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AI Chat Log - Large Graph Path Finder (LGPF)

This log summarizes the development of the Large Graph Path Finder project. Each section documents major features added, the corresponding code files, and purpose.

1. Project Setup

• Created initial folder structure:

• Added .vscode/launch.json for debugging in VS Code.

2. Timer & Memory Monitor

File: core/resource.py

- Added a 60-second countdown timer.
- Tracks system memory usage each second using psutil.
- Runs in a background thread so preprocessing continues.

3. Graph Parser

File: core/parser.py

- Reads graphs generated by graph-generator.py.
- Converts JSON format into an adjacency list: python {0: [(2, 93)], 1: [(2, 24)],
- 2: [(0, 93), (1, 24)]} Handles undirected and directed graphs.
- Raises custom GraphParseError on invalid files.

4. Main Program Driver

File: core/main.py

- **Prompts user** for a graph filename inside input/.
- Starts the **countdown** + **memory tracker** only after a valid file is chosen.
- Loads the graph with parser.py.
- Creates a companion query file output/xxx_node.json containing: json {
 "graph_file": "example_graph.json", "num_nodes": 3, "nodes":
 [0, 1, 2] } Verbose output confirms steps ([INFO], [WARN], [ERROR]).

5. Run Script

File: run.sh

- Ensures the program always runs from project root: bash python3 -m core.main

6. VS Code Debugging

File: .vscode/launch.json

- Configured to run core.main as a Python module.
- Ensures core/imports (core.parser, core.resource) resolve correctly.

Next Planned Steps

- Extend xxx node.json to include query pairs (manual or random).
- Implement pathfinding algorithms (dijkstra, later bidirectional dijkstra).
- Process queries and log results in output/results.log.

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Allow user to choose whether to print (cat) the log at the end.

7. Algorithms Module

File: core/algorithms.py

- Implemented consistent (path, cost) return format across all algorithms.
- Added algorithms: Breadth-First Search (BFS)
- Depth-First Search (DFS)
- Depth-Limited DFS
- Dijkstra's Algorithm
- Greedy Best-First Search (with heuristics)
- A* Search (with heuristics)
- Bidirectional BFS
- Bidirectional Dijkstra (currently falls back to normal Dijkstra for directed graphs).

8. Heuristics Module

File: core/heuristics.py

- Split heuristics from algorithms for modular design.

- Implemented: Zero heuristic
- Manhattan distance
- Euclidean distance
- Chebyshev distance
- Octile distance
- Haversine (for lat/lon geographic graphs)
- File structured for easy extension with new heuristics.

9. Testing

Files: tests/test_algorithms.py, tests/conftest.py

- Created pytest-based unit tests.
- Graph fixtures: Simple graph
- Disconnected graph
- Weighted graph
- Cyclic graph
- Larger graph
- Parameterized tests run all algorithms against all graph types.
- Extended to test A* and Greedy under multiple heuristics (zero, Manhattan, Euclidean).

10. Benchmarking

File: tests/benchmark_test.py

- Benchmarks all algorithms on all sample graphs.
- Logs results to tests/results/benchmark results.txt.
- Provides execution time and path cost for comparison.

11. Documentation

File: README.md

- Added structured README with abstract, installation, prerequisites, environment info, usage, testing, benchmarking, documentation pointer, and project structure.

- Includes Table of Contents for Pandoc PDF conversion.
- Points to detailed development log (docs/ai chat log.md).

Next Planned Steps

- Add more algorithms for completeness and large-graph scenarios:
 - Iterative Deepening DFS (IDDFS)
 - Bellman-Ford (handles negative weights)
 - Floyd-Warshall or Johnson's (all-pairs shortest path)
 - Bidirectional A* (fast for sparse graphs)
- Add support for landmark-based heuristics.
- Expand benchmarking to measure memory usage.

Submission Notes

- **Completeness:** input → preprocessing → query file created.
- **Correctness:** robust parsing + memory/timer enforcement.
- Readability: clear code structure, logging, and error handling.
- Files included:
 - core/main.py
 - core/parser.py
 - core/resource.py

- run.sh
- README.md
- .vscode/launch.json
- sample input/example graph.json

12. Preprocessing & Query Policy

File: core/pathfinder.py

preprocess index(adj) builds:

weight map[u][v] for fast per-edge weight lookups (pretty printing + totals).

Weakly-connected components (reachability pruning).

Basic stats: node_count, edge_count, uniform_weights.

choose algorithm(adj, index):

Small & unweighted → BFS (fast, hop-optimal).

Weighted \rightarrow Dijkstra (optimal).

Very large \rightarrow A* with zero heuristic (safe default).

find path checks reachability cheaply; then runs the chosen algorithm.

13. Result Formatting (new)

File: core/graphing.py

Converts a path like [1,2,3,4] into:

1-[36]->2-[1]->3-[45]->4

and computes total weight (e.g., 82.0).

format result line emits:

src dst | total weight | time sec | path with edge labels

write results writes a header + all result lines to output/results.txt.

14. Interactive Queries (new UX)

If no query file is provided, the program allows interactive pair entry:

src dst (empty to finish):

15. Current Status

Program runs end-to-end:

Prompts for JSON \rightarrow preprocess (\leq 60s target) \rightarrow queries (file or interactive) \rightarrow results/logs/artifacts.

Output artifacts:

output/results.txt – per-query results (weight, time, path).

output/run.log – detailed log (timer, memory, choices, warnings).

output/graph.png – small-graph preview.

output/_nodes.json - node reference.

16. Next Big Feature: Landmark Heuristics (ALT)

Planned files: core/landmarks.py, updates to core/pathfinder.py

ALT (A, Landmarks, Triangle inequality)* for large graphs:

Pick K landmarks (e.g., 8–16) quickly (degree/farthest selection).

Precompute distances: d(L, v) and d(v, L) (via Dijkstra and reversed graph).

Query uses ALT heuristic:

$$h(s,t) = \max_{L} \{ |d(L,t) - d(L,s)|, |d(s,L) - d(t,L)| \}$$

Heuristic is admissible (A* remains optimal when desired).

Major speedups on dense/huge graphs while staying within the 60s preprocessing budget (tune K).

Policy (pathfinder):

If node count \geq = 10k and ALT tables exist \rightarrow use A* + ALT.

Else keep current BFS/Dijkstra/A*0 strategy.

Future: add clustering (e.g., "Tokyo center") for coarse routing + local refinement.

17. How to Run

./run.sh # then follow prompts: # 1) Enter graph JSON in input/ (e.g., example_graph.json) # 2) Enter query file in input/ (e.g., example query.txt) or input pairs interactively

Query file format (input/example_query.txt):

021201202110

18. Roadmap to MVP

Implement core/landmarks.py (selection + tables + ALT heuristic).

Integrate ALT in pathfinder.preprocess_index and choose_algorithm.

Add a large-graph test/benchmark to validate speedups.

Document ALT mode in README (memory/time trade-offs, K tuning).