D_case

Bongseokkim

2021-02-10

차 례

| introduction: Finding Shortest path using TD agent | 1 |
|--|----|
| Problem discription | 1 |
| Preparation | 2 |
| simul_step | 7 |
| TD contol | 9 |
| TD iteration | 10 |
| TD iteration 2 | 11 |
| Limitation | 13 |

introduction: Finding Shortest path using TD agent

In this case, I am trying to solve a path planning problem using TD agent.

Let's say there are 100 states, one of the way to find the optimal path is to find all the cases in trial and error, which will be about 100! it is impossible to compute 100! even if we use all the computer in the world. To solve the real world problem, I would like to learn the agent in a short time to find the optimal path even if it is not strictly optimal

Problem discription

State : set of X,Y coordiantes $\{S_0,S_1,\dots S_{10}\}$, each coordinates are randomly created

Action: agent can do three action at each state, Select and move one of three paths that can go from each node. {go fisrt node, go second node, go third node}

 $P^a_{ss'}$: transition probability with certain action in state is deterministic. It is set randomly and will be explained in detail in matrix form below

reward: Euclidean distance from s to s'

Goal: find the shortest path state 0 to 8 (just randomly selected)

Preparation

```
import numpy as np
import pandas as pd
```

making X,Y coordinate

The coordinates of each 11 state were randomly created.

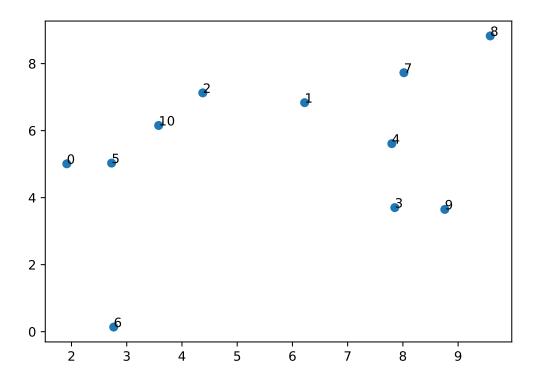
```
np.random.seed(1234)
coordinate= 10*np.random.rand(2,11)
data=pd.DataFrame(coordinate.T, columns=['X','Y'])
data
```

```
##
             Χ
      1.915195 5.009951
## 0
## 1
      6.221088 6.834629
     4.377277 7.127020
## 2
     7.853586 3.702508
## 3
## 4
     7.799758 5.611962
     2.725926 5.030832
## 5
## 6
     2.764643 0.137684
## 7
     8.018722 7.728266
## 8 9.581394 8.826412
## 9
      8.759326 3.648860
## 10 3.578173 6.153962
```

```
import matplotlib.pyplot as plt
states = np.arange(0,11).astype(str)
y = data['Y']
x = data['X']
n = states

fig, ax = plt.subplots()
ax.scatter(x, y)

for i, txt in enumerate(n):
    ax.annotate(n[i], (x[i], y[i]))
plt.show()
```



Computing Euclidean distance

```
distance = np.zeros(shape=(11,11))
for i in range(len(data)):
    for j in range(len(data)):
        distance[i,j]=np.sqrt(np.sum(data.iloc[i]**2+data.iloc[j]**2))
distance = pd.DataFrame(distance,index=states, columns= states)

distance
```

```
##
               0
                          1
                                     2
                                                     8
                                                                           10
        7.585194
                  10.685582
                              9.935923
                                             14.088159
## 0
                                                        10.899888
                                                                     8.913032
       10.685582
                  13.070126
                             12.464713
                                             15.972562
                                                        13.245908
                                                                    11.665704
## 1
## 2
        9.935923
                  12.464713
                             11.828354
                                             15.481073
                                                        12.648911
                                                                    10.983148
## 3
       10.205633
                  12.680752
                             12.055801
                                             15.655543
                                                        12.861856
                                                                    11.227731
       11.004450
                  13.332083
                             12.739125
                                             16.187618 13.504455
## 4
                                                                   11.958466
## 5
        7.842673
                  10.869868
                             10.133850
                                             14.228443 11.080610
                                                                     9.133154
## 6
        6.035709
                   9.647606
                              8.810061
                                             13.318065
                                                         9.884441
                                                                     7.637851
## 7
       12.360970
                  14.472045
                             13.927705
                                             17.138689
                                                        14.630994
                                                                    13.217434
## 8
       14.088159
                  15.972562
                             15.481073
                                             18.423281
                                                        16.116719
                                                                    14.845310
## 9
       10.899888
                  13.245908
                             12.648911
                                             16.116719
                                                        13.419387
                                                                    11.862316
        8.913032 11.665704
                             10.983148
                                             14.845310 11.862316
                                                                   10.067231
## 10
```

```
##
```

```
## [11 rows x 11 columns]
```

