Machine Learning techniques for predicting molecular properties

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Background work

Learning Invariant Representations of Molecules for Atomization Energy Prediction Self-taught Learning: Transfer Learning from Unlabeled Data Information - Theoretic Regret Bounds for Gaussian Process Optimization in the Bandit Setting

OVERVIEW

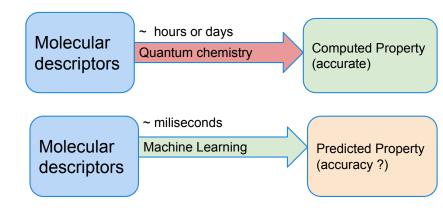
Introduction

Background work

- Compute/predict properties of molecules for materials design
 Examples: drug discovery, water purification, energy transmission and storage
- ► Quantum Chemistry calculations are expensive
- Machine Learning could predict properties of molecules at a fraction of the cost

Background work

Introduction



Overview

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Learning Invariant Representations of Molecules for Atomization Energy Prediction Self-taught Learning: Transfer Learning from Unlabeled Data Information - Theoretic Regret Bounds for Gaussian

Process Optimization in the Bandit Setting

- ► Desired properties of Molecular Descriptor
 - ► invariance to atom indexing
 - ► invariance to rotation and translation
- ► Coloumb Matrix [Rupp2012] scales $O(N^2)$

$$C(i,j) = \begin{cases} 0.5 * Z_i^{2.4} & i == j \\ Z_i * Z_j / ||R_i - R_j|| & i \neq j \end{cases}$$
 (1)

 Z_i is charge of atom i R_i is 3D position of atom i

COLOUMB MATRIX

Coloumb Matrix descriptor

- ▶ invariant to rotation and translation (use $||R_i R_i||$)
- invariance to atom indexing
 - Sorted Coloumb Matrix indexes given by sorting the row norms
 - Random Coloumb Matrix generate multiple Sorted Coloumb Matrices perturbed by noise)

RANDOM COLOUMB MATRIX

Coloumb Matrix descriptor

► Atomization energy prediction for a dataset of 7k samples (5.5k training 1.5k testing) with H,O,C,N,S Multilayer Feed Forward NN with Random Coloumb 3.1 kcal/mol MAE - chemical accuracy level 1kcal/mol

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