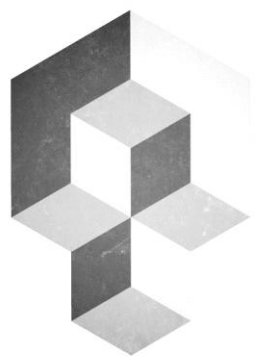


Deep Racing 개요



* 시뮬레이션(ex. Angle Setting in Action Space) 실물 차량이 필요한 경우는 제외.



Course2 - Deep Racer Component



1. 4MP Camera
2. Car Chassis
3. Battery
4. Power Bank
5. Battery Charger
6. Full Size HDMI PORT (Side)
7. Micro USB PORT(Side)
8. Micro USB-C PORT(Side)

* Lidar and Camera (New Model)

*15 pictures in 1second → STEP



Course2 - Deep Racer Component



1. 32gigs of storage(= 32GB)
2. 4GB RAM
3. LED



Course2 – Setting in Action space

Maximum steering angle

degrees

Max values are between 1 and 30.

Steering angle granularity

▼

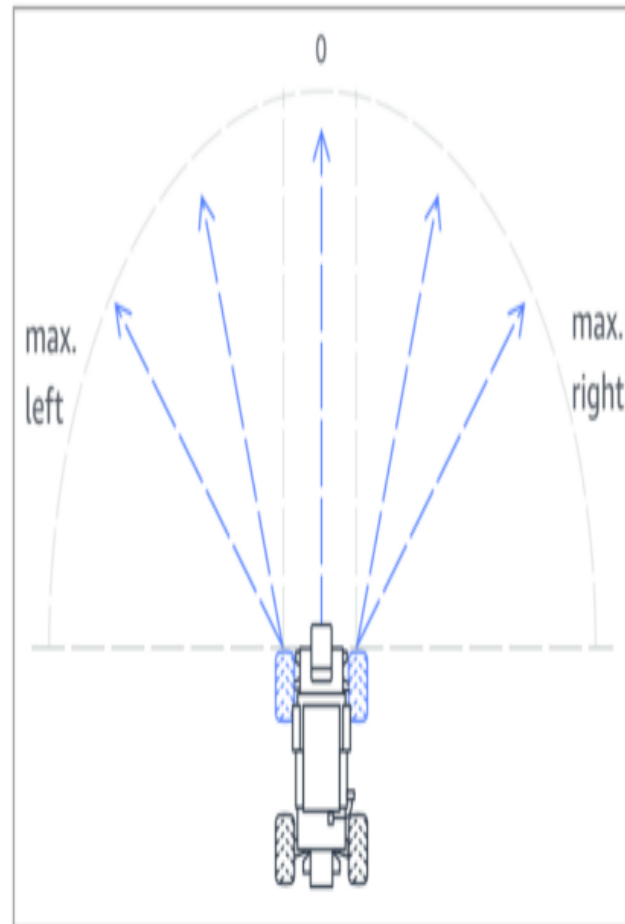
Maximum speed

m/s

Max values are between 0.8 and 8.

