

Module Interface Specification for Re-ProtGNN

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1 Revision History

Date	Version	Notes
Mar 19, 2025	1.0	Initial Draft

2 Symbols, Abbreviations and Acronyms

See SRS Documentation at <https://github.com/Yuanqi-X/Re-ProtGNN/blob/main/docs/SRS/SRS.pdf>.

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3 Introduction

The following document details the Module Interface Specifications for Re-ProtGNN, a re-implementation of an interpretable Graph Neural Network (GNN) Framework.

Complementary documents include the System Requirement Specifications and Module Guide. The full documentation and implementation can be found at <https://github.com/Yuanqi-X/Re-ProtGNN/tree/main>.

4 Notation

The structure of the MIS for modules comes from Hoffman and Strooper (1995), with the addition that template modules have been adapted from Ghezzi et al. (2003). The mathematical notation comes from Chapter 3 of Hoffman and Strooper (1995). For instance, the symbol $:=$ is used for a multiple assignment statement and conditional rules follow the form $(c_1 \Rightarrow r_1 | c_2 \Rightarrow r_2 | \dots | c_n \Rightarrow r_n)$.

The following table summarizes the primitive data types used by Re-ProtGNN.

Data Type	Notation	Description
integer	\mathbb{Z}	a number without a fractional component in $(-\infty, \infty)$
natural number	\mathbb{N}	a number without a fractional component in $[1, \infty)$
real	\mathbb{R}	any number in $(-\infty, \infty)$
boolean	<code>bool</code>	Boolean value: either <code>True</code> or <code>False</code>
string	<code>str</code>	A sequence of Unicode characters
tensor	<code>Tensor</code>	A multi-dimensional array object from PyTorch
graph	<code>Data</code>	A graph object from PyTorch Geometric, with node and edge attributes
dataset	<code>Dataset</code>	A collection of graph objects for training or evaluation
dataloader	<code>DataLoader</code>	A PyTorch Geometric data loader for batching graph data
dictionary	<code>dict[K \rightarrow V]</code>	A mapping from keys of type <code>K</code> to values of type <code>V</code>
list	<code>list[T]</code>	A sequence of elements of type <code>T</code>
function	<code>Customized Function</code>	A self-defined callable function

Re-ProtGNN uses functions, which are defined by the data types of their inputs and outputs. Local functions are described by giving their type signature followed by their specification.

5 Module Decomposition

The following table is taken directly from the Module Guide document for this project.

Level 1	Level 2
Hardware-Hiding	
Behaviour-Hiding Module	Configuration Module Input Format Module Control Module Training Module Output Visualization Module
Software Decision Module	Model Module Inference Module Explanation Module Pytorch Module Pytorch Geometric Module GUI Module

Table 1: Module Hierarchy

6 MIS of Configuration Module

6.1 Module

Configuration

6.2 Uses

None

6.3 Syntax

6.3.1 Exported Constants

None

6.3.2 Exported Access Programs

None

6.4 Semantics

6.4.1 State Variables

- `data_args`: `DataParser` — Stores dataset-level configuration such as name, directory, splitting strategy, and seed.
- `model_args`: `ModelParser` — Stores GNN architecture settings and prototype-related parameters.
- `train_args`: `TrainParser` — Stores training hyperparameters including learning rate, batch size, and epoch count.
- `mcts_args`: `MCTSParser` — Stores Monte Carlo Tree Search and explanation-specific rollout parameters.
- `random_seed`: `int` — Stores the global seed used for generating random numbers.

6.4.2 Environment Variables

None

6.4.3 Access Routine Semantics

None - The state variables in this module are initialized when the system loads and are accessed directly by other modules using:

```
from utils.Configures import data.args, train.args, model.args, mcts.args
```

As such, no explicit accessor routines are exported.

6.4.4 Local Functions

DataParser(name: str, dir: str, split: list[\mathbb{R}], seed: int) \rightarrow DataParser

- output: Returns a configuration object for dataset settings including `name`, `dir`, `split`, and `seed`.

ModelParser(model_name: str, hidden_dim: \mathbb{N} , num_prototypes: \mathbb{N}) \rightarrow ModelParser

- output: Returns a configuration object containing the GNN model name, hidden dimension, and prototype count.

TrainParser(batch_size: \mathbb{N} , lr: \mathbb{R} , epochs: \mathbb{N}) \rightarrow TrainParser

- output: Returns a configuration object with the training hyperparameters: `batch_size`, `lr`, and `epochs`.

