

# F1 Python

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## p.9 Preparation

```
states = np.arange(0,80,10)
P_normal = np.matrix([[0,1,0,0,0,0,0,0],[0,0,1,0,0,0,0,0],
[0,0,0,1,0,0,0,0],[0,0,0,0,1,0,0,0],[0,0,0,0,0,1,0,0],
[0,0,0,0,0,0,1,0],[0,0,0,0,0,0,0,1]])
P_speed = np.matrix([[0.1,0,0.9,0,0,0,0,0],[0.1,0,0,0.9,0,0,0,0],
[0,0.1,0,0,0.9,0,0,0],[0,0,0.1,0,0,0.9,0,0],[0,0,0,0.1,0,0,0.9,0],
[0,0,0,0,0.1,0,0,0.9],[0,0,0,0,0,0.1,0,0.9],[0,0,0,0,0,0,0,1]])
P_normal = pd.DataFrame(P_normal,states,states)
P_speed = pd.DataFrame(P_speed,states,states)
R_s_a = np.matrix([[-1,-1,-1,-1,0,-1,-1,0],[-1.5,-1.5,-1.5,-1.5,-0.5,-1.5,-1.5,0]]).T
R_s_a = pd.DataFrame(R_s_a,states,["n","s"])

pi_speed = np.hstack((np.repeat(0,len(states)).reshape(8,1),np.repeat(1,len(states)).reshape(8,1)))
pi_speed = pd.DataFrame(pi_speed,states,["n","s"])

pi_50 = np.hstack((np.repeat(0.5,len(states)).reshape(8,1),np.repeat(0.5,len(states)).reshape(8,1)))
pi_50 = pd.DataFrame(pi_50,states,["n","s"])
pi_speed.T
```

```
##      0    10    20    30    40    50    60    70
## n      0      0      0      0      0      0      0      0
## s      1      1      1      1      1      1      1      1
```

```
pi_50.T
```

```
##      0      10      20      30      40      50      60      70
## n  0.5  0.5  0.5  0.5  0.5  0.5  0.5  0.5
## s  0.5  0.5  0.5  0.5  0.5  0.5  0.5  0.5
```

## p.11 Simulator $\pi^{speed}$

```

pi = pi_speed
history = []
MC_N = 10000
for MC_i in range(MC_N):
    s_now = 0
    history_i = []
    while(s_now!=70):
        if(np.random.uniform(0,1)<pi.loc[s_now]["n"]):
            a_now="n"
            P=P_normal
        else:
            a_now="s"
            P=P_speed
        r_now = R_s_a.loc[s_now][a_now]
        s_next = states[np.argmin(P.loc[s_now].cumsum())<np.random.uniform(0,1))].item()
        history_i.extend([s_now,a_now,r_now])
        s_now = s_next
    history.append(history_i)

history_speed = history
print(pd.DataFrame(np.matrix(history_speed[:20])).T)

```

```

## 0
## 0 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 1 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 2 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 3 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 4 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 5 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 6 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 7 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 8 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 9 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 10 [0, s, -1.5, 20, s, -1.5, 10, s, -1.5, 30, s, ...
## 11 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 12 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 13 [0, s, -1.5, 20, s, -1.5, 10, s, -1.5, 30, s, ...
## 14 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 30, s, ...

```

```
## 15 [0, s, -1.5, 20, s, -1.5, 10, s, -1.5, 30, s, ...
## 16 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 17 [0, s, -1.5, 0, s, -1.5, 0, s, -1.5, 20, s, -1...
## 18 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 19 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
##
## D:\miniconda3\envs\r-reticulate\lib\site-packages\numpy\matrixlib\defmatrix.py:145: VisibleDeprec
##   arr = N.array(data, dtype=dtype, copy=copy)
```

