# Akhil\_Vallala\_Project\_1

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Akhil Vallala's Project 1

### 1 Importing packages

```
[4]: # Array math
  import numpy as np

# Iteration tracking
  from tqdm import tqdm

# Type hinting
  from typing import Tuple

# Plotting
  import matplotlib.pyplot as plt
  import matplotlib
  import seaborn as sns

# Os traversal
  import os
```

For rendering Greek letters in Latex for plotting

```
[5]: matplotlib.rcParams['font.family'] = 'sans-serif'
matplotlib.rcParams['text.usetex'] = True
matplotlib.rcParams["text.latex.preamble"] = r"\usepackage{cmbright}"
matplotlib.rcParams.update(matplotlib.rcParamsDefault)
```

# 2 Defining the ploting functions

#### 2.1 Array Index to Matplot Coords Converter

#### 2.2 Maze Plotter

```
[7]: def plot_matrix(
         M: np.array,
         goal_coords: list = [],
         hole_coords: list = [],
         start_coords: list = [],
         highlight_coords: list = [],
         img_width: int = 5,
         img_height: int = 5,
         title: str = None,
         filename: str = None,
         ) -> None:
         Plots a matrix as an image.
         height, width = M.shape
         fig = plt.figure(figsize=(img_width, img_width))
         ax = fig.add_subplot(111, aspect='equal')
         for x in range(height):
             for y in range(width):
                 # By default, the (0, 0) coordinate in matplotlib is the bottom_{f \sqcup}
      →left corner,
                 # so we need to invert the y coordinate to plot the matrix correctly
                 matplot_x, matplot_y = array_index_to_matplot_coords(x, y, height)
                 # If there is a tuple of (x, y) in the goal coords list, we color
      ⇔the cell gray
                 if (x, y) in goal coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
      amatplot_y - 0.5), 1, 1, facecolor='lightgreen'))
                 # If there is a tuple of (x, y) in the hole_coords list, we color_
      ⇔the cell salmon
                 elif (x, y) in hole_coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
      →matplot_y - 0.5), 1, 1, facecolor='salmon'))
                 # If there is a tuple of (x, y) in the start_coords list, we color
      ⇔the cell yellow
                 elif (x, y) in start_coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
      →matplot_y - 0.5), 1, 1, facecolor='yellow'))
```

```
# If there is a tuple of (x, y) in the highlight_coords list, we
⇔color the cell lightblue
           elif (x, y) in highlight_coords:
               ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
amatplot_y - 0.5), 1, 1, facecolor='lightblue'))
           ax.annotate(str(M[x][y]), xy=(matplot_x, matplot_y), ha='center', u

ya='center')
  offset = .5
  ax.set_xlim(-offset, width - offset)
  ax.set_ylim(-offset, height - offset)
  ax.hlines(y=np.arange(height+1)- offset, xmin=-offset, xmax=width-offset)
  {\tt ax.vlines(x=np.arange(width+1) - offset, ymin=-offset, ymax=height-offset)}\\
  plt.title(title)
  if filename is not None:
      plt.savefig(filename)
      plt.close(fig)
  else:
      plt.show()
```

```
[8]: def plot_state_visits(
             S: np.array,
             visit_dict: dict,
             img_width: int = 5,
             img_height: int = 5,
             ):
         11 II II
         Plots the states and colors them by the number of visits.
         The more visits a state has, the darker the color.
         n n n
         # Get the number of rows and columns
         n_rows, n_cols = S.shape
         # Create a new matrix to hold the number of visits
         V = np.zeros((n_rows, n_cols))
         # Iterate through the visit dictionary and update the V matrix
         for s, visits in visit_dict.items():
             # Converting the state to an array index
             s_index = np.where(S == s)
             row, col = s_index[0][0], s_index[1][0]
             V[row, col] = visits
```

```
fig = plt.figure(figsize=(img_width, img_height))
ax = fig.add_subplot(111, aspect='equal')

# Ploting the matrix
sns.heatmap(V, cmap='Blues', cbar=False, annot=True, fmt='.0f', ax=ax)
plt.show()
```

