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YEAR: 4TH YEAR

LAB: MACHINE LEARNING

**ASSIGNMENT: 5** 

#### 1. Mountain car trying to go to top hill using Q-learning

```
import gym
import numpy as np
import matplotlib.pyplot as plt
env = gym.make("MountainCar-v0")
#Environment values
print(env.observation space.high) #[0.6 0.07]
print(env.observation space.low) #[-1.2 -0.07]
print(env.action space.n) #3
DISCRETE BUCKETS = 20
EPISODES = 1000
DISCOUNT = 0.95
EPISODE DISPLAY = 100
LEARNING RATE = 0.1
EPSILON = 0.5
EPSILON DECREMENTER = EPSILON/(EPISODES//4)
#Q-Table of size DISCRETE BUCKETS*DISCRETE BUCKETS*env.action space.n
Q TABLE = np.random.randn(DISCRETE BUCKETS, DISCRETE BUCKETS, env.action
space.n)
# For stats
ep rewards = []
ep_rewards_table = {'ep': [], 'avg': [], 'min': [], 'max': []}
def discretised state(state):
  DISCRETE WIN SIZE = (env.observation_space.high-
env.observation space.low)/[DISCRETE BUCKETS]*len(env.observation space
.high)
  discrete_state = (state-env.observation_space.low) // DISCRETE_WIN_SIZE
 return tuple(discrete state.astype(int)) #integer tuple as we need
to use it later on to extract Q table values
for episode in range(EPISODES):
  episode reward = 0
```

```
done = False
 curr discrete state = discretised state(env.reset())
 if episode % EPISODE DISPLAY == 0:
   render state = True
 else:
   render state = False
 while not done:
   if np.random.random() > EPSILON:
     action = np.argmax(Q TABLE[curr discrete state])
     action = np.random.randint(0, env.action space.n)
   new_state, reward, done, _ = env.step(action)
   new discrete state = discretised state(new state)
   if render state:
     env.render()
   if not done:
     max future q = np.max(Q TABLE[new discrete state])
     current q = Q TABLE[curr discrete state+(action,)]
     new q = current q + LEARNING RATE*(reward + DISCOUNT*max future q
 - current q)
     Q TABLE[curr discrete state+(action,)]=new q
   elif new state[0] >= env.goal position:
     Q TABLE[curr discrete state + (action,)] = 0
   curr discrete state = new discrete state
   episode reward += reward
 EPSILON = EPSILON - EPSILON_DECREMENTER
 ep rewards.append(episode reward)
 if not episode % EPISODE DISPLAY:
   avg reward = sum(ep rewards[-EPISODE DISPLAY:])/len(ep rewards[-
EPISODE DISPLAY:])
   ep_rewards_table['ep'].append(episode)
   ep rewards table['avg'].append(avg reward)
   ep rewards table['min'].append(min(ep rewards[-EPISODE DISPLAY:]))
   ep_rewards_table['max'].append(max(ep_rewards[-EPISODE_DISPLAY:]))
   print(f"Episode:{episode} avg:{avg reward} min:{min(ep rewards[-
```

```
env.close()

plt.plot(ep_rewards_table['ep'], ep_rewards_table['avg'], label="avg")
plt.plot(ep_rewards_table['ep'], ep_rewards_table['min'], label="min")
plt.plot(ep_rewards_table['ep'], ep_rewards_table['max'], label="max")
plt.legend(loc=4) #bottom right
plt.title('Mountain Car Q-Learning')
plt.ylabel('Average reward/Episode')
plt.xlabel('Episodes')
plt.show()

[0.6 0.07]
[-1.2 -0.07]
```

```
Episode:0 avg:-200.0 min:-200.0 max:-200.0
Episode:100 avg:-200.0 min:-200.0 max:-200.0
Episode:200 avg:-200.0 min:-200.0 max:-200.0
Episode:300 avg:-199.76 min:-200.0 max:-176.0
Episode:400 avg:-191.17 min:-200.0 max:-119.0
Episode:500 avg:-195.94 min:-200.0 max:-151.0
Episode:600 avg:-176.04 min:-200.0 max:-146.0
Episode:700 avg:-179.34 min:-200.0 max:-121.0
Episode:800 avg:-175.29 min:-200.0 max:-144.0
Episode:900 avg:-178.14 min:-200.0 max:-145.0
                    Mountain Car Q-Learning
   -120
   -130
   -140
   -150
   -160
   -170
   -180
                                                  avg
   -190
                                                  min
    -200
                                     600
          0
                  200
                           400
                                              800
                            Episodes
```

