



KLunar Navigator: AI for Lunar Exploration

Presented by Faiza Abdullah for ITAI-2372-Artificial Intel Applications.



Made with GAMMA

Project Overview

The KLunar Navigator is an AI-driven rover navigation system for lunar exploration, utilizing Q-learning to autonomously navigate a 10x10 grid from a start point (0,0) to a target (9,9) while avoiding obstacles.

Objectives

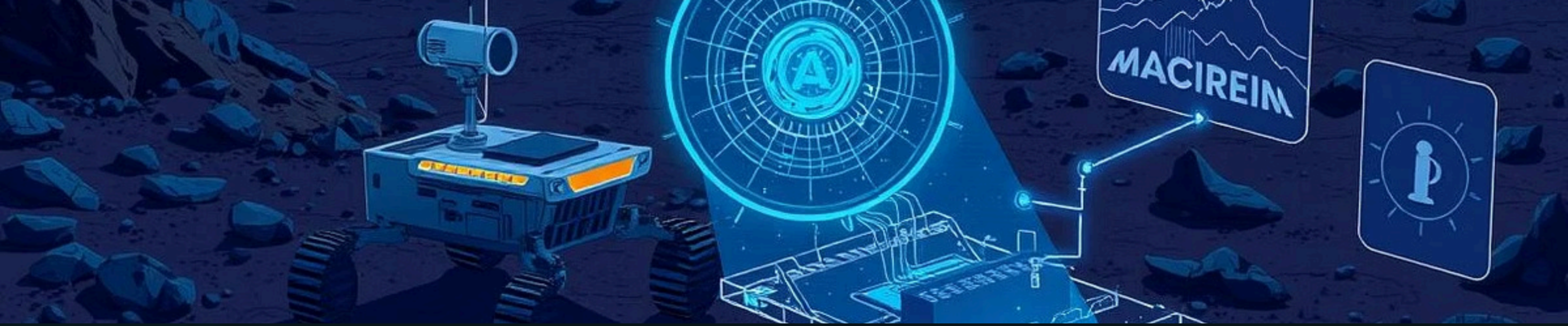
- Develop a functional Q-learning agent.
- Create a Pygame-based visualization.
- Ensure robustness against obstacles.
- Prepare for scalability.

Benefits

- Enhances RL understanding in robotics.
- Provides a testable prototype.
- Supports educational outreach.