#### 2.3 Policy Plotter

```
[9]: # Defining the plotting function for the policy
     def plot policy(
             S: np.ndarray,
             policy: dict,
             goal_coords: tuple,
             wall_coords: list,
             start_coords: tuple,
             optimal_policy_path: list,
             title: str,
             img_width: int = 6,
             img_height: int = 6
             ):
         height, width = S.shape
         fig = plt.figure(figsize=(img_width, img_height))
         ax = fig.add_subplot(111, aspect='equal')
         for x in range(height):
             for y in range(width):
                 matplot_x, matplot_y = array_index_to_matplot_coords(x, y, height)
                 # If there is a tuple of (x, y) in the hole_coords list, we coloru
      → the cell gray
                 if (x, y) in wall_coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
      matplot_y - 0.5), 1, 1, facecolor='gray'))
                 # If there is a tuple of (x, y) in the goal_coords list, we coloru
      → the cell yellow
                 elif (x, y) in goal_coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_

→matplot_y - 0.5), 1, 1, facecolor='lightgreen'))
                 else:
                     try:
                         # If there is a tuple of (x, y) in the start_coords list,
      →we color the cell green
                         if (x, y) in start_coords:
```

```
ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 
⇔0.5, matplot_y - 0.5), 1, 1, facecolor='yellow'))
                        # Adding the arrows to the plot
                        if 0 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, 0, 0.3, head_width⊔
\Rightarrow= 0.05, head length = 0.05)
                        if 1 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, 0, -0.3, head_width⊔
\Rightarrow 0.05, head_length = 0.05)
                        if 2 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, -0.3, 0, head_width_
\Rightarrow 0.05, head_length = 0.05)
                        if 3 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, 0.3, 0, head_width_
\Rightarrow 0.05, head_length = 0.05)
                    elif (x, y) in optimal_policy_path:
                        ax.add_patch(matplotlib.patches.Rectangle((matplot_x -__
⇔0.5, matplot_y - 0.5), 1, 1, facecolor='lightblue'))
                        # Adding the arrows to the plot
                        if 0 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, 0, 0.3, head_width_
\Rightarrow 0.05, head_length = 0.05)
                        if 1 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, 0, -0.3, head_width_
\Rightarrow 0.05, head_length = 0.05)
                        if 2 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, -0.3, 0, head_width_
\Rightarrow 0.05, head_length = 0.05)
                        if 3 == policy[S[x, y]]:
                            plt.arrow(matplot_x, matplot_y, 0.3, 0, head_width_
\Rightarrow 0.05, head_length = 0.05)
                except Exception as e:
                    print(f"Error: {e}")
                    print(f"Current x and y: {x}, {y}")
   offset = .5
   ax.set_xlim(-offset, width - offset)
   ax.set_ylim(-offset, height - offset)
   ax.hlines(y=np.arange(height+1)- offset, xmin=-offset, xmax=width-offset)
   ax.vlines(x=np.arange(width+1) - offset, ymin=-offset, ymax=height-offset)
   plt.title(title)
```

```
plt.show()
```