transition prob

transition prob matrix is created randomly, It can be adjusted later.

```
P_go_first= pd.DataFrame(np.matrix([[0,1,0,0,0,0,0,0,0,0,0],
                                    [0,0,1,0,0,0,0,0,0,0,0],
                                    [0,0,0,1,0,0,0,0,0,0,0],
                                    [0,0,0,0,1,0,0,0,0,0,0]
                                    [0,0,0,0,0,1,0,0,0,0,0],
                                    [0,0,0,0,0,0,1,0,0,0,0]
                                    [0,0,0,0,0,0,0,1,0,0,0],
                                    [0,0,0,0,0,0,0,0,1,0,0],
                                    [0,0,0,0,0,0,0,0,0,1,0],
                                    [0,0,0,0,0,0,0,0,0,0,0,1],
                                    [0,1,0,0,0,0,0,0,0,0,0]
                                                         ]),index= states, columns=states )
P_go_second= pd.DataFrame(np.matrix([[0,0,0,0,0,0,0,0,0,1,0],
                                    [0,0,0,0,0,0,0,0,1,0,0],
                                    [0,0,0,0,0,0,0,1,0,0,0],
                                    [0,0,0,0,0,0,1,0,0,0,0],
                                    [0,0,0,0,0,1,0,0,0,0,0],
                                    [0,0,0,0,1,0,0,0,0,0,0],
                                    [0,0,0,1,0,0,0,0,0,0,0]
                                    [0,0,1,0,0,0,0,0,0,0,0],
                                    [0,1,0,0,0,0,0,0,0,0,0],
                                    [1,0,0,0,0,0,0,0,0,0,0],
                                    [0,0,0,0,0,0,0,0,0,1,0],
                                                         ]),index= states, columns=states )
P_go_third= pd.DataFrame(np.matrix([[0,0,0,1,0,0,0,0,0,0,0],
                                    [0,0,0,0,0,1,0,0,0,0,0]
                                    [0,0,1,0,0,0,0,0,0,0,0],
                                    [0,0,0,0,0,0,1,0,0,0,0],
                                    [0,1,0,0,0,0,0,0,0,0,0]
                                    [0,0,0,1,0,0,0,0,0,0,0],
                                    [0,0,1,0,0,0,0,0,0,0,0]
                                    [0,0,0,0,0,1,0,0,0,0,0],
                                    [0,0,0,0,1,0,0,0,0,0,0]
```

```
[0,0,0,0,0,0,0,0,0,0,1],
[0,0,0,0,0,0,0,0,0,1]
]),index= states, columns=states )
P_go_third
```

```
## 0 1 2 3 4 5 6 7 8 9 10

## 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0

## 1 0 0 0 0 0 1 0 0 0 0 0 0 0 0

## 2 0 0 1 0 0 0 0 0 0 0 0 0 0 0

## 3 0 0 0 0 0 0 0 0 0 0 0 0 0 0

## 4 0 1 0 0 0 0 0 0 0 0 0 0 0 0

## 5 0 0 0 1 0 0 0 0 0 0 0 0 0 0

## 6 0 0 1 0 0 0 0 0 0 0 0 0 0 0

## 7 0 0 0 0 0 0 1 0 0 0 0 0 0 0

## 8 0 0 0 0 0 0 0 0 0 0 0 0 0 0

## 9 0 0 0 0 0 0 0 0 0 0 0 0 0
```

