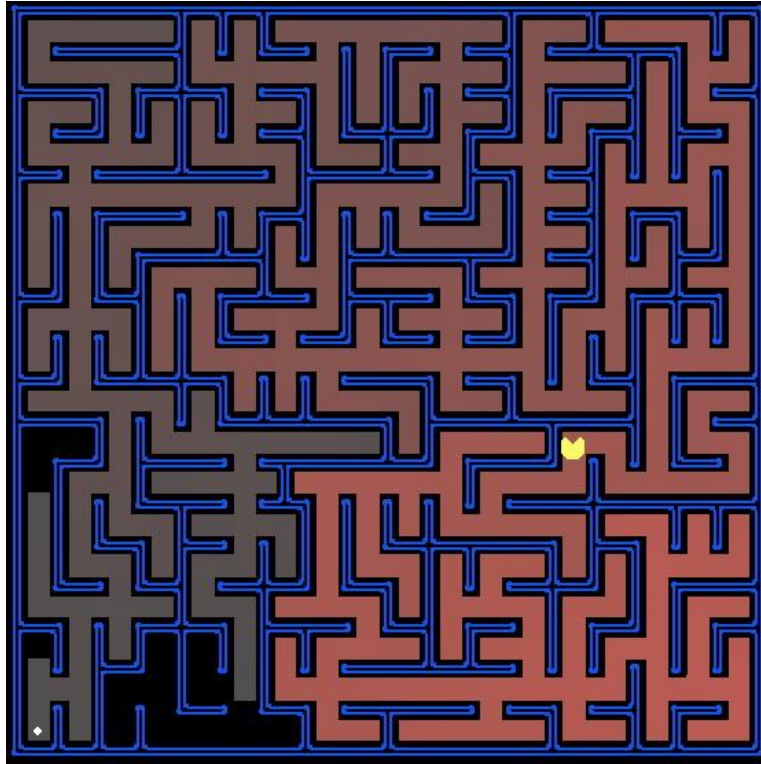


IT097IU: Intro to Artificial Intelligence

Lab#2/Assignment#2

Uninformed Search in Pac-Man



There are two exercises in this lab:

1. Depth-First Search (DFS)
2. Breadth-First Search (BFS)
3. Uniform-Cost Search (UCS)

Introduction

In this assignment, you will continue to work with the archive in the first lab assignment to help your Pac-Man agent find paths through his maze world to reach a particular location. You will build general search algorithms and apply them to many different Pac-Man scenarios.

Files you'll edit:

`search.py` Where all your search algorithms will reside.

`searchAgents.py` Where all your search-based agents will reside.

Files you should look at but NOT edit:

<code>util.py</code>	Useful data structures for implementing search algorithms.
<code>pacman.py</code>	The main file that runs Pac-Man games. This file describes a <code>PacMan GameState</code> type, which you use in this lab.
<code>game.py</code>	The logic behind how the Pac-Man world works. This file describes several supporting types like <code>AgentState</code> , <code>Agent</code> , <code>Direction</code> , and <code>Grid</code> .

Finding a fixed food dot using Uninformed Search

In `searchAgents.py`, you'll find a fully implemented `SearchAgent`, which plans out a path through Pac-Man's world and then executes that path step-by-step. The search algorithms for formulating a plan are not implemented -- that's your job. As you work through the following questions, you might need to refer to the glossary of objects in the code (at the end of this handout).

First, test that the `SearchAgent` is working correctly by running:

```
python pacman.py -l tinyMaze -p SearchAgent -a fn=tinyMazeSearch
```

The command above tells the `SearchAgent` to use `tinyMazeSearch` as its search algorithm, which is implemented in `search.py`. This simply follows a fixed sequence of actions to demonstrate how the code works. Pac-Man should navigate the maze successfully.

Now it's time to write the generic search functions to help Pac-Man plan routes! Pseudocode for the depth-first search, breadth-first search and uniform-cost search algorithms might be useful for the implementation is shown below.

```
function UninformedSearch(problem) returns a list of actions
    initialize the frontier using the initial state of the problem
    initialize exploredSet to empty
    #For explored, use Pacman position as the key with a value True
    #initialize a set of states already explored
    while frontier is not empty do
        choose a leaf node and remove it from the frontier
        if the node contains a goal state
            return list of actions from start state to goal state
        add the state to the exploredSet
        for each successor of the node state
            if the successor state is not in exploredSet
                add node of the successor onto the frontier
    return an empty list (i.e. no solution!)
```

Important note: All your search functions need to return a list of *actions* that will lead the agent from the start to the goal. These actions must be legal moves (valid directions, no moving through walls).

Hint: Algorithms for DFS, BFS and UCS differ only in the details of how the frontier is managed. So, concentrate on getting DFS right and then BFS should be relatively straightforward. Indeed, one possible implementation requires only a single generic search method which is configured with an algorithm-specific queuing strategy. Your implementation need *not* be of this form to receive full credit.

