

Project 1- Pacman Search, 2020

The objective of the lab is to find the search algorithms necessary to perform the behavior of a simple Pacman using Berkely Pacman and determine heuristics derivations.

Introduction

We can find a Pacman Project with all the specifications and the files we need to complete. We had to fill some algorithms in search.py and searchAgents.py, that integrate with our main class pacman.py.

Search

This file stores some of the searching algorithms that we seen in theory classes. Following that, a possible solution that we find was:

1. depthFirstSearch (DFS)

```
174
75 def depthFirstSearch(problem):
76
77     first = (problem.getStartState(), None, 1) #Estado inicial (5,5)
78     s = util.Stack()                          #Creamos una pila s (util.Stack contiene funcio
79     s.push(first)                             #Agregamos origen a la pila s
80     visit = []                                #Nodos visitados
81     path = []                                 #Ruta a devolver
82     adyac = {}                               #Nodos adyacentes
83
84     while not s.isEmpty():
85         v = s.pop()                           #sacamos un elemento de la pila s llamado v
86
87         if not v[0] in visit:
88             if problem.isGoalState(v[0]):      #Si es el objetivo devolvemos la ruta
89
90                 while v[0] != first[0]:
91                     path.append(v[1])          #Introducimos ruta de direcciones en p
92                     v = adyac[v]              #El nodo v ahora sera el padre
93
94                 path.reverse()
95                 return path                   #Devolvemos ruta al reves
96
97         visit.append(v[0])
98         successors = problem.getSuccessors(v[0])
99         for suc in successors:                 #recorremos los nodos sucesores
100             if suc[0] not in visit:
101                 s.push(suc)                   #Anadimos nodos no visitados a la pila
102                 adyac[suc] = v                #Anadimos el nodo a sus hijos
103
104     return []
```

We use a Stack to save the newest nodes before the oldest. The aim is to find the deepest nodes and give a list of actions.

2. breadthFirstSearch (BFS)

```
136 def breadthFirstSearch(problem):
137     """Search the shallowest nodes in the search tree first."""
138     """ YOUR CODE HERE """
139     #BFS es el mismo concepto que el DFS solo que cambia la Pila por la Cola, así hará la
140
141     first = (problem.getStartState(), None, 1) #Estado inicial
142     s = util.Queue() #Creamos una cola s (util.Queue contiene funcio
143     s.push(first) #Agregamos origen a la cola s
144     visit = [] #Nodos visitados
145     path = [] #Ruta a devolver
146     adyac = {} #Nodos adyacentes
147
148     while not s.isEmpty():
149         v = s.pop() #sacamos un elemento de la cola s llamado v
150
151         if not v[0] in visit:
152             if problem.isGoalState(v[0]): #Si es el objetivo delvolvemos la ruta
153
154                 while v[0] != first[0]:
155                     path.append(v[1]) #Introducimos ruta de direcciones en p
156                     v = adyac[v] #El nodo v ahora sera el padre
157
158                 path.reverse()
159                 return path #Devolvemos ruta al reves
160
161             visit.append(v[0])
162             successors = problem.getSuccessors(v[0])
163             for suc in successors: #recorremos los nodos sucesores
164                 if suc[0] not in visit:
165                     s.push(suc) #Anadimos nodos no visitados a la cola
166                     adyac[suc] = v #Anadimos el nodo a sus hijos
167
168     return []
```

This algorithm is very similar to the previous one, the difference is that the visitor nodes in amplitude using a queue instead of a stack.

In practice, we have been created a queue where we will use the queue functions provided by the "util.py" file. The rest of the code is the same.