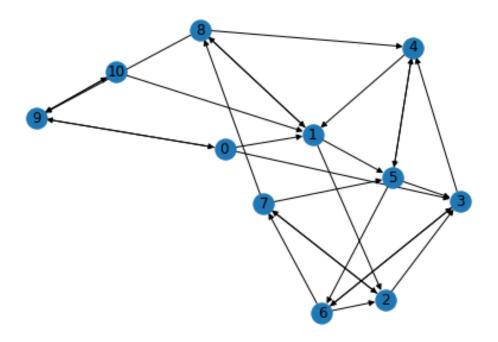


그림 1: nodes

R_s_a

get R_s_a as the distance between the states that arrive when agent act in each state

```
import numpy as np
import pandas as pd
R_s_a = pd.DataFrame(np.c_[np.repeat( 0, len( states ) ), np.repeat( 0, len( states ) ),np.repeat( 0, len( states ) )
R_s_a=R_s_a.astype('float')
for i in range(11):
    R_s_a['first'][i]=-distance.iloc[i,P_go_first.iloc[i][P_go_first.iloc[i].values==1][0]].astype(float)
for i in range(11):
    R s a['second'][i]=-distance.iloc[i,P go second.iloc[i][P go second.iloc[i].values==1][0]].astype(float)
for i in range(11):
    R_s_a['third'][i]=-distance.iloc[i,P_go_third.iloc[i][P_go_third.iloc[i].values==1][0]].astype(float)
R_s_a
           first
                                 third
##
                     second
## 0 -10.685582 -10.685582 -10.685582
## 1 -13.070126 -13.070126 -13.070126
## 2 -12.464713 -12.464713 -12.464713
## 3 -12.680752 -12.680752 -12.680752
## 4 -13.332083 -13.332083 -13.332083
## 5 -10.869868 -10.869868 -10.869868
     -9.647606 -9.647606 -9.647606
## 7 -14.472045 -14.472045 -14.472045
## 8 -15.972562 -15.972562 -15.972562
## 9 -13.245908 -13.245908 -13.245908
## 10 -11.665704 -11.665704 -11.665704
policy
pi_50 = pd.DataFrame( np.c_[np.repeat( 1/3, len( states ) ), np.repeat( 1/3, len( states ) ),np.repeat( 1/3,
                      columns = ['first', 'second','third'] )
pi_50
          first
                              third
##
                   second
       0.333333 0.333333 0.333333
## 0
## 1
       0.333333 0.333333 0.333333
```

