# 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 <a href="https://github.com/Yuanqi-X/Re-ProtGNN/tree/main">https://github.com/Yuanqi-X/Re-ProtGNN/tree/main</a>.

# 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  $(c_1 \Rightarrow r_1 \mid c_2 \Rightarrow r_2 \mid \ldots \mid c_n \Rightarrow r_n)$ .

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

Data Type	Notation	Description
Integer	$\mathbb{Z}$	A whole number in $(-\infty, \infty)$
Natural Number	$\mathbb{N}$	A positive integer in $[1, \infty)$
Real Number	$\mathbb{R}$	A real value in $(-\infty, \infty)$
Boolean	bool	Logical value: True or False
Vector of dimension $d$	$\mathbb{R}^d$	A $d$ -dimensional real-valued vector
Matrix of size $n \times m$	$\mathbb{R}^{n \times m}$	A real-valued matrix with $n$ rows and $m$ columns
Index Vector	$\mathbb{N}^n$	A length- $n$ vector of natural number indices
List of type $T$	${ t list}[T]$	A finite sequence of values of type $T$
String	str	A sequence of characters
Dictionary	$\texttt{dict}[\texttt{K}  \rightarrow  \texttt{V}]$	A mapping from keys of type $K$ to values of type $V$
Tuple	$\texttt{tuple}[T_1, T_2, \ldots]$	An ordered, fixed-length collection of elements where each element can have a different type.
Function	Customized Function	A user-defined function or callable

Functions in Re-ProtGNN are defined using their argument and return types. For instance, a function  $f: \mathbb{N} \to \mathbb{R}$  takes a natural number and returns a real number. 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

#### 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: Dataset,	-	None
	dataloader: dict[str $\rightarrow$		
	DataLoader], gnnNets: GnnNets,		
	output_dim: $\mathbb{N}$ , criterion:		
	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 save_best	info: str ckpt_dir: str, epoch: N, gnnNets: GnnNets, model_name: str, eval_acc: R, is_best: bool	-	FileNotFoundError

# 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

# 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: $\mathbb{N}$ ,	$\operatorname{tuple}[{\tt GnnNets},$	-
	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

evaluation.inference

#### 12.2 Uses

Model Module (11), Output Visualization Module (10), PyTorch Module (14)

# 12.3 Syntax

### 12.3.1 Exported Constants

None

#### 12.3.2 Exported Access Programs

Name	In	Out	Exceptions
$run\_inference$	test_dataloader: DataLoader,	$\operatorname{tuple}[\operatorname{dict}[\operatorname{str} \to \mathbb{R}], \mathbb{R}^{n \times c},$	None
	model: GnnNets, criterion:	$\mathbb{N}^n]$	
	Customized Function		

### 12.4 Semantics

#### 12.4.1 State Variables

None

#### 12.4.2 Environment Variables

• log\_file: str — Path to the log file where final test performance is recorded via append\_record().

#### 12.4.3 Access Routine Semantics

run\_inference(test\_dataloader, model, criterion):

- transition:
  - Appends the final test loss and accuracy to the log\_file using append\_record().
- output:
  - Returns a tuple (test\_state, all\_probs, all\_preds):

- \* test\_state: dict[str  $\to$   $\mathbb{R}$ ] containing keys 'loss' and 'acc'.
- \* all\_probs:  $\mathbb{R}^{n \times c}$  class probability matrix for n test samples and c classes, obtained by passing the data in test\_dataloader into model.
- \* all\_preds:  $\mathbb{N}^n$  vector of predicted class labels.
- exception: None

# 12.4.4 Local Functions

# 13 MIS of Explanation Module

# 13.1 Module

Explanation

# 13.2 Uses

Configuration Module (6), Output Visualization Module (11), PyTorch Geometric Module (15), PyTorch Module (14)

