



Université de Picardie Jules Verne

Licence Métiers du Numérique *3ème année* 2021-2022



Paradis et Enfers

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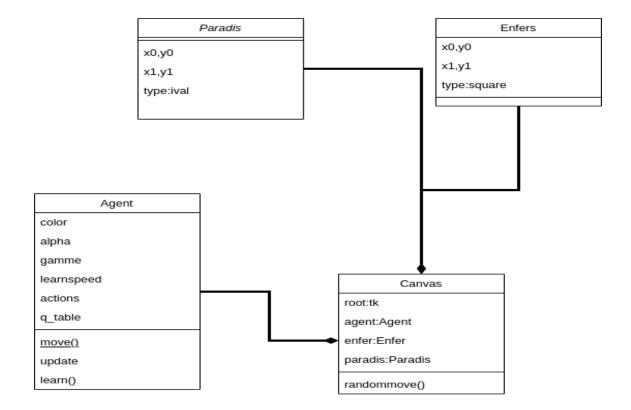
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Chapitre 1

Introduction

The concept of the game is that the agent must find his path to heaven by avoiding the hell in our case the black squares.



Chapitre 2

Implementation of the code

For the implementation i did a class called canvas that contains an array rows and columns and then i put inside it at the first row the agent and in the 2-3 and 3-2 the hell and the heaven in 3-3 then i make a reintialize() that recreate the agent each time he moves after that I generate a method called randomMoves() where i pick random move based on actions which i have it in an array ('top','bottom','right','left') then i test if it hits the hell i will reward it with -1 and if it hits the heaven i will reward it with 1 and I turn it back (the agent) to the terminal which is the first row and if it is an empty square I make 0 in reward and continue moving. In the class agent I implemented the learning strategy with learn-speed of 0.03 and gamma= influence_decay_coef f = 0.9andepsilon = major_prob = 0.9 then I de fined the learn strategy Method that has say trategation, reward, next, tategard be

 $0.9. then I defined the learn strategy Method that has cur_state, action, reward, next_state and base on the action of the property of the p$

2.1 Canvas

```
1
     class Canvas(tk.Tk, object):
2
         def __init__(self):
3
            super(Canvas, self).__init__()
4
            self.title("Canvas")
5
            self.init()
6
7
8
         def init (self):
9
            self.canvas = tk.Canvas(
10
            self, bg='white', height=size height, width=size width)
11
            self.actions = ["left", "right", "top", "bottom"]
12
13
            #colone
14
            for c in range (width+1):
15
                 x0, y0 = c * unit, 0
16
                 x1, y1 = x0, y0 + height * unit
17
18
                 self.canvas.create\_line(x0, y0, x1, y1)
19
            for r in range(width+1):
20
                 x0, y0 = 0, r*unit
21
                 x1, y1 = x0 + width * unit, y0
22
                 self.canvas.create line(x0, y0, x1, y1)
```

• Hell:

```
1
      #First trap HELL 3/2
2
3
            x0, y0 = startPoint[0] + (width - 2) * unit, <math>startPoint[1] + (
        height - 3) * unit
4
            x1, y1 = x0 + 3 * unit / 4, y0 + 3 * unit / 4
6
            self.hell1 = self.canvas.create rectangle(x0, y0, x1, y1, fill=
        black')
      #second trap HELL 2/3
8
            x0, y0 = startPoint[0] + (width - 3) * unit, <math>startPoint[1] + (
9
        height - 2) * unit
            x1, y1 = x0 + 3 * unit / 4, y0 + 3 * unit / 4
10
            self.hell2 = self.canvas.create rectangle(x0, y0, x1, y1, fill=
11
        black')
```

• Heaven and Agent :

```
1
             for i in range (6):
2
                 for j in range(6):
3
                     x0, y0 = startPoint[0] + j*unit, startPoint[1] + i*unit
4
                     x1, y1 = x0+3*unit/4, y0+3*unit/4
5
                     if i = 0 and j = 0:
6
                         #Agent
                          self.rect = self.canvas.create_rectangle(
                                       x0, y0,
9
                                       x1, y1,
10
                                       fill='red')
11
12
                     elif i = 2 and j = 2:
13
                          oval center = startPoint + unit * 2
14
                         #Heaven 3/3
15
                          self.oval = self.canvas.create oval(
16
                                       x0, y0,
17
                                       x1, y1,
18
                                       fill='yellow')
19
            self.canvas.pack()
20
```

• reintialize:

```
1
        def reintialize (self):
2
            self.update()
3
            time. sleep (0.5)
4
            self.canvas.delete(self.rect)
5
            startPoint = np.array([unit/8, unit/8])
6
7
           x0, y0 = startPoint[0]+0*unit, startPoint[1]+0*unit
           x1, y1 = x0+3*unit/4, y0+3*unit/4
8
            self.rect = self.canvas.create rectangle(
```