#### 2.4 Q Table Plotter

```
[10]: # Defining the ploting function for the policy
     def plot_Q_Table(
             S: np.ndarray,
             Q: np.ndarray,
             goal_coords: tuple,
             wall coords: list,
             start_coords: tuple,
             title: str,
             img_width: int = 6,
             img_height: int = 6,
             filename: str = None
             ):
         height, width = S.shape
         fig = plt.figure(figsize=(img_width, img_height))
         ax = fig.add_subplot(111, aspect='equal')
         for x in range(height):
             for y in range(width):
                 matplot_x, matplot_y = array_index_to_matplot_coords(x, y, height)
                 # If there is a tuple of (x, y) in the hole coords list, we color
       ⇔the cell gray
                 if (x, y) in wall_coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
       →matplot_y - 0.5), 1, 1, facecolor='gray'))
                 # If there is a tuple of (x, y) in the goal_coords list, we color
       ⇔the cell yellow
                 elif (x, y) in goal_coords:
                     ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5, ___
       →matplot_y - 0.5), 1, 1, facecolor='lightgreen'))
                 else:
                     try:
                         # If there is a tuple of (x, y) in the start_coords list,
       →we color the cell green
                         if (x, y) in start_coords:
                             ax.add_patch(matplotlib.patches.Rectangle((matplot_x -_u
```

```
plt.arrow(matplot_x, matplot_y, 0, 0.2, head_width = 0.05,__
→head_length = 0.05)
                   ax.annotate(f''\{Q[S[x, y]][0]:.2f\}'', xy=(matplot_x,__

→matplot_y + 0.5), ha='center', va='top')
                   plt.arrow(matplot_x, matplot_y, 0, -0.2, head_width = 0.05,__
head_length = 0.05)
                   ax.annotate(f''\{Q[S[x, y]][1]:.2f\}'', xy=(matplot_x,_

→matplot_y - 0.5), ha='center', va='bottom')
                   plt.arrow(matplot_x, matplot_y, -0.2, 0, head_width = 0.05,__
→head_length = 0.05)
                   ax.annotate(f''{Q[S[x, y]][2]:.2f}", xy=(matplot_x - 0.5, \Box
→matplot_y), ha='left', va='center')
                   plt.arrow(matplot_x, matplot_y, 0.2, 0, head_width = 0.05, u
→head_length = 0.05)
                   ax.annotate(f''{Q[S[x, y]][3]:.2f}", xy=(matplot_x + 0.5,_\text{\text{\text{\text{o}}}}
matplot_y), ha='right', va='center')
               except Exception as e:
                   print(f"Error: {e}")
                   print(f"Current x and y: {x}, {y}")
  offset = .5
  ax.set_xlim(-offset, width - offset)
  ax.set_ylim(-offset, height - offset)
  ax.hlines(y=np.arange(height+1)- offset, xmin=-offset, xmax=width-offset)
  ax.vlines(x=np.arange(width+1) - offset, ymin=-offset, ymax=height-offset)
  plt.title(title)
  if filename is not None:
      plt.savefig(filename)
      plt.close(fig)
  else:
      plt.show()
```

```
title: str,
      img_width: int = 6,
      img_height: int = 6,
      filename: str = None
      ):
  height, width = S.shape
  fig = plt.figure(figsize=(img_width, img_height))
  ax = fig.add_subplot(111, aspect='equal')
  for x in range(height):
      for y in range(width):
          matplot_x, matplot_y = array_index_to_matplot_coords(x, y, height)
          # If there is a tuple of (x, y) in the hole coords list, we color
→ the cell gray
          if (x, y) in wall_coords:
              ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
→matplot_y - 0.5), 1, 1, facecolor='gray'))
          # If there is a tuple of (x, y) in the goal_coords list, we color_

→ the cell yellow
          elif (x, y) in goal_coords:
              ax.add_patch(matplotlib.patches.Rectangle((matplot_x - 0.5,_
else:
              try:
                  # If there is a tuple of (x, y) in the start_coords list,
→we color the cell green
                  if (x, y) in start_coords:
                      ax.add_patch(matplotlib.patches.Rectangle((matplot_x -__
⇔0.5, matplot_y - 0.5), 1, 1, facecolor='yellow'))
                  ax.annotate(f''\{\max(Q[S[x, y]]):.2f\}'', xy=(matplot_x,__
matplot_y), ha='center', va='center')
                  # Adding the arrows to the plot
                  if 0 == np.argmax(Q[S[x, y]]):
                      plt.arrow(matplot_x, matplot_y + 0.1, 0, 0.3, __
→head_width = 0.05, head_length = 0.05)
                  if 1 == np.argmax(Q[S[x, y]]):
                      plt.arrow(matplot_x, matplot_y - 0.1, 0, -0.3,_
head_width = 0.05, head_length = 0.05)
                  if 2 == np.argmax(Q[S[x, y]]):
```

```
plt.arrow(matplot_x - 0.1, matplot_y, -0.3, 0,_
head_width = 0.05, head_length = 0.05)
                  if 3 == np.argmax(Q[S[x, y]]):
                      plt.arrow(matplot_x + 0.1, matplot_y, 0.3, 0, __
head_width = 0.05, head_length = 0.05)
              except Exception as e:
                  print(f"Error: {e}")
                  print(f"Current x and y: {x}, {y}")
  offset = .5
  ax.set_xlim(-offset, width - offset)
  ax.set_ylim(-offset, height - offset)
  ax.hlines(y=np.arange(height+1)- offset, xmin=-offset, xmax=width-offset)
  ax.vlines(x=np.arange(width+1) - offset, ymin=-offset, ymax=height-offset)
  plt.title(title)
  if filename is not None:
      plt.savefig(filename)
      plt.close(fig)
  else:
      plt.show()
```