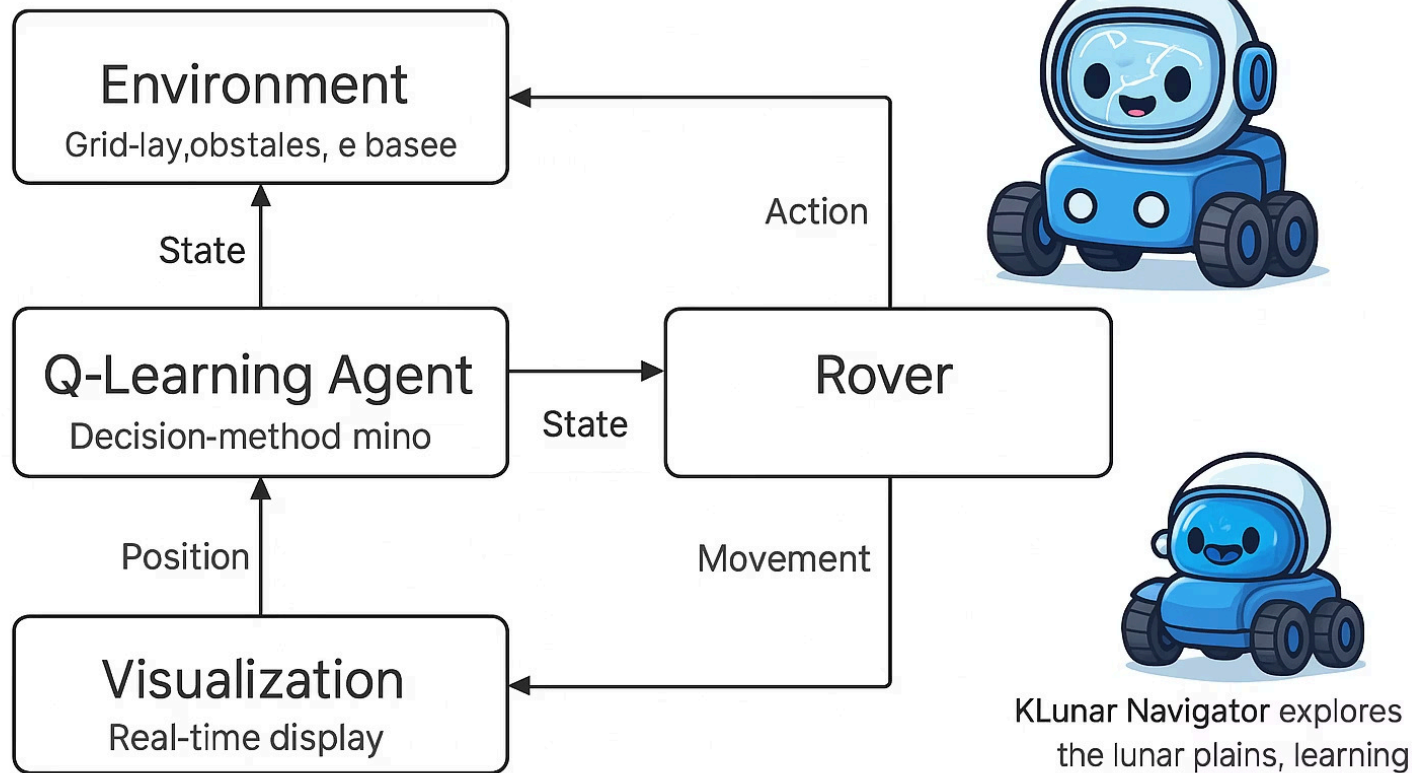


KLunar Navigator
explores the lunar plains,
learning to avoid craters
and reach its base.



System Architecture

The system integrates several key components: LunarEnvironment, Rover, QLearningAgent, and Visualizer, all managed by a main loop for autonomous navigation and real-time feedback.



Architecture



KLunar Navigator explores the lunar plains, learning to avoid craters and reach its base!

Solution Approach: Core Components



Environment Setup

10x10 grid with start (0,0), target (9,9), and random obstacles.



Rover Movement

Four actions (right, left, down, up) with energy tracking and boundary checks.



Q-Learning Agent

Q-table for 10x10x4 states, learning from rewards and saving data.



Visualization

Pygame window rendering rover, target, and obstacles in real-time.

Technical Details & Main Loop

Main Loop (main.py)

- 100 episodes, 200 max steps per episode.
- Manages setup, loop, and Q-table persistence.

Optimizations

- Reduced obstacles for faster convergence.
- Adjusted learning parameters.
- `Max_steps` ensures loop completion.

The system is built with Python 3.12, utilizing Pygame, NumPy, and asyncio libraries for efficient simulation and visualization.

Testing Plan: Validation Objectives

Validate the rover's ability to navigate from (0,0) to (9,9), avoid obstacles, and complete 100 episodes within 200 steps each.

1 Initial Movement

Verify correct movement and reward for basic actions.

2 Obstacle Avoidance

Confirm rover stays at current position with negative reward.

3 Target Reach

Ensure target reached with positive reward and episode termination.