simul_step

```
def simul_step(pi, s_now, P_go_first,P_go_second, P_go_third, R_s_a):
    if np.random.uniform() < pi_50.loc[s_now].cumsum()[0] :</pre>
        a_now ='first'
        P = P_go_first
    elif pi_50.loc[s_now].cumsum()[0]< np.random.uniform() < pi_50.loc[s_now].cumsum()[1]:</pre>
        a_now ='second'
        P = P_go_second
    else :
        a_now = 'third'
        P = P_{go}third
    r_now = R_s_a.loc[s_now , a_now]
    s_next = states[np.argmin( P.loc[s_now].cumsum() < np.random.uniform() )]</pre>
    if np.random.uniform() < pi_50.loc[s_now].cumsum()[0] :</pre>
        a_next ='first'
    elif pi 50.loc[s now].cumsum()[0]< np.random.uniform() < pi 50.loc[s now].cumsum()[1]:</pre>
        a_next ='second'
    else :
        a_next = 'third'
    sarsa = [s_now, a_now, r_now, s_next, a_next]
    return sarsa
```

```
sample_step = simul_step(pi_50, '0', P_go_first,P_go_second, P_go_third, R_s_a )
print( sample_step )
## ['0', 'first', -10.685582447534157, '1', 'second']
test simul step
for i in range(10):
   test_state = str(i)
   sample_step = simul_step(pi_50, test_state, P_go_first,P_go_second, P_go_third, R_s_a )
    print( sample_step )
## ['0', 'third', -10.685582447534157, '3', 'third']
## ['1', 'third', -13.070125529303853, '5', 'third']
## ['2', 'first', -12.464712830709807, '3', 'third']
## ['3', 'first', -12.680751685899999, '4', 'first']
## ['4', 'second', -13.332082835266162, '5', 'first']
## ['5', 'third', -10.869868014054747, '3', 'first']
## ['6', 'second', -9.647605720253912, '3', 'third']
## ['7', 'third', -14.472045062589983, '5', 'third']
## ['8', 'second', -15.97256209643203, '1', 'third']
## ['9', 'third', -13.245907551439055, '10', 'first']
q_s_a
q_s_a_init= pd.DataFrame( np.c_[np.repeat(0, len( states ) ), np.repeat(0, len( states ) ),np.repeat(0, len(
                      columns = ['first', 'second','third'] ).astype(float)
q_s_a_init
       first second third
##
         0.0
## 0
                 0.0
                        0.0
## 1
         0.0
                 0.0
                        0.0
## 2
         0.0
                 0.0
                        0.0
## 3
         0.0
                 0.0
                        0.0
         0.0
                0.0
## 4
                        0.0
## 5
         0.0
                0.0
                        0.0
## 6
         0.0
                 0.0
                        0.0
## 7
         0.0
                 0.0
                        0.0
## 8
         0.0
                 0.0
                        0.0
```

```
## 9 0.0 0.0 0.0
## 10 0.0 0.0 0.0
```

TD contol

```
def pol_eval_TD(sample_step, q_s_a, alpha):
    q_s_a_copy= q_s_a.copy()
    s = sample_step[0]
    a = sample_step[1]
    r = sample_step[2]
    s_next = sample_step[3]
    a_next = sample_step[4]

    q_s_a_copy.loc[s,a] +=alpha*(r+q_s_a_copy.loc[s_next, a_next]-q_s_a_copy.loc[s,a])
    return q_s_a_copy

q_s_a=pol_eval_TD(sample_step, q_s_a_init, alpha = 0.1)
    q_s_a
```

```
##
      first second
                       third
## 0
        0.0
                0.0 0.000000
## 1
        0.0
                0.0 0.000000
               0.0 0.000000
## 2
        0.0
## 3
        0.0
               0.0 0.000000
               0.0 0.000000
## 4
        0.0
        0.0
               0.0 0.000000
## 5
## 6
        0.0
               0.0 0.000000
## 7
        0.0
               0.0 0.000000
## 8
        0.0
                0.0 0.000000
                0.0 -1.324591
## 9
        0.0
## 10
        0.0
                0.0 0.000000
```

```
def pol_imp(pi, q_s_a, epsilon): # epsilon = exploration_rate
    pi_copy =pi.copy()
    for i in range(pi.shape[0]):
        # exploitation
        if np.random.uniform() > epsilon:
            pi_copy.iloc[i] = 0
            pi_copy.iloc[i, np.argmax(q_s_a.iloc[i,])] = 1
```

```
else:
    # exploration
    pi_copy.iloc[i] = 1/q_s_a.shape[1]

return pi_copy

pol_imp(pi_50, q_s_a, epsilon=0)
```

```
##
       first second third
## 0
         1.0
                 0.0
                        0.0
         1.0
                 0.0
                        0.0
## 1
## 2
         1.0
                 0.0
                        0.0
## 3
         1.0
                 0.0
                        0.0
                 0.0
## 4
         1.0
                        0.0
         1.0
                 0.0
                        0.0
## 5
         1.0
                 0.0
                        0.0
## 6
## 7
         1.0
                 0.0
                        0.0
                 0.0
                        0.0
## 8
         1.0
## 9
         1.0
                 0.0
                        0.0
## 10
         1.0
                 0.0
                        0.0
```

