



AWS DEEPRACER

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[Get Started](#)



AWS DEEPRACER

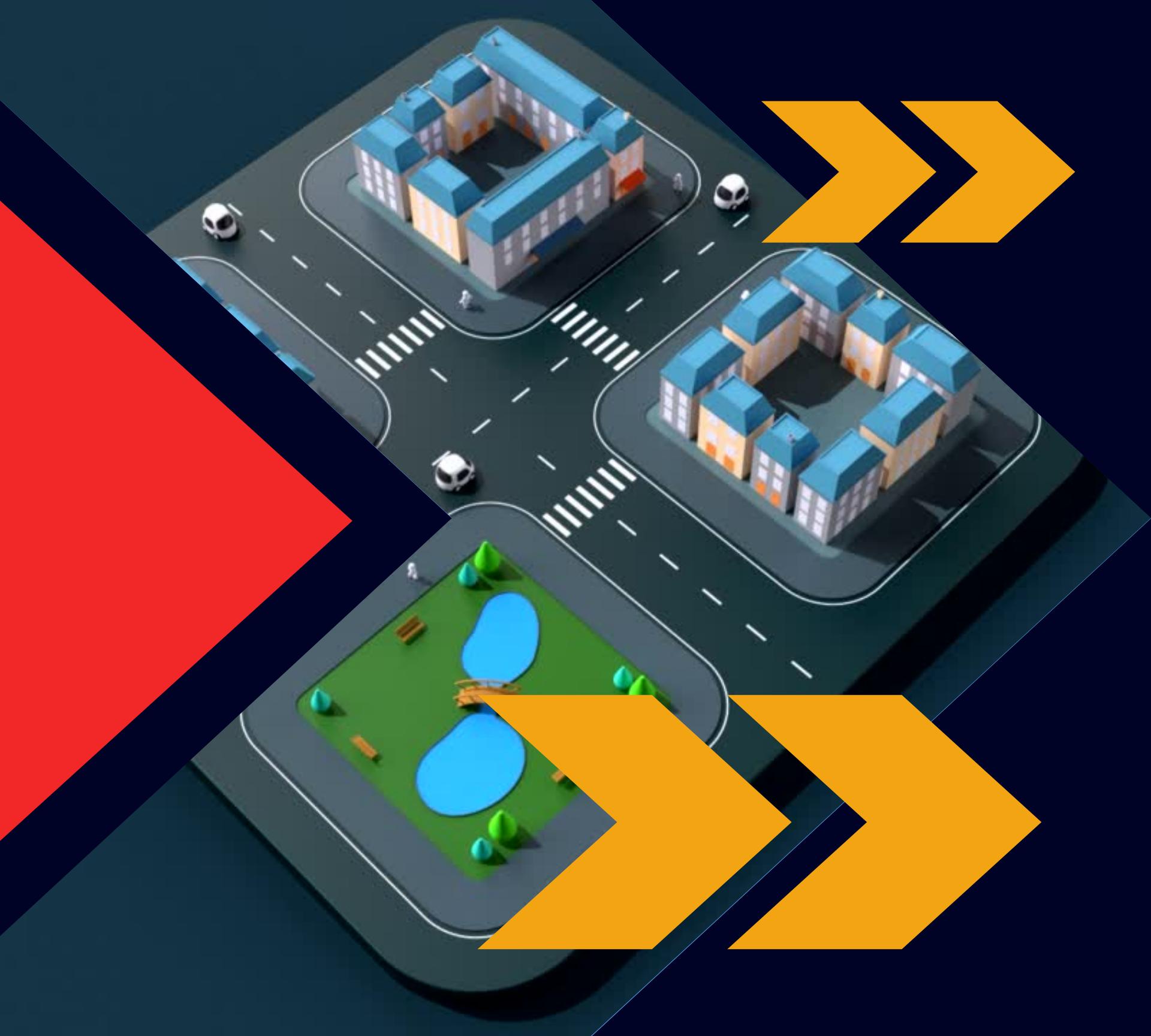
- A fun way to get started with reinforcement learning
- Building and training reinforcement learning model with virtual autonomous car racing
- Compete with autonomous racing league to race for prizes

