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Large Graph Path Finder (LGPF)

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-

Abstract

The **Large Graph Path Finder (LGPF)** project benchmarks and evaluates multiple pathfinding algorithms on arbitrarily large graphs. It supports uninformed search (BFS, DFS, Depth-Limited), weighted search (Dijkstra, Bidirectional Dijkstra), informed search (Greedy Best-First, A*), and advanced heuristics including **ALT (A* with Landmarks and Triangle inequality)**. For graphs containing negative edge weights, the system falls back to **Bellman–Ford**.

Graphs are generated in JSON format and preprocessed to build fast lookup structures, detect uniform weights, and compute landmark tables for ALT on large graphs. During preprocessing, a **real-time timer and memory monitor** report progress directly to the terminal and log file, warning if runtime exceeds 60 seconds or memory usage exceeds 1 GB. Queries can then be executed from file or interactively, producing exact path costs and performance statistics

Algorithms & Heuristics

Used Algorithms & Heuristics

Pathfinding in large graphs requires balancing **accuracy**, **speed**, and **scalability**. In `pathfinder.py`, the algorithm is chosen automatically based on graph properties:

- If landmarks exist → **A* + ALT [USING]**
- If negative weights exist → **Bellman–Ford [USING]**
- If graph is very large → **A* (Zero Heuristic) [USING]**
- If graph is unweighted → **BFS [USING]**
- Otherwise → **Dijkstra [USING]**

This adaptive design ensures that queries are solved efficiently and correctly without the user having to manually choose the algorithm.

Uninformed Search

- **BFS (Breadth-First Search) [USING]**

- *Use* : Unweighted graphs.
 - *Pros* : Finds shortest path in edge count.
 - *Cons* : High memory, ignores weights.
 - **DFS (Depth-First Search)**
 - *Use* : Reachability checks.
 - *Pros* : Simple, low memory.
 - *Cons* : Not optimal, may miss goal in deep graphs.
 - **Depth-Limited DFS**
 - *Use* : Same as DFS but avoids infinite paths.
 - *Pros* : Bound search depth.
 - *Cons* : Still not guaranteed optimal.
-

Weighted Graph Algorithms

- **Dijkstra [USING]**
 - *Use* : Weighted graphs, no negative edges.
 - *Pros* : Guarantees shortest path.
 - *Cons* : Slower for very large graphs.
 - **Bidirectional BFS**
 - *Use* : Unweighted graphs, source–goal pairs.
 - *Pros* : Faster in practice than BFS.
 - *Cons* : Only works on unweighted graphs.
 - **Bidirectional Dijkstra**
 - *Use* : Large weighted graphs.
 - *Pros* : Can halve search space.
 - *Cons* : Not fully optimized in current code.
 - **Bellman–Ford [USING]**
 - *Use* : Graphs with negative edge weights.
 - *Pros* : Correct with negatives.
 - *Cons* : Very slow ($O(VE)$).
-

Informed Search

- **Greedy Best-First Search**

- *Use* : Uses heuristic only (H).
- *Pros* : Fast, goal-directed.
- *Cons* : Not optimal.

- **A* [USING]**

- *Use* : Weighted graphs with heuristics.
 - *Pros* : Optimal if heuristic is admissible.
 - *Cons* : Needs good heuristic, otherwise same as Dijkstra.
-

Heuristics

- **Zero Heuristic [USING]**

- *Pros* : Always admissible, safe.
- *Cons* : Reduces A* to Dijkstra.

- **Manhattan**

- *Use* : Grid-like graphs.
- *Pros* : Simple, consistent.
- *Cons* : Only makes sense on grids.

- **Euclidean**

- *Use* : Spatial graphs.
- *Pros* : Geometrically intuitive.
- *Cons* : Less useful on abstract graphs.

- **Chebyshev / Octile**

- *Use* : Grid movement with diagonals.
- *Pros* : Models 8-directional movement.
- *Cons* : Not needed in non-grid graphs.

- **Haversine**

- *Use* : Geographic (lat/lon).
- *Pros* : Real-world Earth distance.
- *Cons* : Costly, only relevant to maps.

- **ALT (A* + Landmarks + Triangle inequality) [USING]**

- *Use* : Very large graphs.
- *Pros* : Strong admissible heuristic, massive speedups for queries.

- *Cons* : Preprocessing is expensive (runs many Dijkstras)

Uninformed Search

- **BFS (Breadth-First Search)**
 - *Use* : Unweighted graphs.
 - *Pros* : Simple, guarantees shortest path in terms of edges.
 - *Cons* : Inefficient for large weighted graphs.
 - **DFS (Depth-First Search, Depth-Limited)**
 - *Use* : Exploratory only, not used in weighted shortest-path.
 - *Pros* : Low memory footprint, useful for reachability.
 - *Cons* : Does not guarantee shortest path.
-

Weighted Algorithms

- **Dijkstra**
 - *Use* : Weighted graphs with non-negative edges.
 - *Pros* : Guarantees optimal paths, robust for medium graphs.
 - *Cons* : Slower for very large graphs; expands many nodes.
 - **Bidirectional Dijkstra**
 - *Use* : Large graphs with a clear source-target pair.
 - *Pros* : Faster in practice by searching from both ends.
 - *Cons* : More complex implementation; not always optimal in sparse graphs.
 - **Bellman–Ford**
 - *Use* : Graphs with negative edge weights.
 - *Pros* : Handles negative edges safely, guarantees correctness.
 - *Cons* : Very slow ($O(VE)$), not suitable for large graphs unless required.
-

Informed Algorithms

- **Greedy Best-First Search**
 - *Use* : Heuristic-based exploration.
 - *Pros* : Very fast, explores promising nodes first.
 - *Cons* : Not guaranteed to find optimal path.