### p.13 Simulator $\pi^{50}$

```
pi = pi_50
history = []
MC_N = 10000
for MC_i in range(MC_N):
    s_now = 0
    history_i = []
    while(s_now!=70):
        if(np.random.uniform(0,1)<pi.loc[s_now]["n"]):
            a_now="n"
            P=P_normal
        else:
            a_now="s"
            P=P_speed
        r_now = R_s_a.loc[s_now][a_now]
        s_next = states[np.argmin(P.loc[s_now].cumsum())<np.random.uniform(0,1))].item()
        history_i.extend([s_now,a_now,r_now])
        s_now = s_next
    history.append(history_i)

history_50 = history
print(pd.DataFrame(np.matrix(history_50[:20])).T)
```

```
## 0
## 0 [0, n, -1.0, 10, s, -1.5, 0, s, -1.5, 20, n, -...
## 1 [0, n, -1.0, 10, n, -1.0, 20, n, -1.0, 30, n, ...
## 2 [0, s, -1.5, 20, n, -1.0, 30, n, -1.0, 40, n, ...
## 3 [0, n, -1.0, 10, s, -1.5, 30, n, -1.0, 40, s, ...
## 4 [0, s, -1.5, 20, n, -1.0, 30, n, -1.0, 40, n, ...
## 5 [0, n, -1.0, 10, s, -1.5, 30, n, -1.0, 40, s, ...
## 6 [0, n, -1.0, 10, n, -1.0, 20, s, -1.5, 40, s, ...
## 7 [0, n, -1.0, 10, s, -1.5, 30, n, -1.0, 40, s, ...
## 8 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 9 [0, n, -1.0, 10, s, -1.5, 30, s, -1.5, 20, n, ...
## 10 [0, s, -1.5, 20, s, -1.5, 40, s, -0.5, 60, s, ...
## 11 [0, s, -1.5, 20, n, -1.0, 30, n, -1.0, 40, s, ...
## 12 [0, s, -1.5, 20, s, -1.5, 10, n, -1.0, 20, n, ...
## 13 [0, n, -1.0, 10, n, -1.0, 20, n, -1.0, 30, n, ...
## 14 [0, n, -1.0, 10, n, -1.0, 20, s, -1.5, 40, n, ...
```

```

## 15  [0, s, -1.5, 20, n, -1.0, 30, s, -1.5, 50, n, ...
## 16  [0, s, -1.5, 20, n, -1.0, 30, n, -1.0, 40, s, ...
## 17  [0, s, -1.5, 20, s, -1.5, 40, n, 0.0, 50, s, -...
## 18  [0, s, -1.5, 20, s, -1.5, 40, n, 0.0, 50, s, -...
## 19  [0, n, -1.0, 10, n, -1.0, 20, n, -1.0, 30, s, ...
##
## D:\miniconda3\envs\r-reticulate\lib\site-packages\numpy\matrixlib\defmatrix.py:145: VisibleDeprec
##   arr = N.array(data, dtype=dtype, copy=copy)

```

## p.17 Implementation 1 $\pi^{speed}$ (vectorized)

```
pol_eval=pd.DataFrame(np.matrix(np.zeros((len(states)*2))).reshape(len(states),2), index=states, col
pol_eval.T
```

```
##           0    10    20    30    40    50    60    70
## count  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
## sum    0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
```

```
for MC_i in range(MC_N):
    history_i=history_speed[MC_i]

    for j in range(0,len(history_i),3):
        pol_eval.loc[history_i[j]]['count']+=1

        if j < len(history_i) :
            pol_eval.loc[history_i[j]]['sum']+=pd.Series(history_i)[range(j+2,len(history_i)-1,3)].a

        else:
            pol_eval.loc[history_i[j]]['sum']+=0
print(pol_eval.T)
```

```
##           0    10    20    30    40    50    60    70
## count 11272.0 1066.0 10304.0 1905.0 9491.0 2538.0 8546.0 0.0
## sum  -48708.5 -3930.5 -27212.0 -3760.0 -8044.0 -598.0 -1425.5 0.0
```

```
pol_cal=pd.DataFrame(pol_eval['sum']/pol_eval['count'])
print(pol_cal.T)
```

```
##           0    10    20    30    40    50    60    70
## 0 -4.321194 -3.687148 -2.640916 -1.973753 -0.84754 -0.235619 -0.166803 NaN
```

## p.19 Implementation 2 $\pi^{speed}$ (running estimate)

```
pol_eval=pd.DataFrame(np.matrix(np.zeros((len(states)*2))).reshape(len(states),2), index=states, col
pol_eval.T
```

```
##           0    10    20    30    40    50    60    70
## count  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
## est    0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
```

```
for MC_i in range(MC_N):
    history_i=history_speed[MC_i]

    for j in range(0,len(history_i),3):
        # update count
        pol_eval.loc[history_i[j]]['count']+=1
        current_cnt=pol_eval.loc[history_i[j]]['count']

        # return is the new info
        if j < len(history_i):
            new_info=pd.Series(history_i)[range(j+2,len(history_i)-1,3)].astype('float').sum()

        else:
            new_info=0

        # update the last estimate with new info
        alpha=1/current_cnt
        pol_eval.loc[history_i[j]]['est']+=alpha*(new_info-pol_eval.loc[history_i[j]]['est'])

np.round(pol_eval.T,2)
```

```
##           0         10         20         30         40         50         60    70
## count 11272.00  1066.00 10304.00  1905.00  9491.00  2538.00  8546.00  0.0
## est   -4.32   -3.69   -2.64   -1.97   -0.85   -0.24   -0.17  0.0
```



