## CSC411H1S Project 4

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1. The board is represented by a flat 9-element NumPy tuple. Turn denotes whose turn it is (1 for X, 2 for O). Done denotes whether the game is done (True if game is over, False otherwise.)

Below is an example of a sample game played against myself.

```
Python 2.7.14 | Anaconda custom (64-bit) | (default, Oct 5 2017, 02:28:52)
[GCC 4.2.1 Compatible Clang 4.0.1 (tags/RELEASE_401/final)] on darwin
env.render()
====
env.step(0)
Out[3]: (array([1, 0, 0, 0, 0, 0, 0, 0]), 'valid', False)
env.render()
х..
====
env.step(4)
Out[5]: (array([1, 0, 0, 0, 2, 0, 0, 0]), 'valid', False)
env.render()
х..
.0.
====
env.step(8)
Out[7]: (array([1, 0, 0, 0, 2, 0, 0, 0, 1]), 'valid', False)
env.render()
х..
.0.
..x
====
env.step(2)
Out[9]: (array([1, 0, 2, 0, 2, 0, 0, 0, 1]), 'valid', False)
env.render()
x.o
.0.
..x
env.step(6)
Out[11]: (array([1, 0, 2, 0, 2, 0, 1, 0, 1]), 'valid', False)
env.render()
x.o
.0.
x.x
env.step(3)
Out[13]: (array([1, 0, 2, 2, 2, 0, 1, 0, 1]), 'valid', False)
env.render()
x.o
00.
x.x
====
env.step(7)
Out[15]: (array([1, 0, 2, 2, 2, 0, 1, 1, 1]), 'win', True)
env.render()
x.o
00.
xxx
```

```
env.done
Out[17]: True
env.step(1)
Out[18]: (array([1, 0, 2, 2, 2, 0, 1, 1, 1]), 'done', True)
```

2. (a) The following is the new implemented policy

```
Listing 1:
```

```
class Policy (nn. Module):
       The Tic-Tac-Toe Policy
3
4
       def __init__(self, input_size=27, hidden_size=64, output_size=9):
5
6
           super(Policy, self).__init__()
           self.linear1 = nn.Linear(input_size, hidden_size)
8
           self.linear2 = nn.Linear(hidden_size, output_size)
10
       def forward (self, x):
11
           x = F. relu(self.linear1(x))
12
           x = self.linear2(x)
13
           return F.log_softmax(x)
14
```

(b) The 27 dimensions are a flattened encoding of a one-hot encoding of the state of the board. If .view(3,9) is applied to the array, the columns would be the one-hot encoding of each cell in the board (starting from the top left, going across each row, and ending in the bottom right).

If a column contains "1 0 0," the cell is empty.

If a column contains "0 1 0," the cell is occupied by an X.

If a column contains "0 0 1," the cell is occupied by an O.

- (c) The value in each dimension means the chance that making the move (e.g. adding an X into that cell) would result in winning the game. This policy is stochastic because it samples the next move from a distribution, rather than following a deterministic algorithm.
- 3. (a) The implementation of **compute\_returns** is shown below.

## Listing 2:

```
def compute_returns (rewards, gamma=1.0):
2
       Compute returns for each time step, given the rewards
3
         @param rewards: list of floats, where rewards[t] is the reward
                          obtained at time step t
5
         Oparam gamma: the discount factor
6
         @returns list of floats representing the episode's returns
             G_t = r_t + \gamma r_{t+1} + \gamma r_{t+2} + \dots
8
9
      >>>  compute_returns ([0,0,0,1], 1.0)
10
       [1.0, 1.0, 1.0, 1.0]
11
      >>>  compute_returns ([0,0,0,1], 0.9)
12
       [0.7290000000000001, 0.81, 0.9, 1.0]
      >>>  compute_returns ([0, -0.5, 5, 0.5, -10], 0.9)
14
       [-2.5965000000000003, -2.885000000000002, -2.650000000000004,
15
           -8.5, -10.0
16
      # TODO
17
```

```
result = []
18
       for index in range(len(rewards)):
19
           sum_returns = 0
20
           power = 0
           for i in range(index, len(rewards)):
22
                sum_returns = sum_returns + ((gamma ** power) * rewards[i
                   ])
               power = power + 1
           result.append(sum_returns)
25
       return result
26
```

(b) If we update the weights in the middle of an episode, it would lead to erroneous results since the algorithm does not know the terminal state of the episode when it's updating the weights. For instance, if the terminal state for an agent is "loss", then this will cancel out all the rewards that are associating with the agent's previous actions in the episode. Therefore, we should not update the weights in the middle of an episode.

## Listing 3:

- (b) The reward for a valid move is 0, since a single valid move doesn't directly contribute to a win or a loss.
- 5. (a)
  - (b)
  - (c)
  - (d)
- 6.
- 7.
- 8.