MCTSParser(num_rollouts: \mathbb{N} , exploration_const: \mathbb{R}) \rightarrow MCTSParser

- output: Returns a configuration object specifying the number of rollouts and exploration constant for MCTS-based explanation.

7 MIS of Input Format Module

7.1 Module

dataUtils

7.2 Uses

PyTorch Geometric Module (15), PyTorch Module (14), Configuration Module (6), Output Visualization Module (10)

7.3 Syntax

7.3.1 Exported Constants

None

7.3.2 Exported Access Programs

Name	In	Out	Exceptions
load_dataset	-	tuple[Dataset, int, int, dict[str → DataLoader]]	FileNotFoundError, ValueError, NotImplementedError

7.4 Semantics

7.4.1 State Variables

None

7.4.2 Environment Variables

- **dataset_dir:** str — Filesystem path to the dataset root directory, obtained from `data_args.dataset_dir` defined in the Configuration Module.
- **log_file:** str — Path to the log file used by the `append_record()` routine exported from the Output Visualization Module.

7.4.3 Access Routine Semantics

`load_dataset()`:

- transition:
 - Loads the dataset using `data_args.dataset_name` and `data_args.dataset_dir`, where `data_args` are defined in the Configuration Module.

- Logs the dataset name using `append_record(data_args.dataset_name)`, where `append_record()` is a routine exported from the Output Visualization Module
- output:
 - Returns a tuple: `(dataset, input_dim, output_dim, dataloader)` where:
 - * `dataset`: graph dataset object loaded using `_get_dataset()`
 - * `input_dim`: number of node features from `dataset.num_node_features`
 - * `output_dim`: number of output classes from `dataset.num_classes`
 - * `dataloader`: dictionary of DataLoaders split via `_get_dataloader()`
- exception:
 - `FileNotFoundError`: Raised if required dataset files are missing in the specified directory, such as missing raw `‘.pkl‘` or `‘.txt‘` files for the dataset.
 - `ValueError`: Raised if raw data files exist but are empty or malformed (e.g., missing node labels).
 - `NotImplementedError`: Raised if `data_args.dataset_name` does not match any supported dataset (i.e., not MUTAG, BA_2Motifs, or a MoleculeNet dataset).

7.4.4 Local Functions

`_get_dataset(dataset_dir: str, dataset_name: str) → Dataset`

- output: Selects an appropriate dataset loader based on `dataset_name` and returns the resulting dataset loaded from `dataset_dir`. See the Pytorch Geometric Module 15 for the type `Dataset`.

`_get_dataloader(dataset: Dataset, batch_size: \mathbb{N} , data_split_ratio: list[\mathbb{R}]) → dict[str → DataLoader]`

- output: Splits the input `dataset` into train/eval/test sets according to `data_split_ratio`, and returns DataLoaders batched by `batch_size`. See the PyTorch Geometric Module 15 for the type `DataLoader`.

8 MIS of Control Module

8.1 Module

main

8.2 Uses

Configuration Module (6), Input Format Module (7), Model Module (11), Training Module (9), Inference Module (12), Explanation Module (13), PyTorch Module (14)

8.3 Syntax

8.3.1 Exported Constants

None

8.3.2 Exported Access Programs

Name	In	Out	Exceptions
main	clst: \mathbb{R} , sep: \mathbb{R}	-	-

8.4 Semantics

8.4.1 State Variables

None

8.4.2 Environment Variables

- **dataset_dir:** `str` — Filesystem path to the dataset root directory (from `data_args.dataset_dir`).
- **checkpoint_dir:** `str` — Directory path for saving and loading model checkpoints, constructed using `data_args.dataset_name`.
- **device:** `str` — Device identifier used by PyTorch for model training and inference (e.g., 'cpu' or 'cuda').

8.4.3 Access Routine Semantics

`main(clst, sep):`

- transition:
 - Loads the dataset and dataloaders using `load_dataset()`, which references `dataset_dir`.