## 3 The agent

The agent will be an object created by the class Agent. All the uptades of the q table will be done internally in the agent object.

```
alpha: float = 0.1,
       epsilon: float = 0.1,
       seed: int = 42,
       ) -> None:
       11 11 11
       Creates an agent for the maze environment.
       Parameters
       _____
       nrow maze : int
           The number of rows in the maze.
       ncol_maze : int
           The number of columns in the maze.
       actions : list, optional
           A list of actions that the agent can take. The default is [0, 1, 2,\Box
⇔3] .
           0: Up
           1: Down
           2: Left
           3: Right
       rewards : dict, optional
           A dictionary of rewards for the agent. The default is {'step': -1,__

    'wall': -10, 'goal': 10}.

       gamma : float, optional
           The discount factor. The default is 0.9.
       alpha: float, optional
           The learning rate. The default is 0.1.
       epsilon: float, optional
           The exploration rate. The default is 0.1.
       seed: int, optional
           The seed for the random generator. The default is 42.
      self.nrow_maze = nrow_maze
      self.ncol_maze = ncol_maze
      self.rewards = rewards
      self.gamma = gamma
      self.alpha = alpha
      self.epsilon = epsilon
      self.seed = seed
      self.actions = actions
       # By default, the starting index is 0 0
      self.start_state = 0
       # By default, the steps to goal is 0
       self.steps_to_goal = 0
```

```
# Initialize an empty list of steps to goals for all episodes
self.list_steps_to_goals = []
# By default, the goal index is the last index
self.goal_state = nrow_maze * ncol_maze - 1
# Creating the random generator with a fixed seed
self.random_generator = np.random.default_rng(seed)
\# Creating the maze; We will denote it internaly as S
self.init_S_table()
# Initiating the Q-table
self.init_Q_table()
# Saving the initial past_action and past_state
self.past_action = None
self.past_state = None
# Creating the action name dictionary
self.action_name_dict = {
   0: 'up',
   1: 'down',
   2: 'left',
    3: 'right',
}
# Counter for the number of times our agent has seen the terminal state
self.num_goal_reached = 0
# Counter for each state and how many times the agent visited each
self.state_visit_counter = {}
# Empty dictionary of states visition paths
self.state_visit_paths = {}
# Empty dictionary of Q values
self.Q_dict = {}
# Placeholder for the current episode of learning
self.current_episode = 0
# Defining the path to intermediate images
self.path_to_images_V = "q-learning-V-Table_Progress"
# Defining the path to intermediate images
```

```
self.path_to_images_Q = "q-learning-Q-Table_Progress"
  def increment_state_visit(self, state) -> None:
       Increments the state visit counter for the state.
       if state in self.state_visit_counter:
           self.state_visit_counter[state] += 1
       else:
           self.state_visit_counter[state] = 1
  def get_most_recent_action(self) -> str:
      Returns the name of the most recent action.
      return self.action_name_dict[self.past_action]
  def init_S_table(self):
       11 11 11
       Creates an internal S table where the rows of the table are
       the states and the columns are the actions.
       .....
      self.S = np.arange(0, self.nrow_maze * self.ncol_maze).reshape(self.
⇔nrow maze, self.ncol maze)
  def init_Q_table(self):
       Creates an internal q table where the rows of the table are
       the states and the columns are the actions.
       self.Q = np.zeros((self.S.size, len(self.actions)))
  def init_reward_dict(self):
       Creates a dictionary where the keys are the states and the values are \sqcup
\hookrightarrow the rewards for transitioniting to that state.
       if self.rewards.get('step') is None:
           raise ValueError("You must specify a reward for taking a step.")
       else:
           self.reward_dict = {s: self.rewards['step'] for s in self.S.
→flatten()}
       if self.rewards.get('goal') is None:
           raise ValueError ("You must specify a reward for reaching the goal ...
⇔state.")
      else:
```

```
self.reward_dict[self.goal_state] = self.rewards['goal']
       if self.rewards.get('wall') is not None:
           # Setting the reward for the wall states
           for wall_state in self.wall_states:
               self.reward_dict[wall_state] = self.rewards['wall']
  def init_maze(self, maze_density: int = None):
       Creates an array of states in a maze environment.
      Returns
      maze : np.array
          A 2D array of states in a maze environment.
      wall_coords : list
