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# The Beginner Programmer

Shared thoughts, experiments, simulations and simple ideas with Python, R and other languages

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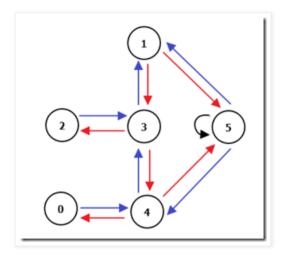
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Monday, 12 September 2016

# Getting AI smarter with Q-learning: a simple first step in Python

Yesterday I found an "old" script I wrote during a morning in the last semester. I remember being a little bored and interested in the concept of **Q-learning**. That was about the time **Alpha-Go** had beaten the world champion of Go and by reading here and there I found out that a bit of Q-learning mixed with deep learning might have been involved.



Indeed Q-learning seems an interesting concept, perhaps even more fascinating than traditional supervised machine learning since in this case the machine is basically learning from scratch how to perform a certain task in oder to optimize future rewards.

If you would like to read a, quote, "Painless Q-learning tutorial", I suggest you to read the following explanation: A Painless Q-learning Tutorial. In this article the concept of Q-learning is explained through a simple example and a clear walk-through. After having read the article I decided to put into code the example shown. The example shows a maze through which the agent should go and find its way up to the goal stage (stage 5). Basically, the idea is to train an algorithm to find the optimal path, given an (often random) initial condition, in order to maximize a certain outcome. In this simple example, as you can see from the picture shown in the article, the possible choices are all known and the outcome of each choice is deterministic. A best path exists and can be found easily regardless of the initial condition. Furthemore, the maze is time invariant. These nice theoretical hypothesis are usually not true when dealing when real world problems and this makes using Q-learning hard in practice, even though the concept behind it is relatively simple.

Before taking a look at the code, I suggest to read the article mentioned above, where you will get familiar with the problem tackled below. I'm not going deep in explaining what is going on since the author of that article has already done a pretty good job doing it and my work is just a (probably horribly inefficient) translation in Python of the algorithm.

Given an initial condition of, say, state 2, the optimal sequence path is clearly 2 - 3 - 1 - 5. Let's see if the algorithm finds it!

#### The Beginner Programmer



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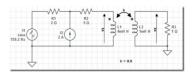
## Featured post

Solving a circuit with a mutual inductor using LTspice

```
# R matrix
    R = np.matrix([[-1,-1,-1,-1,0,-1],
                    [-1, -1, -1, 0, -1, 100],
                    [-1,-1,-1,0,-1,-1],
                    [-1,0,0,-1,0,-1],
                    [-1,0,0,-1,-1,100],
                    [-1,0,-1,-1,0,100]
    # O matrix
    Q = np.matrix(np.zeros([6,6]))
    # Gamma (learning parameter).
    gamma = 0.8
    # Initial state. (Usually to be chosen at random)
    initial_state = 1
    # This function returns all available actions in the state given as an argume
    def available_actions(state):
        current_state_row = R[state,]
        av_act = np.where(current_state_row >= 0)[1]
        return av act
    # Get available actions in the current state
    available_act = available_actions(initial_state)
    # This function chooses at random which action to be performed within the ran
    # of all the available actions.
    def sample_next_action(available_actions_range):
        next_action = int(np.random.choice(available_act,1))
        return next action
    # Sample next action to be performed
    action = sample_next_action(available_act)
38
    # This function updates the Q matrix according to the path selected and the Q
    # learning algorithm
    def update(current_state, action, gamma):
41
        max_index = np.where(Q[action,] == np.max(Q[action,]))[1]
43
        if max index.shape[0] > 1:
45
            max_index = int(np.random.choice(max_index, size = 1))
46
        else:
            max index = int(max index)
        max_value = Q[action, max_index]
        # O learning formula
        Q[current_state, action] = R[current_state, action] + gamma * max_value
    # Update Q matrix
    update(initial_state,action,gamma)
    #-----
    # Train over 10 000 iterations. (Re-iterate the process above).
    for i in range(10000):
        current_state = np.random.randint(0, int(Q.shape[0]))
        available act = available actions(current state)
```

import numpy as np

Mutual inductors can be a lot of fun, and sometimes a bit of an headache if you mess something up or represent them in a complicated way. Ta...