#### Visualization

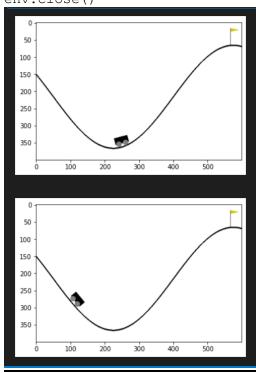
```
%matplotlib inline
```

```
import gym
import numpy as np
from matplotlib import pyplot as plt
env = gym.envs.make("MountainCar-v0")
env.reset()
```

```
plt.figure()
plt.imshow(env.render(mode='rgb_array'))

[env.step(0) for x in range(10000)]
plt.figure()
plt.imshow(env.render(mode='rgb_array'))
```

env.close()



## 1.1.Mountain car trying to go to top hill using SARSA

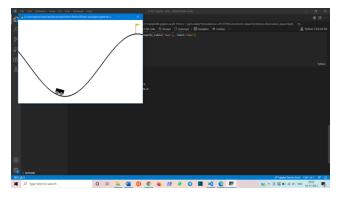
```
import gym
import numpy as np
import matplotlib.pyplot as plt
env = gym.make("MountainCar-v0")
#Environment values
print(env.observation space.high) #[0.6 0.07]
                                 #[-1.2 -0.07]
print(env.observation space.low)
print(env.action space.n)
DISCRETE BUCKETS = 20
EPISODES = 1000
DISCOUNT = 0.95
EPISODE DISPLAY = 100
LEARNING RATE = 0.1
EPSILON = 0.5
EPSILON_DECREMENTER = EPSILON/(EPISODES//4)
```

```
#Q-Table of size DISCRETE BUCKETS*DISCRETE BUCKETS*env.action space.n
Q TABLE = np.random.randn(DISCRETE BUCKETS, DISCRETE BUCKETS, env.action
space.n)
# For stats
ep rewards = []
ep rewards table = {'ep': [], 'avg': [], 'min': [], 'max': []}
def discretised state(state):
      DISCRETE WIN SIZE = (env.observation space.high-
env.observation space.low)/[DISCRETE BUCKETS]*len(env.observation space
.high)
      discrete state = (state-env.observation space.low) // DISCRETE WIN SIZE
     return tuple(discrete state.astype(int)) #integer tuple as we need
  to use it later on to extract Q table values
for episode in range(EPISODES):
      episode reward = 0
      done = False
      if episode % EPISODE DISPLAY == 0:
            render state = True
      else:
            render state = False
      curr discrete state = discretised state(env.reset())
      if np.random.random() > EPSILON:
            action = np.argmax(Q TABLE[curr discrete state])
      else:
             action = np.random.randint(0, env.action space.n)
      while not done:
             new state, reward, done, = env.step(action)
            new_discrete_state = discretised_state(new_state)
            if np.random.random() > EPSILON:
                   new action = np.argmax(Q TABLE[new discrete state])
            else:
                   new action = np.random.randint(0, env.action space.n)
             if render state:
                   env.render()
            if not done:
                   current q = Q TABLE[curr discrete state+(action,)]
                   max_future_q = Q_TABLE[new_discrete_state+(new_action,)]
                   \label{eq:current_q} \texttt{new} \ \ q \ = \ \texttt{current} \_ q \ + \ \texttt{LEARNING} \_ \texttt{RATE*} \ (\texttt{reward+DISCOUNT*max} \_ \texttt{future} \_ q - \texttt{max} \_ q - \texttt{max}
current q)
```

```
Q TABLE[curr discrete state+(action,)]=new q
   elif new state[0] >= env.goal position:
     Q TABLE[curr discrete state + (action,)] = 0
    curr discrete state = new discrete state
    action = new action
   episode reward += reward
  EPSILON = EPSILON - EPSILON DECREMENTER
  ep rewards.append(episode reward)
  if not episode % EPISODE DISPLAY:
   avg reward = sum(ep rewards[-EPISODE DISPLAY:])/len(ep rewards[-
EPISODE DISPLAY:])
   ep rewards table['ep'].append(episode)
   ep rewards table['avg'].append(avg reward)
    ep rewards table['min'].append(min(ep rewards[-EPISODE DISPLAY:]))
   ep rewards table['max'].append(max(ep rewards[-EPISODE DISPLAY:]))
   print(f"Episode:{episode} avg:{avg reward} min:{min(ep rewards[-
env.close()
plt.plot(ep rewards table['ep'], ep rewards table['avg'], label="avg")
plt.plot(ep rewards table['ep'], ep rewards table['min'], label="min")
plt.plot(ep rewards table['ep'], ep rewards table['max'], label="max")
plt.legend(loc=4) #bottom right
plt.title('Mountain Car SARSA')
plt.ylabel('Average reward/Episode')
plt.xlabel('Episodes')
plt.show()
```

```
[-1.2 -0.07]
Episode:0 avg:-200.0 min:-200.0 max:-200.0
Episode:100 avg:-200.0 min:-200.0 max:-200.0
Episode:200 avg:-200.0 min:-200.0 max:-200.0
Episode:300 avg:-192.65 min:-200.0 max:-152.0
Episode:400 avg:-195.16 min:-200.0 max:-150.0
Episode:500 avg:-176.27 min:-200.0 max:-133.0
Episode:600 avg:-170.26 min:-200.0 max:-142.0
Episode:700 avg:-181.62 min:-200.0 max:-124.0
Episode:800 avg:-184.17 min:-200.0 max:-121.0
Episode:900 avg:-196.86 min:-200.0 max:-165.0
                      Mountain Car SARSA
   -120
   -130
 -140
-150
-160
   -170
   -180
   -190
                                                max
```