          A 1D array of wall states in the maze environment.
       start_coords: tuple
           A tuple of start coordinates in the maze environment.
       goal_coords: tuple
          A tuple of goal coordinates in the maze environment.
       # If maze_density is None we will create walls in 20% of the maze
      if maze density is None:
           maze_density = int(self.nrow_maze * self.ncol_maze * 0.2)
       # Creating the wall states
      wall_states = self.random_generator.choice(range(1, self.nrow_maze *_
→self.ncol_maze), size=maze_density, replace=False)
       # Getting the list of wall coordinates
      wall_coords = [self.get_state_coords(s) for s in wall_states]
       # Generating a starting state from the states that are NOT walls
       start_state = self.random_generator.choice(np.setdiff1d(self.S,__
⇔wall_states))
       # Getting the starting coordinates
      start_coords = self.get_state_coords(start_state)
       # Generating a goal state from the states that are NOT walls and NOT_{f L}
⇔the starting state
       goal_state = self.random_generator.choice(np.setdiff1d(np.
⇔setdiff1d(self.S, wall_states), start_state))
       # Getting the goal coordinates
      goal_coords = self.get_state_coords(goal_state)
```

```
# Saving the wall coordinates, start coordinates, and goal coordinates
      self.wall_coords = wall_coords
      self.start_coords = [start_coords]
      self.goal_coords = [goal_coords]
       # Saving the indexes for the wall, start and goal states
      self.wall_states = wall_states
      self.start_state = start_state
      self.goal_state = goal_state
       # Initiating the reward dictionary
      self.init_reward_dict()
       # Initiating the agent
      self.init_agent()
  def get_state_index(self, row: int, col: int) -> int:
      Returns the state index given the state coordinates.
      An inverse function of get_state_coords()
      if (row < 0 or row >= self.nrow maze or col < 0 or col >= self.
⇔ncol maze):
          return -1
      else:
          return self.S[row][col]
  def get_state_coords(self, s) -> tuple:
      Returns the state coordinates given the state index
      s_index = np.where(self.S == s)
      if len(s_index[0]) == 0:
          return -1, -1
      return s_index[0][0], s_index[1][0]
  def get_action(
           self,
           ) -> int:
      Returns a random action from the set of actions
       The actions are:
       0: up
```

```
1: down
    2: left
    3: right
    HHHH
    return self.random_generator.choice(self.actions)
def argmax(self, q_values: np.array):
    """argmax with random tie-breaking
    Args:
        q_values (Numpy array): the array of action values
    Returns:
        action (int): an action with the highest value
    top = float("-inf")
    ties = []
    for i in range(len(q_values)):
        if q_values[i] > top:
            top = q_values[i]
            ties = []
        if q_values[i] == top:
            ties.append(i)
    return self.random_generator.choice(ties)
def get_greedy_action(self, state: int) -> int:
    11 11 11
    Returns the greedy action given the current state
    # Getting the q values for the current state
    q_values = self.Q[state]
    # Getting the greedy action
    greedy_action = self.argmax(q_values)
    # Returning the greedy action
    return greedy_action
def get_epsilon_greedy_action(self, state: int) -> int:
    Returns an epsilon greedy action
    if self.random_generator.random() < self.epsilon:</pre>
        return self.get_action()
    else:
        return self.get_greedy_action(state)
```

```
# Logging of the agent paths
  def log_agent_move(self, state: int):
      Logs the agent's move
      if self.state_visit_paths.get(self.current_episode, None) is None:
          self.state_visit_paths[self.current_episode] = [state]
      else:
          self.state_visit_paths[self.current_episode].append(state)
  def update_Q_table(self, new_state: int):
      Function that applies the RL update function
      # Getting the next_state's reward
      reward = self.reward_dict[new_state]
      # Saving the current Q value
      current_Q = self.Q[self.past_state][self.past_action]
      # If the new state is the terminal state or the wall state, then the
⊶max Q is O
      max_Q = 0
      # Else we get the max Q value for the new state
      if new_state != self.goal_state:
          new_state_Q_values = self.Q[new_state]
          # Getting the max Q value
          max_Q = np.max(new_state_Q_values)
      # Updating inplace the Q value
      self.Q[self.past_state][self.past_action] = current_Q + self.alpha *_
def get_next_state(self, s: int, action: int) -> int:
