Inotes2_MDP

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Implementation	2
Policy Improve	4
policy Iteration	5
Value Improvement	6

Implementation

```
gamma=0.5
states = range(1,8)
P_TL = np.matrix([[1,0,0,0,0,0,0],[1,0,0,0,0,0,0],
[0,1,0,0,0,0,0],[0,0,1,0,0,0,0],[0,0,0,1,0,0,0],
[0,0,0,0,1,0,0],[0,0,0,0,0,1,0]])
P_TR = np.matrix([[0,1,0,0,0,0,0]],
[0,0,1,0,0,0,0],[0,0,0,1,0,0,0],[0,0,0,0,1,0,0],
[0,0,0,0,0,1,0],[0,0,0,0,0,0,1],[0,0,0,0,0,0,1]]
pi_50 = np.hstack((np.repeat(0.5,len(states)).reshape(7,1),np.repeat(0.5,len(states)).reshape(7,1)))
pi_50 = pd.DataFrame(pi_50,states,["Left","Right"])
pi_TL = np.hstack((np.repeat(1,len(states)).reshape(7,1),np.repeat(0,len(states)).reshape(7,1)))
pi_TL = pd.DataFrame(pi_TL,states,["Left","Right"])
pi_TR = np.hstack((np.repeat(0,len(states)).reshape(7,1),np.repeat(1,len(states)).reshape(7,1)))
pi_TR = pd.DataFrame(pi_TR,states,["Left","Right"])
def transition(given_pi, states, P_TL, P_TR):
    P_out = pd.DataFrame(np.zeros((len(states),len(states))),states,states)
    for i,s in enumerate(states):
        action_dist = given_pi.loc[s]
        P = action_dist["Left"]*P_TL + action_dist["Right"]*P_TR
        P_{out.loc[s]} = P[i,:]
    return P_out
R_s_a = np.matrix([[1,1,0,0,0,0,0],[0,0,0,0,0,0,10,10]]).T
R_s_a = pd.DataFrame(R_s_a,states,["Left","Right"])
def reward_fn(given_pi):
    R_pi = np.matrix(given_pi*R_s_a).sum(axis=1)
    R_pi = pd.DataFrame(R_pi,states)
    return(R_pi)
print(reward_fn(pi_TL).T)
```

```
## 1 2 3 4 5 6 7
## 0 1 1 0 0 0 0 0
```

```
def policy_eval(given_pi):
    R=reward_fn(given_pi).values.reshape(7,1)
    P=transition(given_pi, states, P_TL, P_TR)
    gamma=0.5
    epsilon=10**(-8)
    v_old=np.repeat(0,7).reshape(7,1)
    v_new=R+gamma*np.dot(P,v_old)

while(np.max(np.abs(v_new-v_old))>epsilon):
    v_old=v_new
    v_new=R+np.dot(gamma*P,v_old)
    return v_new

V_old=policy_eval(pi_TL)
pi_old=pi_TL
q_s_a=R_s_a+np.hstack((np.dot(gamma*P_TL,V_old),np.dot(gamma*P_TR,V_old)))
print(q_s_a.T)
```

```
## Left 2.0 2.0 1.00 0.500 0.2500 0.12500 0.06250 ## Right 1.0 0.5 0.25 0.125 0.0625 10.03125 10.03125
```