TD iteration

goal is find shortest path S_0 to S_8

```
import time
num_ep = 100
beg_time =time.time()
q_s_a = q_s_a_init
pi=pi_50
exploration_rate = 1

for epi_i in range(1,num_ep) :
    s_now="0"
    while s_now != "8":
        sample_step = simul_step(pi_50, s_now, P_go_first,P_go_second, P_go_third, R_s_a )
        q_s_a = pol_eval_TD(sample_step, q_s_a, alpha = 1/epi_i)
        pi = pol_imp(pi, q_s_a, epsilon= exploration_rate)
```

```
s_now = sample_step[3]
       exploration_rate *=0.9995
end_time =time.time()
print("Time difference of {} sec".format(end_time- beg_time))
## Time difference of 9.77786111831665 sec
print(pi.T)
                            3
                                    5
                                                                 10
         0.0 0.333333 0.0 1.0 0.0 0.0 0.333333 1.0 1.0 1.0 0.333333
## second 1.0 0.333333 1.0 0.0 1.0 1.0 0.333333 0.0 0.0 0.0 0.333333
print(q_s_a.T)
##
                         1
                                  2 ...
                                           8
                                                            10
## first -28.522001 -45.831120 -56.359056 ... 0.0 -7.952490 -15.344732
## second -9.824207 -13.070126 -42.553108 ... 0.0 -8.958847 -6.815551
## third -41.068752 -43.766223 -64.930868 ... 0.0 -9.461634 -10.617875
##
## [3 rows x 11 columns]
```

TD iteration 2

This time, I started with State 0 and looked for the best route to visit all states. aka 한붓그리기

```
import time
num_ep = 1000
beg_time =time.time()
q_s_a = q_s_a_init
pi=pi_50
exploration_rate = 1
```

```
import time
num_ep = 1000
beg_time =time.time()
q_s_a = q_s_a_init
```

```
pi=pi_50
exploration_rate = 1
for epi_i in range(1,num_ep) :
   s now="0"
   history = [int(s_now)]
   while not np.array_equal(history, states.astype(int)):
       sample_step = simul_step(pi_50, s_now, P_go_first,P_go_second, P_go_third, R_s_a )
       q_s_a = pol_eval_TD(sample_step, q_s_a, alpha = 1/epi_i)
       pi = pol_imp(pi, q_s_a, epsilon= exploration_rate)
       s_now = sample_step[3]
       history.append(int(s_now))
       my_set = set(history) #집합set으로 변환
       history = list(my_set) #list로 변환
       exploration rate *=0.9995
       #print(history)
end_time =time.time()
print("Time difference of {} sec".format(end_time- beg_time))
## Time difference of 439.9066162109375 sec
print(pi.T)
                                       6
                                            7
          0.0 0.0 0.0 1.0 0.0 0.0 1.0 1.0 1.0 0.0
## second 1.0 1.0 1.0 0.0 0.0 1.0 0.0 0.0 0.0 1.0
## third
         print(q_s_a.T)
##
                                        2 ...
                                                                            10
## first -166.848426 -234.213716 -241.224002 ... -78.832439 -67.549757 -144.633757
## second -47.990571 -156.676012 -219.220331 ... -190.007449 -98.287234 -39.586080
## third -191.485243 -232.123415 -246.140014 ... -222.407756 -73.269189 -60.129308
##
## [3 rows x 11 columns]
```

Limitation

- 1. Calculations at about 11 states also take longer time than expected (about 10~20 miniutes) It may be because my computer is not good, and the code is not perfect
- 2. It is a randomly generated coordinate, not a coordinate of the real world, and a process set to a task. if the process I have done so far is valid, it would be fun to solve it using the coordinates and distances of the real world.

```
"Done "
```

[1] "Done "