

DeepRacer: The Blind Leads the Blind

David Darling and Ivy Truong

A dark blue, diagonal shape that starts from the bottom left and extends towards the top right, covering the bottom right portion of the slide.

Project Statement:

- Expand knowledge with Machine Learning
- Optimize reward function to provide consistency and versatility



Methods

- Keep DeepRacer driving on track
- Reward DeepRacer for making progress to complete a lap on the track
- Maintain a minimum speed for DeepRacer
- Continue testing model if reward function continues to improve over time
- Testing DeepRacer on a different track

Experiment, Results, and Demo

Reward Function

```
1 # V8-3
2
3 def reward_function(params):
4
5     # Read input parameters
6     track_width = params['track_width']
7     distance_from_center = params['distance_from_center']
8     progress = params['progress']
9     speed = params['speed']
10    all_wheels_on_track = params['all_wheels_on_track']
11
12    # Calculate 3 markers that are at varying distances away from the center line
13    marker_1 = 0.1 * track_width
14    marker_2 = 0.25 * track_width
15    marker_3 = 0.5 * track_width
16
17    # Give higher reward if the car is closer to center line and vice versa
18    if distance_from_center <= marker_1:
19        reward = 10.0
20    elif distance_from_center <= marker_2:
21        reward = 5.0
22    elif distance_from_center <= marker_3:
23        reward = 1.0
24    else:
25        reward = 1e-3 # likely crashed/ close to off track
26
27    # Reward the deepracer based on progress made after completing 25% of the track.
28    if progress >= 25:
29        reward = reward + (progress / 100) # 25% -> reward + 0.25 // 60% -> reward + 0.60
```

```
30
31    # Reward the deepracer if it's speed is higher than 1.0 m/s, otherwise penalize it
32    if speed > 0.65:
33        reward += 0.5
34    else:
35        reward -= 0.5
36
37    # Check if all wheels are on track, if they are reward deepracer, if not give penalty
38    if all_wheels_on_track:
39        reward += 1.0
40    else:
41        reward -= 3.0
42
43    # Check if reward is 0 or below and if so make reward a small decimal value
44    if reward <= 0.0:
45        reward = 1e-6
46
47    return float(reward)
```

Hyperparameters

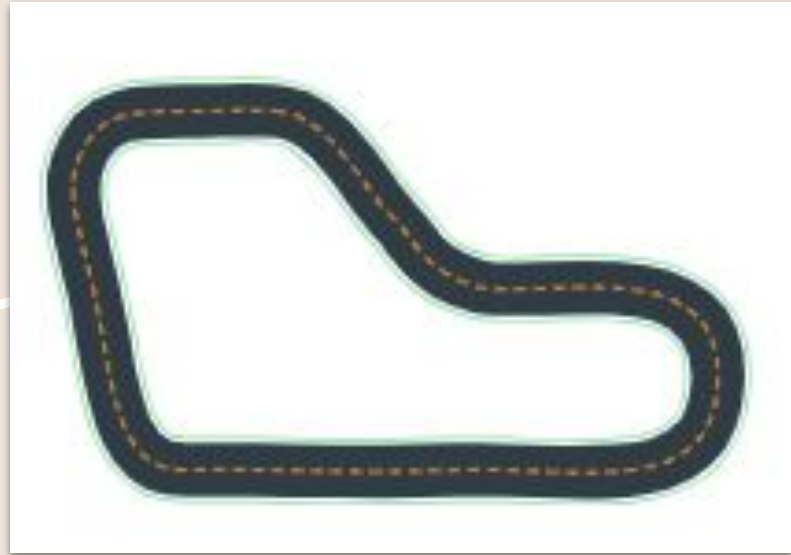
Hyperparameter	Value
Gradient descent batch size	64
Entropy	0.01
Discount factor	0.999
Loss type	Mean squared error
Learning rate	0.0003
Number of experience episodes between each policy-updating iteration	20
Number of epochs	10

