

# Project\_1\_AI

October 14, 2024

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[ ]: import numpy as np
from enum import Enum
import random
from typing import Tuple, List, Set, Dict, Optional
from collections import deque
import matplotlib.pyplot as plt
import matplotlib.colors as mcolors
from IPython.display import clear_output
import time
import ipywidgets as widgets
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[ ]: class grid_cell(Enum):
    block = 0
    open = 1
    fire = 2
    bot = 3
    button = 4
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[23]: class ship:
    # We are creating constructor method to initialize the ship grid and related
    ↪ parameters.
    def __init__(self, size: int = 40, fire_spread: float = 0.6):
        self.size = size # size of the grid (40x40 by default)
        self.grid = np.zeros((self.size, self.size), dtype=int) # We initialise
        ↪ the grid as a 2d Array of Zeroes
        self.q = fire_spread # This represent probability of fire spreading (0.
        ↪ 6 by Default)
        self.bot_pos = None
        self.button_pos = None
        self.start_fire_pos = None
        self.fire_cells = set() # This will store set() of cells which are
        ↪ currently on fire
        #This Method will initialize the grid and set position for bot, button and
        ↪ fire
        def initialize_grid(self):
            self.grid.fill(grid_cell.block.value)
            start_x = random.randint(1, self.size - 2)
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start_y = random.randint(1, self.size - 2)
self.open_cell(start_x, start_y)

while True:
    candidates = self.one_neighbour() #This will get cells with one_
↪open neighbour
    if not candidates:
        break
    x, y = random.choice(list(candidates))
    self.open_cell(x, y)

self.dead_ends() # Handles dead-end cell

open_cells = []
for x in range(self.size):
    for y in range(self.size):
        if self.grid[x][y] == grid_cell.open.value:
            open_cells.append((x, y))

if len(open_cells) < 3:
    raise ValueError("There are no Open Cells")

bot_pos, button_pos, start_fire_pos = random.sample(open_cells, 3)
self.bot_pos = bot_pos # Assign bot position
self.button_pos = button_pos # Assign button position
self.start_fire_pos = start_fire_pos # Assign starting fire position
self.fire_cells.add(start_fire_pos) # Add fire position to fire set

self.grid[bot_pos] = grid_cell.bot.value
self.grid[button_pos] = grid_cell.button.value
self.grid[start_fire_pos] = grid_cell.fire.value

self.bot1_strategy()
self.bot2_strategy()
self.bot3_strategy()
self.bot4_strategy()
# Method to open grid cell by changing its value to open
def open_cell(self, x, y):
    self.grid[x][y] = grid_cell.open.value
# Method to get neighbouring cell
def get_neighbours(self, x: int, y: int) -> List[Tuple[int, int]]:
    neighbours = []
    directions = [(1, 0), (-1, 0), (0, 1), (0, -1)] # all possible_
↪directions (up, down, left, right)
    # we check whether the new coo-ordinates are within the grid boundaries
    for dx, dy in directions:
        nx, ny = x + dx, y + dy

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        if 0 <= nx < self.size and 0 <= ny < self.size:
            neighbours.append((nx, ny)) # Adding valid neighbour
    return neighbours
#Method to find block cells with exactly one neighbour
def one_neighbour(self) -> Set[Tuple[int, int]]:
    candidates = set()
    for x in range(self.size):
        for y in range(self.size):
            if self.grid[x][y] == grid_cell.block.value:
                open_neighbours_count = 0
                for nx, ny in self.get_neighbours(x, y):
                    if self.grid[nx][ny] == grid_cell.open.value:
                        open_neighbours_count += 1
                # if exactly one open neighbour, we will add it to the
↪ candidate set.
                if open_neighbours_count == 1:
                    candidates.add((x, y))
    return candidates
# Method to handle dead-end cells by opening up adjacent block
def dead_ends(self):
    dead = []
    for x in range(self.size):
        for y in range(self.size):
            if self.grid[x][y] == grid_cell.open.value:
                neighbour_count = 0
                for nx, ny in self.get_neighbours(x, y):
                    if self.grid[nx][ny] == grid_cell.open.value:
                        neighbour_count += 1