### **Visualization**



## 2.Car Racing

```
https://www.roboti.us/index.html

ERROR: Failed building wheel for mujoco-py
Failed to build mujoco-py
```

It was the error that I encountered while running the code. I tried a lot to solve it unable to did it finally, I tried to install gym then gym[all] but there was error .

```
import gym
import numpy as np
import random
from scipy import misc
import tensorflow as tf
```

```
env = gym.make('CarRacing-v0')
observation = env.reset()
EPISODES = 1
TIMESTAMP = 5
GAMMA = 0.99
ALPHA = 0.001
explore eps = 1
N = 50
OUT1 = 5
OUT2 = 5
OUT3 = 5
BATCH SIZE = 4
def conv2d(x,W,stride):
  return tf.nn.conv2d(x,W,strides=[1,stride,stride,1],padding='SAME')
def max pool 2x2(x):
  return tf.nn.max pool(x,ksize=[1,2,2,1],strides = [1,2,2,1],padding='
SAME')
class neuralNet:
  def init (self):
    self.sess = tf.InteractiveSession()
    self.X = tf.placeholder(tf.float32, [None, N, N, 1])
    self.C1 = tf.placeholder(tf.float32,[None,OUT1])
    self.C2 = tf.placeholder(tf.float32,[None,OUT2])
    self.C3 = tf.placeholder(tf.float32,[None,OUT3])
    self.Y1 = tf.placeholder(tf.float32,[None,OUT1])
    self.Y2 = tf.placeholder(tf.float32,[None,OUT2])
    self.Y3 = tf.placeholder(tf.float32,[None,OUT3])
    self.W conv1 = tf.Variable(tf.truncated normal([8,8,1,32],stddev =
             # 50 * 50 * 1
0.1))
    self.B conv1 = tf.Variable(tf.zeros([32]))
    self.W conv2 = tf.Variable(tf.truncated normal([5,5,32,64],stddev =
             # 15 * 15 * 32
 0.1))
    self.B conv2 = tf.Variable(tf.zeros([64]))
    self.W conv3 = tf.Variable(tf.truncated normal([3,3,64,64],stddev =
             # 7 * 7 * 64
    self.B conv3 = tf.Variable(tf.zeros([64]))
    self.W fc1 = tf.Variable(tf.truncated normal([ 5*5*64 , 512],stddev
 = 0.1))
           # 5 * 5 * 64
```

```
self.B_fc1 = tf.Variable(tf.zeros([512]))
    self.W fc21 = tf.Variable(tf.truncated normal([512,OUT1],stddev = 0
.1))
    self.B fc21 = tf.Variable(tf.zeros([OUT1]))
    self.W fc22 = tf.Variable(tf.truncated normal([512,OUT2],stddev = 0
.1))
    self.B fc22 = tf.Variable(tf.zeros([OUT2]))
    self.W_fc23 = tf.Variable(tf.truncated_normal([512,OUT3],stddev = 0
.1))
    self.B fc23 = tf.Variable(tf.zeros([OUT3]))
    o conv1 = tf.nn.relu(conv2d(self.X,self.W conv1,3) + self.B conv1)
    o pool1 = \max pool 2x2(o conv1)
    o conv2 = tf.nn.relu(conv2d(o pool1, self.W conv2,2) + self.B conv2)