Policy Improve

```
pi_new_vec=q_s_a.idxmax(axis=1)
pi_new=pd.DataFrame(np.zeros(pi_old.shape), index=pi_old.index, columns=pi_old.columns)
for i in range(len(pi_new_vec)):
   pi_new.iloc[i][pi_new_vec.iloc[i]]=1
def policy_improve(V_old, pi_old = pi_old, R_s_a = R_s_a, gamma = gamma,
P_TL = P_TL, P_TR = P_TR):
   q_s_a=R_s_a+np.hstack((np.dot(gamma*P_TL,V_old),np.dot(gamma*P_TR,V_old)))
   pi_new_vec=q_s_a.idxmax(axis=1)
   pi_new=pd.DataFrame(np.zeros(pi_old.shape), index=pi_old.index, columns=pi_old.columns)
   for i in range(len(pi_new_vec)):
       pi_new.iloc[i][pi_new_vec.iloc[i]]=1
   return pi_new
pi_old = pi_TL
V_old = policy_eval(pi_old)
pi_new = policy_improve(V_old, pi_old = pi_old, R_s_a = R_s_a, gamma = gamma, P_TL = P_TL, P_TR = P_
print(pi_old.T)
##
         1 2 3 4 5 6 7
## Left
         1 1 1 1 1 1 1
## Right 0 0 0 0 0 0
print(pi_new.T)
                     3 4
         1.0 1.0 1.0 1.0 1.0 0.0 0.0
```

Right 0.0 0.0 0.0 0.0 1.0 1.0

policy Iteration

```
pi_old = pi_50
cnt = 0
while True:
   print(cnt,"-th iteration")
   print(pi_old.T)
   V_old = policy_eval(pi_old)
   pi_new = policy_improve(V_old, pi_old = pi_old, R_s_a = R_s_a, gamma = gamma, P_TL = P_TL, P_TR
   if pi_new.equals(pi_old):
      break
   pi_old = pi_new
   cnt += 1
## 0 -th iteration
      1 2 3 4 5 6 7
## Left 0.5 0.5 0.5 0.5 0.5 0.5
## Right 0.5 0.5 0.5 0.5 0.5 0.5
## 1 -th iteration
      1 2 3 4 5
##
                                   7
## Left 1.0 1.0 1.0 0.0 0.0 0.0 0.0
## Right 0.0 0.0 0.0 1.0 1.0 1.0
## 2 -th iteration
       1 2 3 4 5
##
                                 7
## Left 1.0 1.0 0.0 0.0 0.0 0.0 0.0
## Right 0.0 0.0 1.0 1.0 1.0 1.0
print(policy_eval(pi_new))
## [[ 2.
## [ 2.
## [ 2.49999999]
## [ 4.9999999]
## [ 9.99999999]
## [19.9999999]
## [19.9999999]]
```

Value Improvement

```
cnt = 0
gamma=0.5
epsilon = 10**(-8)
V_old = np.zeros(len(states)).T
V_old = pd.DataFrame(V_old,states)
results = V_old.T
while True:
    \label{eq:continuous_problem} $q_s_a=R_s_a+np.hstack((np.dot(gamma*P_TL,V_old),np.dot(gamma*P_TR,V_old)))$
    V_new = np.matrix(q_s_a.apply(max,axis=1)).T
    V_new = pd.DataFrame(V_new,states)
    if(np.max(np.abs(V_new-V_old)).item()<epsilon or cnt>300):
        break
    results = np.vstack((results,V_new.T))
    V_old = V_new
    cnt+=1
results = pd.DataFrame(results, columns = states)
print(results.tail())
##
              2
                   3
                        4
                              5
                                    6
## 26 2.0 2.0 2.5 5.0 10.0 20.0 20.0
## 27 2.0 2.0 2.5 5.0 10.0 20.0 20.0
## 28  2.0  2.0  2.5  5.0  10.0  20.0  20.0
## 29 2.0 2.0 2.5 5.0 10.0 20.0 20.0
## 30 2.0 2.0 2.5 5.0 10.0 20.0 20.0
V_opt=results.tail(1).T
q_s_a=R_s_a+np.c_[np.dot(gamma*P_TL,V_opt), np.dot(gamma*P_TR, V_opt)]
print(q_s_a)
```

```
## Left Right
## 1 2.00 1.00
## 2 2.00 1.25
## 3 1.00 2.50
## 4 1.25 5.00
```

```
## 5 2.50 10.00
## 6 5.00 20.00
## 7 10.00 20.00
```

```
pi_opt_vec=q_s_a.idxmax(axis=1).T
print(pi_opt_vec)
```

1 Left

2 Left

3 Right

4 Right

5 Right

6 Right

7 Right

dtype: object