

Design Document

For a Self-Driving Car

Operations:

The design of this software system of a driverless car focuses around three essential operations: Obstacle Detection and Avoidance, Traffic Sign Recognition, and Automated Parking.

Research Keywords:

The keywords “autonomous vehicle”, “driverless car”, “robotic car”, “radar driverless car”, “GPS driverless car”, “odometry driverless car,” and “ultrasound sensor parking” were selected for the initial research phase of the design of this software, and are an essential part of the technologies that empower the three selected operations. The technologies such as radar, GPS, and odometry are essential for precise positioning and navigation, while ultrasound sensors play a key role in proximity detection (essential for obstacle avoidance and autonomous parking). As I began the design phase of the Obstacle Detection and Avoidance system, I was quickly confronted with the ethical and legal aspect driverless cars. This led me to broaden my research to include key concepts like the "trolley dilemma for autonomous vehicle" and "legal responsibility for autonomous vehicle". These topics are crucial for the design of driverless cars software, especially when the vehicle confronts scenarios where collisions are imminent and inevitable, highlighting these ethical and legal challenges.

Rationale of key elements:

Class: ObstacleDetector

- Attributes:
detectedObjects: Stack
objectType: string
objectID: int
isAvoidable: boolean
fragilityIndex: float
avoidancePriority: Queue
- Operations:
gatherData()
processData()
setIsAvoidable()
getIsAvoidable()
setObjectType()
getObjectType()
setObjectID()
getObjectID()
setFragilityIndex()
getFragilityIndex()
addToAvoidancePriority()
getAvoidancePriority()

avoidObstacle()

The obstacleDetector class detects and manages objects and obstacles the vehicle encounters. In the context of the software, we consider that an object becomes an obstacle when it is in the vehicle's path. It uses a stack (in LIFO order for quick re-evaluation) to keep track of detected objects, allowing for quick updates as objects come into the sensor range. The queue for avoidance priority helps in determining the order in which obstacles should be avoided. A boolean defines if the obstacle is avoidable. If not, the program defines a fragility index of the main obstacle, and defines the fragility index of surrounding objects in order to calculate if swerving the vehicle and hitting other objects would reduce the harm (for example, hitting a public bench to avoid hitting a pedestrian). The methods seen in this class, aligns with the ethical and decision-making considerations discussed in Schäffner's (2020) study on moral decision-making in self-driving car dilemmas.

Class: TrafficSignRecognizer

- Attributes:
recognizedSigns: Queue
isVehicleCompliant: Boolean
- Operations:
recognizeSigns(sensorData)
interpretSigns()
adaptVehicleBehaviour()

The TrafficSignRecognizer class handles the interpretation of traffic signs and signals, ensuring compliance with traffic laws. The queue holds recognized signs for sequential processing, and a boolean indicates compliance status of the vehicle. Traffic Sign Recognition (TSR) systems can reach an accuracy rate of up to 99.91% (Triki, Karray and Ksantini, 2023).

Class: ParkingAssistant

- Attributes:
parkingSpace: AvailableParking
parkingRequested: Boolean
- Operations:
requestParking()
findSpace()
parkVehicle()

The ParkingAssistant class manages the parking operations for the vehicle. It utilizes a boolean to track if parking is requested by the user and launches the parking search as well as the necessary manoeuvres to park the vehicle if a parking space is found.

Class: AvailableParking

- Attributes:
parkingType: string
depth: float
width: float
isAccessible: boolean

- Operations:
 availableParking(parking Type: string, depth: float, width: float, isAccessible: boolean)
 getParkingType(): string
 setParkingType(parkingType: string)
 getDepth(): float
 setDepth(depth: float)
 getWidth(): float
 setWidth(width: float)
 setIsAccessible(isAccessible: boolean)

Rationale: This class represents the potential parking spot available for the vehicle. Attributes include the dimensions and accessibility of the parking space, ensuring that the vehicle can determine if it can fit and comply with any accessibility requirements. The Improved “YOLOV5-OB” algorithm described by Chen et al. (2023), introduces an enhanced algorithm for detecting parking spaces. By encapsulating the properties of a detected parking spot such as type, dimensions, and accessibility, this class is suited for this algorithm.

