

## Motion Planning: Discrete planner

You have been assigned to extend a library of discrete motion planners for an holonomic robot. The robot moves in a static and flat environment and must find a path to a goal. All the planners share the same common interface and should implement the **search** method that has the following signature:

```
search(world_state, robot_pose, goal_pose) return path
```


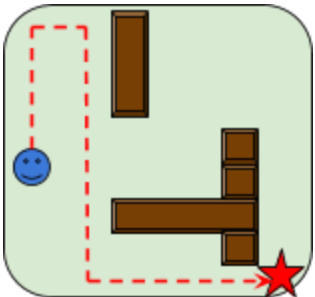
**world\_state** is a 2D-grid representation of the environment where the value 0 indicates a navigable space and the value 1 indicates an occupied/obstacle space.

**robot\_pose** is a tuple of two indices (x, y) which represent the current pose of the robot in **world\_state**.

**goal\_pose** is a tuple of two indices (x, y) which represent the goal in **world\_state** coordinate system.

**path** is a list of tuple (x, y) representing a path from the **robot\_pose** to the **goal\_pose** in **world\_state**, or **None** if no path has been found

Here is a simple example that shows the correspondence between an environment and a representation.

|   |   |
|---|---|
| <p><b>Environment</b></p>  <p>Legend:</p> <ul style="list-style-type: none"><li>Obstacle</li><li>Robot</li><li>Goal</li></ul> | <p><b>Corresponding world_state, robot_pose and goal_pose</b></p> <pre>world_state = [[0, 0, 1, 0, 0, 0],<br/>               [0, 0, 1, 0, 0, 0],<br/>               [0, 0, 0, 0, 1, 0],<br/>               [0, 0, 1, 1, 1, 0],<br/>               [0, 0, 0, 0, 1, 0],<br/>               [0, 0, 0, 0, 0, 0]]</pre> <pre>robot_pose = (2, 0)<br/>goal_pose  = (6, 6)</pre> |
|    | <p>Example of a valid path shown in red on the left figure</p> <pre>path = [(2, 0), (1, 0), (0, 0), (0, 1), (1, 1),<br/>        (2, 1), (3, 1), (4, 1), (5, 1), (6, 1),<br/>        (6, 2), (6, 3), (6, 4), (6, 5), (6, 6)]</pre>   |

We want you to write two discrete planners as described below.

- **Random planner**

The random planner tries to find a path to the goal by randomly moving in the environment (only orthogonal moves are legal). If the planner can not find an acceptable solution in less than `max_step_number`, the search should fail. The random planner, while being erratic, has a short memory, and it will never attempt to visit a cell that was visited in the last `sqrt(max_step_number)` steps except if this is the only available option.

- **Optimal planner:** A planner that goes to the goal with the shortest (non-colliding) path. Again, only orthogonal moves are legal.

Compare the performance of the two planners.

You are expected to **write documentation** and **test correctness** for your code. You should also indicate the complexity of each solution in the documentation. You can either use Python 2.7 and/or C++. For Python, Numpy library may be used. For C++, boost and stl libraries may be use.