Lab 03: Community Detection

CSC14114 Big Data Application 19KHMT

nhom 1

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Task	Completed
From raw data to graph	Percent: 100
Betweenness calculation	Percent: 100
Betweenness calculation comparison with networkx	Percent: 100
Community detection	Percent: 100
Community detection comparision with networkx	Percent: 100

1 Lab 03: Community Detection

1.1 The one do this version is Nguyen Ngoc Phuoc - 19127519

1.2 From raw data to graphs

1.2.1 Graph

You may find immplementation of this in graph. py file. This is a simple abstraction of this class. This is the data structure that will store all data point through out this lab.

```
class Graph:
    # add and edge to the graph
    def add_edge(self, u: str, v: str):
        pass
    # remove and edge from the graph
    def remove_edge(self, u: str, v: str):
        pass
    # get connected components in the graph
    def get_communities(self) -> list[tuple[str]]:
        pass
    # return copy of the graph
    def copy(self):
        pass
    # number of nodes
    def count_nodes(self) -> int:
        return len(self.nodes)
    # number of edges
    def count_edges(self) -> int:
    # whether node_u and node_v is connected by and edge
    def connected(self, node_u, node_v) -> bool:
```

pass

1.2.2 GraphBuilder

This is where I encounter a problem. The graph construction is actually performed on 2 difference objects:

- self-implemmented graph
- networkx graph

To actually keep the business logic compact, not duplicate, I use builder design pattern to do the graph construction phases. You may find this in graph_builder.py.

This is the interface of all builder classes.

```
class GraphBuilder:
    # add and edge to the graph
    @abstractmethod
    def add_edge(self, node_u: str, node_v: str):
        raise NotImplementedError()
    # reset the builder result instance
    @abstractmethod
    def reset(self):
        raise NotImplementedError()

There are 2 implementation for this interface, which is:

class SelfImplementedGraphBuilder(GraphBuilder):
    pass

and:

class NetworkXGraphBuilder(GraphBuilder):
    pass
```

1.2.3 GraphDirector

This is a Director class that would perform on GraphBuilder for graph construction, business logic for graph construction is defined inside this class.

```
class GraphBuilderDirector:
    @staticmethod
    def build_graph_from_csv(file_name: str, graph_builder: GraphBuilder, thresh
        pass
```

Then in the main.py file, you will see the way I use these classes to perform graph constructions.

```
# self implemented_graph
implemented_graph_builder = SelfImplementedGraphBuilder()
GraphBuilderDirector.build_graph_from_csv(data_file, implemented_graph_build
implemented_graph = implemented_graph_builder.build()

# networkx graph
networkx_graph_builder = NetworkXGraphBuilder()
GraphBuilderDirector.build_graph_from_csv(data_file, networkx_graph_builder,
networkx_graph = networkx_graph_builder.build()
```

There's the graph construction phase, the output is 2 graph:

- self-implemented graph
- networkx graph

1.3 Betweenness Calculation

Again, there would be 2 different implemmentation in this phase:

- self-implemented
- networkx

The reason for this is because I want to compare my calculation with networkx one.

1.3.1 Tree

For the betweenness calculation, I go with the most obvious approach: because in the slide, lectures, books use a Tree to do it, so I will do it like that too.

As I present classes, which you can find in tree.py.

```
class TreeNode:
    pass
class Tree:
    pass
```

As each tree is constructed from a graph, the between calculation is done for that node which is the root.

This mean for each graph, I will construct n tree with n is the number of nodes within the graph. The betweenness calculation for each tree (each node as root) is done in a recursive manner, with the intuition from the slide, from top-down requested (bottom-up calculation).

I'm quite proud of this actually. I think that this lab is both hard and interesting compare to the others 2 because of this.

1.3.2 BetweennessCalculator

This is an interface that would perform calculation logic and save it to files. It can also perform benchmark between 2 files for comparison, I use mse for this benchmark. You can find these code in betweenness_calculator.py

And there are 2 classes that implement this:

- class SelfImplementedBetweennessCalculator(BetweennessCalculator):
 - Use Tree data structure to perform edge betweenness fusion for all nodes in graph.
- class NetworkXBetweennessCalculator(BetweennessCalculator):
 - Just call library function.

I run the code in main.py, then save the result to file and perform benchmark. The algorithm output 2 txt file:

- output/edge_betweenness.txt: for self implemented
- output/edge_betweenness_lib.txt: for the library version

The file is almost identical, just with a few different in calculation at a very small fraction level, as the mse score output is:

```
betweenness mse: 1.7621061666021155e-28
```

You may find the code call these function in main.py

```
# betweenness self implemented
implemented_betweenness_calculator = SelfImplementedBetweennessCalculator(im
implemented_betweenness_calculator.calculate_betweenness(betweenness_out_fil)
# betweenness lib
lib_edge_betweenness_calculator = NetworkXBetweennessCalculator(networkx_gra)
lib_edge_betweenness_calculator.calculate_betweenness(betweenness_lib_out_fi)
# benchmark betweenness
BetweennessCalculator.mse_benchmark_betweenness(betweenness_out_file, betweenness_out_file, betweenness_out_file)
```

1.4 Community Detection

For the last part, you may find source code in community_detector.py.

With the same approach used, I present you the base classs:

- It can perform communities detection and save the result to files
- It can perfrom benchmark between 2 different files.

```
class CommunityDetector:
    def detect(self, file_name: str = None) -> list[tuple[str]]:
        pass

    @staticmethod
    def benchmark_communities(community_a_file: str, community_b_file: str):
        pass
```

As for the benchmark in this class, you may find it quite strict, as it actually read both file and compare the content with a single == operator. But nonetheless, it works;). My implementation and library yeild exactly the same result.

There're 2 classes that implement this base:

- class SelfImplementedCommunityDetector(CommunityDetector):
 - Modularity is calculated by the algorithm presented in the assignment pdf.
 - Girvan newman is self-implemented (quite simple, just remove edges till there's new community).
- class NetworkxCommunityDetector(CommunityDetector):
 - Just use modularity and girvan-newman in the library

You may find the code run for this in main.py:

```
# detection self implemented
community_detector = SelfImplementedCommunityDetector(implemented_graph)
community_detector.detect(community_detection_out_file)

# detection networkx
community_detector_lib = NetworkxCommunityDetector(networkx_graph)
community_detector_lib.detect(community_detection_lib_out_file)

# benchmark communities
CommunityDetector.benchmark_communities(community_detection_out_file, community_detection_out_file, community_detection_out_file)
```

This output 2 files:

- output/communities.txt: for the one I implemented
- output/communities_lib.txt: for the library one.

And the benchmark result with == operator used between 2 files:

```
detected communities identical
```

1.5 Reflection

- This lab is very interesting, and hard too than the other 2 lab.
- But nonetheless it bring me satisfaction when completed
- I learn more about girvan-newman, community detection, feel proud because I just wrote some sotisphicated algorithm in about a day or two

1.6 References

- Slides
- Assignment PDF