       Given the current state and the current action, returns the next state\sqcup
\hookrightarrow index
      # Getting the state coordinates
      s_row, s_col = self.get_state_coords(s)
```

```
# Setting the boolean indicating that we have reached the terminal state
       reached_terminal = False
       # Getting the next state
       next_state = -1
       if action == 0:
           next_state = self.get_state_index(s_row - 1, s_col)
       elif action == 1:
           next state = self.get state index(s row + 1, s col)
       elif action == 2:
           next_state = self.get_state_index(s_row, s_col - 1)
       elif action == 3:
           next_state = self.get_state_index(s_row, s_col + 1)
       # If next state is a wall or the agent is out of bounds, we will stay ...
\hookrightarrow in the same state
       if (next_state == -1) or (next_state in self.wall_states):
           return s, reached_terminal
       # Incrementing the number of times we have visited the next state
       self.increment state visit(next state)
       # If next_state is the goal state, we will return to the starting state
       if next_state == self.goal_state:
           # Incrementing the number of times our agent has reached the goal \Box
\hookrightarrowstate
           self.num goal reached += 1
           reached_terminal = True
       # Returning the next state
       return next_state, reached_terminal
  def init_agent(self):
       We will set the past state and past action as the starting state and \sqcup
\Rightarrowaction
       # Setting the previous state as the starting state
       self.past_state = self.start_state
       self.past_action = self.get_epsilon_greedy_action(self.past_state)
       self.num_goal_reached = 0
  def save_steps_to_goal_for_current_episode(self):
       HHHH
       Write your own code here to save the steps to goal for current episode
```

```
#print(f"Episode {self.current episode}: Steps to goal = {self.
⇔steps_to_qoal}")
       #In list_steps_to_goals list, steps_to_goal value is appended to have_
→ list of goals for all episodes
       self.list_steps_to_goals.append(self.steps_to_goal)
      return self.list_steps_to_goals
  def terminal step(self, new state: int):
       \mathit{Updates} the agent one last time and resets the agent to the \mathit{starting}_\sqcup
\neg position
       # Updating the Q table
      self.update_Q_table(new_state)
       # Resetting the agent
      self.past_state = self.start_state
       self.past_action = self.get_epsilon_greedy_action(self.past_state)
       # save the steps taken to reach the goal for this episode, and reset \Box
⇔steps to goal to 0 for next episode
       self.save_steps_to_goal_for_current_episode()
       self.steps_to_goal = 0
       # Incrementing the number of episodes
       self.current_episode += 1
  def move agent(self):
      11 11 11
       The function that moves the agent to the next state
       # Getting the next state
      next_state, reached_terminal = self.get_next_state(self.past_state,__
⇔self.past_action)
       # Adding the next state to the path
      self.log_agent_move(next_state)
       # Updating the Q table
      if not reached_terminal:
           # Checking if the past_state is the same as the next_state; If that_
⇔is true, it means our agent hit a wall
```

```
# or went out of bounds
           if self.past_state != next_state:
               self.update_Q_table(next_state)
           # Setting the past_state as the next_state
           self.past_state = next_state
           # Getting the next action
           self.past_action = self.get_epsilon_greedy_action(self.past_state)
           # add 1 step
           self.steps_to_goal += 1
      else:
           # add 1 step
           self.steps_to_goal += 1
           self.terminal_step(next_state)
  def train_episodes(self, num_episodes: int):
      Function that trains the agent for one episode
       # Calculating the episode number to end the training
      end_episode = self.current_episode + num_episodes - 1
       # Moving the agent until we reach the goal state
      while self.current_episode != end_episode:
           #print(f"Episode {self.current_episode}")
           self.move_agent()
  def create_policy(self):
       Creates a policy dictionary where the key is the state and the value is \sqcup
\hookrightarrow the action
       based on the Q table
       # Creating the policy dictionary
      self.policy = {}
       # Looping through the states
      for state in range(self.S.size):
           # Getting the greedy action
           greedy_action = self.argmax(self.Q[state])
```

