

# Tennis Game (Multi Agent)

I used DDPG algorithm to train the agent so agent can select best action against each state.

## Why DDPG:

DDPG is an actor-critic, model-free algorithm based on the deterministic policy gradient that can operate over continuous action spaces.

## Reference Used:

1. <https://arxiv.org/abs/1509.02971>
2. <https://arxiv.org/pdf/1509.02971.pdf>

## Summary of DDPG Network:

Used an Actor and Critic Network. As per the guidance of paper use Batch normalization as well. I used Adam optimizer to train the network. Fine tune the hyper parameters than mentioned in paper to converge faster. Use the same network as used in previous project but unlikely previous project it is used to train multiple agents.

- **Summary of Actor Network:**

```
In [12]: from torchsummary import summary
summary(agent.actor_local, (state_size,))
```

C:\ProgramData\Anaconda2\envs\udacity\lib\site-packages\torch\nn\functional.  
ted. Use torch.tanh instead.  
warnings.warn("nn.functional.tanh is deprecated. Use torch.tanh instead.")

Layer (type)	Output Shape	Param #
Linear-1	[-1, 128]	3,200
BatchNorm1d-2	[-1, 128]	256
Linear-3	[-1, 128]	16,512
Linear-4	[-1, 2]	258

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Total params: 20,226  
Trainable params: 20,226  
Non-trainable params: 0  
-----  
Input size (MB): 0.00  
Forward/backward pass size (MB): 0.00  
Params size (MB): 0.08  
Estimated Total Size (MB): 0.08  
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- **Summary of Critic Network:**

```
In [13]: summary(agent.critic_local, [(state_size,), (action_size, )])
```

Layer (type)	Output Shape	Param #
Linear-1	[-1, 128]	3,200
BatchNorm1d-2	[-1, 128]	256
Linear-3	[-1, 128]	16,768
Linear-4	[-1, 1]	129

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Total params: 20,353  
Trainable params: 20,353  
Non-trainable params: 0  
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Input size (MB): 0.00  
Forward/backward pass size (MB): 0.00  
Params size (MB): 0.08  
Estimated Total Size (MB): 0.08  
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## Summary Hyper parameter:

- **Maximum steps per episode:** 1000
- **Replay Buffer size:** 100000
- **Batch Size:** 128
- **Gamma:** 0.99
- **Tau:** 0.001
- **Actor Learning Rate:** 0.0002
- **Critic Learning Rate:** 0.0002
- **Weight Decay:** 0.0

## Rewards Function Performance During Training:

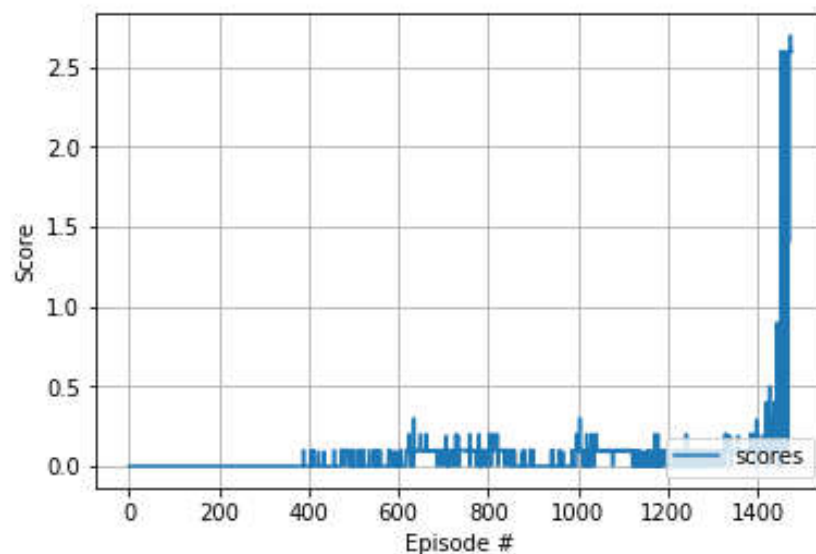
- **Average Score after every 100 episodes:**

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Episode 100	Average Score: 0.0000
Episode 200	Average Score: 0.0000
Episode 300	Average Score: 0.0000
Episode 400	Average Score: 0.0010
Episode 500	Average Score: 0.0260
Episode 600	Average Score: 0.0194
Episode 700	Average Score: 0.0811
Episode 800	Average Score: 0.0882
Episode 900	Average Score: 0.0624
Episode 1000	Average Score: 0.0099
Episode 1100	Average Score: 0.0890
Episode 1200	Average Score: 0.0524
Episode 1300	Average Score: 0.0340
Episode 1400	Average Score: 0.0973
Episode 1475	Average Score: 0.5219

Environment solved in 1475 episodes!      Average Score: 0.5219

- **Plot shows average rewards against each episode:**



### Rewards Function Performance During Prediction:

Episode 92	Average Score: 2.13
Episode 93	Average Score: 2.11
Episode 94	Average Score: 2.11
Episode 95	Average Score: 2.12
Episode 96	Average Score: 2.10
Episode 97	Average Score: 2.10
Episode 98	Average Score: 2.11
Episode 99	Average Score: 2.11
Episode 100	Average Score: 2.12
Average Score Of 100 Consecutive Episodes: 2.1279026303537676	

### Networks want to try in future:

1. I want to try MADDPG.