Class: SensorManager

- Attributes:
 sensorsData: Queue
- Operations:
 gatherData()
 processData()

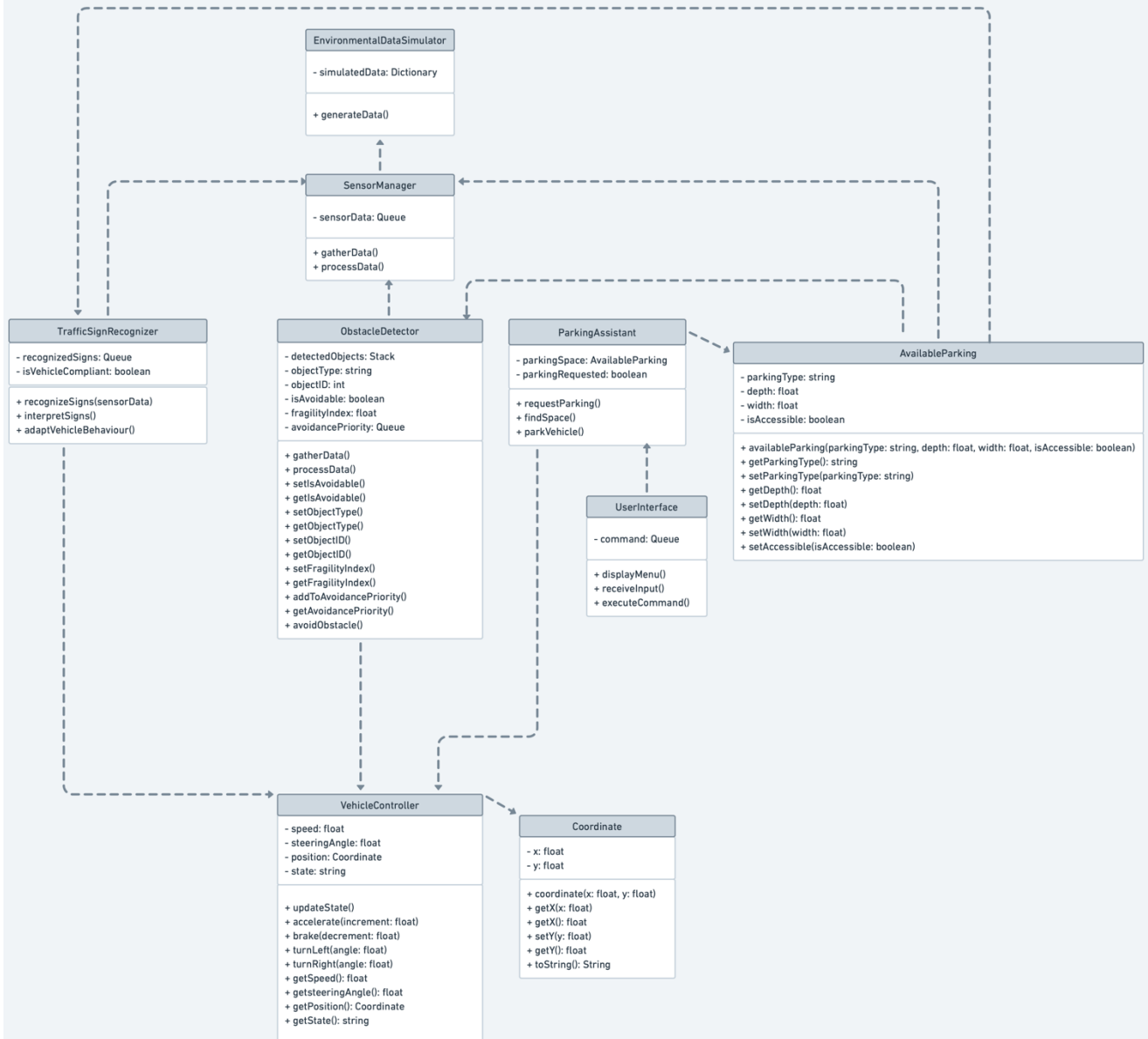
The SensorManager class Manages the data collected from sensors. Using a queue (First In, First Out) ensures that data is processed in the order it's received, which is crucial for real-time applications like autonomous driving.

