

Project: Molecular Energy Prediction

ModIA 2024

The goal of this project is to model the inter-atomic potential energy surface of small organic molecules [1]. We denote by $\mathbf{r} = \{r_1, r_2, \dots, r_N\}$ the positions of atoms in a molecule in 3D space, and denote by $E(\mathbf{r})$ the atomization energy of the configuration \mathbf{r} . One should define a unified model for all the molecules to predict $E(\mathbf{r})$, based on the geometric information represented by \mathbf{r} and some extra information about the atoms. This amounts to a high dimensional regression problem.

The most challenging component of this problem is to respect the symmetry constraint. In other words, we define the translational, rotational, and permutational operations of \mathbf{r} as:

$$T_b(\mathbf{r}) = \mathbf{r} + b, \quad T_U(\mathbf{r}) = U\mathbf{r}, \quad T_\sigma(\mathbf{r}) = \{r_{\sigma(1), \sigma(2), \dots, \sigma(N)}\}$$

The symmetry constraint implies that

$$E(T_b(\mathbf{r})) = E(T_U(\mathbf{r})) = E(T_\sigma(\mathbf{r})) = E(\mathbf{r})$$

for all possible translation b , rotation U and permutation σ (acting on the N particles in the 3D domain).

Data description

We use a subset of QM7-X, which contains 4739 structures of molecules, with various number of particles (atoms). QM7-X is an extension of the dataset QM7, and there are many works to address the same regression problem for QM7, c.f. quantum-machine.org. Further details about QM7-X can be found in [1].

The data source contains 2 folders: atoms and energies. In the folder of atoms, the training and test data are further separated. The file is ordered by the id of each molecule configuration, written in the extended xyz format. You may use the python package ase to read the coordinates \mathbf{r} in each xyz file. This file also contains the type of each atom. In the folder of energies, there is one csv file, with two columns: id and energy. The energy is real-valued and it is the atomization energy of the molecules in the training set. In this challenge, the goal is produce another csv file in the same format, computed from the test data.

Metric description

The metric used in this challenge to decide the winner is the root mean square error. Assume you model predicts $\tilde{E}(\mathbf{r}_{id})$ on D test configurations $\{\mathbf{r}_{id}\}_{id \leq D}$, then the error is

$$\sqrt{\frac{1}{D} \sum_{id=1}^D (E(\mathbf{r}_{id}) - \tilde{E}(\mathbf{r}_{id}))^2}$$

Group work, presentation and report

Group work The project is to be realized in group en [binome](https://binome.com). During the TPs of this project, we are going to apply scattering 3d to this problem, by following a tutorial from Kymatio website [2]. Meanwhile, you may also try other methods to address this problem.

Presentation Each group should present their work in 5 minutes, followed by a 5 minute discussion.

Report Each group should also write a report which contains the following elements:

- Data preprocessing: visualizaton of dataset
- Model description: discuss invariant properties of each proposed model. Guve details on how you address the problem and its novelty.
- Analysis and evaluation of the results: clarity of main results, training plots, parameter turning, and cross validation. It should contain details in order to reproduce your results (e.g hyper-parameters).
- Perspective of the work: state-of-the-art, future work. Conclude with a brief summary of the main idea and your results, including your insights on the topic.
- Writing Quality: it should be concise by highlighting your main ideas (motivation, state-of-the-art, precise question and solution).

References

- [1] Johannes Hoja, Leonardo Medrano Sandonas, Brian G Ernst, Alvaro Vazquez-Mayagoitia, Robert A DiStasio Jr, and Alexandre Tkatchenko. Qm7-x, a comprehensive dataset of quantum-mechanical properties spanning the chemical space of small organic molecules. *Scientific data*, 8(1):1–11, 2021.
- [2] 3d scattering quantum chemistry regression. Available at https://www.kymat.io/gallery_3d/scattering3d_qm7_torch.html#sphx-glr-gallery-3d-scattering3d-qm7-torch-py.