reInvent: 2018

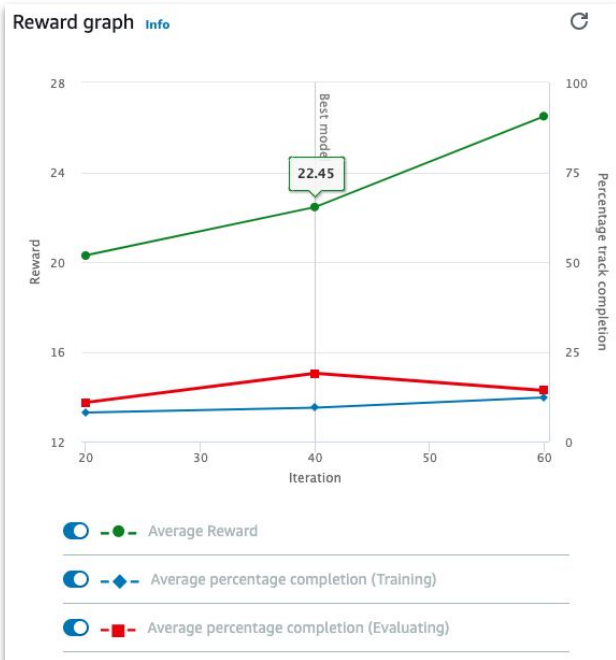
Track Info

Length: 17.6 m

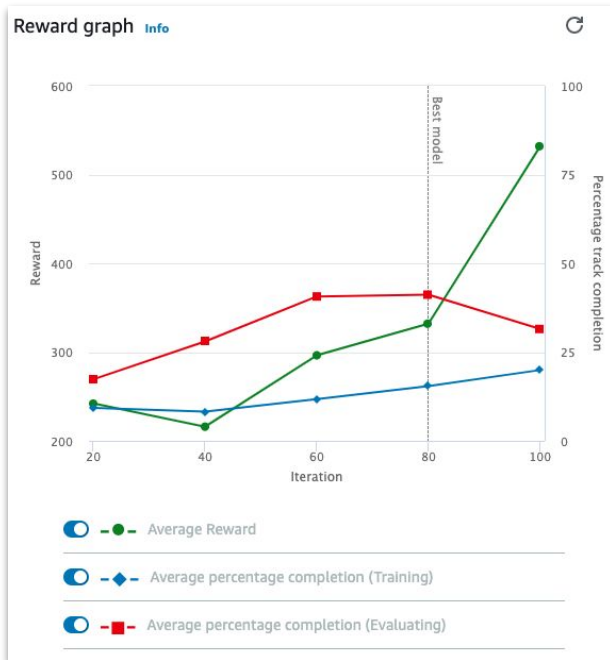
Width: 76 cm



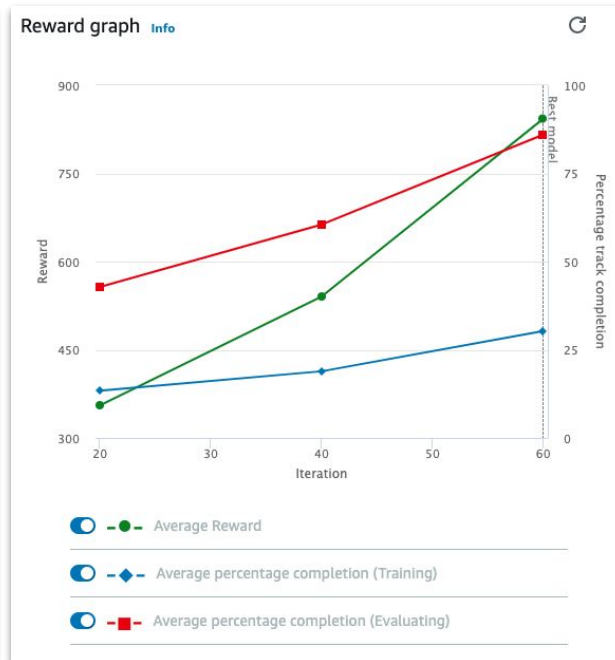
Training Reward Graphs



V2

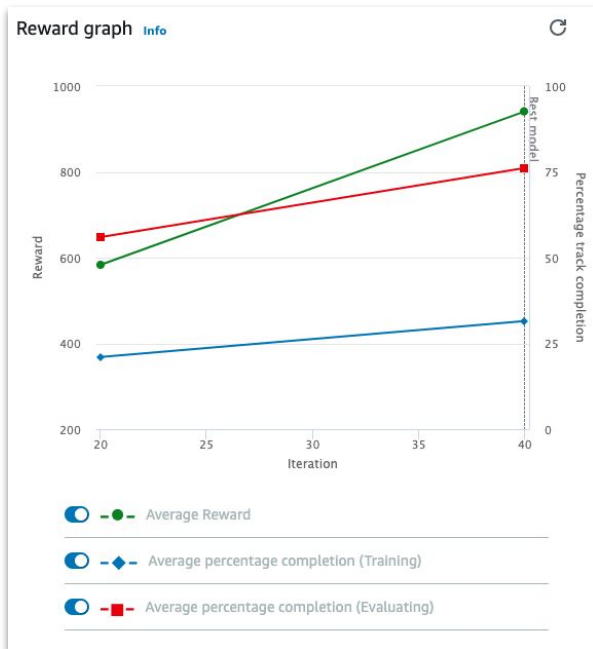


V8



V8-1

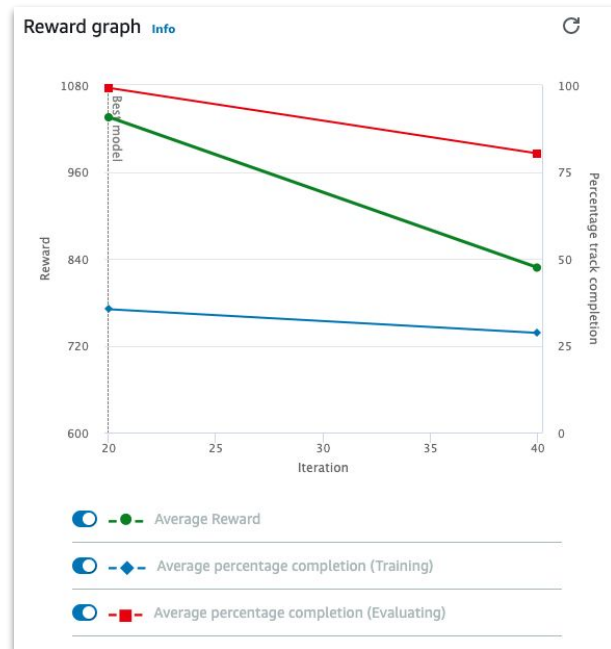
Training Reward Graphs Cont.