Speed granularity

▼



1. Maximum_steering_angle
→ 0~30 degree

2. Steering_granularity

3. Maximum_speed
→ 0.8m/sec ~ 8.0m/sec

4. Speed_granularity
→ 1,2,3 (select)

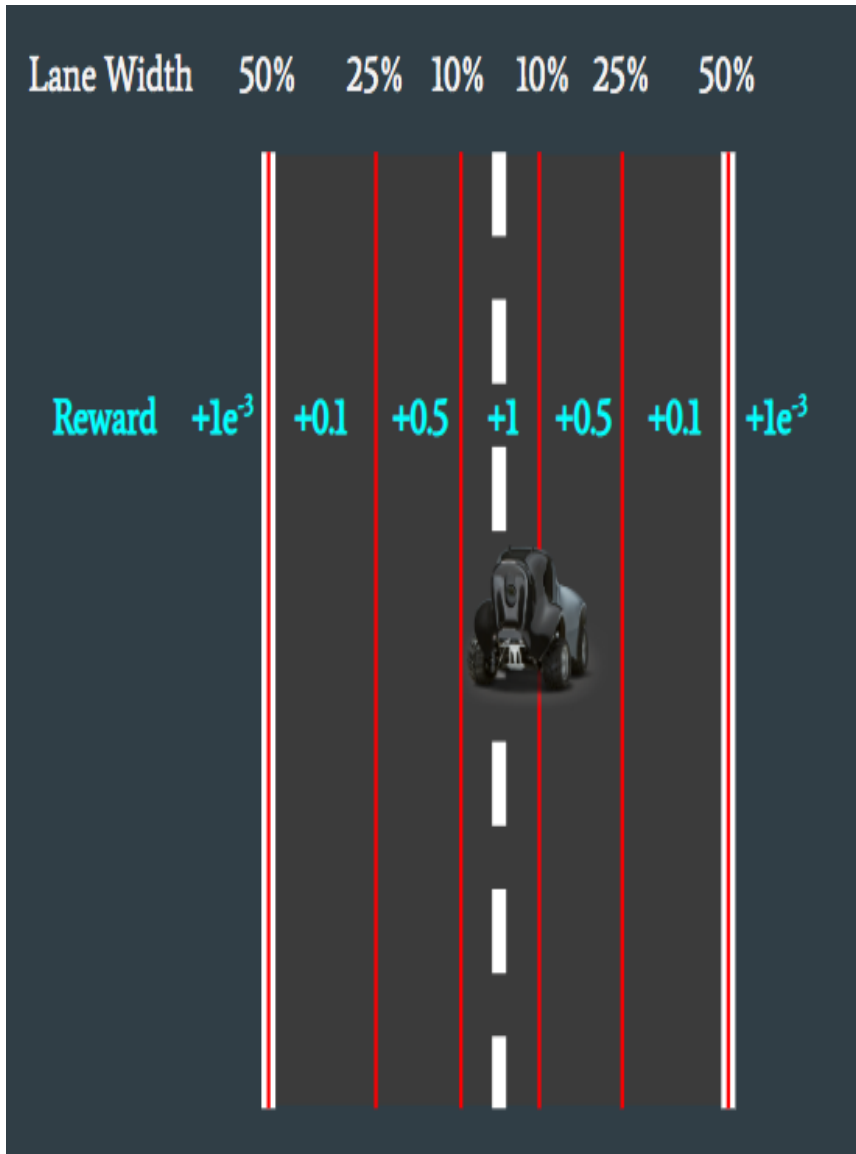
* **Action Number** = steering_granularity * speed_granularity

Deep Racer Assembly, Steering Calibration, Throttle Calibration

<https://classroom.udacity.com/courses/ud014/lessons/336b89e3-f4e8-4ec5-951c-dd67da63a089/concepts/50f3e7f8-778c-47d6-b424-8d3a4abd6074>



Course3 – Reward Function – distance from Center



```
def reward_function(params):  
    '''  
    Example of rewarding the agent to follow center line  
    '''  
  
    # Read input parameters  
    track_width = params['track_width']  
    distance_from_center = params['distance_from_center']  
  
    # Calculate 3 markers that are at varying distances away from the center line  
    marker_1 = 0.1 * track_width  
    marker_2 = 0.25 * track_width  
    marker_3 = 0.5 * track_width  
  
    # Give higher reward if the car is closer to center line and vice versa  
    if distance_from_center <= marker_1:  
        reward = 1.0  
    elif 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  
  
    return float(reward)
```



Course3 – Reward Function - Example



Track Width = 100m
Distance_from_Center = 23m

```
marker_1 = 0.1 * track_width  
marker_2 = 0.25 * track_width  
marker_3 = 0.5 * track_width
```

marker_1 = 10m
marker_2 = 25m
marker_3 = 50m

```
elif distance_from_center <= marker_2:  
    reward = 0.5
```

* In Track Reward Function, Not Zig-Zag Reward Function



Course4 - Reinforcement learning



1. Agent
2. Action
3. Environment
4. State
5. Reward



1. Deep Racer
2. Driving behavior(Turn, Accelerate)
3. Track
4. Point on the track at a given time
5. how far from the center of the track
(In basic Reward Function)



Course4 - Reinforcement learning

- * Step(tuple) → state, action, reward, new state
- * Episodes → 2types(Finish or not) , Set of steps
- * Experience buffer → Set of steps in several episodes.
- * Policy
- * Value Function