- **A ***
 - *Use* : Weighted graphs with admissible heuristics.
 - *Pros* : Guarantees optimal paths, faster than Dijkstra if heuristic is strong.
 - *Cons* : Requires a good heuristic; fallback is same as Dijkstra if heuristic=0.
-

Heuristics

- **Zero Heuristic**
 - *Pros* : Safe, guarantees correctness.
 - *Cons* : Reduces A* to Dijkstra.
- ****ALT (A* with Landmarks and Triangle inequality)* ***
 - *Use* : Very large graphs ($\geq 10k$ nodes).
 - *Pros* : Strong admissible heuristic, greatly reduces search space; queries become very fast after preprocessing.
 - *Cons* : Preprocessing cost is high (multiple Dijkstra runs for each landmark).

Installation

No installation step is required — just ensure prerequisites are installed see Prerequisites and tested Environment & Versions all other OS is risk of the user.

Quick Start

note: this program is designed to run on Linux/Linux VMs OS, other OS is not fully supported, certain commandline is gear thorse the OS mentioned see Prerequisites

1. See Usage and its subsection 3. Required files to run then 4. Run the Path Finder *main program*

Prerequisites

The program runs on **Python 3.13+** and does not require installation beyond having the correct dependencies available.

- **Python** : ≥ 3.13 (tested on 3.13.5)
- **pip** : ≥ 25.0 (tested with 25.2)

- **Operating Systems Tested :**
 - Windows 11 Pro for Workstations
 - Ubuntu 25.04 (inside VirtualBox)
- **Required Libraries :**
 - `networkx` (graph handling)
 - `matplotlib` (visualization)
 - `psutil` (timer & memory monitor)
 - `pytest` (testing framework)

Optional but supported:

- `pandas / scipy` for extended analysis
 - `graphviz` for extra visualization format
-

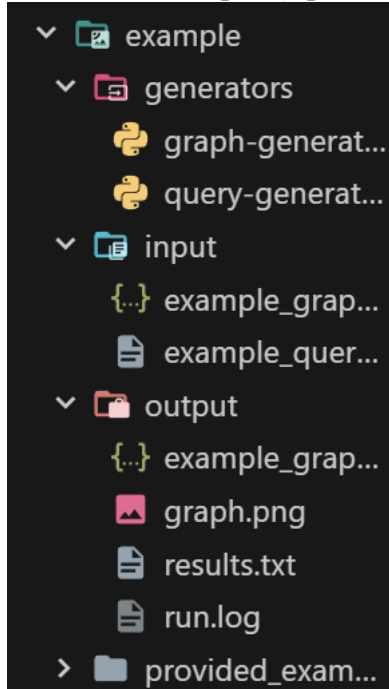
Environment & Versions

- **Python Version :** 3.13.5
 - **pip Version :** 25.2
 - **OS :** Windows 11 Pro for Workstations; Ubuntu 25.04 (VirtualBox)
 - **Key Packages :**
 - `networkx` 3.5
 - `matplotlib` 3.10.6
 - `psutil` 7.1.0
 - `pytest` 8.4.2
 - **Memory Tested :** Preprocessing up to ~1 GB RAM (monitored with `psutil`)
 - **Hardware Used :** Intel Core i9 13980HX(24/32), 96 GB RAM
-

Usage

Run Main Program

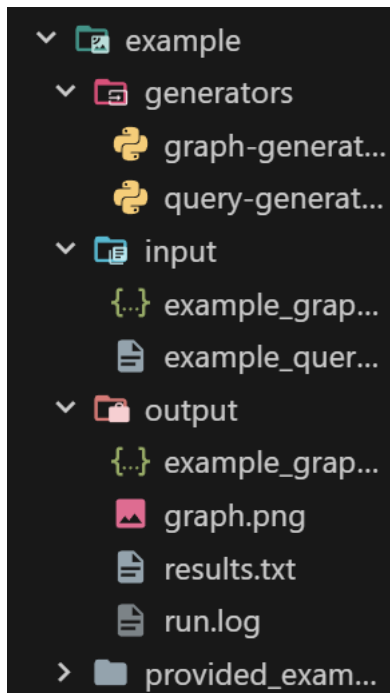
1. Generate Graphs (Optional, if and only if sub section 3. Required files to run are satisfied)



