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

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

OutputVisualize

## 10.2 Uses

Hardware-Hiding Module

# 10.3 Syntax

## 10.3.1 Exported Constants

None

## 10.3.2 Exported Access Programs

Name	In	Out	Exceptions
PlotUtils	dataset_name: str	PlotUtils instance	
plot	graph: Data, nodelist: list[int], figname: str,	-	-
append_record	kwargs: dict info: str	-	FileNotFoundError

## 10.4 Semantics

## 10.4.1 State Variables

None

## 10.4.2 Environment Variables

• Filesystem: the file system for saving log files and outputting explanation images.

## 10.4.3 Assumptions

None

#### 10.4.4 Access Routine Semantics

## $PlotUtils(dataset\_name)$ :

• transition: None

- output: A PlotUtils object with methods for graph visualization.
- exception: None

## plot(graph, nodelist, figname, kwargs):

- transition: Generates explanation images and saves them to the specified path.
- output: None
- exception: None

## append\_record(info):

- transition: Appends the info string to the log file located in the given log directory.
- output: None
- exception:
  - FileNotFoundError: if the directory does not exist.

#### 10.4.5 Local Functions

None

# 11 MIS of Model Module

## 11.1 Module

GnnNets

#### 11.2 Uses

None

# 11.3 Syntax

## 11.3.1 Exported Constants

## 11.3.2 Exported Access Programs

Name	In	Out	Exceptions
GnnNets	input_dim: int,	GnnNets	NotImplementedError
	output_dim: int,		
	model_args: dict		
forward	data: Data, prot-	logits: Tensor,	-
	gnn_plus: bool, similar-	prob: Tensor, emb1:	
	ity: Tensor	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:	${\tt GnnNets},$	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

# 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

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

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

- 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:	DataLoader	-
	$\mathbb{N}$		
$to\_networkx$	data: Data	${\tt networkx.Graph}$	-

## 15.4 Semantics

## 15.4.1 State Variables

None

## 15.4.2 Environment Variables

None

## 15.4.3 Assumptions

#### 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	-	<del>-</del>
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

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