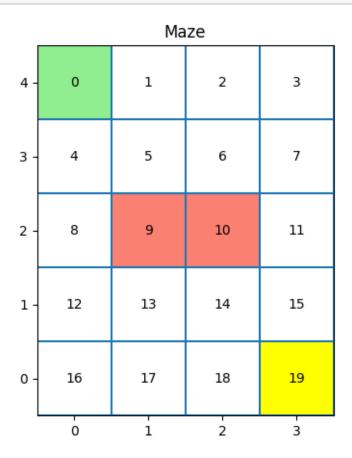
```
# Adding the state and action to the policy dictionary
          self.policy[state] = greedy_action
  def create_optimal_policy_path(self):
      Creates the path of the optimal policy, starting from the starting state
      # Creating the policy path
      self.optimal_policy_path = [self.start_state]
      self.optimal_policy_path_coords = [self.get_state_coords(self.
⇔start_state)]
      # Getting the current state
      current_state = self.start_state
      # Looping through the states
      while current_state != self.goal_state:
          # Getting the next state
          next_state, _ = self.get_next_state(current_state, self.
→policy[current state])
          # Adding the next state to the path
          self.optimal_policy_path.append(next_state)
          self.optimal_policy_path_coords.append(self.
→get_state_coords(next_state))
          # Setting the current state as the next state
          current_state = next_state
```