                if neighbour_count == 1:
                    dead.append((x, y))

    half_dead = random.sample(dead, len(dead) // 2)
    for x, y in half_dead:
        blocked_neighbours = []

        for nx, ny in self.get_neighbours(x, y):
            if self.grid[nx][ny] == grid_cell.block.value:
                blocked_neighbours.append((nx, ny))

        if blocked_neighbours:
            nx, ny = random.choice(blocked_neighbours)
            self.open_cell(nx, ny)

    def bfs(self, start: Tuple[int, int], goal: Tuple[int, int], fire_escape:
↪ bool = False, escape_adjacent_fire: bool = False) ->
↪ Optional[List[Tuple[int, int]]]:

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queue = deque([start]) # Queue for BFS with start position
visited = {start: None} # This will track the visited cells and
↳previous nodes

while queue:
    current = queue.popleft()

    if current == goal:
        return self.reconstruct_path(visited, start, current)

    for neighbour in self.get_neighbours(*current):
        if neighbour not in visited and self.grid[neighbour] !=
↳grid_cell.block.value:
            if fire_escape and self.grid[neighbour] == grid_cell.fire.
↳value:
                continue # This will avoid fire if fire_escape is
↳enabled

                if escape_adjacent_fire and any(self.grid[nx,ny] ==
↳grid_cell.fire.value for nx, ny in self.get_neighbours(*neighbour)):
                    continue # Avoid adjacent_fire if the
↳escape_adjacent_fire is enabled

            queue.append(neighbour) #This will append valid neighbours
↳to the queue

            visited[neighbour] = current

    return None

# This will reconstruct the path from start to goal based on visited nodes
def reconstruct_path(self, visited: Dict[Tuple[int, int],
↳Optional[Tuple[int, int]]], start: Tuple[int, int], goal: Tuple[int, int])
↳-> List[Tuple[int, int]]:
    path = []
    current = goal
    while current != start:
        path.append(current)
        current = visited[current]
    return path[::-1]

def bot1_strategy(self):
    self.bot1 = self.bfs(self.bot_pos, self.button_pos, fire_escape=False)

def bot2_strategy(self):
    self.bot2 = self.bfs(self.bot_pos, self.button_pos, fire_escape=True)

def bot3_strategy(self):

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        self.bot3 = self.bfs(self.bot_pos, self.button_pos, fire_escape=True,
↪escape_adjacent_fire=True)
        if not self.bot3:
            self.bot3 = self.bfs(self.bot_pos, self.button_pos,
↪fire_escape=True)

    def bot4_strategy(self):
        open_set = [(0, self.bot_pos)] # Priority Queue: (f-score, node)
        came_from = {self.bot_pos: None} # To reconstruct the path
        g_score = {self.bot_pos: 0} # It is Cost from start to the current node
        f_score = {self.bot_pos: self.heuristic(self.bot_pos, self.button_pos)}
↪# It is Cost from start to the goal + Heuristic i.e cost(g+h)

        while open_set:
            current = min(open_set, key=lambda x: x[0])[1]
            open_set = [x for x in open_set if x[1] != current] # It will
↪remove the current node from the open set

            if current == self.button_pos:
                self.bot4 = self.reconstruct_path(came_from, self.bot_pos,
↪current)
                return

            for neighbour in self.get_neighbours(*current):
                if self.grid[neighbour] == grid_cell.block.value:
                    continue

                # We calculate the current path cost i.e tentative g score by
↪assuming the uniform cost for each step
                current_path = g_score[current] + 1