Simulator

League

Speed





REINFORCEMENT LEARNING

- A type of machine learning technique ideal for autonomous driving
- Learning the optimal behavior in an environment to obtain maximum reward





REWARD FUNCTION

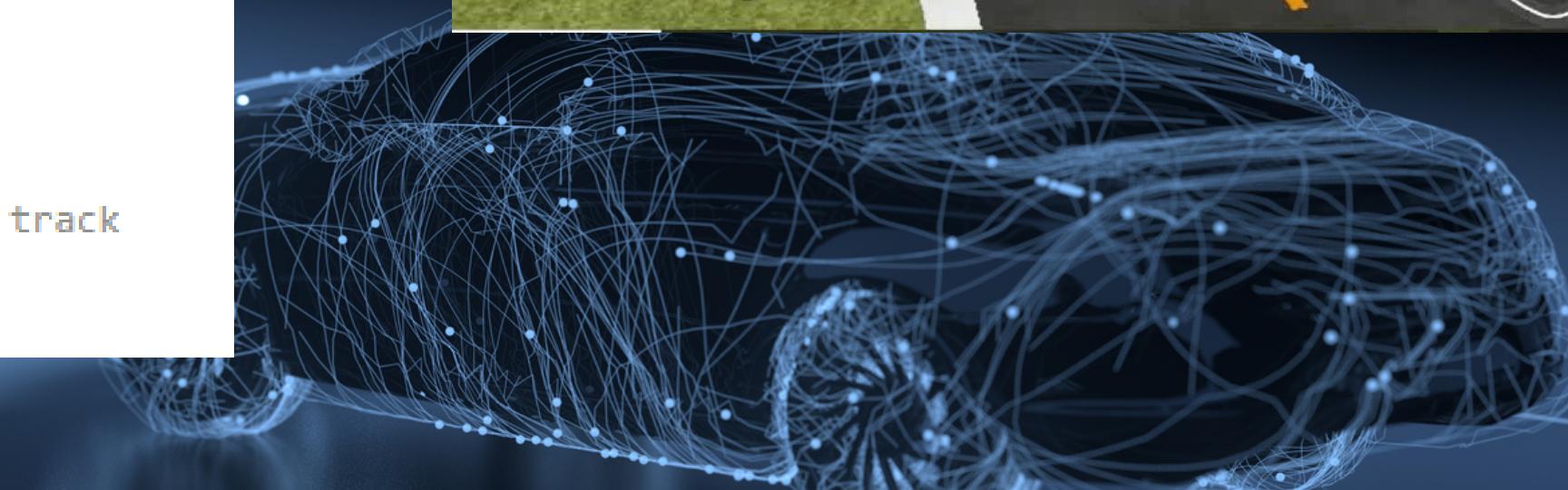
- The core of reinforcement learning.
- Provides a numerical score based on the state of the environment.
- Incentivize our virtual car to take specific actions to finish a lap as fast as possible



OUR FIRST MODEL

Rewarding the agent to follow center line

```
1 def reward_function(params):
2     """
3         Example of rewarding the agent to follow center line
4     """
5
6     # Read input parameters
7     track_width = params['track_width']
8     distance_from_center = params['distance_from_center']
9
10    # Calculate 3 markers that are at varying distances away from
11    # the center line
12    marker_1 = 0.1 * track_width
13    marker_2 = 0.25 * track_width
14    marker_3 = 0.5 * track_width
15
16    # Give higher reward if the car is closer to center line and
17    # penalize for being too far away
18    if distance_from_center <= marker_1:
19        reward = 1.0
20    elif distance_from_center <= marker_2:
21        reward = 0.5
22    elif distance_from_center <= marker_3:
23        reward = 0.1
24    else:
25        reward = 1e-3 # likely crashed/ close to off track
26
27    return float(reward)
```



