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 reimplementation 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 $(c_1 \Rightarrow r_1|c_2 \Rightarrow r_2|...|c_n \Rightarrow r_n)$.

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	bool	Boolean value: either True or False
string	str	A sequence of Unicode characters
tensor	Tensor	A multi-dimensional array object from PyTorch
graph	Data	A graph object from PyTorch Geometric, with node and edge attributes
dataset	Dataset	A collection of graph objects for training or evaluation
dataloader	DataLoader	A PyTorch Geometric data loader for batching graph data
dictionary	$\texttt{dict}[\texttt{K} \rightarrow \texttt{V}]$	A mapping from keys of type K to values of type V
list	list[T]	A sequence of elements of type T
function	Customized Function	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
Software Decision Module	Output Visualization 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

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.

$ModelParser(model_name: str, hidden_dim: \mathbb{N}, num_prototypes: \mathbb{N}) \rightarrow Model-Parser$

• 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	-	<pre>tuple[Dataset, int,</pre>	FileNotFoundError,
		int, dict[str $ ightarrow$	${\tt ValueError},$
		DataLoader]]	${\tt NotImplementedError}$

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

 $_get_dataset(dataset_dir:\ str,\ dataset_name:\ str) \rightarrow Dataset$

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

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

• 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

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: -	None
	Dataset, dataloader: dict[str	
	ightarrow DataLoader], gnnNets:	
	$GnnNets$, $output_dim$: \mathbb{N} , cri -	
	terion: Customized Function,	
	$\operatorname{ckpt_dir}$: str	

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) \rightarrow dict[str \rightarrow 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".

 $log_dataset_stats(dataset: Dataset) \rightarrow None$

- 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}) \to 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

output Utils

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: str	ExpPlot instance	-
draw	graph: networkx.Graph, nodelist: list[int], figname: str, kwargs: dict	-	NotImplementedError
$append_record$	info: str	-	${ t FileNotFoundError}$
save_best	ckpt_dir: str, epoch: \mathbb{N} , gnnNets: GnnNets, model_name: str, eval_acc: \mathbb{R} , is_best: bool	-	_

10.4 Semantics

10.4.1 State Variables

None

10.4.2 Environment Variables

- \bullet 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

11 MIS of Model Module

11.1 Module

GnnNets

11.2 Uses

PyTorch Module (14), Output Visualization Module (10)

11.3 Syntax

11.3.1 Exported Constants

None

11.3.2 Exported Access Programs

Name	In		Out	Exceptions
setup_model	input_dim: \mathbb{N} , output_dim:	$\mathbb{N},$	$\operatorname{tuple}[{ t GnnNets},$	-
	model_args: dict		Customized	
			Function]	

11.4 Semantics

11.4.1 State Variables

None

11.4.2 Environment Variables

- device: str Target device (e.g., 'cuda' or 'cpu'), used to move the model after initialization.
- log_file: str Path to the log file used in append_record().

11.4.3 Access Routine Semantics

setup_model(input_dim, output_dim, model_args):

- transition: Instantiates a GNN model with specified input_dim, output_dim, and model_args. Moves the model to device. Writes the model name to log_file using append_record().
- output: A tuple containing:

- ${\tt gnnNets:}$ a model instance supporting GNN forward/inference

- criterion: a cross-entropy loss function

• exception: None

11.4.4 Local Functions

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:	-	RuntimeError
	DataLoader				

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,	gnnNet:	coalition: list[int], P:	-
	${\tt GnnNets},$		prototype:	\mathbb{R} , embedding: Tensor	
	Tensor				

13.4 Semantics

13.4.1 State Variables

None

13.4.2 Environment Variables

None

13.4.3 Assumptions

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

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

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

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

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

 $\operatorname{child_scores}(\operatorname{score_func:}\ \operatorname{Customized}\ \operatorname{Function},\operatorname{children:}\ \operatorname{list}[\operatorname{\texttt{MCTSNode}}]) o \operatorname{\texttt{list}}[\mathbb{R}]$

- 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:	Data	-
	Tensor		
MoleculeNet	root: str, name: str	Dataset	${\tt FileNotFoundError}$
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

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

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(figname):

- transition: Sets figure_path := figname and saves the current figure to the specified path using plt.savefig(figname).
- exception: FileNotFoundError if figname 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

References

Carlo Ghezzi, Mehdi Jazayeri, and Dino Mandrioli. Fundamentals of Software Engineering. Prentice Hall, Upper Saddle River, NJ, USA, 2nd edition, 2003.

Daniel M. Hoffman and Paul A. Strooper. Software Design, Automated Testing, and Maintenance: A Practical Approach. International Thomson Computer Press, New York, NY, USA, 1995. URL http://citeseer.ist.psu.edu/428727.html.