                # We are applying fire risk penalty to the g-score
                risk_penalty = self.fire_risk(neighbour)
                current_path += risk_penalty

                if neighbour not in g_score or current_path <
↪g_score[neighbour]:
                    came_from[neighbour] = current
                    g_score[neighbour] = current_path
                    f_score[neighbour] = current_path + self.
↪heuristic(neighbour, self.button_pos)
                    open_set.append((f_score[neighbour], neighbour))

        self.bot4 = None # If no path is found

    def heuristic(self, node: Tuple[int, int], goal: Tuple[int, int]) -> int:

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    return abs(node[0] - goal[0]) + abs(node[1] - goal[1])

def fire_risk(self, node: Tuple[int, int]) -> float:
    neighbour = self.get_neighbours(*node)
    burning_neighbours = 0

    for nx, ny in neighbour:
        if self.grid[nx,ny] == grid_cell.fire.value:
            burning_neighbours += 1

    return burning_neighbours * 0.5

def spread_fire(self):
    new_fire_cell = set()
    for x, y in list(self.fire_cells):
        for nx, ny in self.get_neighbours(x, y):
            if self.grid[nx][ny] == grid_cell.open.value:
                if random.random() < self.q:
                    new_fire_cell.add((nx, ny))

    self.fire_cells.update(new_fire_cell)
    for x, y in self.fire_cells:
        self.grid[x,y] = grid_cell.fire.value

def movement(self, bot_type: int):
    if bot_type == 1:
        path = self.bot1
    elif bot_type == 2:
        self.bot2_strategy()
        path = self.bot2
    elif bot_type == 3:
        self.bot3_strategy()
        path = self.bot3
    elif bot_type == 4:
        self.bot4_strategy()
        path = self.bot4
    else:
        return False

    if not path:
        return False

    next_pos = path.pop(0)
    self.grid[self.bot_pos] = grid_cell.open.value
    self.bot_pos = next_pos
    self.grid[self.bot_pos] = grid_cell.bot.value

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        return True

def bot_step(self, bot_type: int) -> bool:
    if not self.movement(bot_type):
        return False # Bot is trapped or reached to the button

    # check if bot reached button
    if self.bot_pos == self.button_pos:
        return True

    #Spread fire
    self.spread_fire()

    #check if bot is burned
    if self.bot_pos in self.fire_cells:
        return False

    return None

def reset(self):
    #Reset the grid and reinitialize the ship.
    self.__init__(self.size, self.q)
    self.initialize_grid()

def visualise(self):
    # Visualize the grid.
    plt.figure(figsize=(10, 10))
    colors = ['black', 'white', 'red', 'blue', 'green']
    cmap = mcolors.ListedColormap(colors)
    plt.imshow(self.grid, cmap=cmap)
    plt.axis('off')
    plt.show()

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[ ]: def run_simulations(simulation: ship, fire_spread_values: List[float],
    ↪ num_trials: int = 100) -> Dict[int, List[float]]:
    bot_success_rates = {1: [], 2: [], 3: [], 4: []} # Success rates for bots
    ↪ 1-4

    for q in fire_spread_values:
        simulation.q = q
        print(f"Running simulations for fire spread probability: {q}")
        for bot_type in range(1, 5):
            successes = 0
            for trial in range(num_trials):
                simulation.reset() # Reset the simulation
                result = None
                while result is None:

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        result = simulation.bot_step(bot_type)

        # Count successes
        if result:
            successes += 1

        success_rate = successes / num_trials
        bot_success_rates[bot_type].append(success_rate)
        print(f"Bot {bot_type} success rate for q={q}: {success_rate}")

    return bot_success_rates

# Function to plot results
def plot_results(fire_spread_values: List[float], success_rates: Dict[int, List[float]]):
    """Plot the success rates of all bots."""
    plt.figure(figsize=(10, 6))
    for bot_type in range(1, 5):
        plt.plot(fire_spread_values, success_rates[bot_type], label=f'Bot {bot_type}', marker='o')

    plt.xlabel('Fire Spread Probability (q)')
    plt.ylabel('Success Rate')
    plt.title('Success Rate vs Fire Spread Probability for Different Bots')
    plt.legend()
    plt.grid(True)
    plt.ylim(0, 1)
    plt.show()

