

CSE518 : Artificial Intelligence Project Final Report

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Abstract

This project develops an intelligent rover navigation system that combines Reflex-based real-time control with Hierarchical A* path planning. The rover continuously adapts to environmental conditions such as obstacles, cliffs, traps, battery limits, and terrain cost. A utility-based terrain evaluation model is integrated to balance safety with scientific value by adjusting movement cost through terrain utility gains. Multiple heuristics, including Manhattan, Euclidean, Adaptive Cost, Terrain-based, and Obstacle-Aversion, were implemented and compared for efficiency and optimality. The overall system produces safe, battery-feasible, and mission-oriented paths while maintaining responsiveness through reflex actions. This approach helps enhance performance in dynamic and uncertain terrain environments.

1. Implementation of the reflex module :

- The reflex module continuously monitors real-time conditions such as battery level, terrain cost, and nearby obstacles during path execution.
- Each reflex is triggered instantly when its condition (e.g., Battery < 20%, or ROCK tile ahead) is detected, ensuring immediate corrective action.
- The module supports five reflex types - Obstacle, Low Battery, Terrain, Obstacle-Aversion Heuristic, and Recharge - each modifying movement or heuristic dynamically.
- Reflex actions include rerouting, skipping unsafe cells, penalizing risky areas, or initiating recharging sequences to maintain safe navigation.
- The reflex logic operates independently of A*, allowing real-time adaptability without restarting the full path planning process.

Based upon our Analysis, the Reflex mechanisms implemented are as follows:

Reflex Type	Condition	Actions
Obstacle Reflex	Rock tile in next step	Avoid move
Low Battery Reflex	Battery < 20%	Stop expansion
Terrain Reflex	High cost	Costly move
Obstacle-Aversion Heuristic Reflex	Near Rock obstacle	Add penalty to heuristic
Recharge Reflex	Battery insufficient	Move to nearest recharge station or abort

Cliff Reflex	Cliff ahead	Block direction + reroute
Hazard Reflex	Hazard at current cell	STOP → BACKTRACK to last safe cell
Trap Reflex	Rover stuck / trap tile	STOP + BACKTRACK + avoid zone

2. Implementation of Path planning module :

The path planning of the rover is based on a Hierarchical Strategy to generate battery-conscious and safe route planning based upon the A* algorithm.

I. Base-Level Planning (Battery-aware A* Function):

Function objective : The cost of the path segment between two nodes is minimized to find a safe path (battery-conscious).

Constraints :

1. Rock Avoidance: Completely excludes Rock tiles from the search space.
2. Battery Safety: Prunes any path branch if the projected cost would cause the remaining battery to drop below a 20% threshold.

II. High level Planning (Plan_with_Recharges module) :

Initial Check: Always first attempts a Direct Path to the goal using A* with the current battery level.

Iterative Chaining (If Direct one Fails):

1. Identify Reachable sites : Finds all Recharge Sites (RC's) accessible with the current battery (considering 20% threshold safety check).
2. Selection: Chooses the reachable RC that is closest to the Final Goal (greedy algorithm using the heuristic).
3. Segment Execution: Appends the found A* segment to the RC, simulates Recharge (100% battery), and updates the current position.
4. Backtracking Check: If, during segment execution, the rover encounters a Hazard/Trap tile not predicted by A*, it: Stops immediately, Backtracks to the last known safe cell, Permanently blocks the hazardous cell, and Replans the current segment from the safe position.
5. Repeat: It loops back to attempt a path from the newly charged RC (or the backtracked safe node) to the Goal.

Goal: Returns a full, cost-optimized, and battery-feasible path, which may consist of multiple segments between Start, RC's, and Goal.

III. Integration of both modules :

Condition	Action Taken	Planning Outcome
B < 20%	Immediate Stop and reroute to find nearby RC.	Priority for battery over goal.
20% ≤ B ≤ 25% & RC nearby (≤ 2 tiles)	Reroute to nearby RC : Plans ahead to stop at the adjacent recharge for efficiency.	Saves run time by optimizing for charging opportunities.
Next Step Cost > B (Unsafe Move)	Replans:Calls the high-level planning module to insert a necessary recharge stop before the greater cost tile.	It ensures that the rover never attempts a move that is not feasible, even if initially planned.
Rover enters Hazard / Trap during execution	Stop, backtrack to last safe cell, permanently block that hazardous tile	Ensures safe recovery and prevents the planner from using the hazardous cell again.

Note : For the code detailed flow understanding, 2 flowcharts are attached in the google drive link.