Use the provided generator to create test graphs in JSON:

```
# Example: 20k vertices, 60k edges
python3 example/generators/graph-generator.py \
  --num_vertices 20000 \
  --num_edges 60000 \
  --min_weight 1 \
  --max_weight 100 \
  --json_out input/large_graph.json \
  --img_out input/large_graph.png \
  --seed 42
```

2. Generate Queries Where to locate generator Python files under program root *example/generators*:



Create source–target pairs to test algorithms:

```
python3 example/generators/query-generator.py \  
  --graph input/medium_graph.json  
  --out input/medium_queries.txt  
  --count 500  
  --seed 0
```

or generate nodes that are all reachable *note: good luck on tring to perform 10K node on personal machines. generator will provide node pairs:*

```
python3 example/generators/query-generator.py \  
  --graph input/medium_graph.json  
  --out input/medium_reach.txt  
  --count 500  
  --seed 0  
  --reachable-only
```

3. Required files to run *note: exsample files can be obtained in exsample/input or generate see this subsection 1 and 2*

1. <graph>.json file required and insert it in input/
2. query.txt file optional and insert it in input/

4. Run the Path Finder main program Launch the main program *note: this program is designed to run on Linux/Linux VMs OS, other OS is not fully supported, certain commandline is gear thorse the OS mentioned see Prerequisites:*

```
python3 -m core.main
```

or use quick script *note: user must be under root directory.*

```
./run.sh
```

```
delinquentNULL → Documents\CYBR439AI\LargeGraphPathFinder → (dev) 23ms 11:12 PM
null >> ./run.sh
/home/null/.local/lib/python3.13/site-packages/numpy/_core/getlimits.py:551: UserWarning: Signature b'\x00\x00\x00\x00\x00\x00\x00\x00' for <class 'numpy.longdouble'> does not match any known type: falling back to type object
This warnings indicates broken support for the dtype!
  machar = _get_machar(dtype)
[INFO] Large Graph Path Finder starting...
[INFO] JSON files in input/: alt_graph.json, big_graph.json, example_graph.json
Enter graph filename (inside input/, *.json): big_graph.json
[TIMER] 0s elapsed | Memory: 64.32 MB
[TIMER] 1s elapsed | Memory: 526.36 MB
[TIMER] 2s elapsed | Memory: 601.80 MB
[TIMER] 3s elapsed | Memory: 617.30 MB
[TIMER] 4s elapsed | Memory: 617.30 MB
[TIMER] 5s elapsed | Memory: 632.30 MB
[TIMER] 6s elapsed | Memory: 632.30 MB
```

You'll be prompted to select:

- A graph file (e.g. input/large_graph.json)
- A query file after preprocessing is complete(e.g. input/large_queries.txt)

```
[INFO] Graph loaded with 250000 nodes.
[TIMER] Preprocessing finished at 34s, stopping timer.
[INFO] Node listing saved: output/big_graph_nodes.json
[INFO] Files in input/: .gitignore, alt_graph.json, alt_queries.txt, big_graph.json, example_graph.json,
achable.txt
Enter query filename (inside input/), or press Enter to input pairs interactively: example_query.txt
[INFO] Algorithm selected: AStar_ALT
0 2 | 157.0 | 0.858080 | 0-[24]->221512-[12]->191623-[1]->169695-[5]->174894-[2]->190672-[2]->136616-[11]->190624-[23]->230458-[4]->93533-[1]->231143-[26]->207399-[2]->82118-[15]->238825-[2]->2
1 2 | 176.0 | 2.441432 | 1-[12]->28927-[3]->52273-[14]->103654-[21]->3930-[8]->12358-[53]->54378-[50]->2
0 1 | 159.0 | 0.680813 | 0-[24]->221512-[26]->193227-[5]->140531-[5]->243685-[24]->102249-[2]->241782-[1]->11728-[10]->52273-[3]->28927-[12]->1
2 0 | 157.0 | 0.287673 | 2-[2]->238825-[15]->82118-[2]->207399-[26]->231143-[1]->93533-[4]->230458-[23]->99519-[11]->136616-[2]->190672-[2]->174894-[5]->169695-[1]->191623-[12]->221512-[24]->0
2 1 | 176.0 | 1.143685 | 2-[15]->205169-[50]->54378-[53]->12358-[8]->3930-[21]->103654-[14]->52273-[3]->
1 0 | 159.0 | 0.551669 | 1-[12]->28927-[3]->52273-[10]->11728-[11]->248450-[29]->164615-[7]->21826-[1]->140531-[5]->193227-[26]->221512-[24]->0
[INFO] Results saved: output/results.txt
[INFO] Run complete. Exiting.
```