# Main simulation and analysis
fire_spread_values = [0.1 * i for i in range(10)] # From 0 to 0.9 in steps of 0.1
num_trials = 1 # Run 1 trials per bot per fire spread value
simulation = ship()

# Run simulations
success_rates = run_simulations(simulation, fire_spread_values, num_trials)

# Plot the results
plot_results(fire_spread_values, success_rates)

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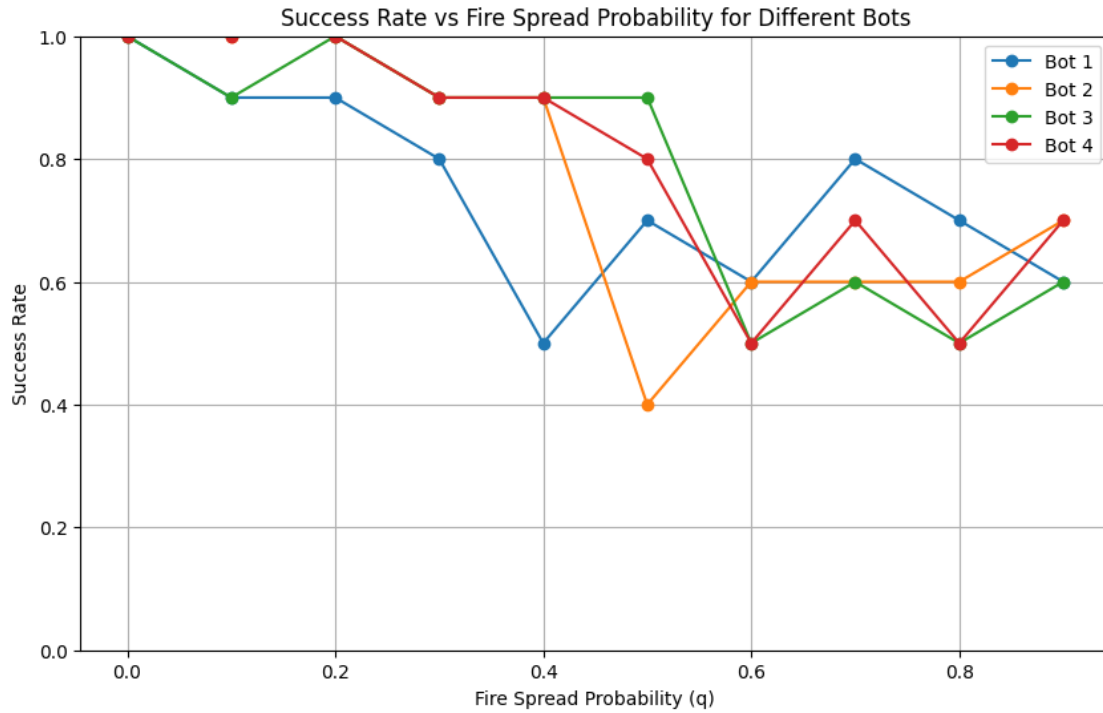
```

Running simulations for fire spread probability: 0.0
Bot 1 success rate for q=0.0: 1.0
Bot 2 success rate for q=0.0: 1.0
Bot 3 success rate for q=0.0: 1.0
Bot 4 success rate for q=0.0: 1.0
Running simulations for fire spread probability: 0.1

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Bot 1 success rate for  $q=0.1$ : 0.9  
 Bot 2 success rate for  $q=0.1$ : 1.0  
 Bot 3 success rate for  $q=0.1$ : 0.9  
 Bot 4 success rate for  $q=0.1$ : 1.0  
 Running simulations for fire spread probability: 0.2  
 Bot 1 success rate for  $q=0.2$ : 0.9  
 Bot 2 success rate for  $q=0.2$ : 1.0  
 Bot 3 success rate for  $q=0.2$ : 1.0  
 Bot 4 success rate for  $q=0.2$ : 1.0  
 Running simulations for fire spread probability: 0.30000000000000004  
 Bot 1 success rate for  $q=0.30000000000000004$ : 0.8  
 Bot 2 success rate for  $q=0.30000000000000004$ : 0.9  
 Bot 3 success rate for  $q=0.30000000000000004$ : 0.9  
 Bot 4 success rate for  $q=0.30000000000000004$ : 0.9  
 Running simulations for fire spread probability: 0.4  
 Bot 1 success rate for  $q=0.4$ : 0.5  
 Bot 2 success rate for  $q=0.4$ : 0.9  
 Bot 3 success rate for  $q=0.4$ : 0.9  
 Bot 4 success rate for  $q=0.4$ : 0.9  
 Running simulations for fire spread probability: 0.5  
 Bot 1 success rate for  $q=0.5$ : 0.7  
 Bot 2 success rate for  $q=0.5$ : 0.4  
 Bot 3 success rate for  $q=0.5$ : 0.9  
 Bot 4 success rate for  $q=0.5$ : 0.8  
 Running simulations for fire spread probability: 0.6000000000000001  
 Bot 1 success rate for  $q=0.6000000000000001$ : 0.6  
 Bot 2 success rate for  $q=0.6000000000000001$ : 0.6  
 Bot 3 success rate for  $q=0.6000000000000001$ : 0.5  
 Bot 4 success rate for  $q=0.6000000000000001$ : 0.5  
 Running simulations for fire spread probability: 0.7000000000000001  
 Bot 1 success rate for  $q=0.7000000000000001$ : 0.8  
 Bot 2 success rate for  $q=0.7000000000000001$ : 0.6  
 Bot 3 success rate for  $q=0.7000000000000001$ : 0.6  
 Bot 4 success rate for  $q=0.7000000000000001$ : 0.7  
 Running simulations for fire spread probability: 0.8  
 Bot 1 success rate for  $q=0.8$ : 0.7  
 Bot 2 success rate for  $q=0.8$ : 0.6  
 Bot 3 success rate for  $q=0.8$ : 0.5  
 Bot 4 success rate for  $q=0.8$ : 0.5  
 Running simulations for fire spread probability: 0.9  
 Bot 1 success rate for  $q=0.9$ : 0.6  
 Bot 2 success rate for  $q=0.9$ : 0.7  
 Bot 3 success rate for  $q=0.9$ : 0.6  
 Bot 4 success rate for  $q=0.9$ : 0.7