4

```
x0, y0,

x1, y1,

fill='red')

return self.canvas.coords(self.rect)
```

• random move:

```
1
     def random move(self, action):
2
3
               cur state = self.canvas.coords(self.rect)
4
               base action = np.array([0,0])
5
               if action == 'top': # haut
6
                   if cur state[1] > unit:
7
                        base\_action[1] = base\_action[1] - unit
8
               elif action = 'bottom': # bas
9
                   if \operatorname{cur} state [1] < (\operatorname{height}-1)*\operatorname{unit}:
10
                        base action[1] = base action[1] + unit
11
               elif action == 'right': # droite
12
                   if \operatorname{cur\_state}[0] < (\operatorname{width}-1)*\operatorname{unit}:
13
                        base_action[0] = base_action[0] + unit
14
               elif action == 'left': # gauche
15
                   if cur_state[0] > unit:
16
                        base action[0] = base action[0] - unit
17
18
               self.canvas.move(self.rect, base_action[0], base_action[1])
19
              next move = self.canvas.coords(self.rect)
20
21
22
               if next move == self.canvas.coords(self.oval):
23
                   reward = 1
24
                   over = True
25
                   next move = 'terminal'
26
27
                   # print("Point gagne!")
28
               elif next move == self.canvas.coords(self.hell1) or next move =
29
          self.canvas.coords(self.hell2):
                   reward = -1
30
                   over = True
31
                   next move = 'terminal'
32
                   # print("Point perdu!")
33
               else:
34
                   reward = 0
35
                   over = False
36
               return next move, reward, over
37
```

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2.2 Agent

• Learning Strategy:

```
def learn strategy (self, cur state, action, reward, next state):
2
             self.check state(next state)
3
4
             q_predict = self.q_table.loc[cur_state, action]
5
             #if the next state is not equal to the first case (terminal)
6
             if next_state != 'terminal':
7
                 q_{target} = reward + self.gamma * 
8
                     self.q table.loc[next state, :].max()
10
                 q_target = reward
11
             self.q table.loc[cur state, action] = self.q table.loc[cur state
12
                                                                       action]+
13
        self.speed * (q_target - q_predict)
```

• chooseAction Methode:

```
1
          def chooseAction(self, observation):
2
              self.check state (observation)
3
              if \quad np.\, random.\, uniform\, (\,) \ < \ self.\, epsilon:
4
                   s\_action = self.q\_table.loc[observation, :]
5
                   action = np.random.choice(
6
                        s = action [s = action = np.max(s = action)].index)
8
              else:
9
                   action = np.random.choice(self.actions)
10
              return action)
```

```
Paradis_Enfers > 🦆 agent.py > ધ AgentRL > 😭 learn_strategy_
                                                                                      X
                                                       Canvas
      class AgentRL:
           def init (se
                                                                                          major
               self.action
               self.speed
               self.gamma
               self.epsile
               self.q tab
           def chooseAction
               self.check
               if np.rand
                   s actio
                   action
               else:
                   action
               return act
           def learn stra
               self.check
 27
               q predict
               if next state !=
                                  · terminat:
                   q target = reward + self.gamma * \
                        self.q table.loc[next state, :].max()
               else:
                   q target = reward
               self a table locicur state action = self a table locicur state
PROBLEMS (2)
              OUTPUT
                                     TERMINAL
[15.0, 375.0, 105.0, 465.0]
                               0.000000e+00
                                             0.000229
                                                        0.000000
                                                                  0.000000e+00
[135.0, 375.0, 225.0, 465.0]
                               5.904900e-07
                                             0.007625
                                                        0.000000
                                                                  0.000000e+00
[255.0, 375.0, 345.0, 465.0]
                               0.000000e+00
                                             0.000000
                                                        0.141266
                                                                  0.000000e+00
 [375.0, 375.0, 465.0, 465.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000
                                                                  0.000000e+00
[375.0, 15.0, 465.0, 105.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000
                                                                  2.187000e-05
[375.0, 135.0, 465.0, 225.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000
                                                                  2.381400e-03
[375.0, 255.0, 465.0, 345.0]
                               8.732700e-02
                                             0.000000
                                                        0.000000
                                                                  0.000000e+00
                                       left
                                                 right
                                                             top
                                                                        bottom
[15.0, 15.0, 105.0, 105.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000
                                                                  0.000000e+00
 [15.0, 135.0, 105.0, 225.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000
                                                                  3.140816e-08
[135.0, 135.0, 225.0, 225.0]
                               0.000000e+00 -0.087327
                                                        0.000000 -3.000000e-02
[135.0, 15.0, 225.0, 105.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000
                                                                 0.000000e+00
 [255.0, 15.0, 345.0, 105.0]
                               0.000000e+00
                                             0.000000
                                                        0.000000 -3.000000e-02
```