# Module Interface Specification for Re-ProtGNN

Yuanqi Xue

April 18, 2025

# 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.

# Contents

1	Rev	vision History					
2	Symbols, Abbreviations and Acronyms						
3	Introduction						
4	Not	tation					
5	Mo	dule Decomposition					
6	MIS	S of Configuration Module					
	6.1	Module					
	6.2	Uses					
	6.3	Syntax					
		6.3.1 Exported Constants					
		6.3.2 Exported Access Programs					
	6.4	Semantics					
		6.4.1 State Variables					
		6.4.2 Environment Variables					
		6.4.3 Access Routine Semantics					
		6.4.4 Local Functions					
7	MIS	S of Input Format Module					
	7.1	Module					
	7.2	Uses					
	7.3	Syntax					
		7.3.1 Exported Constants					
		7.3.2 Exported Access Programs					
	7.4	Semantics					
		7.4.1 State Variables					
		7.4.2 Environment Variables					
		7.4.3 Access Routine Semantics					
		7.4.4 Local Functions					
8	MIS	S of Control Module					
	8.1	Module					
	8.2	Uses					
	8.3	Syntax					
		8.3.1 Exported Constants					
		8.3.2 Exported Access Programs					
	8.4	Semantics					
		8.4.1 State Variables					

		8.4.2	Environment Va	riables													7
		8.4.3	Access Routine	Semantics													7
		8.4.4	Local Functions													•	8
9	MIS	of Tra	nining Module														9
	9.1	Module	e														9
	9.2	Uses .															9
	9.3	Syntax															9
		9.3.1	Exported Const	ants													9
		9.3.2	Exported Access	s Programs													9
	9.4	Semant	ics														9
		9.4.1	State Variables														9
		9.4.2	Environment Va	riables													9
		9.4.3	Access Routine	Semantics													9
		9.4.4	Local Functions														10
<b>10</b>	MIS	of Ou	tput Visualiza	tion Modi	ıle	)											11
			· 9														11
	10.2	Uses .															11
																	11
		10.3.1	Exported Const	ants													11
			Exported Access														11
	10.4																11
			State Variables														11
			Environment Va														11
			Access Routine														12
			Local Functions														12
11	МТ	of Mo	del Module														13
			e														13
																	13
																	13
	11.5		Exported Const.														13
			Exported Consta Exported Access														13
	11 /		cics														13
	11.4		State Variables														13
			Environment Va														13 13
																	13
			Access Routine Local Functions														13 14
		11.4.4	Local Functions		•		•	 •	 •	•	 •	 •	•	 •	•	•	14
<b>12</b>			erence Module														15
			9														15
	12.2	Uses .															15

12.	3 Syntax	15
	12.3.1 Exported Constants	15
4.0	12.3.2 Exported Access Programs	15
12.	4 Semantics	15
	12.4.1 State Variables	15
	12.4.2 Environment Variables	15
	12.4.3 Access Routine Semantics	15
	12.4.4 Local Functions	16
<b>13</b> M	IS of Explanation Module	17
	1 Module	17
13.	2 Uses	17
13.	3 Syntax	17
	13.3.1 Exported Constants	17
	13.3.2 Exported Access Programs	17
13.	4 Semantics	17
	13.4.1 State Variables	17
	13.4.2 Environment Variables	17
	13.4.3 Access Routine Semantics	17
	13.4.4 Local Functions	18
1 4 N.T	IS of PyTorch Module	19
	1 Module	19
	2 Uses	19 19
	3 Syntax	19
14.	14.3.1 Exported Constants	19
	14.3.2 Exported Access Programs	19
1/1	4 Semantics	19
14.	14.4.1 State Variables	19
	14.4.2 Environment Variables	19
	14.4.3 Assumptions	19
	14.4.4 Access Routine Semantics	19
	14.4.5 Local Functions	20
	11.1.0 Booki Fullouidillo	20
15 M	IS of PyTorch Geometric Module	<b>20</b>
15.	1 Module	20
15.	2 Uses	20
15.	3 Syntax	20
	15.3.1 Exported Constants	20
	15.3.2 Exported Access Programs	20
15.	4 Semantics	20
	15.4.1 State Variables	20
	15.4.2 Environment Variables	20

		15.4.3	Assumptions	21
		15.4.4	Access Routine Semantics	21
		15.4.5	Local Functions	21
<b>16</b>	MIS	of GU	UI Module	21
	16.1	Modul	e	21
	16.2	Uses .		21
	16.3	Syntax	C	21
		16.3.1	Exported Constants	21
		16.3.2	Exported Access Programs	22
	16.4	Seman	tics	22
		16.4.1	State Variables	22
		16.4.2	Environment Variables	$2\overline{2}$
		16.4.3	Assumptions	$2\overline{2}$
		16.4.4	Access Routine Semantics	22
		16 4 5	Local Functions	2.5

# 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

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

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	model: GnnNets, criterion:	tuple[dict[str $\to \mathbb{R}$ ], $\mathbb{R}^{n \times c}$ , $\mathbb{N}^n$ ]	None
	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, dat-	_	${\tt FileNotFoundError}$
	aloader: 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: 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	-	<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

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.