- Initializes a GNN model and loss function using `setup_model(input_dim, output_dim, model_args)` from Model Module (11).
 - Constructs `checkpoint_dir := "./src/checkpoint/{data_args.dataset_name}/"`.
 - Trains the model using `train(clst, sep, dataset, dataloader, gnnNets, output_dim, criterion, checkpoint_dir)` from Training Module (9).
 - Loads the best checkpoint from `checkpoint_dir`, and updates model weights using `update_state_dict()`.
 - Evaluates the trained model via `test(dataloader['test'], gnnNets, criterion)` from Inference Module (12).
 - Generates explanations using `exp_visualize(dataset, dataloader, gnnNets, output_dim)` from Explanation Module (13).
- output: None
 - exception: None

8.4.4 Local Functions

None

9 MIS of Training Module

9.1 Module

Train

9.2 Uses

Configuration Module (6), Model Module (11), Explanation Module (13), Output Visualization Module (10), PyTorch Module (14)

9.3 Syntax

9.3.1 Exported Constants

None

9.3.2 Exported Access Programs

Name	In	Out	Exceptions
train	clst: \mathbb{R} , sep: \mathbb{R} , dataset: <code>-</code> Dataset, dataloader: <code>dict[str</code> <code>→ DataLoader]</code> , gnnNets: GnnNets, output_dim: \mathbb{N} , cri- terion: <code>Customized Function</code> , ckpt_dir: <code>str</code>	-	None

9.4 Semantics

9.4.1 State Variables

None

9.4.2 Environment Variables

- `checkpoint_dir: str` — Path to the directory for saving model checkpoints.
- `device: str` — Target computation device, used to allocate model weights and prototype vectors.

9.4.3 Access Routine Semantics

`train(clst, sep, dataset, dataloader, gnnNets, output_dim, criterion, ckpt_dir):`

- transition:

- Initializes the optimizer using parameters from `gnnNets` and `train_args`.
 - Logs statistics for `dataset` using `_log_dataset_stats(dataset)`.
 - Iteratively trains the model using batches from `dataloader['train']` with cluster/separation losses weighted by `clst` and `sep`.
 - Periodically projects prototypes onto embedding space using `_project_prototypes(gnnNets, dataset, ...)`.
 - Evaluates performance on the validation set using `_evaluate(dataloader['eval'], gnnNets, criterion)`.
 - Saves model checkpoints to `ckpt_dir`.
- output: None
 - exception: None

9.4.4 Local Functions

`_evaluate(eval_dataloader: DataLoader, model: GnnNets, criterion: Customized Function) → dict[str → float]`

- transition: None
- output: Runs model evaluation on `eval_dataloader` and computes loss/accuracy basing on `criterion`. Returns a dictionary with keys "loss" and "acc".

`_log_dataset_stats(dataset: Dataset) → None`

- transition: Computes average number of nodes and edges from `dataset`, and prints the result.
- output: None

`_project_prototypes(model: GnnNets, dataset: Dataset, indices: list[\mathbb{N}], output_dim: \mathbb{N}) → None`

- transition: Updates each prototype vector in `model` with a real example from `dataset` using `get_explanation()` from Explanation Module (13).
- output: None

10 MIS of Output Visualization Module

10.1 Module

outputUtils

10.2 Uses

PyTorch Module (14), PyTorch Geometric Module (15), GUI Module (16)

10.3 Syntax

10.3.1 Exported Constants

None

10.3.2 Exported Access Programs

Name	In	Out	Exceptions
ExpPlot	dataset_name: <code>str</code>	ExpPlot instance	-
draw	graph: <code>networkx.Graph</code> , nodelist: <code>list[int]</code> , figname: <code>str</code> , kwargs: <code>dict</code>	-	NotImplementedError
append_record	info: <code>str</code>	-	FileNotFoundError
save_best	ckpt_dir: <code>str</code> , epoch: <code>N</code> , gnnNets: <code>GnnNets</code> , model_name: <code>str</code> , eval_acc: \mathbb{R} , is_best: <code>bool</code>	-	-

10.4 Semantics

10.4.1 State Variables

None

10.4.2 Environment Variables

- `log_file: str` — Hardcoded path to the log file: `./results/log/hyper_search`.
- `device: str` — Computation device (e.g., ‘cuda’ or ‘cpu’) used to store model after saving.

