Documentation

The generic search algorithm.

The idea for the algorithm can be found here:

https://artint.info/2e/html/ArtInt2e.Ch3.S4.html

The frontier is a set of paths. Initially, the frontier contains the path of zero cost consisting of just the start node. At each step, the algorithm removes a path

if it has **found a solution** and returns the path

. Otherwise, the path is extended by one more arc by finding the neighbors and add to the frontier. This step is known as **expanding** the path

This algorithm has a few features that should be noted:

- Which path is selelected defines the search strategy. The selection of a path can affect the efficiency; see the box for more details on the use of "select".
- It is useful to think of the *return* as a temporary return, where a caller can **retry** the search to get another answer by continuing the while loop. This can be implemented by having a class that keeps the state of the search and a search()

method that returns the next solution.

• If the procedure returns

("bottom"), there are no solutions, or no remaining solutions if the search has been retried.

• The algorithm only tests if a path ends in a goal node *after* the path has been selected from the frontier, not when it is added to the frontier. There are two important reasons for this. There could be a costly arc from a node on the frontier to a goal node. The search should not always return the path with this arc, because a lower-cost solution may exist. This is crucial when the lowest-cost path is required. A second reason is that it may be expensive to determine whether a node is a goal node, and so this should be delayed in case the computation is not necessary.

If the node at the end of the selected path is not a goal node and it has no neighbours, then extending the path means removing the path from the frontier. This outcome is reasonable because this path could not be part of a path from the start node to a goal node.

```
class Node:
 def __init__(self, state, cost=0, action=None, parentNode=None, ):
 self.state = state
self.action = action
self.parentNode = parentNode
self.cost = cost
 def setParentNode(self, state):
  self.parentNode = state
 def getParentNode(self):
  return self.parentNode
def getCost(self):
  return self.cost
def getAction(self):
return self.action
```

```
def getState(self):
   return self.state
   def __eq_ (self, other):
   if isinstance(other, Node):
   if type(self.state) == tuple:
         return self.state[0] == other.state[0] and
len(self.state[1]) == len(other.state[1])
   return self.state == other.getState()
   return False
   def _ hash_ (self) -> int:
   return abs(hash(self.state)) % (10 ** 8)
   def __str__(self):
   return repr(self.state)
   def genericSearchAlgorithm(problem, frontier):
   expanded = []
   firstNode = Node(problem.getStartState())
   frontier.push(firstNode)
```

```
while not frontier.isEmpty():
       current = frontier.pop()
      if problem.isGoalState(current.getState()):
             list = []
            list.append(current.getAction())
           current = current.getParentNode()
      list.reverse()
      if current.getState() not in expanded:
      expanded.append(current.getState())
              for (nst, act, cost) in
problem.getSuccessors(current.getState()):
               node = Node(nst, current.getCost() + cost, act, current)
                frontier.push(node)
```

Question 1 - DFS - this algorithm is based on a LIFO approach because we use a stack to keep track of the nodes - as long as we still have nodes to process, we take the first item from the stack and add it to the "visited" list - each node is stored in a tuple with its actions $\frac{1}{2}$

It uses a queue for the generic search algorithm.

```
frontier = util.Queue()

return genericSearchAlgorithm(problem, frontier)
```

Question 2 - BFS - this algorithm is based on a FIFO approach because we use a queue to keep track of the nodes - as long as we still have nodes to process, we take the first item from the queue and add it to the "visited" list - each node is stored in a tuple with its actions

```
frontier = util.Queue()

return genericSearchAlgorithm(problem, frontier)
```

Question 3 - UCS - this algorithm uses a priority queue to keep track of the nodes - it expands the node with the lowest cost - it searches for a goal state with the lowest cost path

```
fringe = util.PriorityQueue()

visitedList = []

fringe.push((problem.getStartState(), [], 0), 0)

(state, toDirection, toCost) = fringe.pop()

visitedList.append((state, toCost))

while not problem.isGoalState(state):
```

```
successors = problem.getSuccessors(state)
      for son in successors:
       visitedExist = False
          total_cost = toCost + son[2]
      for (visitedState, visitedToCost) in visitedList:
              if (son[0] == visitedState) and (total cost >= visitedToCost):
        break
      if not visitedExist:
toCost + son[2])
       visitedList.append((son[0], toCost + son[2]))
    (state, toDirection, toCost) = fringe.pop()
   return toDirection
   Question 4 - A* - this algorithm uses a priority queue to keep track of the nodes - it
```

is basically the UCS algorithm but we have to keep track on the heuristics given for each node - it searches for a goal state with the lowest cost path it uses. It uses a priority queue for the generic search algorithm

```
print(heuristic)
    new heuristic = lambda x: heuristic(x.getState(), problem) + x.getCost()
    return genericSearchAlgorithm(problem,
util.PriorityQueueWithFunction(new heuristic))
```

Question 7 - Eating all the dots/foodHeuristic - at first we have a function that finds the parent of a set 3 - the second function that makes the union between two sets and finds the common parent by comparing the rank of both sets - these function are used to create a path that will collect all the food - then we create a list of lists which contains the pairs of unvisited foods and the Manhattan distance between them -the next step is to perfom Kruskal algorithm for minimum spanning trees in order to find the shortest path between foods

foodposition = foodGrid.asList()

heuristic = [0]

for pos in foodposition:
 heuristic.append(mazeDistance(position, pos, problem.startingGameState))

return max(heuristic)