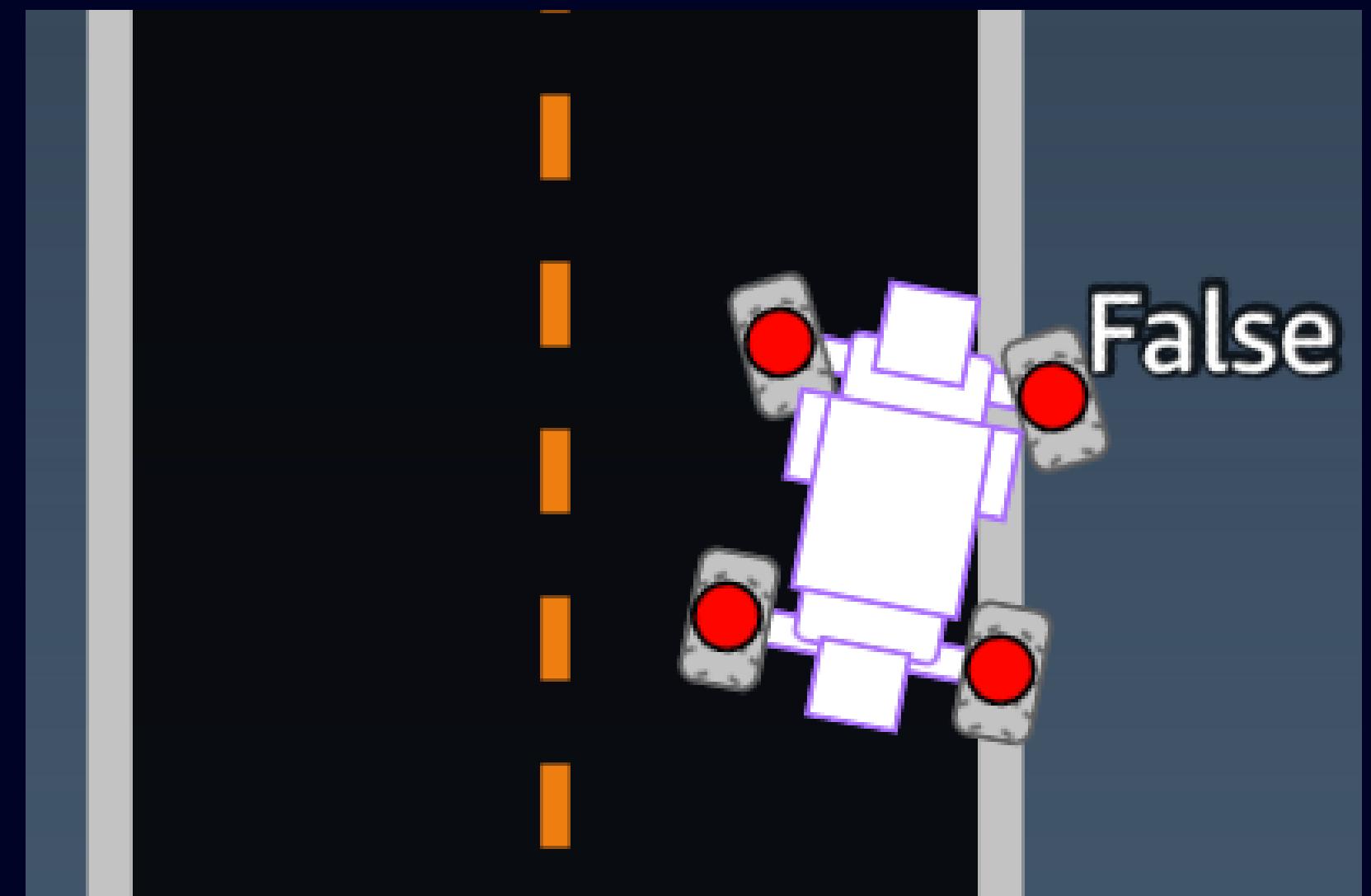
ALL_WHEELS_ON_TRACK

Type: Boolean

Variable:

```
all_wheels_on_track = params['all_wheels_on_track']
```

```
# Give higher reward if the car is closer to center line and vice versa
if all_wheels_on_track and distance_from_center <= marker_1:
    reward = 1.0
elif all_wheels_on_track and distance_from_center <= marker_2:
    reward = 0.5
elif distance_from_center <= marker_3:
    reward = 0.1
else:
    reward = 1e-3 # likely crashed/ close to off track
```



STEERING_ANGLE

Type: float

Range: -30:30

Variable:

```
abs_steering = abs(params['steering_angle'])
```

```
# Steering penalty threshold, change the number if you want
ABS_STEERING_THRESHOLD = 15

# Penalize reward if the car is steering too much
if abs_steering > ABS_STEERING_THRESHOLD:
    reward *= 0.8
```