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# Taylor series with Python and Sympy

Here I am again using my beloved Python and doing

maths stuff. Today I'd like to post a short piece of code I made after a review of Taylor...



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Yesterday I found an

"old" script I wrote during a morning in the last semester. I remember being a little bored and interested in the conce...



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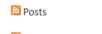
By making some assumptions, I am

going to simulate the flow of heat through an ideal rod. Suppose you have a cylindrical rod whose ends are ...

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10th MilanoR Meeting: Information Retrieval and Price **Positioning** 

```
action = sample next action(available act)
        update(current_state,action,gamma)
     # Normalize the "trained" Q matrix
66
     print("Trained Q matrix:")
     print(Q/np.max(Q)*100)
69
     # Testing
 72
     # Goal state = 5
 74
     # Best sequence path starting from 2 -> 2, 3, 1, 5
 75
     current_state = 2
     steps = [current_state]
 78
     while current_state != 5:
80
81
        next_step_index = np.where(Q[current_state,] == np.max(Q[current_state,])
 82
83
        if next_step_index.shape[0] > 1:
           next_step_index = int(np.random.choice(next_step_index, size = 1))
 85
86
            next_step_index = int(next_step_index)
87
        steps.append(next_step_index)
        current_state = next_step_index
89
90
     # Print selected sequence of steps
     print("Selected path:")
93
     print(steps)
95
     #-----
96
                                OUTPUT
98
99
     # Trained Q matrix:
              0. 0. 0. 80.
     #[[ 0.
                                       0. ]
              0. 0. 64. 0. 100.]
     # F 0.
102
     # [
         0.
             0. 0. 64. 0. 0.]
               80. 51.2 0. 80.
     # [
          0.
                                       0.]
     # [ 0.
               80. 51.2 0. 0. 100.]
105
     # [ 0. 80. 0. 0. 80. 100.]]
     # Selected path:
108
     # [2, 3, 1, 5]
q-learning.py hosted with \bigcirc by GitHub
                                                                 view raw
```

And sure enough there it is! Bear in mind this is a very simplified problem. At the same time though, keep in mind that Alpha Go is powered in part by a similar concept. I wonder what other application will come from Q-learning.

The concept of Q-learning is more vast than what I showed here, nevertheless I hope my post was an interesting insight.

Posted by Mic

Labels: Machine Learning, python

# 28 comments:

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LoggingFusion



#### Vishwas Pai 27 June 2017 at 09:10

Hey, I am getting error in Q learning formula. The error is " too many indices for array". How to solve that?

Reply

## **Replies**



#### Mic 30 June 2017 at 21:26

Hi, there seems to be an error in the indexing process, try to start debugging there.

## Reply



# August 2017 at 16:08 بنت النور 12

In my example I have 50000 states  $\dots$  and I got memory error.

Reply

## **Replies**



# Mic 13 September 2017 at 19:27

mmm.. not sure I understand. Sorry.

# Reply



# Manuel Amunategui 11 September 2017 at 19:34

Thanks for the nice python implementation - works great on new graphs. One minor nitpick, your graph image doesn't correspond to your reward matrix, it shouldn't go from 3 to 5 and back, instead 3 only goes to points 4 and 1. 5 loops onto itself for 100 points. Thanks for posting this!

Manuel

Reply

# Replies



# Mic 13 September 2017 at 19:35

Hi, thanks for reading! Oh, I didn't notice that! I didn't want to take the picture from the original article so I made one myself in a hurry. Thanks for pointing out the mistake, fixed it!:)

# Reply



# Nina 30 December 2017 at 10:36

Hi Mic, can you tell me why the next action is chosen at random (lines 29-33). I thought in Q learning the next action is chosen based on the highest Q value. Best, Nina

# Reply

# Replies



# Mic 30 December 2017 at 11:52

Hi Nina, the next action is chosen at random only during the training of the agent in order to explore the environment (and build the Q matrix). In testing (lines 71->89) the next action is chosen according to the highest Q value as you expect.

# Reply



In the update function, why are we selecting the Q max index based on the randomly selected action going into what I thought was the state field of Q? Such that Q(state, action) returns the Q-value for that specific pair, why are we inserting action into state field? (Q[action,])? Shouldn't we want to insert state there to get index from that?