3. uniformCostSearch (UCS)

```

175 def uniformCostSearch(problem):
176     """Search the node of least total cost first."""
177     """ YOUR CODE HERE """
178
179     #El objetivo es encontrar una ruta al nodo de destino que tenga el coste acumulativo mas b
180
181
182     first = ((problem.getStartState(), None, 0), 0) #Estado inicial, costo inicial 0
183
184     s = util.PriorityQueue() #Creamos una cola de prioridades s
185     s.push(first,0) #Agregamos origen a la cola s y prioridad
186
187     visit = [] #Nodos visitados
188     path = [] #Ruta a devolver
189     adyac = {}
190
191
192     while not s.isEmpty():
193
194         v, p = s.pop() #sacamos un elemento de la cola y su prioridad p
195
196         if not v[0] in visit:
197             if problem.isGoalState(v[0]): #Si es el objetivo delvolvemos la ruta
198
199                 while v[0] != first[0][0]:
200                     path.append(v[1]) #Introducimos ruta de direcciones en p
201                     v = adyac[v] #El nodo v ahora sera el padre
202
203                 path.reverse()
204                 return path #Devolvemos ruta al reves
205
206             visit.append(v[0])
207             successors = problem.getSuccessors(v[0])
208             for suc in successors: #recorremos los nodos sucesores
209                 if suc[0] not in visit:
210
211                     cost = suc[2] + p #Coste sucesor + coste nodo padre
212                     s.push((suc,cost),cost) #Anadimos nodos no visitados a la cola y su coste aci
213                     adyac[suc] = v #Anadimos el nodo a sus hijos
214
215     return []
216

```

For this algorithm we have to find a path of cost for the deepest nodes, so the main challenge compared to the other algorithms is to store this value.

We use a queue that will contain the state and its priority. At the beginning, the priority will be 0 and as we go to the successor nodes, the cost will increase and add to the queue.

4. aStarSearch (A*)

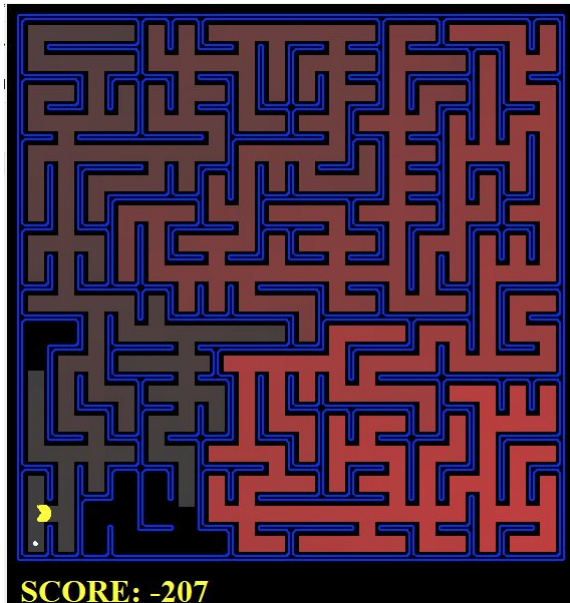
```

228 def aStarSearch(problem, heuristic=nullHeuristic):
229     """Search the node that has the lowest combined cost and heuristic first."""
230     """ YOUR CODE HERE """
231
232     #El algoritmo tiene en cuenta el coste del camino recorrido y el coste de la heuristica (esti
233
234     first = ((problem.getStartState(), None, 0), 0) #Estado inicial, costo inicial 0
235
236     s = util.PriorityQueue() #Creamos una cola de prioridades s
237     s.push(first,0) #Agregamos origen a la cola s y prioridad
238
239     visit = [] #Nodos visitados
240     path = [] #Ruta a devolver
241     adyac = {}
242
243
244     while not s.isEmpty():
245
246         v, p = s.pop() #sacamos un elemento de la cola y su prioridad p
247
248         if not v[0] in visit:
249             if problem.isGoalState(v[0]): #Si es el objetivo delvolvemos la ruta
250
251                 while v[0] != first[0][0]:
252                     path.append(v[1]) #Introducimos ruta de direcciones en p
253                     v = adyac[v] #El nodo v ahora sera el padre
254
255                 path.reverse()
256                 return path #Devolvemos ruta al reves
257
258             visit.append(v[0])
259             successors = problem.getSuccessors(v[0])
260             for suc in successors: #recorremos los nodos sucesores
261                 if suc[0] not in visit:
262                     cost = suc[2] + p #Coste sucesor + coste nodo padre
263                     s.push((suc,cost),cost + heuristic(suc[0],problem)) #Anadimos nodos no visita
264                     adyac[suc] = v #Anadimos el nodo a sus hijos
265
266     return []
267

```

As with the uniformCostSearch algorithm, the main issue is to store not only the cost of the actions as well as to store the value of the heuristic function.

