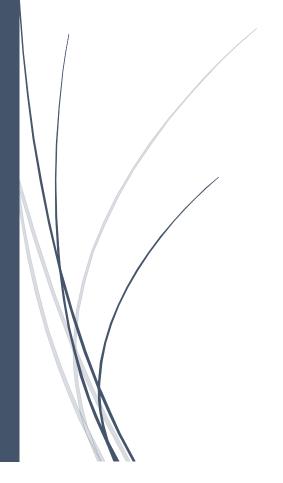
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# Deep Learning

Project 3 – Deep Reinforcement Learning



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### 1. Atari (PongNoFrameskip-v4)

#### 1.1 Environment

As input to the Neural Network I used 2 consecutive frames of the game so as to capture the movement of the ball. The images were resized to lower dimensions and scaled from 0 to 1.

#### 1.2 Replay Buffer

The Replay Buffer was used to store the current states, actions, rewards, next states and whether or not a terminal state was reached. The total memory size was set to 20000, while when this value was initially set to 50000 I ran out of memory. The replay buffer memory was sampled with a batch size of 32.

#### 1.3 Model Architecture

The model consists of 3 convolutional layers with 32, 64 and 64 filters accordingly and ReLU activations. The kernel sizes and strides were set to 8 and 4, 4 and 2, 3 and 1 accordingly. The convolutional layers were followed by 2 fully-connected layers consisted of 512 units each with ReLU and Linear activations accordingly. In fact, 2 networks with the aforementioned architecture were used. One that was trained and one that was updated every 1000 iterations.

#### 1.4 Methods

4 RL methods were implemented in order to solve this RL problem:

- Deep Q-Learning
- Double Deep Q-Learning
- Dueling Deep Q-Learning
- Dueling Double Deep Q-Learning

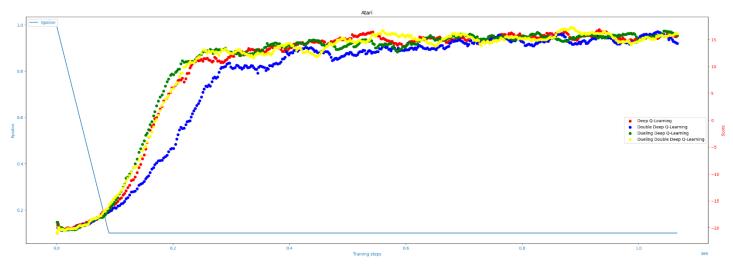
#### 1.4 Hyper-parameters

Number of games, learning rate and y were set to 500,  $10^{-4}$  and 0.99 accordingly.

#### 1.5 Epsilon

It was initiated to 1 and decreased by  $10^{-5}$  in each iteration until 0.1. After this value was reached it was kept constant.

#### 1.6 Results



#### 1.7 Observations

As we can see from the figure above, learning actually starts when the value of  $\varepsilon$  has dropped to its minimum value of 0.1. Dueling Deep Q-Learning presents the fastest learning, while Deep Q-Learning is clearly the slowest as far as learning is concerned. However, overall it seems that all 4 methods achieve almost the same score ( $\cong 15$ ) after 500 game played which indicates the game is solved by all of them.