```
[ ]: def run_simulations(simulation: ship, fire_spread_values: List[float],
    ↳ num_trials: int = 100) -> Dict[int, List[float]]:
    bot_success_rates = {1: [], 2: [], 3: [], 4: []} # Success rates for bots
    ↳ 1-4

    for q in fire_spread_values:
        simulation.q = q
        print(f"Running simulations for fire spread probability: {q}")
        for bot_type in range(1, 5):
            successes = 0
            for trial in range(num_trials):
                simulation.reset() # Reset the simulation
                result = None
                while result is None:
                    result = simulation.bot_step(bot_type)

                # Count successes
                if result:
                    successes += 1

            success_rate = successes / num_trials
            bot_success_rates[bot_type].append(success_rate)
            print(f"Bot {bot_type} success rate for q={q}: {success_rate}")
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    return bot_success_rates

# Function to plot results
def plot_results(fire_spread_values: List[float], success_rates: Dict[int, List[float]]):
    plt.figure(figsize=(10, 6))
    for bot_type in range(1, 5):
        plt.plot(fire_spread_values, success_rates[bot_type], label=f'Bot {bot_type}', marker='o')

    plt.xlabel('Fire Spread Probability (q)')
    plt.ylabel('Success Rate')
    plt.title('Success Rate vs Fire Spread Probability for Different Bots')
    plt.legend()
    plt.grid(True)
    plt.ylim(0, 1)
    plt.show()