Class: EnvironmentalDataSimulator

- Attributes:
 simulateData: Dictionary
- Operations:
 generateData()

The EnvironmentalDataSimulator class is responsible for simulating environmental data that the driverless car might encounter, for test purposes, which is crucial in the context of driverless cars, as highlighted by Zhou and Sun (2019). This class also facilitates the implementation of the “safety score” as proposed by Zhao et al. (2020), by generating simulated environmental data and ensuring the autonomous vehicle's computing system can make safe, timely decisions in varied real-world scenarios.

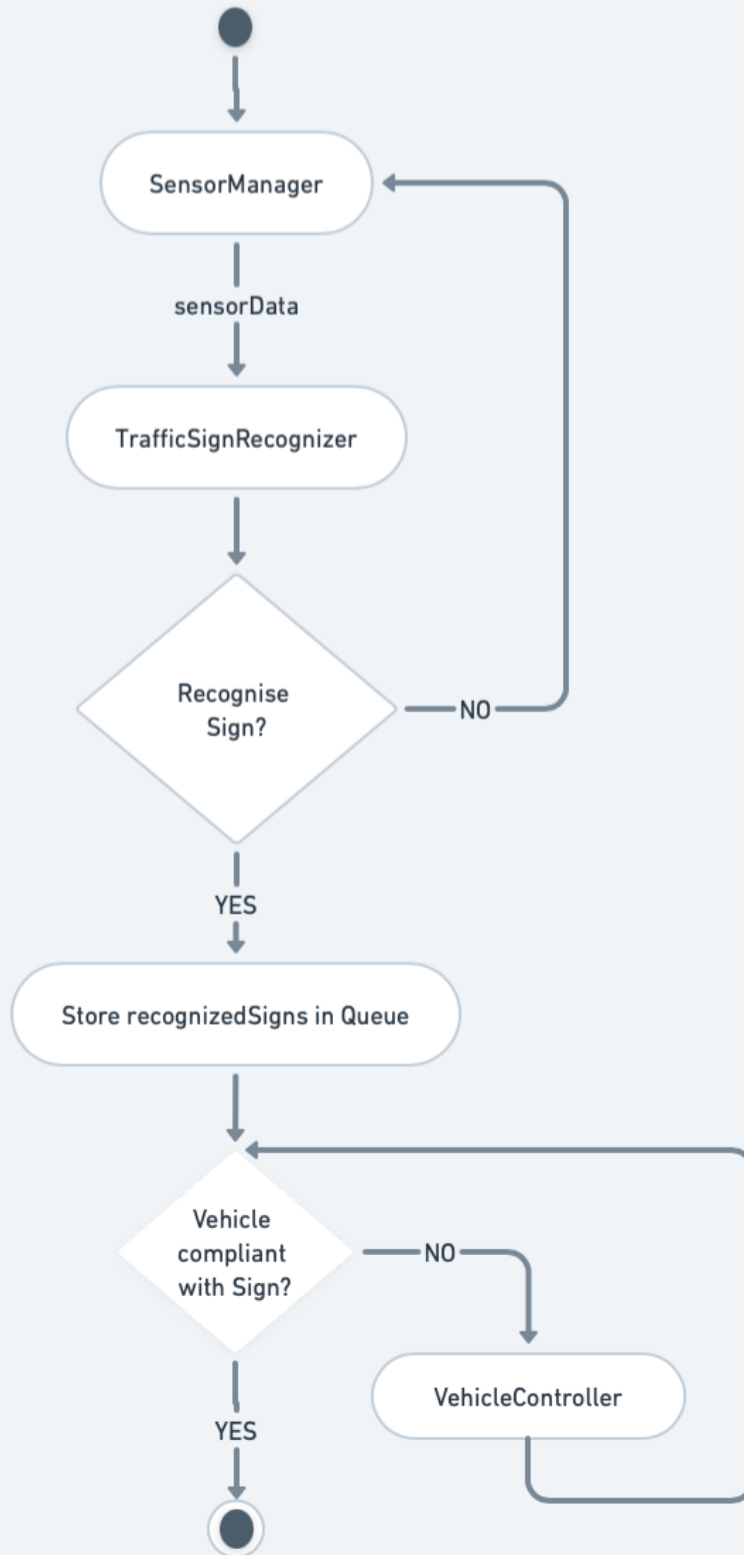
Class Diagram



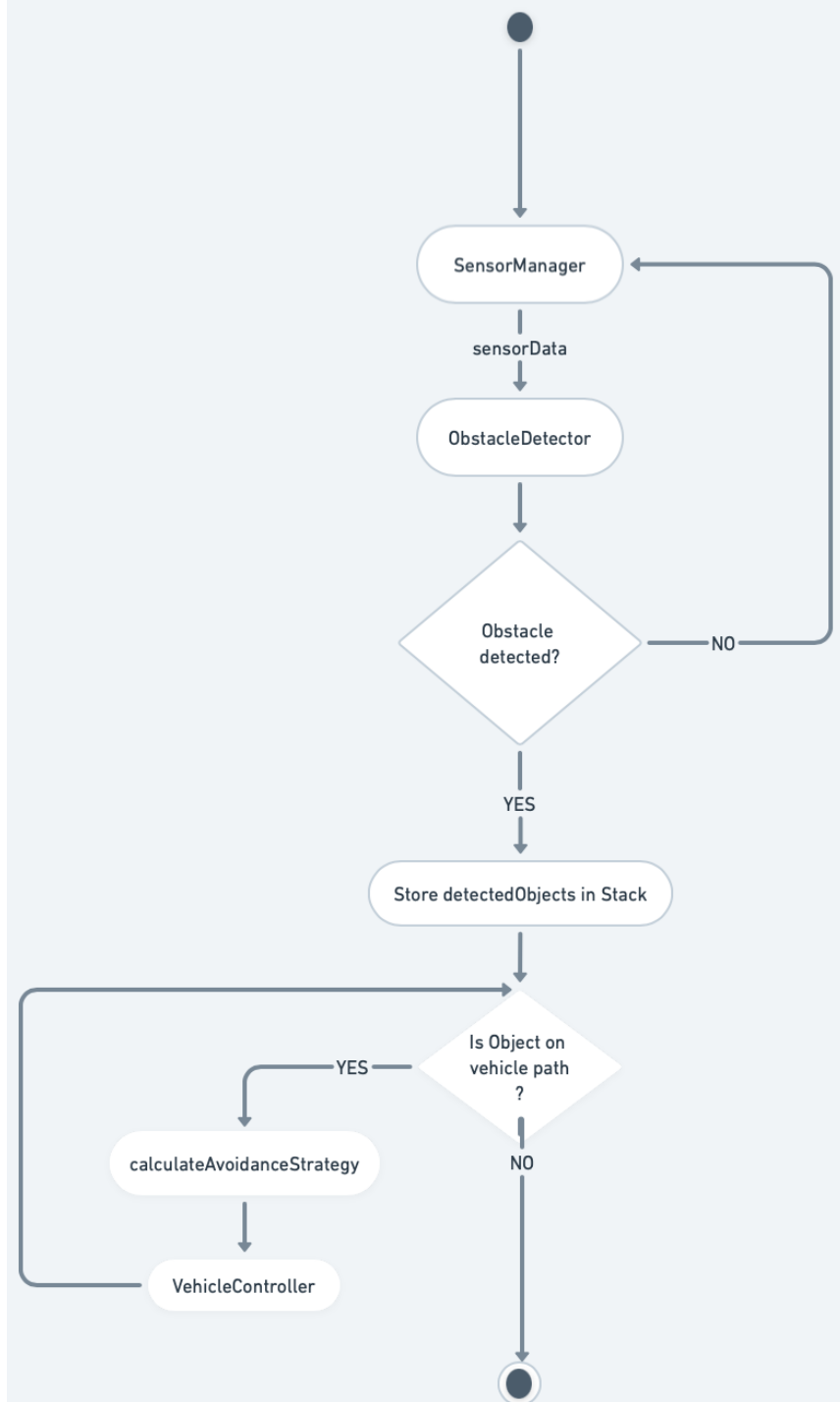