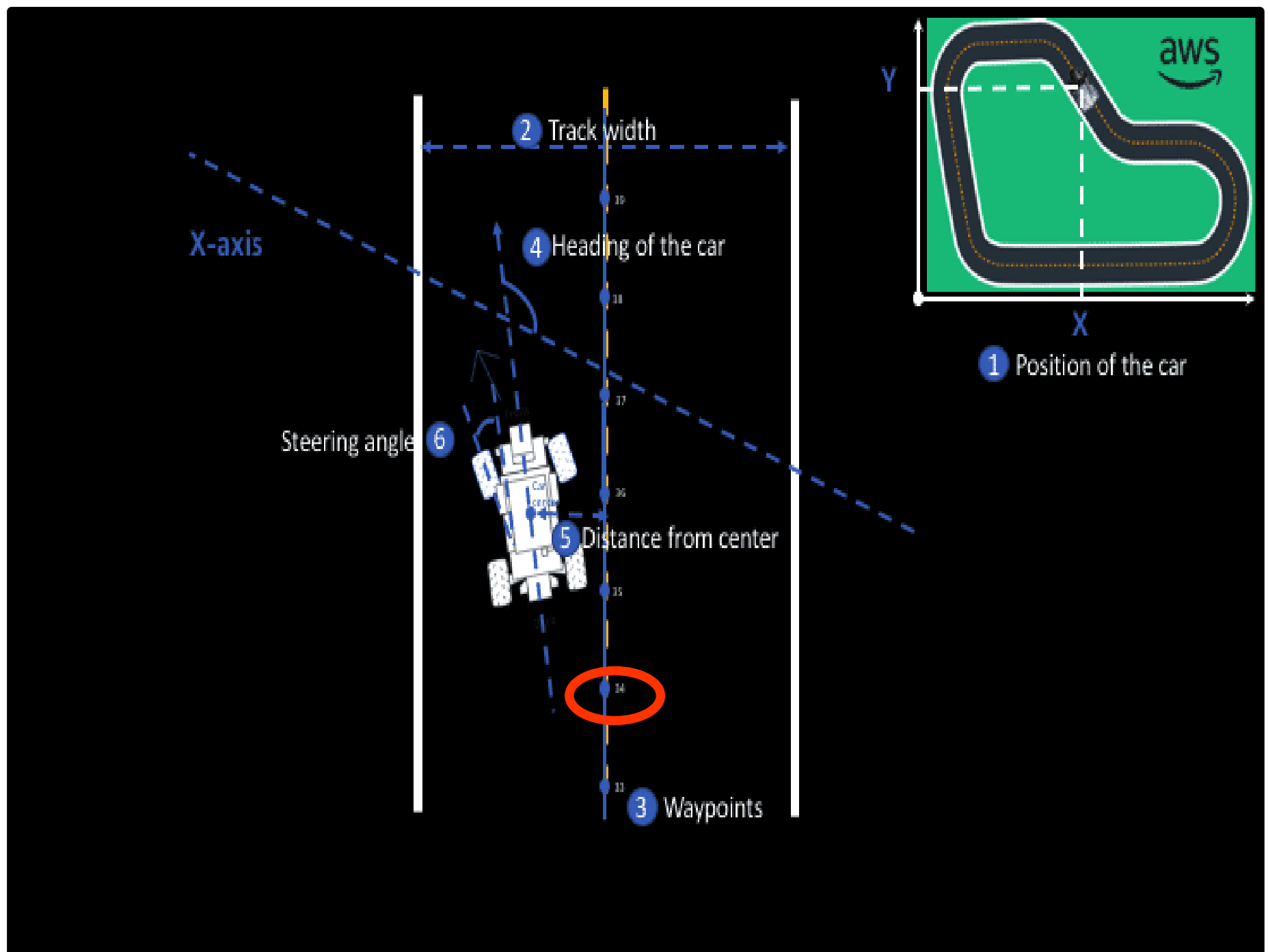


Course5 – parameter

1. all_wheels_on_track (Bool)
2. x,y (float)
3. distacne_from_center(float)
4. is_left_of_center(Bool)
5. is_reversed(Bool)
6. Heading (float) → -180 ~ 180 degree
7. progress(float) → 0 ~ 100
8. Steps (intger)
9. Speed(float)
10. Steering angle(float) → + : going left , - : going right
11. track_width(float) → unit in meter
12. **waypoints(List)**
13. **closet waypoints(int)**