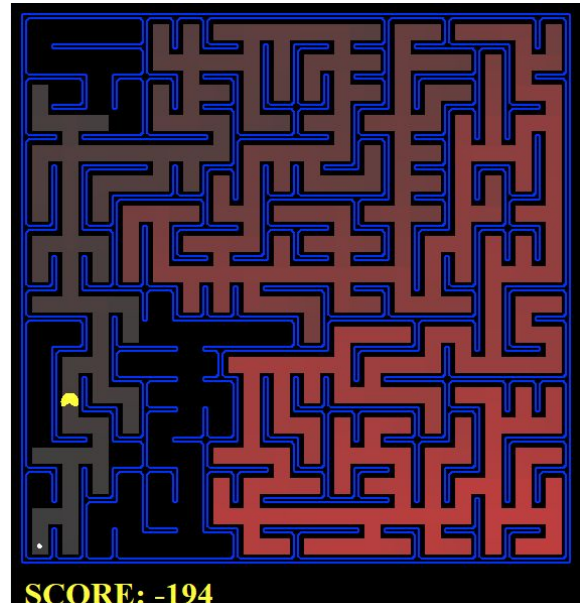
Uniform Cost Search (UCS)



Search nodes expanded: 620

```
[SearchAgent] using function ucs
[SearchAgent] using problem type PositionSearchProblem
Path found with total cost of 210 in 0.1 seconds
Search nodes expanded: 620
Pacman emerges victorious! Score: 300
Average Score: 300.0
Scores:      300.0
Win Rate:    1/1 (1.00)
Record:      Win
```

A* Search



Search nodes expanded: 549

```
[SearchAgent] using function astar and heuristic manhattanHeuristic
[SearchAgent] using problem type PositionSearchProblem
Path found with total cost of 210 in 0.1 seconds
Search nodes expanded: 549
Pacman emerges victorious! Score: 300
Average Score: 300.0
Scores:      300.0
Win Rate:    1/1 (1.00)
Record:      Win
```

SearchAgents

This file stores all the agents that can be controlled on the program. SearchAgent finds a route using algorithms of search and returns actions to follow that path.

5. Finding All the corners

Our goal now is to find the shortest path between 4 corners, whether with food or not. To do this, we implement the class *CornersProblem* that is responsible for finding the path of the 4 corners. The algorithm used in this section is the BFS of the search.

Our implementation looks like:

a. startState getter

```
292 def getStartState(self):
293     """
294     Returns the start state (in your state space, not the full Pacman state
295     space)
296     """
297     """ YOUR CODE HERE """
298     posIni = self.startingPosition #Posicion inicial
299     posCorners = self.corners      #Posiciones de las esquinas del juego
300
301     return posIni, posCorners
302
303     util.raiseNotDefined()
```

Make a return of the start state of the class.

b. isGoalState

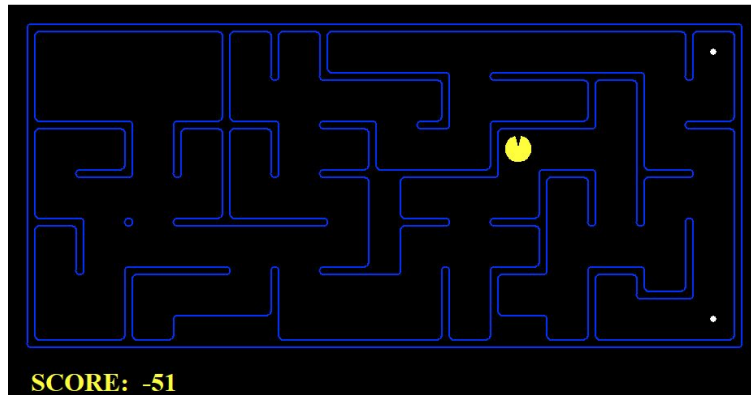
```
305 def isGoalState(self, state):
306     """
307     Returns whether this search state is a goal state of the problem.
308     """
309     """ YOUR CODE HERE """
310
311     if not state[1]:
312         return True
313     return False
314
315     util.raiseNotDefined()
```