STEPS

Type: int

Variable:

```
steps = params['steps']
progress = params['progress']
```

```
# Total num of steps we want the car to finish the lap, it will vary depending on the track
TOTAL_NUM_STEPS = 300

# Initialize the reward with typical value
reward = 1.0

# Give additional reward if the car pass every 100 steps faster than expected
if (steps % 100) == 0 and progress > (steps / TOTAL_NUM_STEPS) * 100 :
    reward += 10.0
```

OUR FIRST MODEL

Rewarding the agent to follow center line

Evaluation results			
Trial	Time (MM:SS.mmm)	Trial results (% track completed)	Status
1	00:20.057	47%	Off track
2	00:20.474	47%	Off track
3	00:26.942	62%	Off track

WAYPOINTS

Type: list

Variable:

```
waypoints = params['waypoints']
closest_waypoints = params['closest_waypoints']
heading = params['heading']
```



CLOSEST_WAYPOINTS

```
waypoints = params['waypoints']
closest_waypoints = params['closest_waypoints']
heading = params['heading']
```

```
# Calculate the direction of the center line based on the closest waypoints
next_point = waypoints[closest_waypoints[1]]
prev_point = waypoints[closest_waypoints[0]]

# Calculate the direction in radius, arctan2(dy, dx), the result is (-pi, pi) in radians
track_direction = math.atan2(next_point[1] - prev_point[1], next_point[0] - prev_point[0])
# Convert to degree
track_direction = math.degrees(track_direction)

# Calculate the difference between the track direction and the heading direction of the car
direction_diff = abs(track_direction - heading)
if direction_diff > 180:
    direction_diff = 360 - direction_diff

# Penalize the reward if the difference is too large
DIRECTION_THRESHOLD = 10.0
if direction_diff > DIRECTION_THRESHOLD:
    reward *= 0.5
```



PROGRESS

Type: float

Range: 0:100

Variable:

```
progress = params['progress']
is_offtrack=params['is_offtrack']
```

```
#Get reward if completes the lap
if progress == 100:
    reward += 100
elif is_offtrack:
    reward-=50
```

EVALUATION

Evaluation results			
Trial	Time (MM:SS.mmm)	Trial results (% track completed)	Status
1	00:50.254	100%	Lap complete
2	00:31.937	63%	Off track
3	00:51.005	100%	Lap complete



5 MARKRS

```
# Calculate 5 markers that are at varying distances away from the center line
marker_1 = 0.01 * track_width
marker_2 = 0.1 * track_width
marker_3 = 0.2 * track_width
marker_4 = 0.3 * track_width
marker_5 = 0.5 * track_width

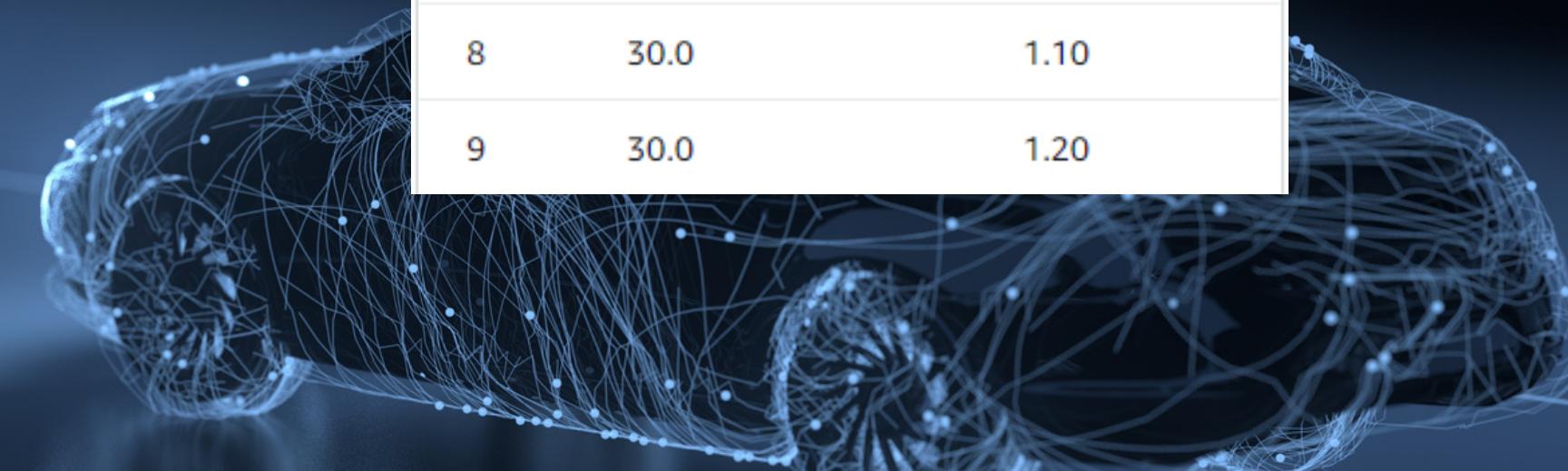
# Give higher reward if the car is closer to center line and vice versa
if all_wheels_on_track and distance_from_center <= marker_1:
    reward = 1.0
elif all_wheels_on_track and distance_from_center <= marker_2:
    reward = 0.5
elif distance_from_center <= marker_3:
    reward = 0.2
elif distance_from_center <= marker_4:
    reward = 0.1
elif distance_from_center <= marker_5:
    reward = 0.001
else:
    reward = 1e-3 # likely crashed/ close to off track
```