10.4.3 Access Routine Semantics

ExpPlot(dataset_name):

- transition: None
- output: Constructs an instance for drawing explanations for `dataset_name`. Returns an `ExpPlot` object.
- Note: Please see in PyTorch Geometric Module (15) for the type `networkx.Graph`.

draw(graph, nodelist, figname, kwargs):

- transition: Calls the drawing routine and uses GUI Module (16) to generate and save a figure to `figname`.
- output: None
- exception: `NotImplementedError` if `dataset_name` is unsupported.

append_record(info):

- transition: Writes `info` as a new line to the file located at `log_file`.
- output: None
- exception: `FileNotFoundError` if the parent directory of `log_file` does not exist.

save_best(ckpt_dir, epoch, gnnNets, model_name, eval_acc, is_best):

- transition:
 - Saves model weights and training metadata to `ckpt_dir`.
 - If `is_best=True`, copies this file to `ckpt_dir`.
 - Moves model between ‘cpu’ and device `:= model_args.device`.
- output: None
- exception: None

10.4.4 Local Functions

None

11 MIS of Model Module

11.1 Module

GnnNets

11.2 Uses

None

11.3 Syntax

11.3.1 Exported Constants

None

11.3.2 Exported Access Programs

Name	In	Out	Exceptions
GnnNets	input_dim: int, output_dim: int, model_args: dict	GnnNets	NotImplementedError
forward	data: Data, prot- gnn_plus: bool, similar- ity: Tensor	logits: Tensor, - prob: Tensor, emb1: Tensor, emb2: Tensor, min_distances: Tensor	-
update_state_dict	state_dict: dict	-	-
to_device	-	-	-

11.4 Semantics

11.4.1 State Variables

- **self.model**: the internal GNN encoder consisting of learnable layers.
- **self.prototype_vectors**: a tensor containing learnable prototype embeddings, where each prototype represents a latent concept tied to a specific class.
- **self.device**: the computing device (e.g., 'cpu' or 'cuda') on which the model is running.

11.4.2 Environment Variables

GPU/CPU hardware for model training and inference.

11.4.3 Assumptions

None

11.4.4 Access Routine Semantics

GnnNets(input_dim, output_dim, model_args):

- transition: None
- output: Returns an instance of the **GnnNets** class with specified input/output dimensions and model hyperparameters.
- exception:
 - **NotImplementedError**: if the specified model name in **model_args** is unsupported.

forward(data, protgnn_plus, similarity):

- transition: Moves graph data to the correct device and performs a forward pass through the model.
- output:
 - **logits**: raw output scores for each class.
 - **prob**: predicted class probabilities for each input graph, obtained by applying softmax to logits.
 - **emb1**: intermediate representation from an early layer of the model.
 - **emb2**: deeper-level embedding capturing higher-level graph features after additional processing layers.
 - **min_distances**: for each input graph, the minimum distance to each prototype vector.
- exception: None

update_state_dict(state_dict):

- transition: Loads and updates model parameters from a dictionary of saved weights.
- output: None
- exception: None

to_device():

- transition: Moves all model components to the device.
- output: None
- exception: None

11.4.5 Local Functions

None

12 MIS of Inference Module

12.1 Module

Test

12.2 Uses

Model Module ([11](#)), Output Visualization Module ([10](#))

12.3 Syntax

12.3.1 Exported Constants

None

12.3.2 Exported Access Programs

Name	In	Out	Exceptions
test	model: GnnNets, DataLoader	dataloader: -	RuntimeError

12.4 Semantics

12.4.1 State Variables

None

12.4.2 Environment Variables

None

12.4.3 Assumptions

The model has been trained and its best checkpoint has been loaded.

12.4.4 Access Routine Semantics

test(model, dataloader):

- transition: Evaluates the trained model on the test set. Computes loss and accuracy, and uses the Output Visualization Module to log results.
- output: None
- exception:
 - RuntimeError: if inference fails due to an invalid model state or shape mismatch

12.4.5 Local Functions

None

13 MIS of Explanation Module

13.1 Module

Explanation

13.2 Uses

Configuration Module (6)

13.3 Syntax

13.3.1 Exported Constants

None

13.3.2 Exported Access Programs

Name	In	Out	Exceptions
get_explanation	data: Data, GnnNets, Tensor	gnnNet: coalition: list[int], P: prototype: \mathbb{R} , embedding: Tensor	-

13.4 Semantics

13.4.1 State Variables

None

13.4.2 Environment Variables

None

13.4.3 Assumptions

None

13.4.4 Access Routine Semantics

get_explanation(data, gnnNet, prototype):

- transition: None
- output:
 - coalition: list of node indices forming the explanation.
 - P: float score indicating similarity to the prototype.
 - embedding: matrix of floats representing the masked subgraph embedding.
- exception: None

13.4.5 Local Functions

MCTSNode(coalition: list[int], data: Data, ori_graph: networkx.Graph, c_puct: \mathbb{R} , W: \mathbb{R} , N: \mathbb{R} , P: \mathbb{R}) \rightarrow MCTSNode