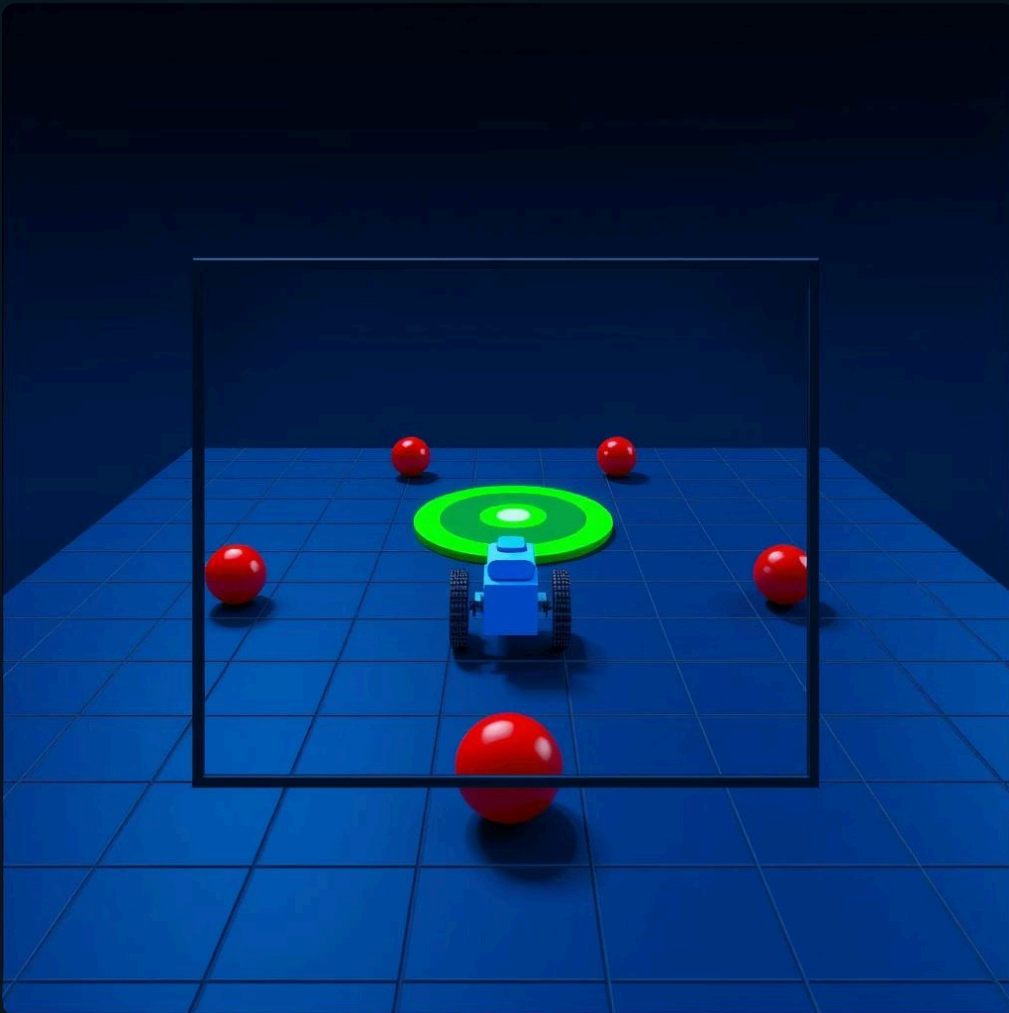
4 Boundary Check

Test edge cases to confirm rover stays within grid.

Testing Success Criteria

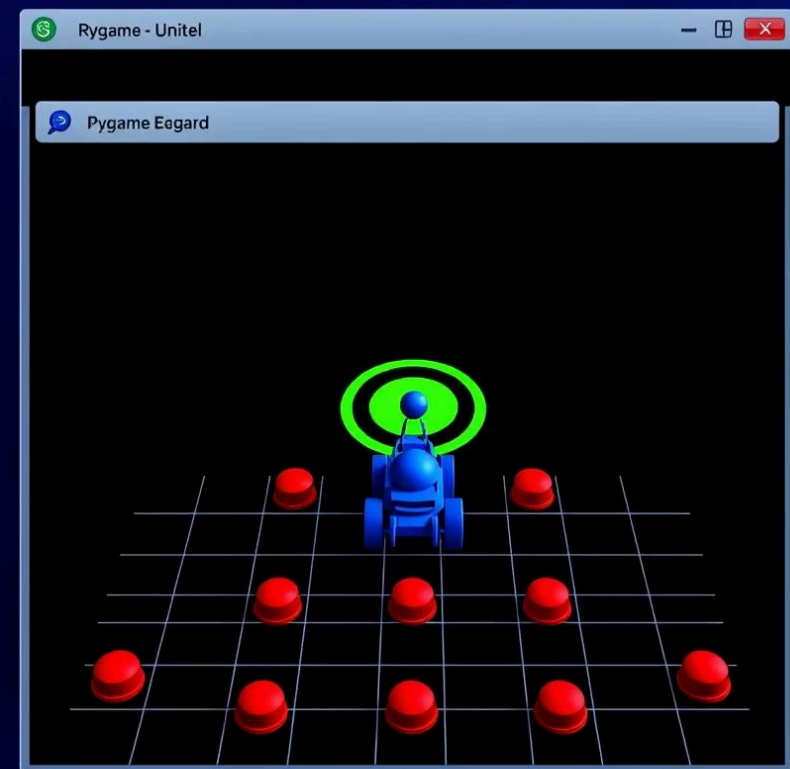
Visual Confirmation

Rover moves visibly toward (9,9) in Pygame.



Episode Completion

100 episodes complete within 20,000 total steps.



A generated `q_table.npy` with non-zero values confirms successful learning.



Future Enhancements

Dynamic Obstacles

Implement real-time obstacle generation for complex scenarios.

Multi-Rover Coordination

Develop algorithms for multiple rovers to cooperate.

3D Terrain Simulation

Upgrade visualization and navigation to a realistic 3D environment.

These enhancements will further bridge the gap between conceptual simulation and real-world space exploration.