    o conv3 = tf.nn.relu(conv2d(o conv2,self.W conv3,1) + self.B conv3)
    o fconv3 = tf.reshape(o conv3, [-1,5*5*64])
    o fc1 = tf.nn.relu(tf.matmul(o fconv3, self.W fc1) + self.B fc1)
    self.o fc21 = tf.matmul(o fc1,self.W fc21) + self.B fc21
    self.o fc22 = tf.matmul(o fc1,self.W fc22) + self.B fc22
    self.o fc23 = tf.matmul(o fc1, self.W fc23) + self.B fc23
    self.L1 = tf.reduce sum(tf.square(self.Y1 - tf.mul(self.o fc21,self
.C1)))
    self.L2 = tf.reduce sum(tf.square(self.Y2 - tf.mul(self.o fc22,self
    self.L3 = tf.reduce sum(tf.square(self.Y3 - tf.mul(self.o fc23,self
.C3)))
    self.optimizer = tf.train.AdamOptimizer(ALPHA)
    self.train step1 = self.optimizer.minimize(self.L1)
    self.train step2 = self.optimizer.minimize(self.L2)
    self.train_step3 = self.optimizer.minimize(self.L3)
    self.sess.run(tf.initialize all variables())
  def forward pass(self,x):
    with self.sess.as default():
      out1, out2, out3 = self.sess.run([self.o_fc21,self.o_fc22,self.o_
fc23], feed dict={self.X:x})
```

```
# print out
    return np.argmax(out1),np.argmax(out2),np.argmax(out3),np.max(out1)
, np.max(out2), np.max(out3)
  def train(self,x,y1,y2,y3,c1,c2,c3):
    with self.sess.as default():
      self.sess.run([self.train step1,self.train step2,self.train step3
],feed dict={
        self.X:x , self.Y1:y1, self.Y2:y2, self.Y3:y3, self.C1:c1, self
.C2:c2, self.C3:c3})
def sanity check():
 observation = env.reset()
 print (observation.shape)
 print(env.action space)
 print(env.action space.sample())
 print(env.observation space)
  # print(env.observation space.high)
  # print(env.observation space.low)
 print(env.action space.high)
 print(env.action space.low)
def process image(ot):
 ot = misc.imresize(ot, (N,N,3))
 ot = 0.299 * ot[:,:,0] + 0.587 * ot[:,:,1] + 0.114 * ot[:,:,2]
  ot = np.reshape(ot, (1,N,N,1))
  return ot
def create new data(ot,re,ot2,reset,done,a1,a2,a3):
  c1 = np.zeros((1,OUT1))
  c1[0][a1] = 1
  c2 = np.zeros((1,OUT1))
 c2[0][a2] = 1
  c3 = np.zeros((1,OUT1))
 c3[0][a3] = 1
  yval1 = np.zeros((1,OUT1))
  yval2 = np.zeros((1,OUT2))
  yval3 = np.zeros((1,OUT3))
 b1,b2,b3,bv1,bv2,bv3 = nnet.forward pass(ot2)
 yval1[0][a1] = re
  yval2[0][a2] = re
 yval3[0][a3] = re
  if not done:
    yval1[0][a1] = re + GAMMA*bv1
    yval2[0][a2] = re + GAMMA*bv2
    yval3[0][a3] = re + GAMMA*bv3
  data batch['C1'] = c1
  data batch['C2'] = c2
```