5. Real-Time Timer & Memory Monitor During preprocessing:

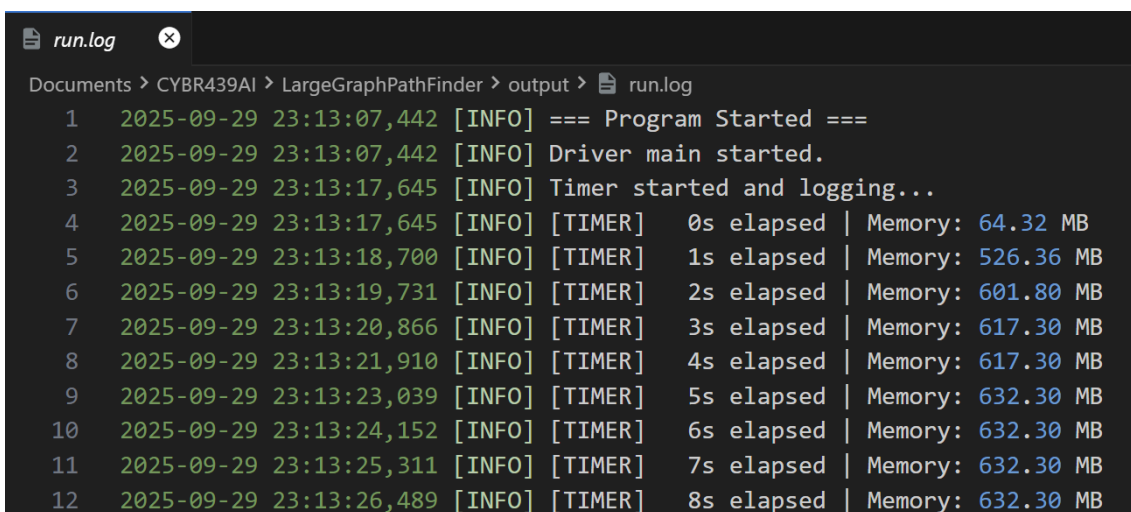
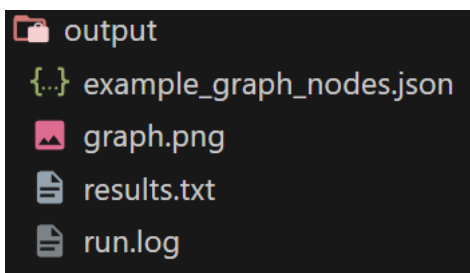
- timer counts up in seconds
- memory usage is displayed
- warnings appear if:
 - Runtime exceeds 60s
 - Memory usage exceeds 1024 MB

Example:

```
[TIMER]    0s elapsed | Memory: 64.31 MB
[TIMER]    1s elapsed | Memory: 531.61 MB
...
[TIMER] Preprocessing finished at 35s, stopping timer.
```

6. Results Output contains after a success or partial successful run:

- Query results are printed in terminal and saved to `output/results.txt`. Each line shows source, target, path cost, and runtime. If no path exists, cost will be inf.
- Query outputs are written to `output/results.txt`
- Program logs as `run.log`
- A visual graph if and only if node is no more than 1000 nodes.
- A json file showing all the processed nodes



```

run.log
Documents > CYBR439AI > LargeGraphPathFinder > output > run.log
38 2025-09-29 23:13:54,568 [INFO] Graph loaded with 250000 nodes. Memory now 754.67 MB
39 2025-09-29 23:13:55,195 [INFO] [TIMER] Preprocessing finished at 34s, stopping timer.
40 2025-09-29 23:13:57,159 [INFO] Graph too large to visualize (>100 nodes)
41 2025-09-29 23:13:57,350 [INFO] Node listing saved: output/big_graph_nodes.json
42 2025-09-29 23:13:57,351 [INFO] Preprocessing finished in 39.71s
43 2025-09-29 23:15:05,424 [INFO] Loaded 6 queries from input/example_query.txt
44 2025-09-29 23:15:05,424 [INFO] Algorithm selected by policy: AStar_ALT
45 2025-09-29 23:15:06,283 [INFO] 0 2 | 157.0 | 0.858080 | 0-[24]->221512-[12]->191623-[1]->1
46 2025-09-29 23:15:08,724 [INFO] 1 2 | 176.0 | 2.441432 | 1-[12]->28927-[3]->52273-[14]->103
47 2025-09-29 23:15:09,405 [INFO] 0 1 | 159.0 | 0.680813 | 0-[24]->221512-[26]->193227-[5]->1
48 2025-09-29 23:15:09,853 [INFO] 2 0 | 157.0 | 0.287673 | 2-[2]->238825-[15]->82118-[2]->207
49 2025-09-29 23:15:10,997 [INFO] 2 1 | 176.0 | 1.143685 | 2-[15]->205169-[50]->54378-[53]->1
50 2025-09-29 23:15:11,549 [INFO] 1 0 | 159.0 | 0.551669 | 1-[12]->28927-[3]->52273-[10]->117
51 2025-09-29 23:15:11,550 [INFO] Results saved: output/results.txt
52 2025-09-29 23:15:11,550 [INFO] Final memory usage: 782.42 MB
53 2025-09-29 23:15:11,550 [INFO] === LGPF session finished ===