INCREASE ACTION SPACE SPEED

Action space		
No.	Steering angle (°)	Speed (m/s)
0	-30.0	1.25
1	-30.0	2.50
2	-15.0	1.25
3	-15.0	2.50
4	0.0	1.25
5	0.0	2.50
6	15.0	1.25
7	15.0	2.50
8	30.0	1.25
9	30.0	2.50



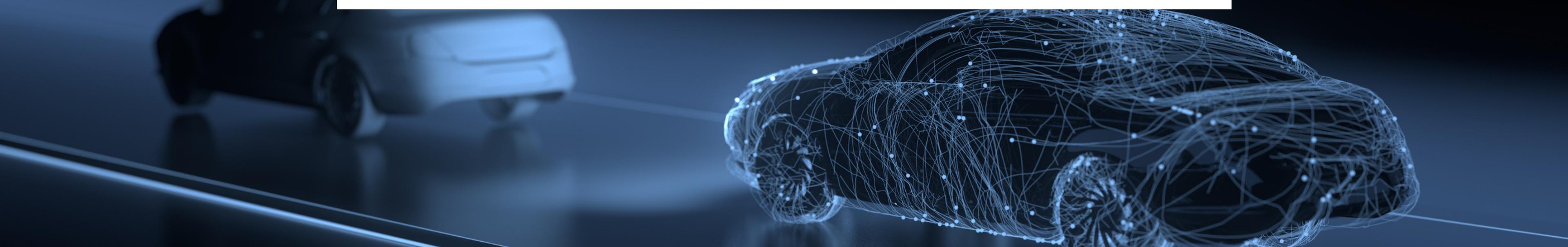
Action space		
No.	Steering angle (°)	Speed (m/s)
0	-30.0	1.10
1	-30.0	1.20
2	-15.0	1.50
3	-15.0	1.90
4	0.0	3.00
5	0.0	2.90
6	15.0	1.70
7	15.0	2.00
8	30.0	1.10
9	30.0	1.20



EVALUATION

Evaluation results

Trial	Time (MM:SS.mmm)	Trial results (% track completed)	Status
1	00:45.129	100%	Lap complete
2	00:44.004	100%	Lap complete
3	00:28.728	64%	Off track



PENALIZE REWARD

Variable:

```
speed = params['speed']
SPEED_THRESHOLD = 1.5
```

```
# Penalize reward if the car goes too slow
if speed < SPEED_THRESHOLD:
    reward *= 0.8
```

EVALUATION

Evaluation results			
Trial	Time (MM:SS.mmm)	Trial results (% track completed)	Status
1	00:26.195	62%	Off track
2	00:42.800	100%	Lap complete
3	00:42.602	100%	Lap complete