V8-2

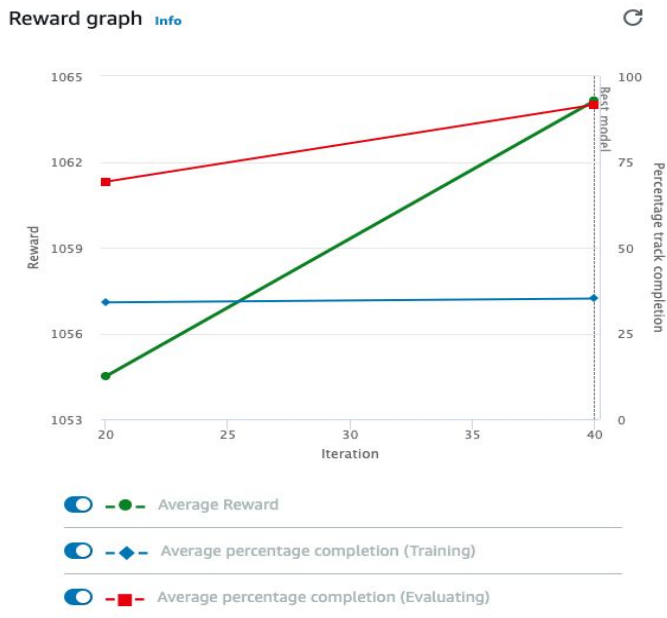


V8-3



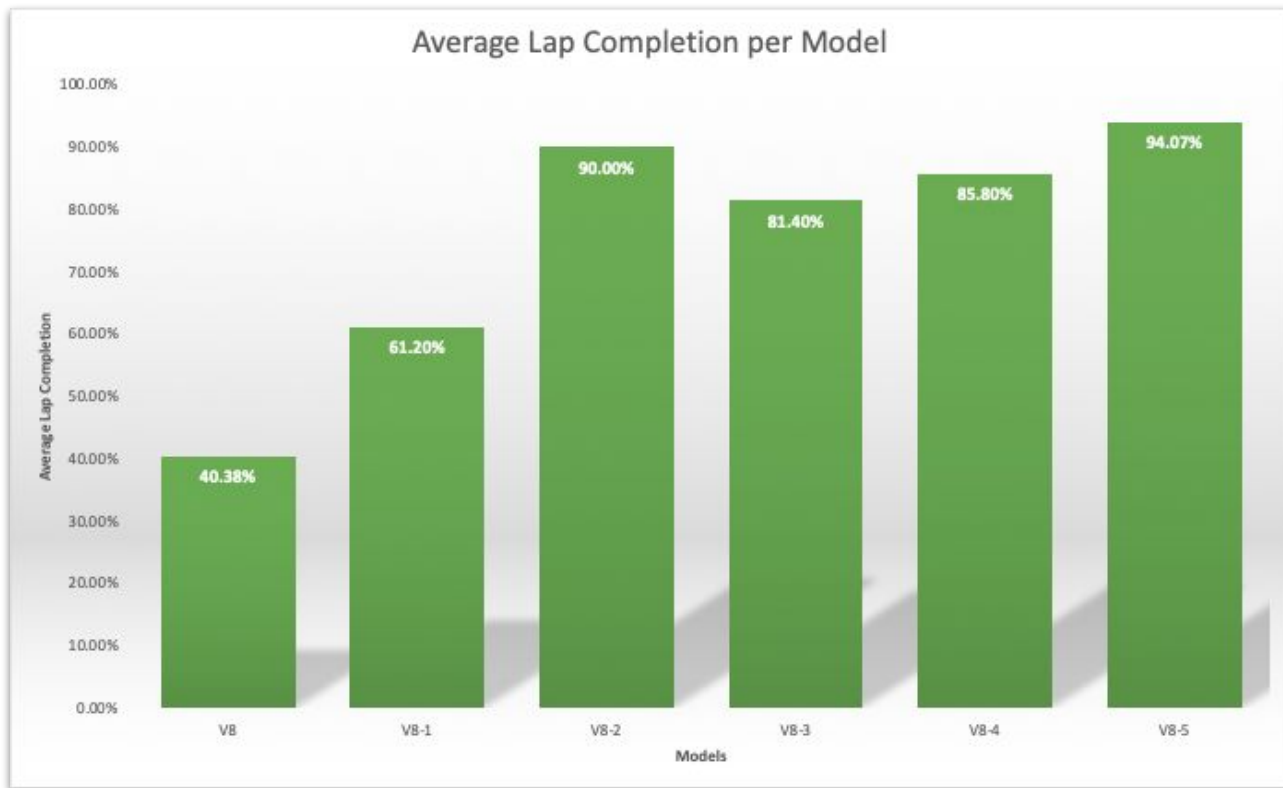
V8-4

Training Reward Graphs Cont.



V8-5

Evaluations



Simulation video stream

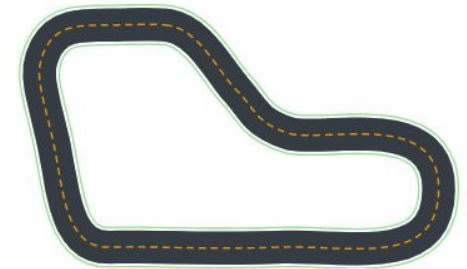


Evaluation results

Trial	Time	Trial results (% track completed)	Status
1	00:00:28.485	100%	Lap complete