# Main simulation and analysis
fire_spread_values = [0.1 * i for i in range(10)] # From 0 to 0.9 in steps of 0.1
num_trials = 25 # Run 25 trials per bot per fire spread value
simulation = ship()

# Run simulations
success_rates = run_simulations(simulation, fire_spread_values, num_trials)

# Plot the results
plot_results(fire_spread_values, success_rates)

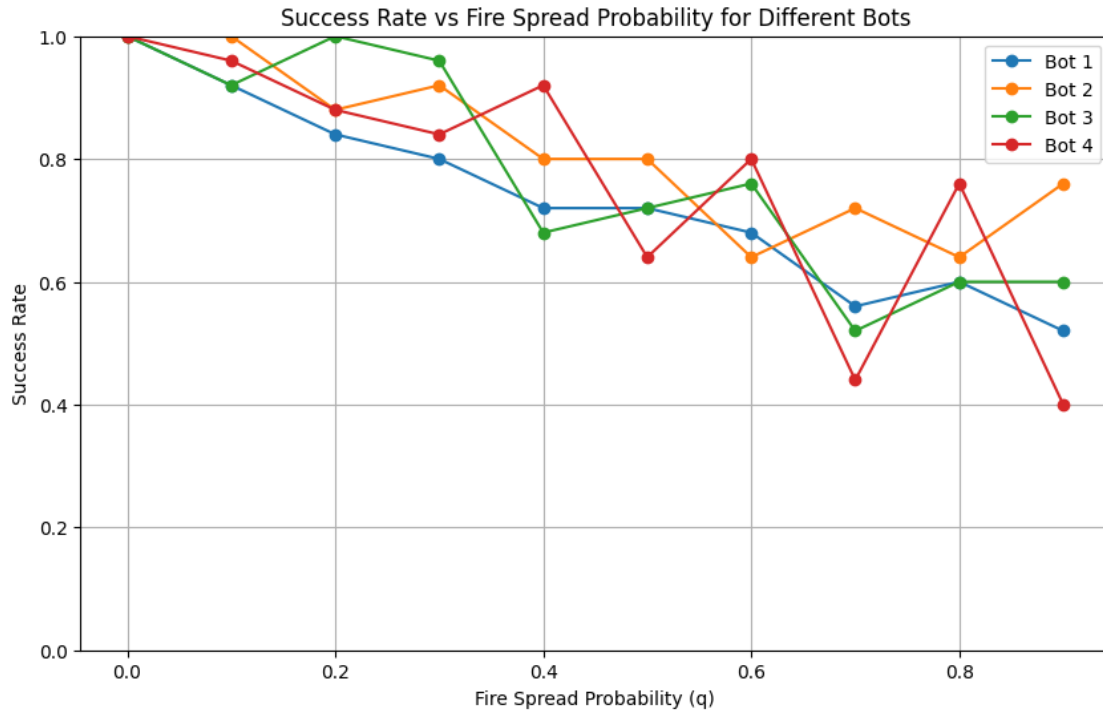
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Running simulations for fire spread probability: 0.0
Bot 1 success rate for q=0.0: 1.0
Bot 2 success rate for q=0.0: 1.0
Bot 3 success rate for q=0.0: 1.0
Bot 4 success rate for q=0.0: 1.0
Running simulations for fire spread probability: 0.1
Bot 1 success rate for q=0.1: 0.92
Bot 2 success rate for q=0.1: 1.0
Bot 3 success rate for q=0.1: 0.92
Bot 4 success rate for q=0.1: 0.96
Running simulations for fire spread probability: 0.2
Bot 1 success rate for q=0.2: 0.84
Bot 2 success rate for q=0.2: 0.88
Bot 3 success rate for q=0.2: 1.0
Bot 4 success rate for q=0.2: 0.88
Running simulations for fire spread probability: 0.30000000000000004
Bot 1 success rate for q=0.30000000000000004: 0.8