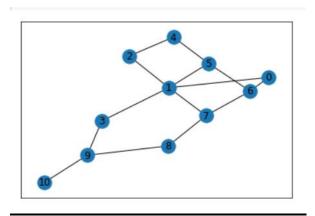
```
data batch['C3'] = c3
  if reset:
    data batch['X'] = ot
    data batch['Y1'] = yval1
    data batch['Y2'] = yval1
    data batch['Y3'] = yval1
  else:
    data batch['X'] = np.append(data batch['X'],ot,axis=0)
    data batch['Y1'] = np.append(data batch['Y1'], yval1, axis=0)
    data batch['Y2'] = np.append(data batch['Y2'], yval2, axis=0)
    data batch['Y3'] = np.append(data batch['Y3'], yval3, axis=0)
nnet = neuralNet()
data batch = {}
sanity check()
ans = np.zeros((12))
anssum = np.zeros((12))
for ep in range (EPISODES):
  observation = env.reset()
 observation = process image(observation)
 reward = 0
 sum reward = 0
 data batch = {}
 reset = True
 for t in range(TIMESTAMP):
   env.render()
    x = np.array(observation)
    a1,a2,a3,av1,av2,av3 = nnet.forward_pass(x)
    tempvar = random.random()
    if tempvar < max((500/(ep+1)), explore eps) and ep < 9000:
                                                                     # do
nt explore for last 1000 episodes
      a1 = np.random.randint(0, 5, size=1)
      a2 = np.random.randint(0,5,size=1)
      a3 = np.random.randint(0,5,size=1)
    action = [-1.0 + a1*0.4, a2*0.2, a3*0.2]
    observation, reward, done, info = env.step(action)
    observation = process image(observation)
    create new data(x,reward,np.array(observation),reset,done,a1,a2,a3)
    print (data_batch['X'].shape , data_batch['Y'].shape , data_batch['
C'].shape)
    reset = False
    if data batch['X'].shape[0] == BATCH SIZE:
      nnet.train(data batch['X'] , data batch['Y'], data batch['C'])
      reset = True
```

```
sum reward = sum reward + reward
    if done or t == TIMESTAMP-1:
      nnet.train(data batch['X'] , data batch['Y'], data batch['C'])
      print("Episode {0} finished after {1} timesteps.".format(ep+1,t+1
) )
      ans[int(ep/5000)] = \max(ans[int(ep/5000)],t)
      anssum[int(ep/5000)] += anssum[int(ep/5000)]
      break
for i in range(3):
 print (i*5000 , " -
- ", (i+1)*5000 , " == " , ans[i] , (anssum[i]/5000))
3.Roulette
import gym
import numpy as np
import matplotlib.pyplot as plt
import gym toytext
env = gym.make('Roulette-v0')
EPS = 0.05
GAMMA = 1.0
Q = \{ \}
agentSumSpace = [i for i in range(0,37)]
actionSpace = [i for i in range(0, 38)]
stateSpace = []
returns = {}
pairsVisited = {}
for total in agentSumSpace:
    for action in actionSpace:
        Q[(total, action)] = 0
        returns[(total, action)] = 0
        pairsVisited[(total, action)] = 0
    stateSpace.append(total)
policy = {}
for state in stateSpace:
    policy[state] = np.random.choice(actionSpace)
numEpisodes = 1000000
for i in range(numEpisodes):
    statesActionsReturns = []
    memory = []
    if i % 100000 == 0:
        print('starting episode', i)
    observation = env.reset()
    done = False
while not done:
    action = policy[observation]
```

```
observation , reward, done, info = env.step(action)
    memory.append((observation, action, reward))
    observation = observation
memory.append((observation, action, reward))
G = 0
last = True
for observed, action, reward in reversed (memory):
    if last:
        last = False
    else:
        statesActionsReturns.append((observed, action, G))
    G = GAMMA*G + reward
statesActionsReturns.reverse()
statesActionsVisited = []
for observed, action, G in statesActionsReturns:
    sa = (observed, action)
    if sa not in statesActionsVisited:
        pairsVisited[sa] += 1
        returns[(sa)] += (1 / pairsVisited[(sa)]) * (G-returns[(sa)])
        Q[sa] = returns[sa]
        rand = np.random.random()
        if rand < 1 - EPS:
            state = observed
            values = np.array([Q[(state, a)] for a in actionSpace ])
            best=np.random.choice(np.where(values==values.max())[0])
            policy[state] = actionSpace[best]
            policy[state] = np.random.choice(actionSpace)
        statesActionsVisited.append(sa)
if EPS - 1e-7 > 0:
   EPS -= 1e-7
else:
    EPS = 0
numEpisodes = 1000
rewards = np.zeros(numEpisodes)
totalReward = 0
wins = 0
losses = 0
print('getting ready to test policy')
for i in range(numEpisodes):
    observation = env.reset()
    done = False
    while not done:
        action = policy[observation]
        observation , reward, done, info = env.step(action)
```

```
observation = observation
           totalReward += reward
           rewards[i] = totalReward
if reward >= 1:
     wins += 1
elif reward == -1:
     losses += 1
wins /= numEpisodes
losses /= numEpisodes
print('win rate', wins, 'loss rate', losses)
plt.plot(rewards)
plt.show()
 starting episode 0
 starting episode 100000
 starting episode 200000
 starting episode 300000
 starting episode 400000
 starting episode 500000
 starting episode 600000
 starting episode 700000
 starting episode 800000
 starting episode 900000
 getting ready to test policy
 win rate 0.001 loss rate 0.0
   -500
  -1000
  -1500
  -2000
  -2500
                        600
                              800
```