- re:Invent 2018
Inspired by Monza, re:Invent 2018 was the first Championship Cup track. This short, classic speedway remains a perennial rookie favorite.

Length: 17.6 m (57.97')
Width: 76 cm (30")



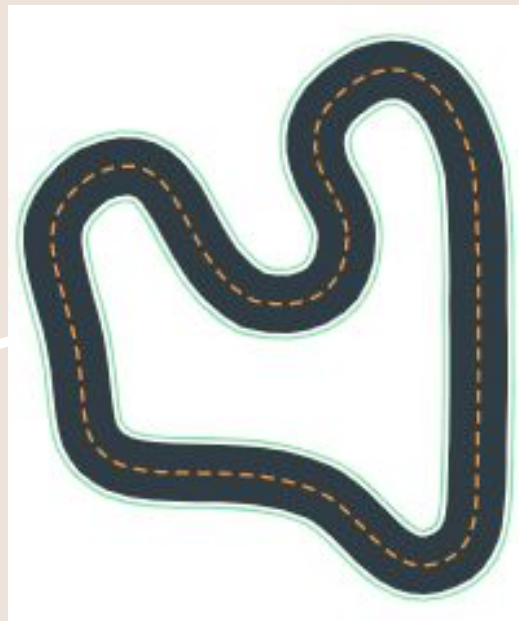
Video of the DeepRacer driving on the re:Invent 2018 track with the most updated reward function.