INCREASE ACTION SPACE SPEED & INCREASE SPEED THRESHOLD

Action space		
No.	Steering angle (°)	Speed (m/s)
0	-30.0	1.10
1	-30.0	1.10
2	-15.0	2.00
3	-15.0	2.50
4	0.0	3.60
5	0.0	4.00
6	15.0	2.00
7	15.0	2.50
8	30.0	1.10
9	30.0	1.20

```
# Set the speed threshold based your action space  
SPEED_THRESHOLD = 2.0
```

REWARD FUNCTION

```
8 import math
9
10 def reward_function(params):
11     """
12     Example of rewarding the agent to follow center line
13     """
14
15     # Read input parameters
16     all_wheels_on_track = params['all_wheels_on_track']
17     speed = params['speed']
18
19     track_width = params['track_width']
20     distance_from_center = params['distance_from_center']
21     steps = params['steps']
22     progress = params['progress']
23     is_offtrack=params['is_offtrack']
24
25     abs_steering = abs(params['steering_angle']) # Only need the absolute steering angle
26
27     waypoints = params['waypoints']
28     closest_waypoints = params['closest_waypoints']
29     heading = params['heading']
30
31     # Total num of steps we want the car to finish the lap, it will vary depends on the track length
32     TOTAL_NUM_STEPS = 800
33
34     # Initialize the reward with typical value
35     reward = 1.0
36
37     # Set the speed threshold based your action space
38     SPEED_THRESHOLD = 2.0
39
40     #Get reward if completes the lap
41     if progress == 100:
42         reward += 100
43     elif is_offtrack:
44         reward-=50
45
46     # Calculate the direction of the center line based on the closest waypoints
47     next_point = waypoints[closest_waypoints[1]]
48     prev_point = waypoints[closest_waypoints[0]]
49
50     # Calculate the direction in radius, arctan2(dy, dx), the result is (-pi, pi) in radians
51     track_direction = math.atan2(next_point[1] - prev_point[1], next_point[0] - prev_point[0])
52     # Convert to degree
53     track_direction = math.degrees(track_direction)
54
```

```
55     # Calculate the difference between the track direction and the heading direction of the car
56     direction_diff = abs(track_direction - heading)
57     if direction_diff > 180:
58         direction_diff = 360 - direction_diff
59
60     # Penalize the reward if the difference is too large
61     DIRECTION_THRESHOLD = 10.0
62     if direction_diff > DIRECTION_THRESHOLD:
63         reward *= 0.5
64
65     # Give additional reward if the car pass every 100 steps faster than expected
66     if (steps % 100) == 0 and progress > (steps / TOTAL_NUM_STEPS) * 100 :
67         reward += 10.0
68
69     # Calculate 5 markers that are at varying distances away from the center line
70     marker_1 = 0.01 * track_width
71     marker_2 = 0.1 * track_width
72     marker_3 = 0.2 * track_width
73     marker_4 = 0.3 * track_width
74     marker_5 = 0.5 * track_width
75
76     # Give higher reward if the car is closer to center line and vice versa
77     if all_wheels_on_track and distance_from_center <= marker_1:
78         reward = 1.0
79     elif all_wheels_on_track and distance_from_center <= marker_2:
80         reward = 0.5
81     elif distance_from_center <= marker_3:
82         reward = 0.2
83     elif distance_from_center <= marker_4:
84         reward = 0.1
85     elif distance_from_center <= marker_5:
86         reward = 0.001
87     else:
88         reward = 1e-3 # likely crashed/ close to off track
89
90     # Steering penalty threshold, change the number based on your action space setting
91     ABS_STEERING_THRESHOLD = 15
92
93     # Penalize reward if the car is steering too much
94     if abs_steering > ABS_STEERING_THRESHOLD:
95         reward *= 0.8
96
97     # Penalize reward if the car goes too slow
98     if speed < SPEED_THRESHOLD:
99         reward *= 0.8
100
101 return float(reward)
```

FINAL MODEL

- Lap complete X 2
- Best time(MM:SS.mmm):
00:40.539



EVALUATION

Evaluation results

Trial	Time (MM:SS.mmm)	Trial results (% track completed)	Status
1	00:41.397	100%	Lap complete
2	00:27.139	65%	Off track
3	00:40.539	100%	Lap complete

THANK YOU