Activity Diagram - Automated Parking



Activity Diagram - Traffic Sign Recognition

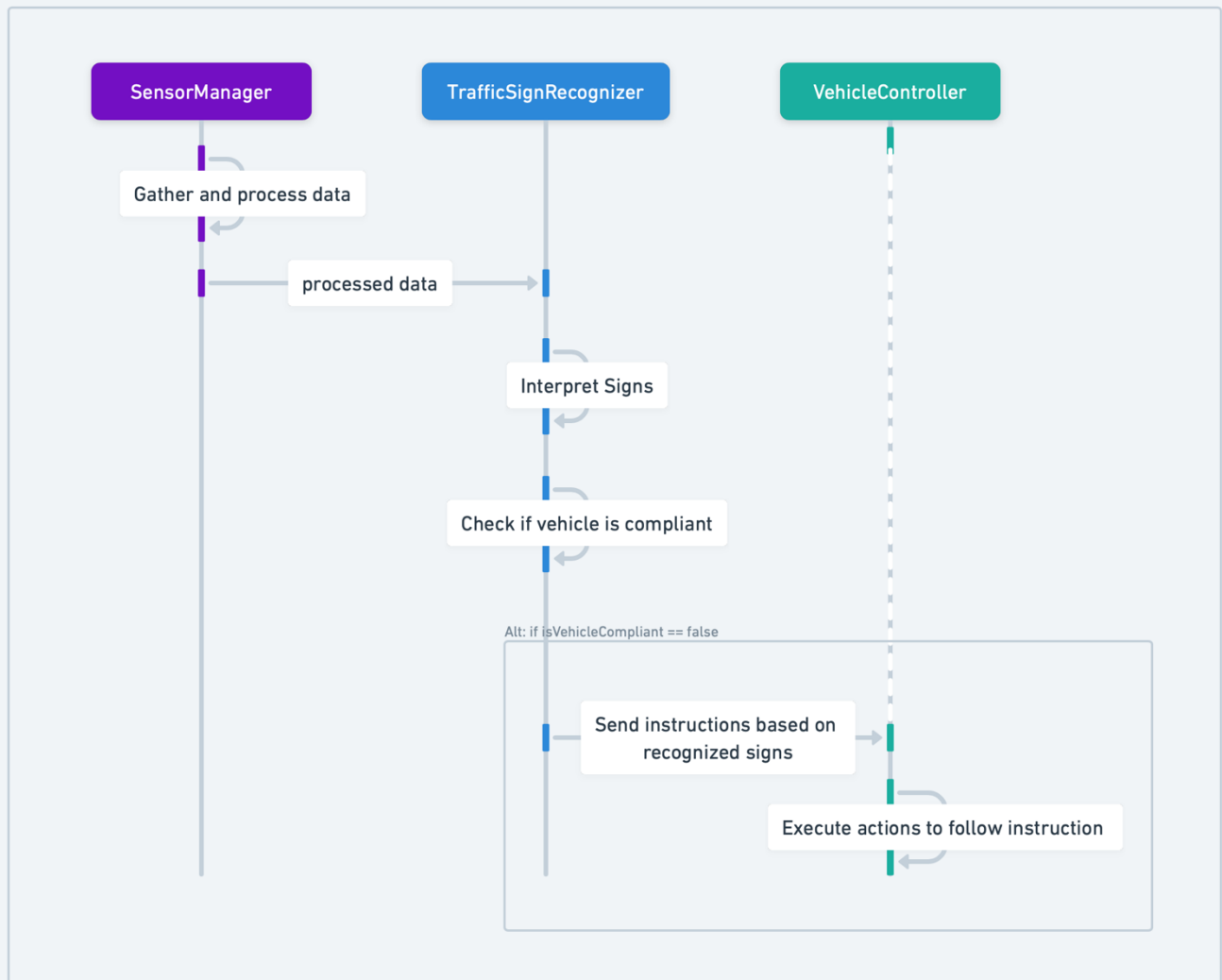


Activity Diagram - Obstacle Detection and Avoidance

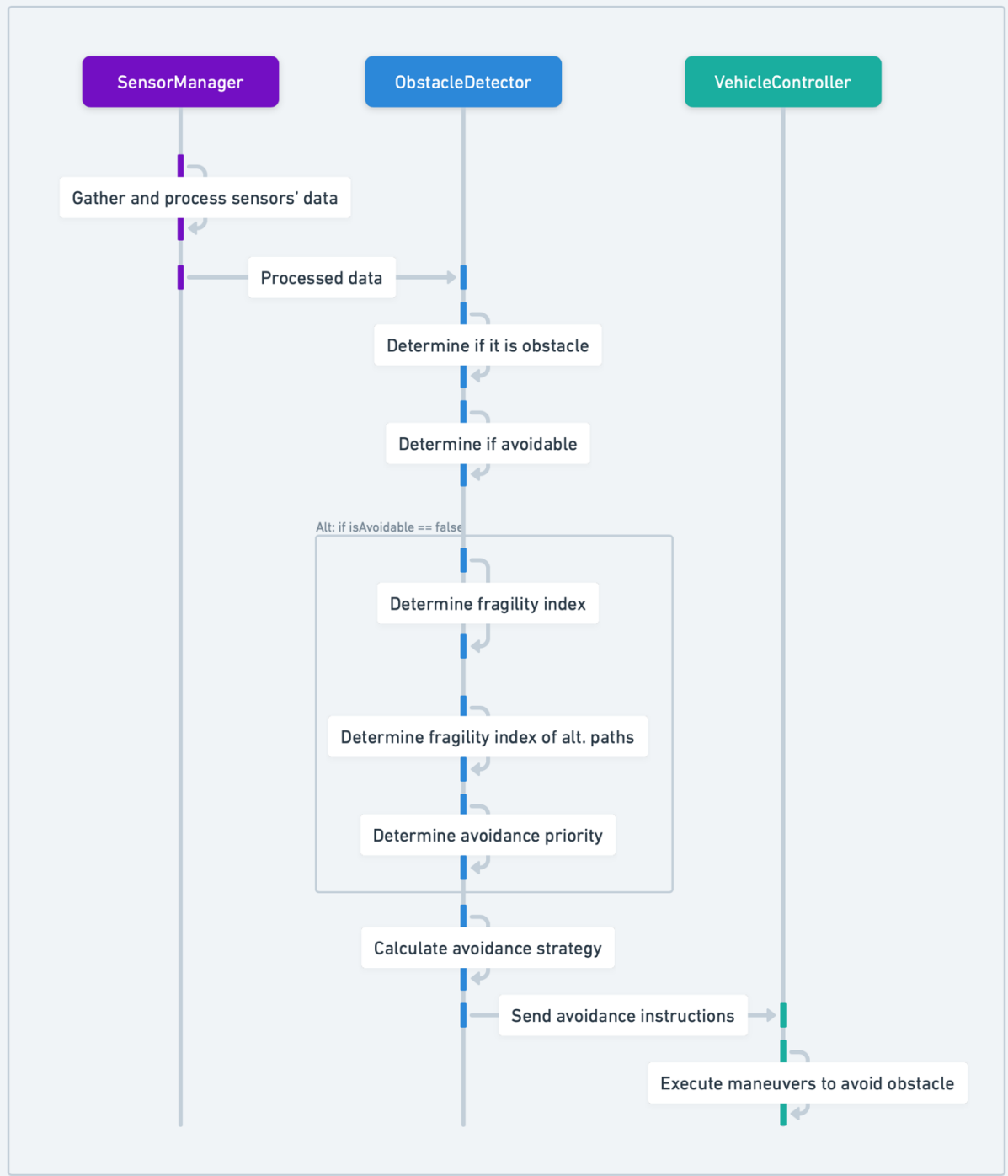