## 4 Agent learning

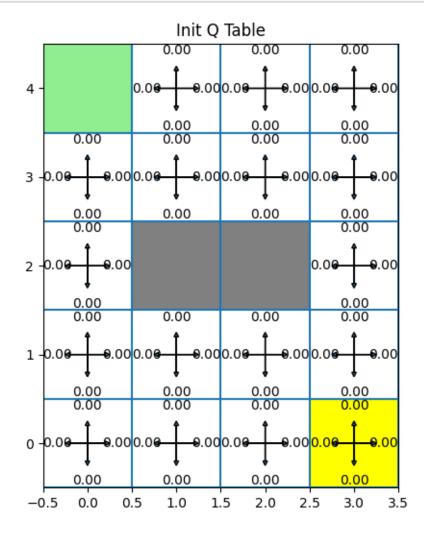
First, let us create the agent object and initiate the maze environment for it.

- 5 Q1. Design your own grid world problem: you can design arbitrary grid world problem but should be bigger than 4\*3; the grid world must include obstacles or walls
  - Grid world Configuration:
    - Size: 5x4 maze
    - Walls: 2 walls placed within the maze
- 6 Q2: Design your states, actions, and rewards.
  - States:

- Initial: on coords (3,0) which is in yellow color
- Goal: in (0,4) which is in green color
- Actions: There are four actions at every state: up, down, right, left.
- Rewards:
  - Step: 0Wall: -5Goal: 10



### Display the initial Q table



7 Q3. Implement your Q-learning algorithm.

```
[15]: # Training the agent for one episode agent.train_episodes(num_episodes=100)
```

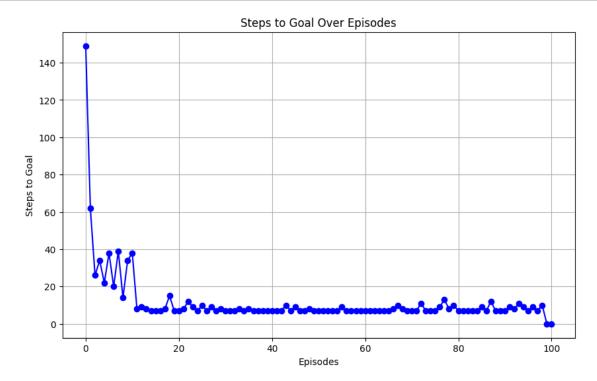
8 Q4. Show your goal searching process with step-to-go curve, and learned Q-table (good to show some intermediate Q-tables).

#step-to-go curve

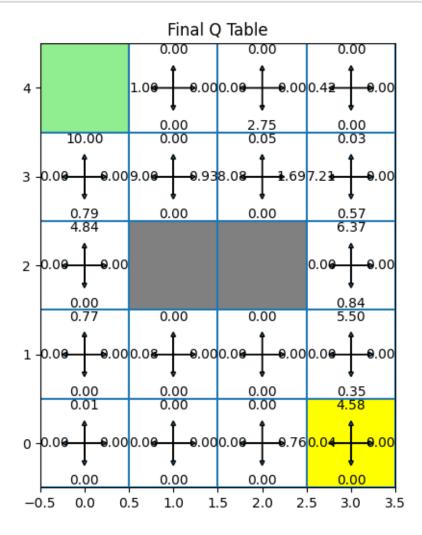
```
[24]: steps_to_goal_list_for_all = agent.save_steps_to_goal_for_current_episode()

[26]: import matplotlib.pyplot as plt

# Plotting the graph
plt.figure(figsize=(10, 6))
plt.plot(steps_to_goal_list_for_all, marker='o', linestyle='-', color='b')
plt.title('Steps to Goal Curve')
plt.xlabel('Episodes')
plt.ylabel('Steps to Goal')
plt.grid(True)
plt.show()
```



#learned Q-table

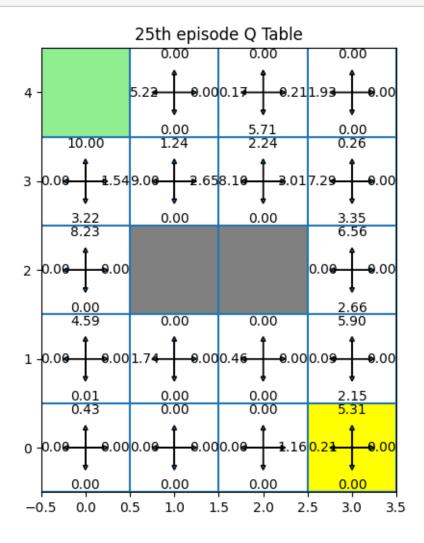


#Intermediate Q tables

```
Q=agent.Q,
goal_coords=agent.goal_coords,
wall_coords=agent.wall_coords,
start_coords=agent.start_coords,
title=f"{num_episodes}th episode Q Table"
)
```

##Q learning for 25th Episode

### [35]: q\_learning\_for\_intermediate\_episodes(25)



##Q learning for 50th Episode

[36]: q\_learning\_for\_intermediate\_episodes(50)