```

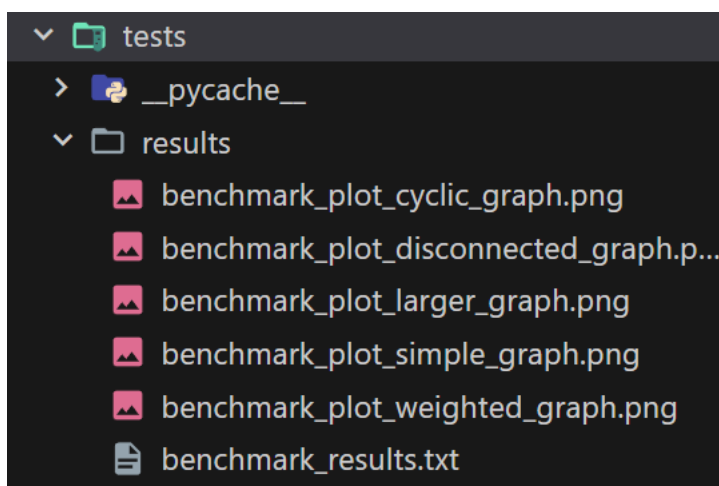
7. Benchmarking (Optional) *note: see Benchmarking for deeper details*

1. Run all unit tests and benchmarks it will create `tests/results/benchmark_result.txt` see Testing on how to run tests.

2. or Generate plots at root directory:

```
python3 tests/plot_benchmarks.py
```

3. Charts are saved in `tests/results/benchmark_plot_<name>_graph.png`.



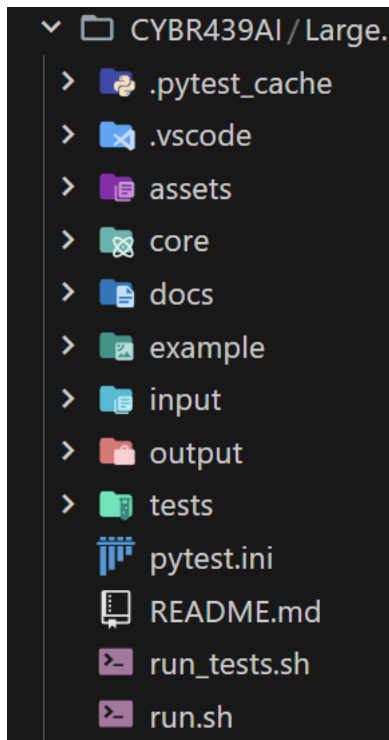
Testing

Unit tests are provided under `tests/<test_test_file_name>.py`. Run with:

This project uses **pytest** for automated unit tests and benchmarks.

Running All Tests

Run all unit tests and benchmarks with *note: must be under root folder*:



```
pytest -v
```

or use the helper script:

```
./run_tests.sh
```

This will execute all test cases under *tests/* and run benchmark comparisons across algorithms.

Save benchmark results into *tests/results/benchmark_results.txt*

Benchmark Plotting After benchmarks are generated, you can visualize results see [Benchmarking](#).

Benchmarking

Benchmarking evaluates how each algorithm performs across different graph types. This helps us compare runtime efficiency, path costs, and algorithm suitability.

Running Benchmarks

Run the benchmarks directly with Python:

```
python3 tests/benchmark.py
```

Or use the convenience script:

```
./run_tests.sh
```

Both commands will execute all algorithms against predefined graphs (simple_graph, weighted_graph, cyclic_graph, etc.) and store results in: *output/benchmark_results.txt*

Example Results

Below is a snippet of results from *benchmark_results.txt*:

23	weighted_graph	GreedyZero	Time: 0.000005s	Cost: None
24	cyclic_graph	GreedyZero	Time: 0.000004s	Cost: None
25	larger_graph	GreedyZero	Time: 0.000005s	Cost: None
26	simple_graph	AStarZero	Time: 0.000005s	Cost: 1.0
27	disconnected_graph	AStarZero	Time: 0.000005s	Cost: inf
28	weighted_graph	AStarZero	Time: 0.000005s	Cost: 4.0
29	cyclic_graph	AStarZero	Time: 0.000009s	Cost: 4.0
30	larger_graph	AStarZero	Time: 0.000006s	Cost: 4.0
31	simple_graph	BidirectionalBFS	Time: 0.000005s	Cost: None
32	disconnected_graph	BidirectionalBFS	Time: 0.000005s	Cost: None
33	weighted_graph	BidirectionalBFS	Time: 0.000004s	Cost: None
34	cyclic_graph	BidirectionalBFS	Time: 0.000003s	Cost: None
35	larger_graph	BidirectionalBFS	Time: 0.000004s	Cost: None