Reply



## emetss 28 January 2018 at 14:24

It's a fantastic example for designing MDP based algorithms for many randomly generated environments

May I ask a question? Do you have like this example for multi agents. like team-q (friend-q) learning for cooperative mission

Reply

## **Replies**



Mic 29 January 2018 at 12:37

Thanks! I'm sorry, as of now I haven't got any examples for multi agents.



## Soumaila Fomba 9 February 2018 at 00:32

Bonjour,

comment adapter un algorithme de Q learning aux jeux de shogi ou échecs?

Reply



# Louis Bazireau 25 April 2018 at 14:30

Hi ! Great articles combined with the painless Q-learning tutorial ! I'm interested to develop a reinforcement learning algorithm for a flash game, if i'm correct, you don't create any walls or rooms, your agent are just wandering around. So i just need to get data from the flash game and change the python algorithm?

Reply

# Replies



Mic 29 April 2018 at 11:16

Hi, as long as you can build the R matrix you can adapt this example by simply using your R matrix. However I think it is going to be a bit slow with larger R matrices.

Reply



# Chathurangi Shyalika 17 October 2018 at 05:48

Hi,

Thanks for this great article!. Btw how many layers are in the NN you have implemented here? Can we specify the number of hidden layers in the network in reinforcement learning?

Reply



# resma k 1 December 2018 at 07:15

I like your blog, I read this blog please update more content on python, further check it once at python online training

Reply



# MAK 27 January 2019 at 05:31

def sample\_next\_action(available\_actions\_range):
 ... next\_action = int(np.random.choice(available\_act,1))
File "", line 2
next\_action = int(np.random.choice(available\_act,1))

Λ

IndentationError: expected an indented block

any idea why?

Reply



## Shailendra 14 March 2019 at 12:32

Good Post. I like your blog. Thanks for Sharing!

Machine Learning with Python Training in Gurgaon

Reply



# manishagaur 22 March 2019 at 13:01

Thanks a lot for sharing marvellous information on sap course. Thanks for sharing this valuable post with us.

Python Training in Gurgaon

Reply



# Manisha singh 1 April 2019 at 13:45

Thank you for sharing such great information very useful to us. Python Training institute in Noida

Reply



# Pankaj Singh 5 April 2019 at 05:59

Thankful to you for this amazing information sharing with us. Get website designing and development services by Ogen Infosystem.

Website Designing Company in Delhi

Reply



# Neha Sharma 13 April 2019 at 05:12

Rice Bags Manufacturers
Pouch Manufacturers
wall putty bag manufacturers
Lyrics with music

Reply



# Neha Sharma 13 April 2019 at 05:13

we have provide the best ppc service.
ppc agency in gurgaon
website designing company in Gurgaon
PPC company in Noida
seo services in gurgaon

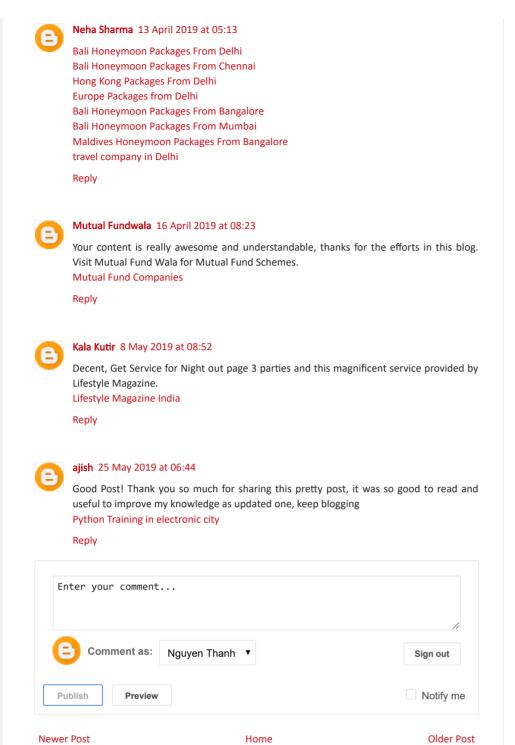
Reply



# Neha Sharma 13 April 2019 at 05:13

we have provide the best fridge repair service. fridge repair in faridabad
Videocon Fridge Repair in Faridabad
Whirlpool Fridge Repair in Faridabad
Washing Machine Repair in Noida
godrej washing machine repair in noida
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