Phase 2: Innovation & Problem Solving

<u>Title: Autonomous Vehicle and Robotic System</u>

Innovation in Problem Solving

The objective of this phase is to explore and implement innovative solutions to modern transportation and automation challenges. This project aims to enhance mobility, safety, and operational efficiency by developing Alpowered autonomous vehicles and robotic systems capable of real-time decision-making and smart interaction in dynamic environments.

Core Problems to Solve

- 1. <u>Real-Time Decision Making</u>: Autonomous systems should accurately process real-time data plus also quickly make safe decisions within unpredictable environments such as in traffic or in crowded areas.
- 2. <u>Navigation in Dynamic Environments</u>: Robots and vehicles must navigate through terrains, cityscapes, changing, and unstructured. They must also consistently avoid various obstacles.
- 3. <u>Interoperability and Communication</u>: Coordination between autonomous units is essential for smooth operations and traffic flow.
- 4. <u>Trust and Safety</u>: High reliability, behavior that is transparent, and communication of system decisions that is effective are required for gaining public trust in fully autonomous systems.

Innovative Solutions Proposed

1. Al-Based Perception and Navigation System

- -Solution Overview: Implement computer vision and sensor fusion techniques using AI to enable accurate perception of surroundings.
- <u>Innovation</u>: Real-time object detection and tracking using deep learning models, integrated with LiDAR and ultrasonic sensors.

- Technical Aspects:

- CNN-based vision for lane, object, and pedestrian detection.
- Sensor fusion from camera, GPS, and IMU.
- Adaptive route planning and path correction.

2. Intelligent Decision-Making with Reinforcement Learning

- <u>Solution Overview</u>: Equip the autonomous system with reinforcement learning agents for adaptive, experience-based decision-making.
- <u>Innovation</u>: The vehicle/robot learns from simulation environments and real-world feedback to improve navigation efficiency.
- Technical Aspects:
 - Deep Q-learning algorithms.
 - Simulation-based training using virtual environments.
 - Real-time decision system deployed on edge devices.

3. <u>Multi-Robot Coordination via IoT and V2V</u> <u>Communication</u>

- <u>Solution Overview</u>: Enable communication between multiple autonomous vehicles or robots for collision avoidance and task distribution.
- Innovation: Use IoT protocols (like MQTT) and vehicle-to-vehicle communication to manage traffic flow and collaboration.
- Technical Aspects:
 - ESP32 or Raspberry Pi for network communication.
 - Cloud dashboard for telemetry and control.
 - Real-time status broadcast and decision syncing.

4. Trust-Building Through Transparent Al Behavior

- <u>Solution Overview</u>: Display real-time system decisions and justifications through a user interface or voice feedback.
- <u>Innovation:</u> Build user trust by explaining why a robot or vehicle made a specific move or stop.

- Technical Aspects:

- Simple GUI or dashboard for feedback.
- Integration of NLP for voice-based system explanations.
- Logging decisions for audit and training improvement.

Implementation Strategy

1. Prototype Development

- Assemble a model autonomous vehicle using Arduino or Raspberry Pi, sensors, and motors.
 - Develop software stack: perception, path planning, control.

2. Al Training & Simulation

- Train models using open datasets (e.g., KITTI, Carla Simulator).
- Test edge-case scenarios in simulation before real-world trials.

3. Communication Network

- Implement IoT protocols for inter-robot communication.
- Set up a cloud-based or local server to collect and analyze telemetry.

Challenges and Solutions

- <u>Sensor Noise and Uncertainty</u>: Use Kalman Filters and data smoothing techniques for better sensor fusion.
- <u>Complex Environments</u>: Train AI models on diverse datasets and simulate real-world complexity before deployment.
- <u>Constraints Power and Processing</u>: Optimize models using edge AI techniques (e.g., TensorFlow Lite).

Expected outcomes:

- 1. <u>With Improved Mobility</u>: Robots and vehicles are capable of navigating cleverly and with minimal human intervention.
- 2. <u>Scalable Deployment</u>: Small bots benefit through modular design. Modular designs also give full-size vehicles scalability.
- 3. <u>Increased Safety</u>: Al-powered perception considerably cuts down on accidents by reacting more quickly than human drivers.
- 4. <u>Public Trust and Transparency</u>: Individual users and operators gain a comprehension of system actions, so individual users and operators have trust for autonomous decisions.

Next steps:

1. Prototype Testing

- Test in certain campuses or obstacle courses inside controlled environments.

2. Data-Driven Improvements

- Collect logs from test runs refining models and route planning.

3. Full-Scale Deployment

- Deployment in use-cases such as campus shuttles, delivery bots, or warehouse automation is prepared for.