## 1 Overview

This project demonstrates the use of deep Q-Networks for reinforcement learning. Reinforcement learning is a branch of Machine learning where an agent is trained to correctly behave in a particular environment. The Agent interacts with the environment by choosing some action a. In response to the action the environment returns a new environment state s and a reward r. The goal of the agent is to maximize the cumulative reward.

In this particular project the agent uses Q-Networks to be able to easily deal with a big set of possible states S. Q-Network's are deep neural networks that serve as a nonlinear function approximator which are trying to find the best action value function. This means the Q-Network tries to estimate the expected sum of future rewards (discounted by  $\gamma$ ) for a given state action pair.

$$Q_{\pi}(s, a) = \mathbb{E}[\gamma^{0}R_{1} + \gamma^{1}R_{2} + \gamma^{2}R_{3} + \dots | S_{0} = s, A_{0} = a, \pi]$$

Due to the unstable nature of nonlinear functions there are multiple additions of the Deep Q-learning Algorithm to increase the stability. This project implements the following two:

**Replay Buffer:** Use a buffer to store N most recent Experiences (S, A, R, S') tuples. Sample n random elements from the buffer at xth every time step and learn from them.

**Double DQN:** Use two neural networks instead of one to select the best action in order to prevent the algorithm from propagating incidental high rewards received by chance. The target network is updated

## 2 Results

## 3 Ideas for future improvements

One way to improve the Model is always to fine tune the hyper parameters. Additionally, it is possible to implement additional algorithms that help to improve and stabilize the DQN algorithm. Such additions would be "Dueling DQN", "Distributional DQN" and "Noisy DQN".

Moreover I noticed that with the given environment states from the unity engine it seems that the agent is not able to detect bananas that are very far away. Thus, using actual pixel values as input could also improve performance in this particular case.