Hint: Make sure to check out the `Stack`, `Queue`, and `PriorityQueue` types provided to you in `util.py`.

EXERCISE 1: Implement the depth-first search algorithm in the `depthFirstSearch` function in `search.py`. Although DFS and BFS ignore the costs, you'll need them for later search methods. Your code should quickly find a solution for:

```
python pacman.py -l tinyMaze -p SearchAgent -a fn=dfs
python pacman.py -l mediumMaze -p SearchAgent -a fn=dfs
python pacman.py -l bigMaze -p SearchAgent -z .5 -a fn=dfs
```

The Pac-Man board will show an overlay of color for the states explored and the order in which they were explored (brighter red means earlier exploration). Is the exploration order what you would have expected? Does Pac-Man actually go to all the explored squares on his way to the goal?

Hint: The solution found by your DFS algorithm for `mediumMaze` should have a length of 130 (provided you push successors onto the frontier in the order provided by `getSuccessors()`; you might get 244 if you push them in the reverse order). Is this a least cost solution? If not, think about what depth-first search is doing wrong.

EXERCISE 2: Implement the breadth-first search algorithm in the `breadthFirstSearch` function in `search.py`. Use the same algorithm as shown in the above pseudocode. Test your code the same way you did for depth-first search.

```
python pacman.py -l tinyMaze -p SearchAgent -a fn=bfs
python pacman.py -l mediumMaze -p SearchAgent -a fn=bfs
python pacman.py -l bigMaze -p SearchAgent -z .5 -a fn=bfs
```

Does BFS find a least cost solution?

Hint: If Pac-Man moves too slowly for you, try the option `--frameTime 0`.

EXERCISE 3: Implement the uniform-cost search algorithm in the `uniformCostSearch` function in `search.py`. Does UCS find a least cost solution? How many nodes are expanded? You should now observe successful behavior in all three of the following layouts, where the agents below are all UCS agents that differ only in the cost function they use (the agents and cost functions are written for you):

```
python pacman.py -l mediumMaze -p SearchAgent -a fn=ucs
python pacman.py -l mediumDottedMaze -p StayEastSearchAgent
python pacman.py -l mediumScaryMaze -p StayWestSearchAgent
```

Note: You should get very low and very high path costs for the `StayEastSearchAgent` and `StayWestSearchAgent` respectively, due to their exponential cost functions (see `searchAgents.py` for details).

Object Glossary

Here's a glossary of the key objects in the code base related to search problems, for your reference:

SearchProblem (`search.py`)

A `SearchProblem` is an abstract object that represents the state space, successor function, costs, and goal state of a problem. You will interact with any `SearchProblem` only through the methods defined at the top of `search.py`.

PositionSearchProblem (`searchAgents.py`)

A specific type of `SearchProblem` that you will be working with --- it corresponds to searching for a single pellet in a maze.

Search Function

A search function is a function which takes an instance of `SearchProblem` as a parameter, runs some algorithm, and returns a sequence of actions that lead to a goal. Example of search functions are `depthFirstSearch` and `breadthFirstSearch`, which you have to write. You are provided `tinyMazeSearch` which is a very bad search function that only works correctly on `tinyMaze`.

SearchAgent

`SearchAgent` is a class which implements an `Agent` (an object that interacts with the world) and does its planning through a search function. The `SearchAgent` first uses the search function provided to make a plan of actions to take to reach the goal state, and then executes the actions one at a time.

What to submit

1. Fill out the table below:

	Depth-First Search			Breadth-First Search			Uniform-Cost Search		
Maze	#nodes explored	Solution length	Is it optimal?	#nodes explored	Solution length	Is it optimal?	#nodes explored	Solution length	Is it optimal?
tiny	15	10	No	15	8	Yes	15	8	Yes
medium	146	130	No	269	68	Yes	269	68	Yes
big	390	210	No	620	210	Yes	620	210	Yes

2. Based on the above, a short discussion/reflection of how the searches compare.
3. Source code + README (how to compile and run your code)
4. Please create a folder called "yourname_IT097IU_Lab2" that includes all the required files and generate a zip file called "yourname_IT097IU_Lab2.zip".
5. Please submit your work (.zip) to Blackboard.

2.

- The UCS is implemented by a priority queue data structure. Each inserted item has a priority associated with it and the client is usually interested in quick retrieval of the lowest-priority item in the queue. This data structure allows $O(1)$ access to the lowest-priority item. Which seems like get the optimal most of the time.

- The DFS is implemented by a stack data structure. Which is a container with a last-in-first-out (LIFO) queuing policy, and seem to give the complete results, but not optimal all the time.

- The BFS is implemented by a queue data structure. Which is a container with a first in first out (FIFO) queuing policy and give complete result which also optimal in these 3 cases.