# 13.3 Syntax

# 13.3.1 Exported Constants

None

# 13.3.2 Exported Access Programs

Name	In	Out	Exceptions
get_explanation	data: Data, gnnNet:	$tuple[list[\mathbb{N}], \mathbb{R}, \mathbb{R}^d]$	None
	GnnNets, prototype: $\mathbb{R}^d$		
$\exp_{\text{-}}$ visualize	dataset: Dataset,	-	FileNotFoundError
	dataloader: dict[str $\rightarrow$		
	DataLoader], gnnNets:		
	${\tt GnnNets},  {\tt output\_dim} \colon  \mathbb{N}$		

# 13.4 Semantics

#### 13.4.1 State Variables

None

#### 13.4.2 Environment Variables

• save\_dir: str — Filesystem path for saving explanation plot images.

# 13.4.3 Access Routine Semantics

get\_explanation(data, gnnNet, prototype):

- transition: None
- output:

- Uses data, gnnNet, and prototype to invoke a Monte Carlo Tree Search (MCTS) via the local function \_mcts().
- MCTS explores subgraph coalitions within the graph structure of data and evaluates their similarity to the provided prototype using \_prot\_similarity\_scores().
- Returns:
  - \* coalition: list of node indices selected as the best explanation subgraph.
  - \* P: similarity score  $(\in \mathbb{R})$  between the selected subgraph and the prototype.
  - \* embedding: a masked subgraph embedding  $\in \mathbb{R}^{1 \times d}$  extracted by applying gnnNet to the subgraph induced by coalition.
- exception: None

## exp\_visualize(dataset, dataloader, gnnNets, output\_dim):

- transition:
  - Computes  $K := \text{output\_dim} \times \text{model\_args.num\_prototypes\_per\_class}$  random prototype vectors of dimension d, fixed for visualization.
  - Samples 16 graphs from dataloader ['train'].
  - For each selected graph and for each of the first 10 prototype vectors, calls get\_explanation(data, gnnNets, prototype) to compute explanations.
  - Saves the explanations to image files in the environment variable save\_dir.
  - Sets save\_dir := './results/plots/<dataset>\_<model>\_' using values from data\_args.dataset\_name and model\_args.model\_name, imported from Configuration Module (6).
- output: None
- exception:
  - FileNotFoundError: if save\_dir cannot be created due to missing parent directories, invalid paths, or insufficient filesystem permissions.

#### 13.4.4 Local Functions

 $\_mcts(data: ext{Data}, ext{gnnNet}: ext{GnnNets}, ext{prototype}: \mathbb{R}^d) o ext{tuple}[ ext{list}[\mathbb{N}], \, \mathbb{R}, \, \mathbb{R}^d]$ 

• output: Applies a multi-step rollout procedure to search for the optimal node coalition in data, maximizing similarity to the reference vector prototype  $\in \mathbb{R}^d$ . Returns the selected node indices, similarity score, and final embedding.

 $\_$ prot $\_$ similarity $\_$ scores(coalition: list $[\mathbb{N}]$ , data: Data, gnnNet: GnnNets, prototype:  $\mathbb{R}^d) \to \mathbb{R}$ 

• output: Computes a scalar similarity score  $\in \mathbb{R}$  between the subgraph embedding of nodes in coalition (produced by gnnNet) and the given reference vector prototype  $\in \mathbb{R}^d$ , based on squared Euclidean distance.

# 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: list[Tensor],	Optimizer	-
	$\operatorname{lr} : \mathbb{R}$		

# 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 1r.

# 14.4.5 Local Functions

# 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	<del>-</del>
	Tensor		
MoleculeNet	root: str, name: str	Dataset	${ t FileNotFoundError}$
DataLoader	dataset: Dataset, batch_size:	DataLoader	-
	$\mathbb{N}$		
$to\_networkx$	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

# 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

$\mathbf{Name}$	${f In}$	$\mathbf{Out}$	Exceptions
axis	axis_choice: str	-	<del>-</del>
title	title_sentence: str	-	-
$save\_fig$	figname: str	-	${ t FileNotFoundError}$
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.