

# Implementation of Machine Learning Models to Predict Chemical Properties

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**Next Steps and Advice for** 

**Future SHINE Students** 

### Introduction

# Machine learning application for chemical properties prediction

Density Functional Theory (DFT) is widely used in the field of quantum chemistry because of its high accuracy and reasonable computational cost. However, in recent years, machine learning models have been proved to reduce computational cost while maintaining an acceptable level of accuracy.

### **Bioinspired Dicopper Complexes Screening** by DFT for Methane Hydroxylation:

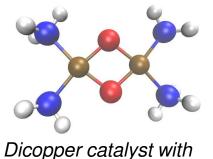
Dicopper Complexes inspired by pMMO is promising to catalyze methane activation at low temperature and pressure.

# **Objective & Impact of Professor's** Research

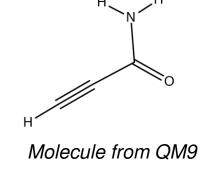
Professor Sharada is implementing state-of-theart machine learning techniques in combination with chemical research. Her main goals are to:

- Predict chemical properties more efficiently in a sensible, low-cost method
- Predict the connectivity between atoms based on previously trained datasets
- Comprehend the mechanisms of catalyzed reactions, thus enabling the design of new and high-performance catalysts.

The overall impact of this work is to allow chemical research to be done more efficiently, and design selective catalysts in order to save energy in the chemical industry, therefore contributing to the reduction of greenhouse gas



NH<sub>3</sub> ligand



## **Computational Methods**

### **Computational Methods:**

Properties (cal/mol

Properties (cal/mol

LR MAE

**Properties** 

**LR MAE** 

**DFT MAE** 

**KRR MAE** 

1.6

**ZPVE** 

(eV)

0.27

0.34

1.84

Regression (KRR)

**Linear Regression vs DFT** 

- Dataset: QM9 Google Database (rotational constant(GHz), dipole moment(Debye), isotropic polarizability(Bohr^3), energy of HOMO(Ha), etc.)
- Model: Linear Regression, Logistic Regression, Support Vector Machine (SVM), Regression Tree, etc.)

$$\hat{y}(w, x) = w_0 + w_1 x_1 + \dots + w_p x_p$$

$$\min_{w} ||Xw - y||_2^2$$

**Results and Discussion** 

Table 1: Linear Regression Mean Absolute Error

(MAE) of Chemical Properties

0.84

U<sub>0</sub> (Ha)

16.04

0.1

0.49

Table 2: Benchmark MAE<sup>1</sup>

(Debye) (Bohr<sup>3</sup>)

(Debye) (Bohr<sup>3</sup>)

3.65

 $U_{298}$ 

(Ha)

16.04

0.4

4.17

Linear regression has greater accuracy

when predicting HOMO and LUMO

Linear regression favored over KRR

**Linear Regression vs Kernel Ridge** 

when predicting  $C_v$  and  $\alpha$ 

HOMO

(eV)

0.27

 $H_{298}$ 

(Ha)

16.04

HOMO

(eV)

2.0

0.124

LUMO

(eV)

0.54

G<sub>298</sub>

(Ha)

16.04

LUMO

(eV)

2.6

0.133

**Representation: Fingerprint Molecule** 

### **Skills Learned**

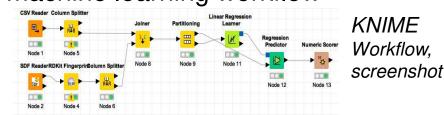
### **Machine Learning:**

- Basic ideas and principles
- Different models (ex: linear regression, , logistic regression, etc.)
- Optimization (regularization, cross validation)

### **Programming in Python:**

Use Spyder program in Anaconda **KNIME** 

Build machine learning workflow



# screenshot

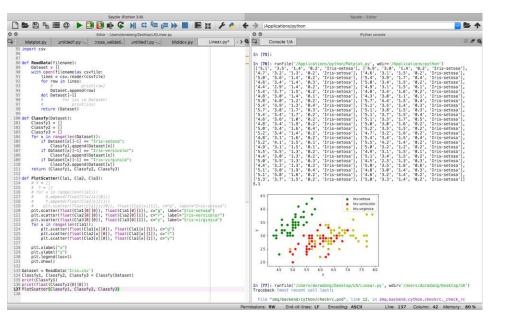
# **How This Relates to My STEM** Coursework

### **Computer Science:**

I am currently learning Java at school, and the similarities between Python and Java will help me be more prepared to take AP Computer Science A

### **Chemistry:**

Knowledge of different chemical attributes (heat capacity, HOMO/LUMO) will help me with my future science courses



kNN algorithm coded in Python using Spyder, screenshot

#### **Advice for Future Students:**

- I was worried about being one of the only sophomores in the program, but quickly grew to realize that age doesn't even matter, because we are all here for the same purpose.
- Don't be afraid of not knowing enough, you are here to learn!
- Keep in touch with the SHINE cohort and your mentors.

### **Future Steps:**

- Continue programming and learning more about machine learning algorithms
- Enter project into science fair
- Pursue more research programs

### Reference

1. Faber, F. A., Hutchison, L., Huang, B., Gilmer, J., Schoenholz, S. S., Dahl, G. E., ... & von Lilienfeld, O. A. (2017). Machine learning prediction errors better than DFT accuracy. arXiv preprint arXiv:1702.05532.

# **Acknowledgements**

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# levels in the environment.