# 4. Implement both RL and DRL for finding the shortest path in any user-input graph.



```
R=np.matrix(np.zeros(shape=(11, 11)))
for x in G[10]:
 R[x, 10] = 100
Q=np.matrix(np.zeros(shape=(11, 11)))
0-=100
for node in G.nodes:
 for x in G[node]:
  Q[node,x]=0
  Q[x, node] = 0
import pandas as pd
pd.DataFrame(R)
  0 1 2 3 4 5 6 7 8 9
0.0
0.0
0.0
```

#### pd.DataFrame(Q)

```
2
0 -100.0 0.0 -100.0 -100.0 -100.0 -100.0 0.0 -100.0 -100.0 -100.0 -100.0
     0.0 -100.0
                 0.0 0.0 -100.0
                                    0.0 -100.0
                                                0.0 -100.0 -100.0 -100.0
2 -100.0 0.0 -100.0 -100.0 0.0 -100.0 -100.0 -100.0 -100.0 -100.0
3 -100.0
           0.0 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0
                                                            0.0 -100.0
4 -100.0 -100.0 0.0 -100.0 -100.0 0.0 -100.0 -100.0 -100.0 -100.0 -100.0
5 -100.0
           0.0 -100.0 -100.0
                             0.0 -100.0
                                          0.0 -100.0 -100.0 -100.0 -100.0
    0.0 -100.0 -100.0 -100.0 -100.0 0.0 -100.0 0.0 -100.0 -100.0
           0.0 -100.0 -100.0 -100.0 -100.0
                                          0.0 -100.0
7 -100.0
                                                      0.0 -100.0 -100.0
8 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0 0.0 -100.0 0.0 -100.0
9 -100.0 -100.0 -100.0 0.0 -100.0 -100.0 -100.0 -100.0 0.0 -100.0
10 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0 -100.0
```

 8
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
 0.0
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```
random value=random.uniform(0, 1)
   if random value<er:</pre>
      sample=G[start]
   else:
      sample=np.where(Q[start,] == np.max(Q[start,]))[1]
   next node=int(np.random.choice(sample, 1))
   return next node
def updateQ(node1, node2, lr, discount):
  max index=np.where (Q[node2, ]==np.max(Q[node2, ]))[1]
   if max index.shape[0]>1:
     max index=int(np.random.choice(max index,size=1))
   else:
     max index=int(max index)
  max value=Q[node2,max index]
   Q[node1, node2] = int (1-
lr)*Q[node1, node2]+lr*(R[node1, node2]+discount*max value)
def learn(er, lr, discount):
   for i in range (50000):
      start=np.random.randint(0,11)
      next node=next number(start,er)
      updateQ(start,next node,lr,discount)
learn(0.5, 0.8, 0.8)
pd.DataFrame(Q)
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def shortest path(begin, end):
  path=[begin]
   next node=np.argmax(Q[begin,])
  path.append(next node)
   while next node!=end:
      next node=np.argmax(Q[next node,])
     path.append(next node)
   return path
shortest path (0,10)
[0, 1, 3, 9, 10]
```