Benchmark notations and meaning

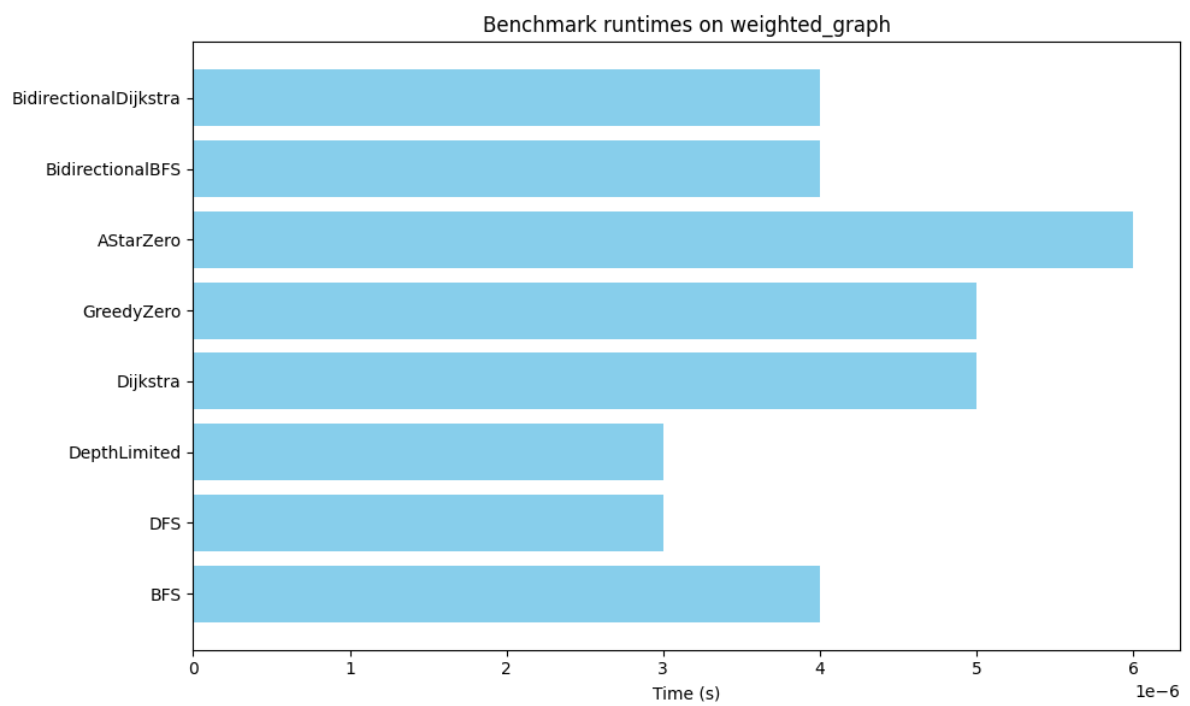
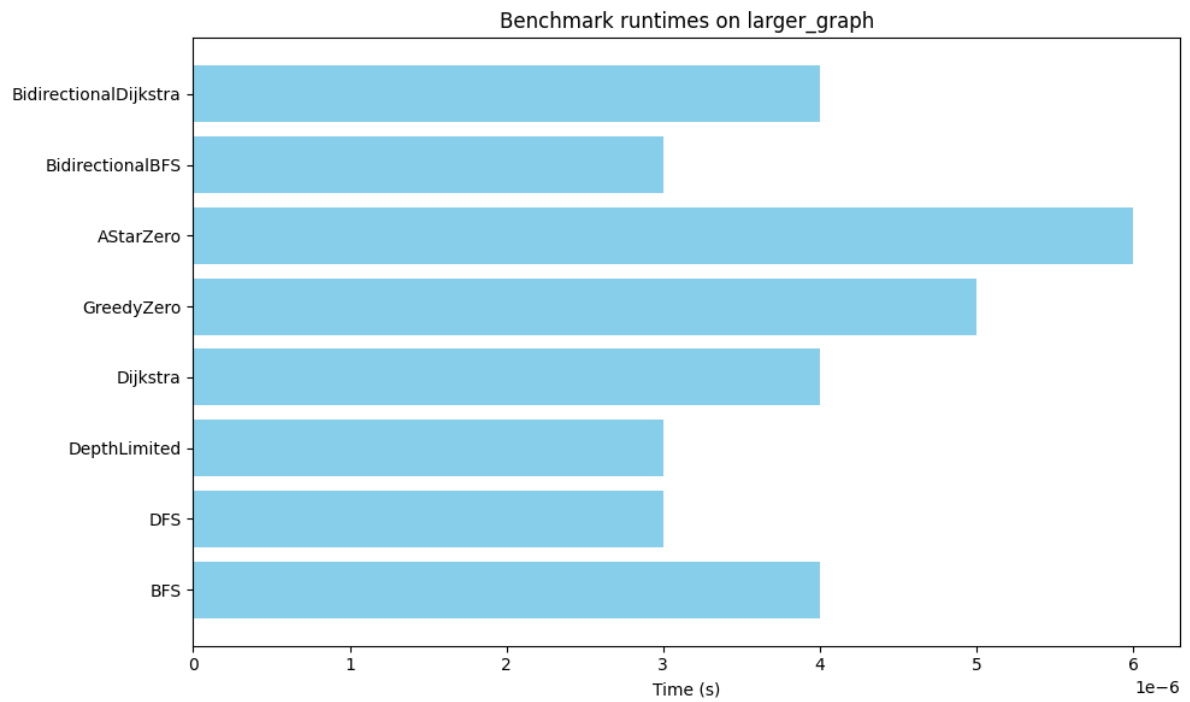
- Time → Runtime of the algorithm on that graph.
- Cost → Path cost from source to target:
- A number (e.g., 4.0) means a path was found.
- inf means no path exists between source and goal.
- None means the algorithm doesn't compute costs in that context (e.g., plain BFS for reachability).