Shanghai Sudu

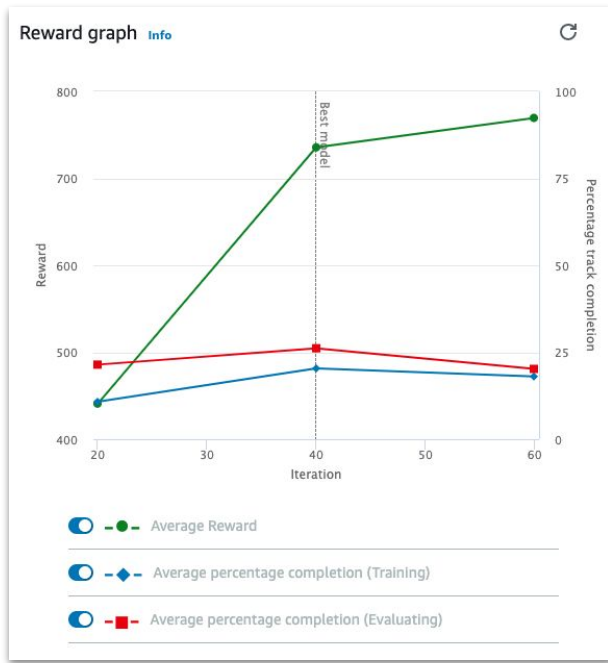
Track Info

Length: 22.92 m

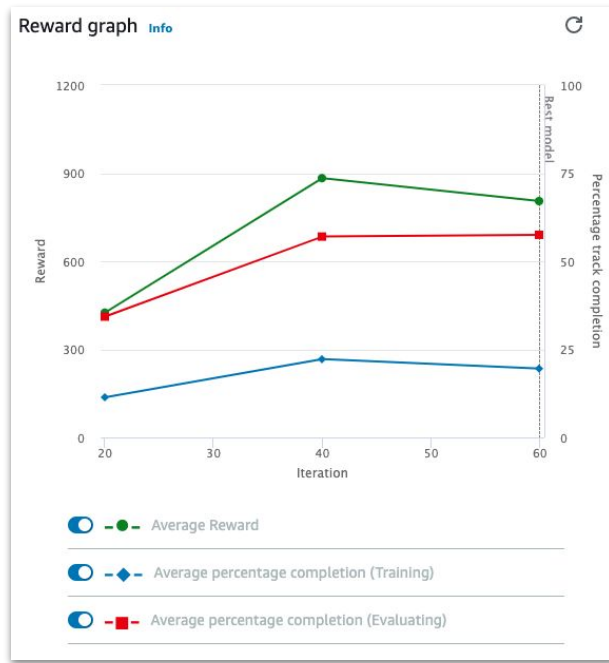
Width: 76 cm



Training Reward Graphs



V8-6

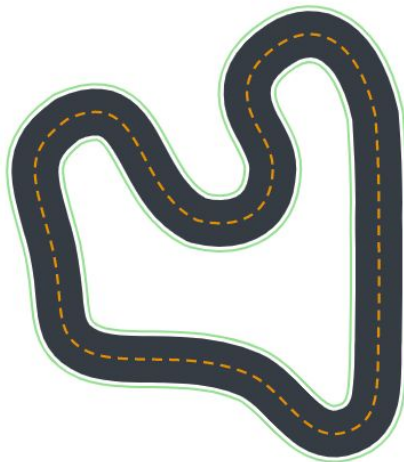


V8-7



- Shanghai Sudu Training
August's training track for the 2019 Virtual Circuit World Tour, the Shanghai Sudu is an ideal speedway for experimenting with bold corners and basic straightaways.

Length: 22.92 m (75.2')
Width: 76 cm (30")



Video of DeepRacer driving on Shanghai Sudu Training track with most updated reward function.

Limitations and Future Works

Limitations

Local Environment

- Difficult to set up
 - Many libraries to set up
 - Deprecated githubs
- Space Requirement
 - Running another OS
 - Linux
 - Storage for data
- Favors certain GFX cards
- Virtualbox

AWS Console

- Flexibility in environment design
 - RL Algorithm
- Cost
- Time

Future Work

- More training time for our model
- Improve speed and lap times
- Try out more parameters
 - Waypoints
 - Steering angle
- Optimizing hyperparameters
- Watch a physical DeepRacer run using our reward function

Questions?