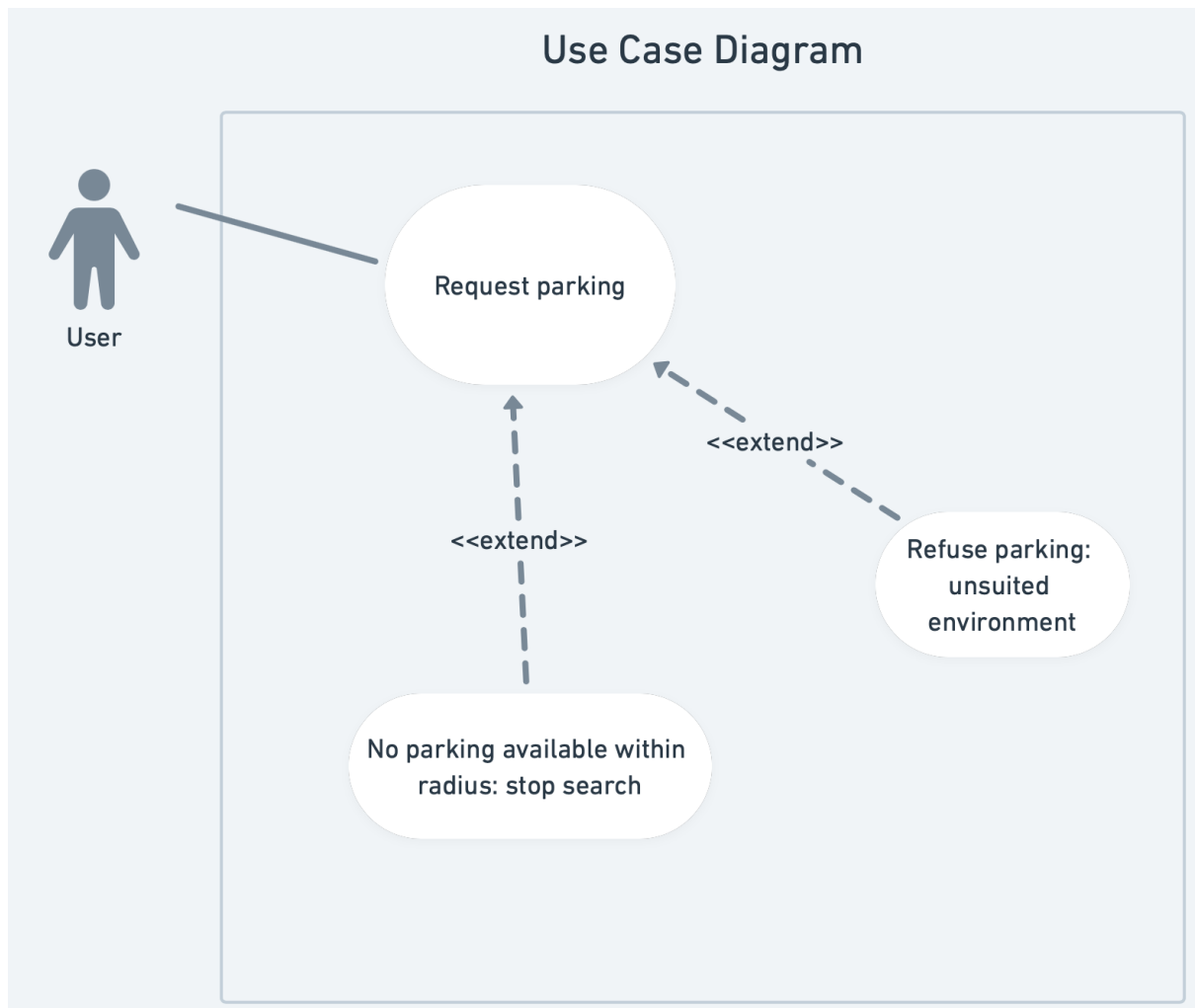


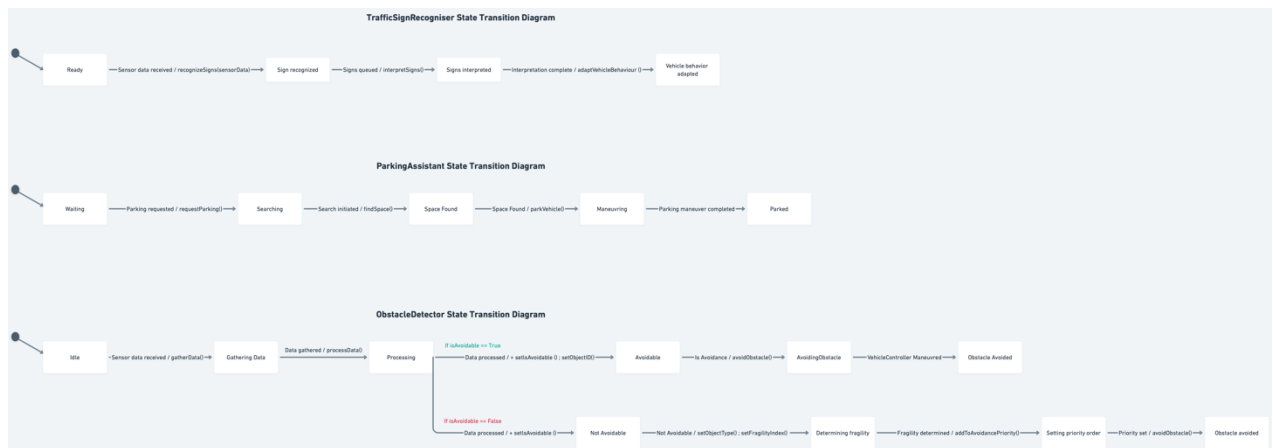
Traffic Sign Recognition Sequence Diagram



Obstacle Sequence Diagram







Reference list

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