▣ CSE518_Artificial_Intelligence_Code_Flowchart

3. Explanation of the 4 heuristics :

3.1. Manhattan Distance

- Counts the number of grid steps between any 2 points in both horizontal directions and vertical direction.
- Most effective on motion 4 ways (no diagonals).
- **Formula:** $|x_1 - x_2| + |y_1 - y_2|$

3.2. Euclidean Distance

- Measures the straight-line (diagonal) distance between two points.
- More expensive due to the square root term in the calculation
- **Formula:** $\sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$

3.3. Terrain Aggressive Heuristic :

- It calculates the terrain cost between two points, not just the distance.
- It gathers multiple points along the line in order to find the average terrain cost.
- It penalizes the routes passing through expensive terrains like rocks or sand.
- **Formula :** Distance * (Average terrain cost along the path)

3.4. Adaptive Cost Manhattan(H1) :

- It extends Manhattan by scaling it with the minimum terrain cost (5).
- Ensures the heuristic is admissible and consistent.
- Provides stronger guidance than raw Manhattan while remaining optimal.
- Easy and fast to compute.
- **Formula:** $5.0 * (|x_1 - x_2| + |y_1 - y_2|)$

3.5. Obstacle-Aversion (H2) :

- It works keeping in mind the H1 but adds extra cost when near to the rocks.
- It naturally keeps the path finding rover to stay away from the rock or obstacle.
- It follows an inverse relation i.e. higher penalty for the nearby rocky area.
- **Formula:** $(5.0 * \text{Manhattan Distance}) + (10 / (\text{Min distance to rock})^2)$

4. Descriptive Comparison Report :

Heuristic	Efficiency (Nodes)	Optimality (Cost)	Speed (Computation Time)	Consistent	Admissible
Adaptive Cost (H1)	Moderate	Moderate	Fast	Yes	Yes
Euclidean	Low (many nodes)	Moderate	Fast	Yes	Yes
Manhattan	Low (many nodes)	Moderate	Fast	Yes	Yes
Obstacle Aversion (H2)	Moderate	Moderate	Very Slow	No	No
Terrain (Aggressive)	High(fewer)	Lowest (least optimal)	Slow	No	No

Interpretation Summary :

- **Adaptive Cost (H1):** Best overall balance: fast, efficient, and near-optimal with reliable consistency.
- **Euclidean / Manhattan:** Most optimal but explore more nodes (less efficient).
- **Obstacle Aversion (H2):** Safe but very slow and not optimal due to inconsistency.

- **Terrain (Aggressive):** Very efficient but produces the least optimal paths; slow and unreliable.

Utility Based Cost Enhancement for comparison:

1. Cost-Only Model (Previous Approach)

- Move cost = $\text{terrain_cost} \times \text{multiplier}$
- Planners always preferred the cheapest path.
- No consideration of the scientific value of terrains.

2. Cost- Utility Model (Current Approach)

- New cost = $(\text{terrain_cost} \times \text{multiplier}) - \text{terrain_utility}$.
- High-utility terrains become more favorable. E.g. Hazardous = +10 for better scientific exploration.
- Negative-utility terrains are naturally avoided. E.g. Traps = -5 for avoidance.

Key Benefits :

- The planner now balances safety + scientific value.
- Encourages meaningful exploration while avoiding risky zones.
- Produces more mission-focused and intelligent paths.

Conclusion :

The proposed navigation framework successfully integrates reflex mechanisms with a hierarchical planning strategy to achieve safe and reliable rover movement. Reflexes provide immediate responses to hazards, low battery, cliffs, and traps, while the planning module ensures long-term feasibility through battery-aware A* search and recharge routing. The other uniquely added utility-based cost model improves decision-making by encouraging exploration of scientifically valuable terrains and avoiding low-utility or dangerous regions. Experimental comparison of heuristics shows that Adaptive Cost (H1) offers the best balance of speed and near-optimal paths. Overall, our model demonstrates strong adaptability, efficient exploration, and goal-focused navigation suitable for real-world rover applications.

Credits : Nitant Jain :

- Implemented high-level planner and segment-based planning.
- Developed a utility-based cost model.
- Implemented Obstacle Aversion (H2) and merged it with A*.

Credits : Even Patel :

- Implemented Base level planner A* with battery-aware constraints.
- Designed the Reflex Module (Obstacle, Battery, Terrain, Hazard, Trap, Recharge).
- Implemented Adaptive Cost Manhattan (H1) and Terrain Aggressive Heuristic.