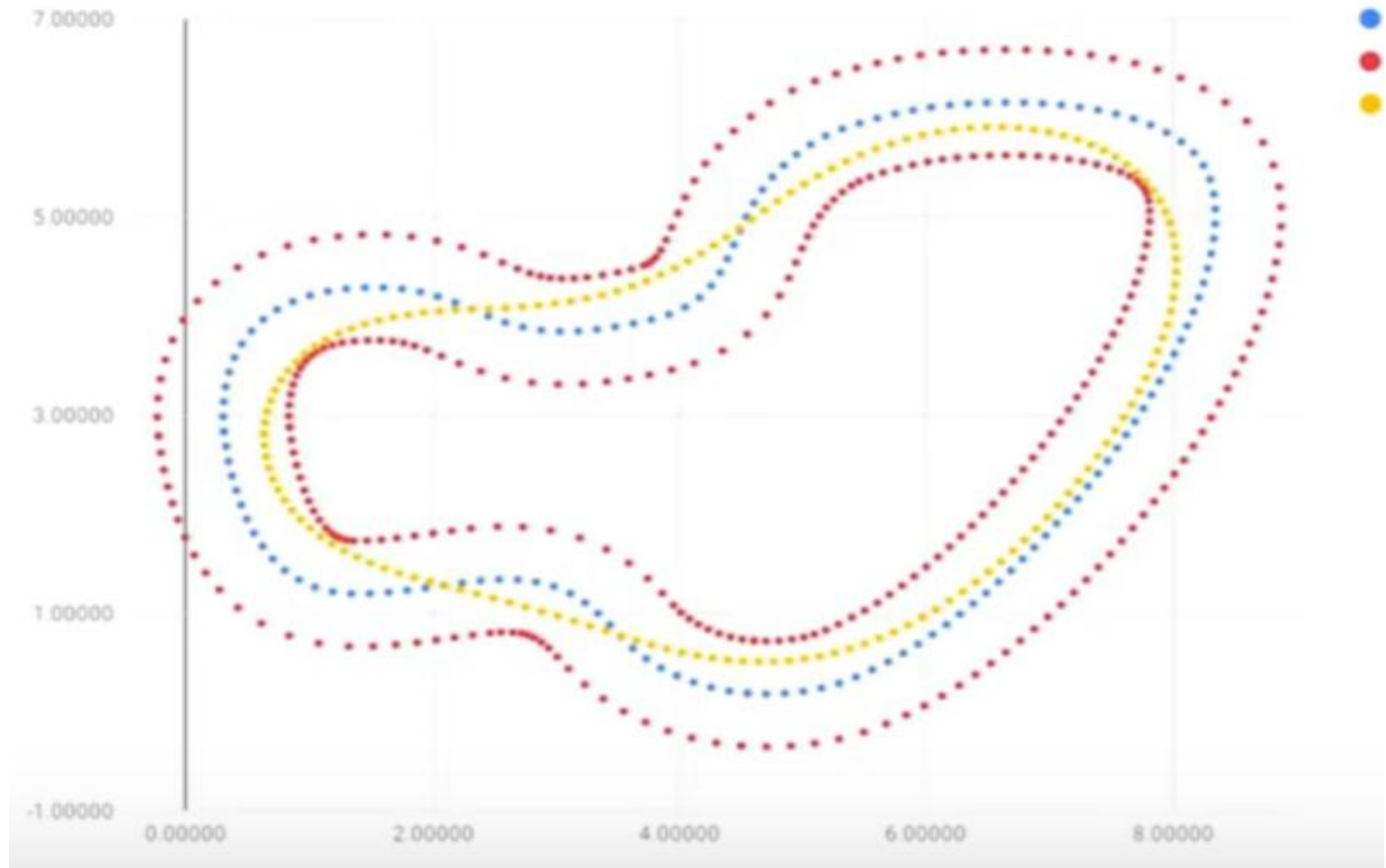
Course5 – parameter



<https://classroom.udacity.com/courses/ud014/lessons/0afdbee8-cf7a-4966-b8ed-f860f3d68cb6/concepts/4208eda1-1094-4e66-94d2-9450a7afc6f9>



Course5 – Waypoint





Course5 – Hyperparameter

- * Batch Size
- * Number of epochs
- * LEARNING RATE
- * EXPLORATION
- * ENTROPY
- * DISCOUNT FACTOR
- * LOSS TYPE
- * EPISODES



Course5 – Hyperparameter

Hyperparameter	Value
Gradient descent batch size	64
Entropy	0.01
Discount factor	0.888
Loss type	Huber
Learning rate	0.0003
Number of experience episodes between each policy-updating iteration	20
Number of epochs	10

* Course6

ROS(Deep Racing을 위해 필수적으로 공부해야할 부분은 아닐 것으로 보임),
DeepRacer를 직접 Track에서 돌리는 방법을 설명.(실물이 있어야 한다고 판단)



Conclusion

1. 시뮬레이터 사용 여부 판단 필요.
2. 반복적인 시뮬레이션이 요구됨.
3. 어떤 식으로 Reward를 주는 것이 좋을 지 생각.
4. Waypoint의 활용방안 찾기
5. HyperParameter는 초기의 설정을 유지하는 것이 좋음.
만약, 미세한 조정이 필요하다고 느껴질 시 , 그 때 수정.