Reinforcement Learning Exercise 10

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1 DQN on the Cart-Pole (7P)

In this exercise you will implement the deep Q-learning (DQN) algorithm. The code template can be found on github (https://github.com/humans-to-robots-motion/rl-course) in ex10-dqn/ex10-dqn.py. Please update the repository using "git pull". You will need tensorflow 2 python3 -m pip install tensorflow in order to run the code. For this exercise we will again use the Cart-Pole environment from gym: https://gym.openai.com/envs/CartPole-v1/c). What are the point issues of Q Learning when approximating the Q function with a neural network? How does

- a) What are the main issues of Q-Learning when approximating the Q-function with a neural network? How does DQN tackle them? (1P)
- b) The code template does naive Q-Learning with a neural net. It always uses the latest (s, a, r, s') pairs to compute the loss and update the weights. Your task is to implement DQN:
 - Implement a replay buffer. Store the observed pairs and randomly sample a batch for updating.
 - Implement fixed target weights. Keep another network with a separate set of weights for computing the target. Only update the target weights delayed. Hint: you can use the model.set_weights and get_weights functions

Compare the learning performance of the naive implementation with DQN. Do this by running it for 1000 iterations and plot the episode rewards. (4P)

- c) The code template uses Adam as optimizer. How does the performance change with SGD? Repeat your comparison from task b). (1P)
- d) Consider the "Breakout-v0" environment: https://gym.openai.com/envs/Breakout-v0/. What kind of neural network architecture would you use for this environment? (1P)

2 DDPG vs TRPO (3P)

In this exercise you will compare the performance of DDPG and TRPO using standard implementations. We recommend to use the *stable-baselines* package (https://stable-baselines.readthedocs.io/en/master/guide/quickstart.html). It can be installed using pip: *python3 -m pip install 'stable-baselines3[mpi]'*. Care, it needs an older tensorflow version as was used in task 1 (*python3 -m pip install tensorflow==1.15.0*), you can, for example, use a python virtual environment to have both installed.

- a) Run the TRPO and DDPG algorithms on the continuous mountain car environment "MountainCarContinuous-v0" (https://gym.openai.com/envs/MountainCarContinuous-v0/). You can try different action noise types for DDPG (e.g. Ornstein Uhlenbeck). (2P)
- b) Compare the performances of the algorithms. Which algorithm performs better on this environment, why? (1P)