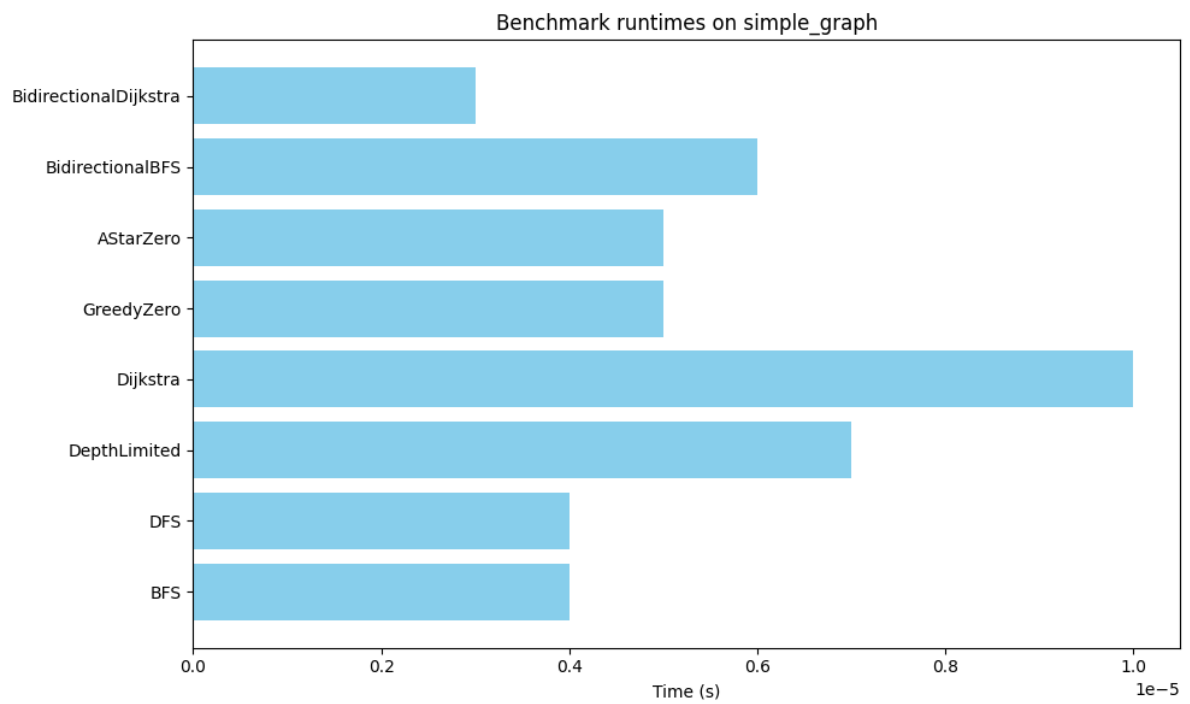
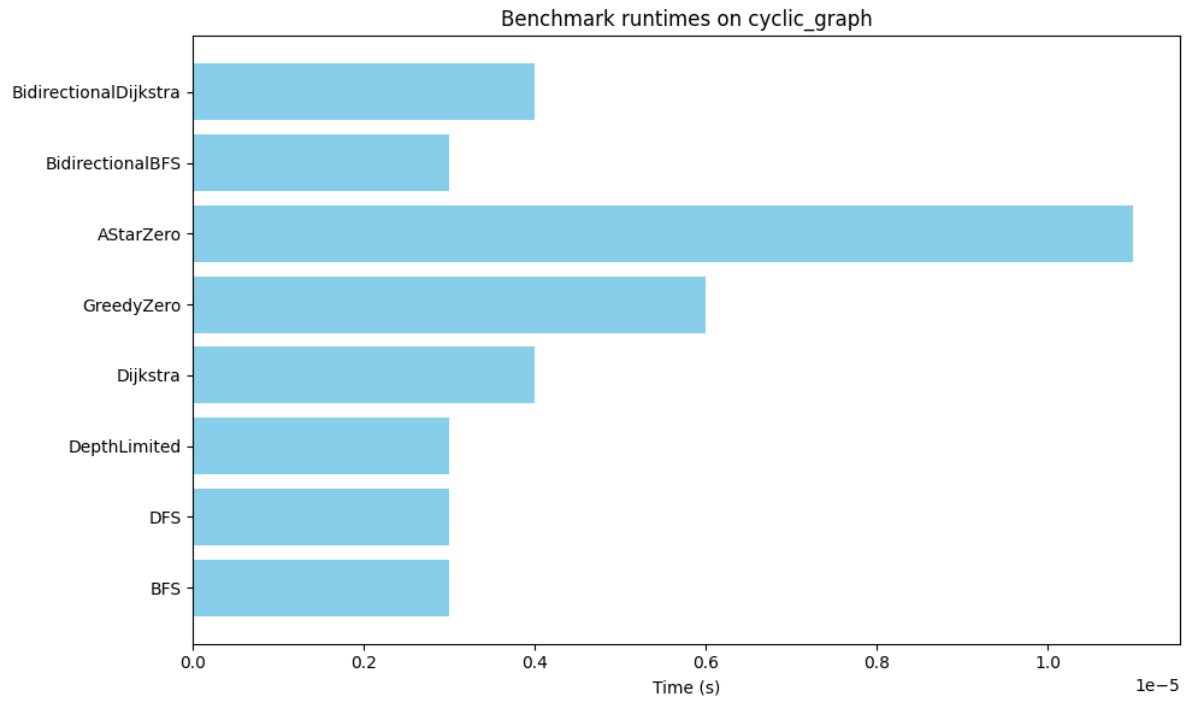
Visualization