### p.21 Implementation 3 $\pi^{50}$ (vectorized)

```
pol_eval=pd.DataFrame(np.matrix(np.zeros((len(states)*2))).reshape(len(states),2), index=states, col
pol_eval.T
```

```
##           0    10    20    30    40    50    60    70
## count  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
## sum    0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
```

```
for MC_i in range(MC_N):
    history_i=history_50[MC_i]

    for j in range(0,len(history_i),3):
        pol_eval.loc[history_i[j]]['count']+=1

        if j < len(history_i) :
            pol_eval.loc[history_i[j]]['sum']+=pd.Series(history_i)[range(j+2,len(history_i)-1,3)].a

        else:
            pol_eval.loc[history_i[j]]['sum']+=0
pol_eval.T
```

```
##           0          10          20          30          40          50          60    70
## count 10845.0   5912.0   8058.0   7058.0   7546.0   7324.0   7074.0  0.0
## sum  -50262.0 -22547.5 -22345.0 -14418.0 -5489.0 -4871.0  -714.0  0.0
```

```
pol_cal=pd.DataFrame(pol_eval['sum']/pol_eval['count'])
print(pol_cal.T)
```

```
##           0          10          20          30          40          50          60    70
## 0 -4.634578 -3.813853 -2.773021 -2.042788 -0.727405 -0.665074 -0.100933 NaN
```

## p.23 Implementation 4 $\pi^{50}$ (running estimate)

```
pol_eval=pd.DataFrame(np.matrix(np.zeros((len(states)*2))).reshape(len(states),2), index=states, col
pol_eval.T
```

```
##           0    10    20    30    40    50    60    70
## count  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
## est    0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
```

```
for MC_i in range(MC_N):
    history_i=history_50[MC_i]

    for j in range(0,len(history_i),3):
        # increment count
        pol_eval.loc[history_i[j]]['count']+=1
        current_cnt=pol_eval.loc[history_i[j]]['count']

        # return is the new info
        if j < len(history_i):
            new_info=pd.Series(history_i)[range(j+2,len(history_i)-1,3)].astype('float').sum()

        else:
            new_info=0

        # update the last estimate with new info
        alpha=1/current_cnt
        pol_eval.loc[history_i[j]]['est']+=alpha*(new_info-pol_eval.loc[history_i[j]]['est'])

np.round(pol_eval.T,2)
```

```
##           0      10      20      30      40      50      60      70
## count 10845.00 5912.00 8058.00 7058.00 7546.00 7324.00 7074.0 0.0
## est   -4.63   -3.81   -2.77   -2.04   -0.73   -0.67   -0.1  0.0
```

### p.35 Implementation 5 $\pi^{speed}$

```
pol_eval=pd.DataFrame(np.matrix(np.zeros((len(states)*2))).reshape(len(states),2), states, ['count',
pol_eval.T
```

```
##          0    10    20    30    40    50    60    70
## count  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
## est    0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
```

```
for MC_i in range(MC_N):
    history_i=history_speed[MC_i]

    for j in range(0,len(history_i),3):
        pol_eval.loc[history_i[j]]['count']+=1
        current_cnt=pol_eval.loc[history_i[j]]['count']

        if j < len(history_i)-3 :
            new_info = history_i[j+2]+pol_eval.loc[history_i[j+3]]['est']

        else:
            new_info =0

        alpha=1/current_cnt
        pol_eval.loc[history_i[j]]['est']+=alpha*(new_info-pol_eval.loc[history_i[j]]['est'])

print(np.round(pol_eval.T,2))
```

```
##          0          10          20          30          40          50          60    70
## count 11272.00 1066.00 10304.00 1905.00 9491.00 2538.00 8546.00 0.0
## est   -4.28   -3.71   -2.64   -1.98   -0.85   -0.23   -0.17  0.0
```

### p.37 Implementation 5 $\pi^{50}$

```
pol_eval=pd.DataFrame(np.matrix(np.zeros((len(states)*2))).reshape(len(states),2), states, ['count',  
pol_eval.T
```

```
##           0    10    20    30    40    50    60    70  
## count  0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0  
## est    0.0  0.0  0.0  0.0  0.0  0.0  0.0  0.0
```

```
for MC_i in range(MC_N):  
    history_i=history_50[MC_i]  
  
    for j in range(0,len(history_i),3):  
        pol_eval.loc[history_i[j]]['count']+=1  
        current_cnt=pol_eval.loc[history_i[j]]['count']  
  
        if j < len(history_i)-3 :  
            new_info = history_i[j+2]+pol_eval.loc[history_i[j+3]]['est']  
  
        else:  
            new_info =0  
  
        alpha=1/current_cnt  
        pol_eval.loc[history_i[j]]['est']+=alpha*(new_info-pol_eval.loc[history_i[j]]['est'])  
  
print(np.round(pol_eval.T,2))
```

```
##           0         10         20         30         40         50         60         70  
## count 10845.00  5912.00  8058.00  7058.00  7546.00  7324.00  7074.00  0.0  
## est   -4.62   -3.79   -2.77   -2.03   -0.71   -0.66   -0.11  0.0
```