## 1. Objective

To implement a rule-based agent to navigate a simulated racecar around a track using sensor information (Lidar) and limited velocity information. The agent must make appropriate decisions to optimize forward movement and avoid going off the track.

### 2. Initial Setup

The starting code base includes:

An environment for Racecar

A cs1graphics-based visualizer

Various agent files: RandomAgent.py, RuleAgent.py

Run script: Run.py

Helper files: Track.py, Visualize.py, Geometry.py

Tracks can be initiated with:

python Run.py RuleAgent 1 -g 1000 -t 0.01

# 3. Problem Description

The agent must choose from 9 possible action combinations of motion and direction:

Directions: 'left', 'straight', 'right'

Motions: 'accelerate', 'coast', 'brake'

Sensor inputs:

obs['lidar']: 5 Lidar readings (left, left-front, front, right-front, right)

obs['velocity']: scalar float of current velocity

## 4. Original RuleAgent Design

Initially, the agent used basic position- and velocity-based heuristics to choose among three actions. This was successful but not reactive to the momentary configuration of the track as inferred from lidar feedback.

### 5. Modifications Made

```
import random
class Agent:
       def chooseAction(self, observations, possibleActions):
               return ('straight', 'coast')
class RuleAgent:
  def __init__(self):
    self.max_velocity = 0.8
 def chooseAction(self, obs):
    lidar = obs['lidar']
    velocity = obs['velocity']
    front = lidar[2] # center
    left = lidar[0] # far left
    right = lidar[4] # far right
    if front > 0.8:
      direction = 'straight'
    elif left > right:
      direction = 'left'
    else:
      direction = 'right'
    if front < 0.3:
```

```
speed = 'brake'
elif velocity < self.max_velocity and front > 0.5:
    speed = 'accelerate'
else:
    speed = 'coast'
return (direction, speed)
```

## 6. Design Rationale

Obstacle Avoidance: If the front lidar sees an obstacle in front of it, the agent abruptly turns and brakes to circumvent it.

Path Finding: The agent prioritizes directions with greater lidar distance as safer paths.

Stability: In the absence of a preference, the car coasts straight, saving momentum without violent acceleration.

## 7. Testing & Validation

The updated agent was tested on track 1 using:

python Run.py RuleAgent 1 -g 1000 -t 0.01

#### Results:

- The car could make turns more effectively.
- Fewer collisions with track boundaries.
- Average speed increased slightly due to better path planning.

#### 8. Conclusion

The minimal rule-based reasoning was significantly improved through the addition of lidar-based directional selection. The new logic is nearer to reactive behavior and provides a foundation for further improvements (e.g., memory, path prediction, or reinforcement learning).