- transition: None
- output: A node object representing a state in the search tree.
- exception: None

mcts_rollout(tree_node: MCTSNode, state_map: dict, data: Data, graph: networkx.Graph, score_func: Customized Function) $\rightarrow \mathbb{R}$

- transition: None
- output: Scalar value representing the reward from this rollout.
- exception: None

child_scores(score_func: Customized Function, children: list[MCTSNode]) \rightarrow list[\mathbb{R}]

- transition: None
- output: List of float scores, one for each child.
- exception: None

prot_score(coalition: list[int], data: Data, gnnNet: GnnNets, prototype: Tensor)
 $\rightarrow \mathbb{R}$

- transition: None
- output: A float similarity score (higher = more aligned with prototype).
- exception: None

14 MIS of PyTorch Module

14.1 Module

Torch

14.2 Uses

None

14.3 Syntax

14.3.1 Exported Constants

None

14.3.2 Exported Access Programs

Name	In	Out	Exceptions
Tensor	shape: list[int], dtype: str	Tensor	-
cross_entropy	logits: Tensor, labels: Tensor	Tensor	-
Adam	parameters: iterable, lr: \mathbb{R}	Optimizer	-

14.4 Semantics

14.4.1 State Variables

None

14.4.2 Environment Variables

None

14.4.3 Assumptions

None

14.4.4 Access Routine Semantics

Tensor(shape, dtype):

- output: Returns a tensor initialized with zeros of the given **shape** and **dtype**.

cross_entropy(logits, labels):

- output: Computes the cross-entropy loss between **logits** and **labels**.

Adam(parameters, lr):

- output: Returns an Adam optimizer configured with the given **parameters** and learning rate **lr**.

14.4.5 Local Functions

None

15 MIS of PyTorch Geometric Module

15.1 Module

PyG

15.2 Uses

Torch

15.3 Syntax

15.3.1 Exported Constants

None

15.3.2 Exported Access Programs

Name	In	Out	Exceptions
Data	x: Tensor, edge_index: Tensor	Data	-
MoleculeNet	root: str, name: str	Dataset	FileNotFoundError
DataLoader	dataset: Dataset, batch_size: N	DataLoader	-
to_networkx	data: Data	networkx.Graph	-

15.4 Semantics

15.4.1 State Variables

None

15.4.2 Environment Variables

None

15.4.3 Assumptions

None

15.4.4 Access Routine Semantics

Data(x, edge_index):

- output: Constructs and returns a PyG graph object using **x** as node features and **edge_index** as edge indices.

MoleculeNet(root, name):

- output: Loads the dataset specified by **name** from directory **root** and returns a **Dataset** object.
- exception: **FileNotFoundError** if **root** does not exist.

DataLoader(dataset, batch_size):

- output: Returns a **DataLoader** that batches data from the given **dataset** with batch size **batch_size**.

to_networkx(data):

- output: Converts the input PyG **data** object into a NetworkX graph.

15.4.5 Local Functions

None

16 MIS of GUI Module

16.1 Module

Matplotlib

16.2 Uses

None

16.3 Syntax

16.3.1 Exported Constants

None

16.3.2 Exported Access Programs

Name	In	Out	Exceptions
axis	axis_choice: str	-	-
title	title_sentence: str	-	-
save_fig	figname: str	-	FileNotFoundError
close	choice: str	-	-

16.4 Semantics

16.4.1 State Variables

None

16.4.2 Environment Variables

- **figure_path:** **str** — Path where the current figure will be saved.
- **axis_visible:** **bool** — Whether axes are displayed in the active figure.
- **figure_title:** **str** — Title of the current figure.
- **figure_open:** **bool** — Whether there are any open figures.

16.4.3 Assumptions

None.

16.4.4 Access Routine Semantics

axis(axis_choice):

- **transition:** If `axis_choice == 'off'`, sets `axis_visible := False` and disables axes using `plt.axis('off')`. Otherwise sets `axis_visible := True`.

title(title_sentence):

- transition: Sets `figure_title := title_sentence` and updates the title of the current figure using `plt.title()`.

save_fig(filename):

- transition: Sets `figure_path := filename` and saves the current figure to the specified path using `plt.savefig(filename)`.
- exception: `FileNotFoundError` if `filename` refers to a non-existent directory.

close(choice):

- transition: Closes all active figure windows using `plt.close(choice)` and sets `figure_open := False`.

16.4.5 Local Functions

None

References

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