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Bot 2 success rate for  $q=0.30000000000000004$ : 0.92  
 Bot 3 success rate for  $q=0.30000000000000004$ : 0.96  
 Bot 4 success rate for  $q=0.30000000000000004$ : 0.84  
 Running simulations for fire spread probability: 0.4  
 Bot 1 success rate for  $q=0.4$ : 0.72  
 Bot 2 success rate for  $q=0.4$ : 0.8  
 Bot 3 success rate for  $q=0.4$ : 0.68  
 Bot 4 success rate for  $q=0.4$ : 0.92  
 Running simulations for fire spread probability: 0.5  
 Bot 1 success rate for  $q=0.5$ : 0.72  
 Bot 2 success rate for  $q=0.5$ : 0.8  
 Bot 3 success rate for  $q=0.5$ : 0.72  
 Bot 4 success rate for  $q=0.5$ : 0.64  
 Running simulations for fire spread probability: 0.6000000000000001  
 Bot 1 success rate for  $q=0.6000000000000001$ : 0.68  
 Bot 2 success rate for  $q=0.6000000000000001$ : 0.64  
 Bot 3 success rate for  $q=0.6000000000000001$ : 0.76  
 Bot 4 success rate for  $q=0.6000000000000001$ : 0.8  
 Running simulations for fire spread probability: 0.7000000000000001  
 Bot 1 success rate for  $q=0.7000000000000001$ : 0.56  
 Bot 2 success rate for  $q=0.7000000000000001$ : 0.72  
 Bot 3 success rate for  $q=0.7000000000000001$ : 0.52  
 Bot 4 success rate for  $q=0.7000000000000001$ : 0.44  
 Running simulations for fire spread probability: 0.8  
 Bot 1 success rate for  $q=0.8$ : 0.6  
 Bot 2 success rate for  $q=0.8$ : 0.64  
 Bot 3 success rate for  $q=0.8$ : 0.6  
 Bot 4 success rate for  $q=0.8$ : 0.76  
 Running simulations for fire spread probability: 0.9  
 Bot 1 success rate for  $q=0.9$ : 0.52  
 Bot 2 success rate for  $q=0.9$ : 0.76  
 Bot 3 success rate for  $q=0.9$ : 0.6  
 Bot 4 success rate for  $q=0.9$ : 0.4

