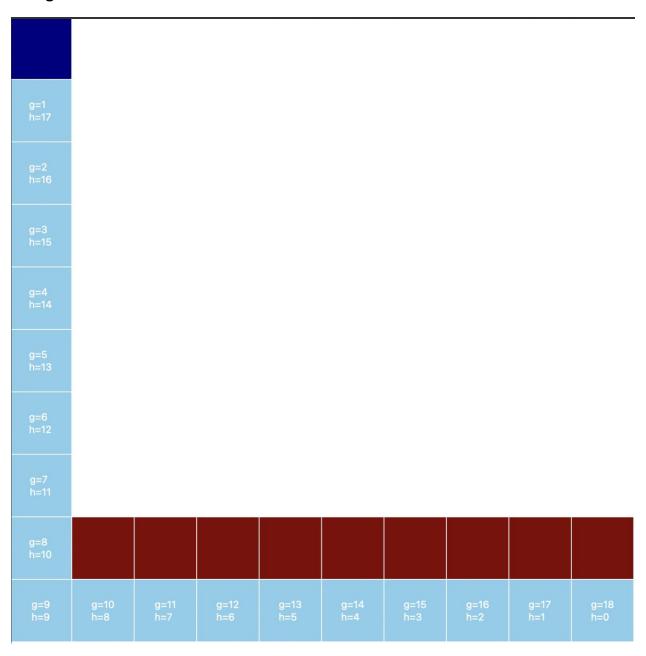
## Question 1:

## A\* Algorithm



**Explanation:** In a A\* search, because it takes into consideration both path cost and future cost, A\* takes a more balanced approach. It expands nodes that balance shortest path so far g(n) with closeness to goal h(n).

## **Greedy Best-First**

	g=0 h=17	g=0 h=16	g=0 h=15	g=0 h=14	g=0 h=13	g=0 h=12	g=0 h=11	g=0 h=10	g=0 h=9
									g=0 h=8
									g=0 h=7
									g=0 h=6
									g=0 h=5
									g=0 h=4
									g=0 h=3
g=0 h=11	g=0 h=10	g=0 h=9	g=0 h=8	g=0 h=7	g=0 h=6	g=0 h=5	g=0 h=4	g=0 h=3	g=0 h=2
g=0 h=10									
g=0 h=9	g=0 h=8	g=0 h=7	g=0 h=6	g=0 h=5	g=0 h=4	g=0 h=3	g=0 h=2	g=0 h=1	g=0 h=0

**Explanation:** In a Greedy Best-First search, it ignores path cost and only takes into consideration future cost. In doing so, the search attempts to "beeline" to the goal but is blocked by a wall. It is forced to travel along the wall until it isn't blocked.

## Removal of g(n)

```
#### Greedy Best-First Algorithm
def find path(self):
    #### Add the start state to the queue
open_set.put((0, self.agent_pos))
     #### Continue exploring until the queue is exhausted
while not open_set.empty():
          current_cost, current_pos = open_set.get()
current_cell = self.cells[current_pos[0]][current_pos[1]]
           #### Stop if goal is reached
if current_pos == self.goal_pos:
                 self.reconstruct_path()
          #### Agent goes E, W, N, and S, whenever possible for dx, dy in [(0, 1), (0, -1), (1, 0), (-1, 0)]: new_pos = (current_pos[0] + dx, current_pos[1] + dy)
                 if 0 <= new_pos[0] < self.rows and 0 <= new_pos[1] < self.cols and not self.cells[new_pos[0]][new_pos[1]].is_wall:
                      #### Eliminates cost of moving to a new position
new_g = current_cell.g
                      if new_g < self.cells[new_pos[0]][new_pos[1]].g:
    ### Update the path cost g()</pre>
                            ### Update the path cost g()
self.cells[new_pos[0]][new_pos[1]].g = new_g
                            self.cells[new_pos[0]][new_pos[1]].h = self.heuristic(new_pos)
                            ### Update the evaluation function for the cell n: f(n) = h(n) self.cells[new_pos[0]][new_pos[1]].f = new_g + self.cells[new_pos[0]][new_pos[1]].h ## g(n) = 0 in search self.cells[new_pos[0]][new_pos[1]].parent = current_cell
                            #### Add the new cell to the priority queue
open_set.put((self.cells[new_pos[0]][new_pos[1]].f, new_pos))
```

**Explanation:** To achieve a Greedy Best-First search, I altered the A\* Algorithm in AStarMaze by changing this snippet of code:

```
new g = current cell.g + 1 \rightarrow new g = current cell.g
```

By doing so, g(n) is constant and does not accumulate path cost and therefore is not calculated into the f(n) function. The function now looks like this:

$$f(n) = g(n) + h(n)$$
  $\rightarrow$   $f(n) = h(n)$ 

The algorithm no longer takes past cost g(n) into consideration, only concerning itself with the estimate of future cost h(n), making it greedy.

**Conclusion:** A\* and Greedy algorithms differ in their best path to the goal. A\* will take a balanced approach and consider both future and past costs. Greedy Best-First searches will result in a "beeline" to the goal, often resulting in inference with obstacles and resulting in a longer path. By making a small tweak to the code, by removing accumulation of g(n), will result in a Greedy Best-First algorithm.