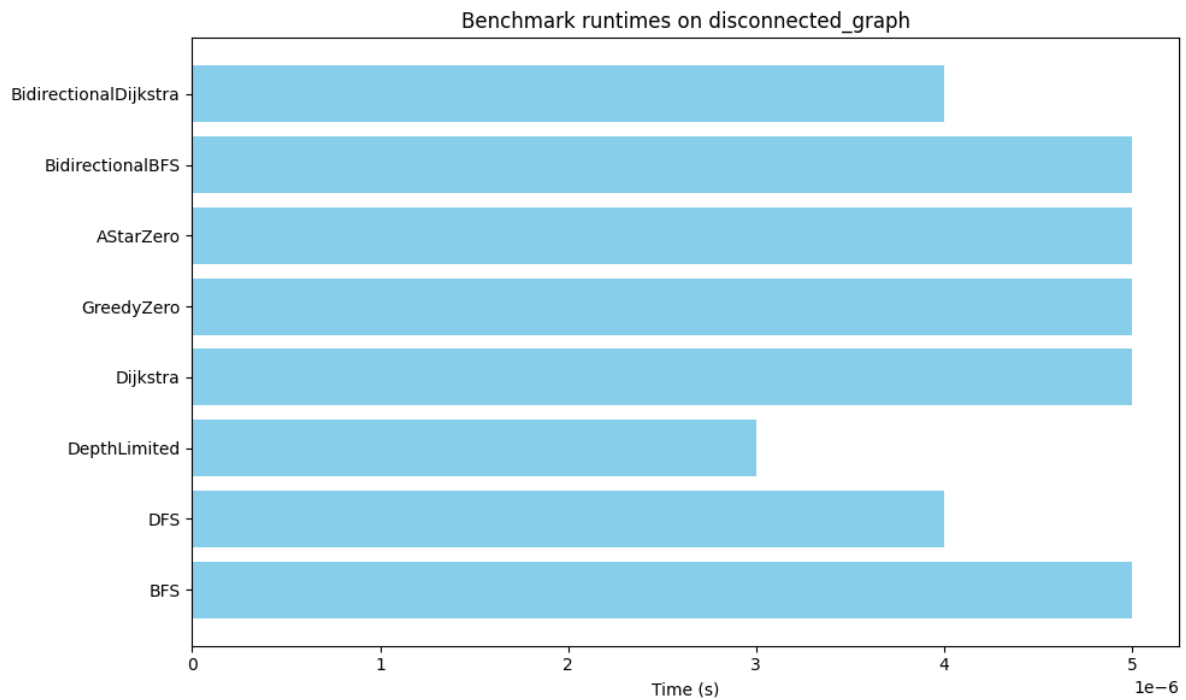
You can visualize algorithm performance (runtime vs. cost) using matplotlib:

```
python3 tests/plot_benchmarks.py
```

This produces a chart comparing algorithms across test graphs, saved under *tests/results/benchmark_plot.png*.







Documentation

- Development log: `docs/ai_chat_log.md` Contains incremental notes on implementation choices, code summaries, and next steps.
-

AI Usage

- AI is used through out the development process. Mostly auto-complete feature with VS Code GitHub Copilot
- For code intergration and advice Open-AI ChatGPT 5 is used see `project_root/docs/ai_chat_log.md` for prompt and scripts.

20250930_000502_ai_chat_log.md

Project Structure

LargeGraphPathFinder/

```
├── .pytest_cache/    # Pytest cache
├── .vscode/          # VS Code configuration
├── assets/           # Used to store README photos
```

— core/	# Main source code
— docs/	# Documentation (ai_chat_log.md, etc.)
— example/	# input and out generator is located here too
— input/	# User-provided JSON graphs
— output/	# Program-generated query files & logs
— tests/	# Unit tests + benchmarks
— pytest.ini	# Pytest configuration
— README.md	# This file
— README.pdf	# Pandoc converted PDF version of this file
— run_test.sh	# Launcher script for all tests
— run.sh	# Launcher script
— pytest.ini	# Pytest configuration

Future Work

- Increase performance across the board
- Code refactoring and efficiency
- Scale ALT preprocessing to millions of nodes with smarter landmark selection
- Parallel preprocessing (multi-core Dijkstra)
- Cluster-based heuristics for very large graphs
- Web-based visualization for graphs >10k nodes
- Incremental memory-usage profiling