
TS-DiffuGen

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INTRODUCTION:

This file contains simple documentation regarding the TS-DiffuGen package.

DATASETS:

The dataset classes can be found within the data directory.

2.1 Setting up file:

`data.Dataset_W93.setup_dataset_files.create_pdb_from_file_path(path_to_raw_log, path_to_final)`

Create a .XYZ file from a .log DFT file containing atomic coordinates.

This function reads a DFT .log file, extracts atomic symbols and coordinates, and writes them into a .PDB file format.

Parameters

- **path_to_raw_log** (*str*) – Path to the input DFT .log file.
- **path_to_final** (*str*) – Path to the output .PDB file to be created.

Returns

None

`data.Dataset_W93.setup_dataset_files.extract_geometry(logs)`

Extract geometry from DFT .log files.

This function extracts atomic symbols and coordinates from a list of lines obtained from a DFT .log file. It searches for the section indicating the standard nuclear orientation and extracts the atom labels and corresponding coordinates in the next line.

Parameters

logs (*list*) – List of lines from the DFT .log file.

Returns

A tuple containing a list of atomic symbols and a numpy array of atomic coordinates.

Return type

tuple

`data.Dataset_W93.setup_dataset_files.process_reactions(dataframe, directory)`

Process reaction data from .log files and create various output files.

This function processes reaction data stored in .log files and organizes the data into clean geometry files in .xyz format. It also generates images of the reactants and products and saves them, along with the geometry files, for each reaction.

Parameters

- **dataframe** (*pd.DataFrame*) – DataFrame containing reaction data, including reactant and product SMILES.

- **directory** (*str*) – Path to the main directory containing the raw log files.

Returns

None

2.2 Simple W93 Dataset without Graphs:

2.3 W93 Dataset with Graphs:

2.4 TX1 Dataset:

2.5 RGD1 Dataset:

`data.Dataset_RGD1.parse_data.checking_duplicates()`

Check for duplicate reactants and products based on SMILES strings.

This function reads reaction data from an HDF5 file, parses the SMILES strings of reactants and products, and identifies duplicate reactions. It keeps track of duplicate reaction information and their counts.

Returns

None

`data.Dataset_RGD1.parse_data.main()`

Main Function to save reaction geometries in required XYZ format.

This function reads a dataset containing reaction geometries from an HDF5 file, parses the elements and geometries of reactants, products, and transition states, and saves them as XYZ files.

Returns

None

`data.Dataset_RGD1.parse_data.save_reactions_with_multiple_ts(save_even_single=False)`

Save reactions with multiple TS conformers to separate directories.

This function reads reaction data from an HDF5 file along with activation energy data from a CSV file. It identifies reactions with multiple transition state (TS) conformers based on identical reaction SMILES strings. It then organizes and saves these reactions and their geometries into separate directories.

Parameters

save_even_single (*bool*, *optional*) – Save reactions with only one TS conformer as well.
(default: False)

Returns

None

`data.Dataset_RGD1.parse_data.save_xyz_file(path, atoms, coordinates)`

Save XYZ file with atomic elements and corresponding coordinates.

Parameters

- **file_path** (*str*) – Path to the output XYZ file.
- **elements** (*list*) – List of atomic symbols.
- **coordinates** (*numpy.ndarray*) – Array of atomic coordinates.

Returns
None

THE EQUIVARIANT GRAPH NEURAL NETWORK:

Background about equivariance and invariance as well as the architecture of the model is described in Sacha's thesis. The main EGNN files used for the diffusion process are the dynamics files:

3.1 Dynamics without Graphs:

3.2 Dynamics with Graphs:

The main difference being that the EGNN takes extra inputs of edge attributes.

3.3 Utility Functions:

The Utility functions used in the EGNN process are outlined bellow:

EXAMPLE FUNCTIONS AND CLASSES FROM THE PACKAGE INCLUDE:

See [src.evaluate_samples](#) for details.

`src.evaluate_samples.calc_cov_mat(results_matrix, cov_threshold=0.1)`

Calculate COV and MAT scores based on D-MAE matrix.

Parameters

- **results_matrix** (*np.ndarray*) – D-MAE matrix.
- **cov_threshold** (*float, optional*) – COV threshold. Defaults to 0.1.

Returns

Calculated MAT-R mean, median, and COV-R scores.

Return type

tuple

`src.evaluate_samples.calculate_DMAE(gen_mol, true_mol)`

Calculate D-MAE between inter-atomic distance matrices.

Parameters

- **gen_mol** (*list*) – Inter-atomic distance matrix of generated molecule.
- **true_mol** (*list*) – Inter-atomic distance matrix of true molecule.

Returns

D-MAE value.

Return type

float

`src.evaluate_samples.calculate_best_rmse(gen_mol, ref_mol, max_iters=100000, use_hydrogens=False)`

Calculate Best RMSD between RDKit Molecule Objects.

Parameters

- **gen_mol** (*Chem.Mol*) – RDKit molecule object representing generated molecule.
- **ref_mol** (*Chem.Mol*) – RDKit molecule object representing reference molecule.
- **max_iters** (*int, optional*) – Maximum atom matches. Defaults to 100_000.
- **use_hydrogens** (*bool, optional*) – True to include hydrogens. Defaults to False.

Returns

Best RMSD value.

Return type

float

`src.evaluate_samples.calculate_distance_matrix(coordinates)`

Calculate pairwise distance matrix from 3D coordinates.

Parameters**coordinates** (*list*) – List of 3D coordinates for each atom.**Returns**

Pairwise distance matrix.

Return type

np.ndarray

`src.evaluate_samples.create_lists(original_path, RMSD=False)`

Create lists of true and generated molecules from the given path.

Parameters

- **original_path** (*str*) – Path to the original directory containing molecule files.
- **RMSD** (*bool, optional*) – True if RMSD format, False if standard XYZ format.

Returns

Two lists containing RDKit molecule objects.

Return type

tuple

`src.evaluate_samples.create_table(true_mols, gen_mols, max_iters=1, metric='RMSD')`

Create comparison table between molecules.

Parameters

- **true_mols** (*list*) – List of true molecule RDKit objects.
- **gen_mols** (*list*) – List of lists containing generated molecule RDKit objects.
- **max_iters** (*int*) – Maximum atom matches for RMSD calculation.
- **metric** (*str*) – Metric choice, “RMSD” or “DMAE”.

Returns

DataFrame of comparison metrics.

Return type

pd.DataFrame

`src.evaluate_samples.evaluate(sample_path, evaluation_type, cov_threshold=0.1)`

Evaluate generated samples using RMSE or D-MAE metrics.

Parameters

- **sample_path** (*str*) – Path to sample directory.
- **evaluation_type** (*str*) – Metric choice, “RMSD” or “DMAE”.
- **cov_threshold** (*float, optional*) – COV threshold. Defaults to 0.1.

`src.evaluate_samples.get_paths(sample_path)`

Load molecule files from a sample path and organize them into true and generated samples.

Parameters**sample_path** (*str*) – Path to the sample directory containing molecule files.

Returns

Two lists containing true and generated sample file paths.

Return type

tuple

`src.evaluate_samples.import_xyz_file(molecule_path, RMSD=False)`

Import an XYZ file as an RDKit molecule object.

Parameters

- **molecule_path** (*str*) – File path of the XYZ file to be imported.
- **RMSD** (*bool*, *optional*) – True if RMSD format, False if standard XYZ format.

Returns

RDKit molecule object or None if loading fails.

Return type

Chem.Mol or None

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