To figure out if the the state inserted as a parameter is a goal

c. getSuccessors

```
317 def getSuccessors(self, state):
318
319     successors = []
320     for action in [Directions.NORTH, Directions.SOUTH, Directions.EAST, Directions.WEST]:
321         # Add a successor state to the successor list if the action is legal
322         # Here's a code snippet for figuring out whether a new position hits a wall:
323         # x,y = currentPosition
324         # dx, dy = Actions.directionToVector(action)
325         # nextx, nexty = int(x + dx), int(y + dy)
326         # hitsWall = self.walls[nextx][nexty]
327
328         """ YOUR CODE HERE """
329
330
331         x,y = state[0] #Estado actual
332         dx, dy = Actions.directionToVector(action) #Direccion de la accion
333         nextx, nexty = int(x + dx), int(y + dy) #Nueva posicion
334         hitsWall = self.walls[nextx][nexty]
335         if not hitsWall: #Miramos que la siguiente pos. no s
336
337             food = filter(lambda a: a != (nextx, nexty), state[1]) #Filramos y devolvemos la
338             successors.append(((nextx, nexty), food), action, 1)
339
340     self._expanded += 1 # DO NOT CHANGE
341     return successors
```

Where we must find the successors of a state, the actions and the cost.



6. Corners Heuristic

```
357 def cornersHeuristic(state, problem):
358
359     corners = problem.corners # These are the corner coordinates
360     walls = problem.walls # These are the walls of the maze, as a Grid (game.py)
361
362     """ YOUR CODE HERE """
363
364     posState = state[0]
365     food = state[1]
366     h = 0
367     while food: #Mientras haya comida
368
369         near=999999
370         pos=0
371
372         for f in food: #Para cada pos. donde hay
373             dist = util.manhattanDistance(posState, f) #Calculamos la distancia (
374             if dist < near:
375                 near = dist
376                 pos = f
377
378         h = h + near #Incrementamos el coste p
379         posState = pos #Posicion comida
380         food = filter(lambda el: el != posState, food) #filtaramos y eliminamos
381
382     return h #Devuelve el coste del camino mas corto
```

Returns a number that will always be lower from the current state to the goal. The returned heuristic will always be admissible.

7. Eating All The dots


```
457 def foodHeuristic(state, problem):
458
459     position, foodGrid = state
460
461     """ YOUR CODE HERE """
462
463     heuristic = 0
464     foodList = foodGrid.asList() #Lista de coordenadas food
465
466     if not foodList:
467         return 0
468
469     maxDist = 0
470     for f in foodList: #para cada pos. de la comida, cojemos la distancia mas c
471         dist = mazeDistance(position, f, problem.startingGameState) #Distancia e
472         if dist > maxDist:
473             maxDist = dist #Actualizamos si la distancia de esta food es mayor
474
475     return maxDist #Devolvemos distancia de la comida mas Lejana
```

To find the further capsule, using List and a heuristic value.

8. Suboptimal Search

a. findPathToClosestDot

```
524 def findPathToClosestDot(self, gameState):
525     """
526     Returns a path (a list of actions) to the closest dot, starting from
527     gameState.
528     """
529     # Here are some useful elements of the startState:
530     startPosition = gameState.getPacmanPosition()
531     food = gameState.getFood()
532     walls = gameState.getWalls()
533     problem = AnyFoodSearchProblem(gameState)
534
535
536     """ YOUR CODE HERE """
537
538     return search.aStarSearch(problem)
539     util.raiseNotDefined()
```

Using our previous aStarSearch algorithm, we return a path to find the closest dot (or capsule).

A* is capable of finding the solution having to expand slightly less nodes than BFS.

Used why findPathToClosestDot, is necessary to implement a class called AnyFoodSearchProblem to find out paths to any food, as the name can suggest.

b. isGoalState

```
567 def isGoalState(self, state):
568     """
569     The state is Pacman's position. Fill this in with a goal test that will
570     complete the problem definition.
571     """
572     x,y = state
573
574     """ YOUR CODE HERE """
575
576     food = self.food.asList()
577
578     return (x, y) in food
579     util.raiseNotDefined()
```

Similar to the implementation for the previous class.

Conclusions and issues

For search.py file we have to implement algorithms that we see briefly in previous theory sessions.