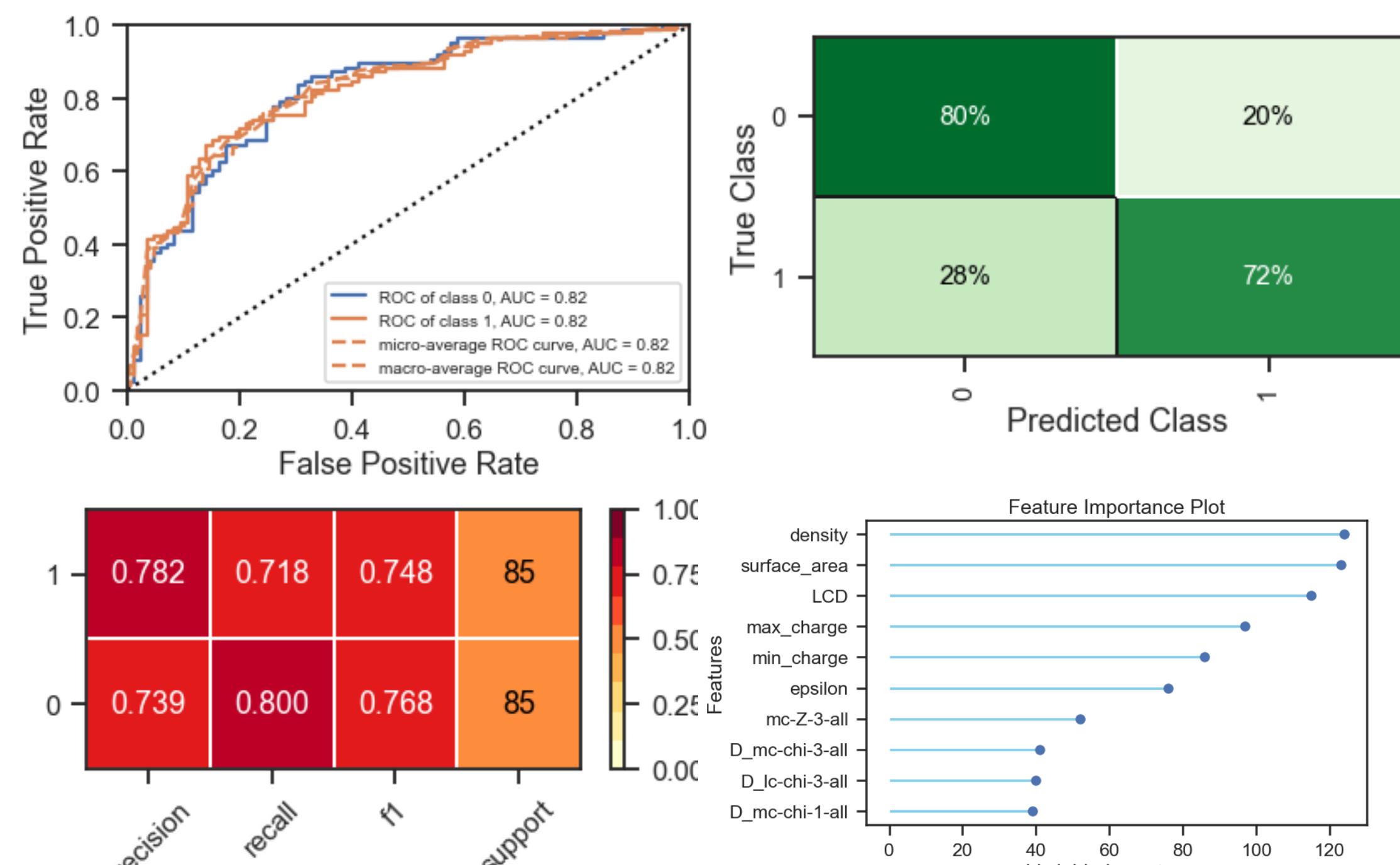
▪ This is a different ranking of top MOFs for flexible and rigid MOF force-fields

## Machine Learning (ML) Classifier



- 11 different classification ML models were evaluated for different MOF featurization techniques
- All the features were customized (denoted by '\_cf' in plot) with geometric features such as surface area, pore size, electrostatics, and dispersion term
- Customized RAC\_cf (revised auto-correlation) featurization yielded the best accuracy of all the models
- Light Gradient Boosting model (LGBM) algorithm performed the best

## ML Results



- ML model has an average accuracy of 75%, which is how often the model correctly identifies a MOF as 0 (low CO<sub>2</sub> adsorbing) or 1 (high CO<sub>2</sub> adsorbing)

## Conclusions

- Flexible force-fields yielded CO<sub>2</sub> sorption for MOFs with pore size < 3 Å
- Ranking of MOFs based on CO<sub>2</sub> sorption was different for flexible models and rigid models, which underscores the importance of incorporating flexibility for estimation of CO<sub>2</sub> sorption
- LGBM yielded about 75% accuracy in identifying low CO<sub>2</sub> adsorbing MOFs and high CO<sub>2</sub> adsorbing MOFs
- MOF geometric features (pore size, density, surface area) along with electrostatic and dispersion terms were found to be the most important features for the ML model

## References

1. Jawahery et. al., *Journal of Chemical Theory and Computation* 2019 15 (6), 3666-3677
2. Rosen et. al., *Matter* 2021, 4, 1-20
3. Vanduyghuys et. al., *J Comput Chem* 2018, 39(16), 999-1011
4. S. Plimpton, *Journal of Computational Physics* 1995, 117, 1-19
5. Dubbeldam et. al., *Molecular Simulation* 2016, 42(2), 81-101
6. Moez Ali, <https://www.pycaret.org>, 2020

Disclaimer - This project was funded by the United States Department of Energy, National Energy Technology Laboratory, in part, through a site support contract. Neither the United States Government nor any agency thereof, nor any of their employees, nor the support contractor, nor any of their employees, makes any warranty, express or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or any agency thereof. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or any agency thereof.