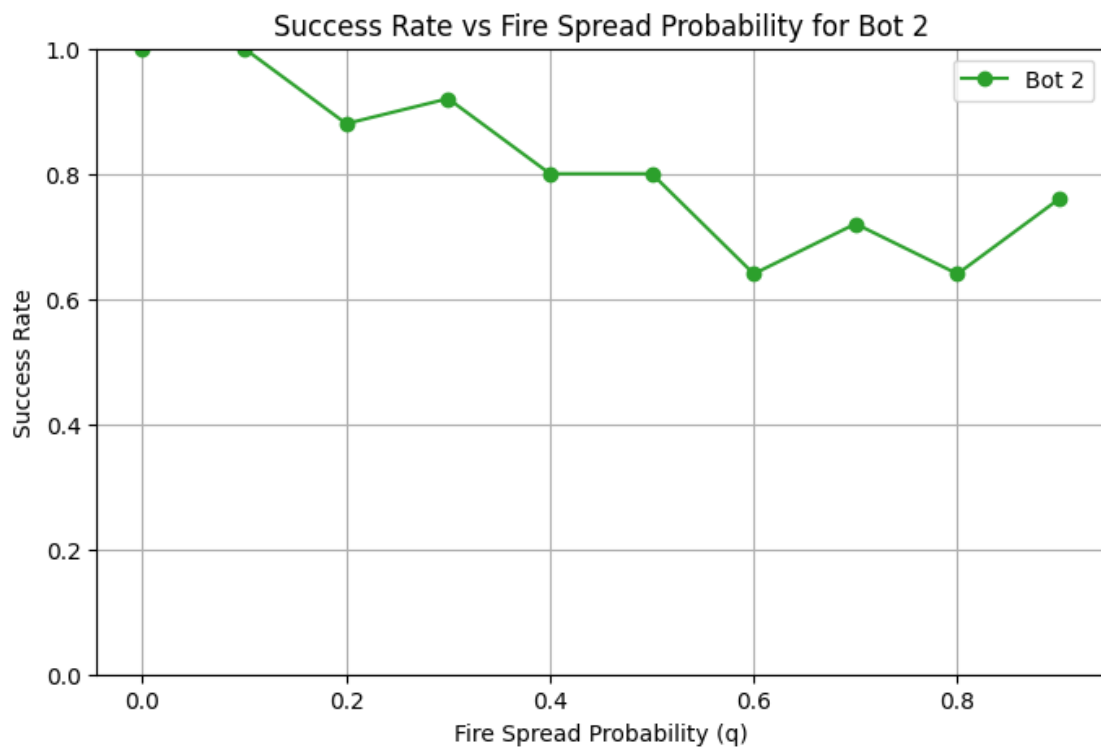
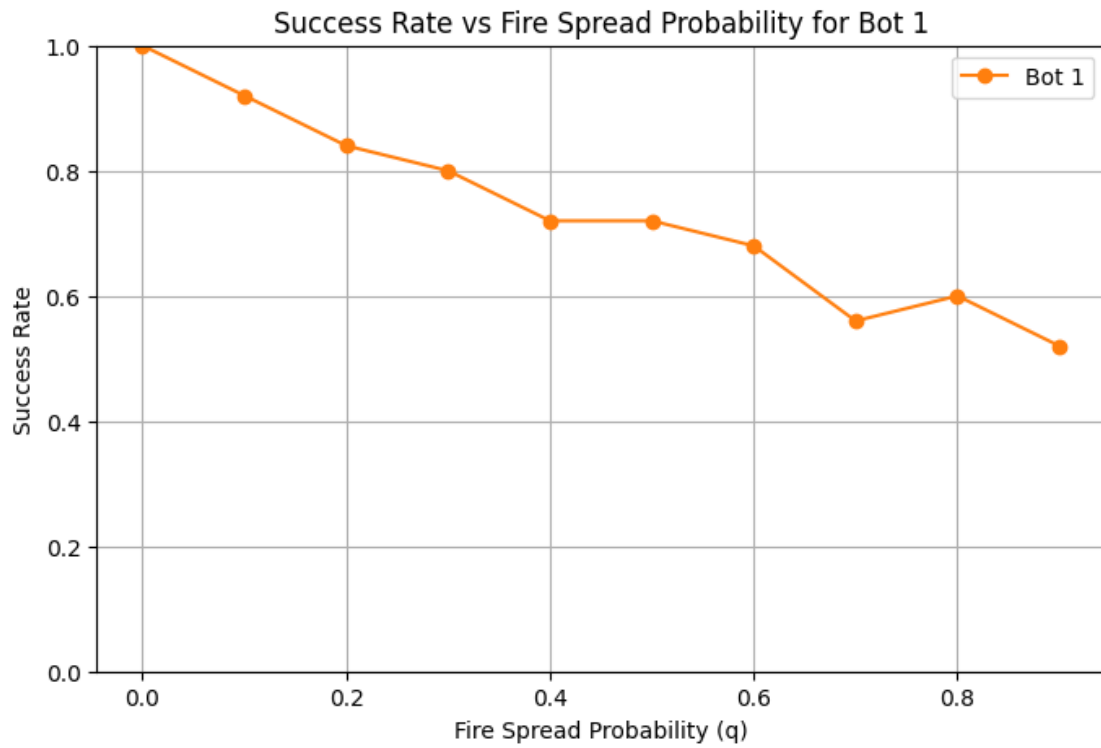


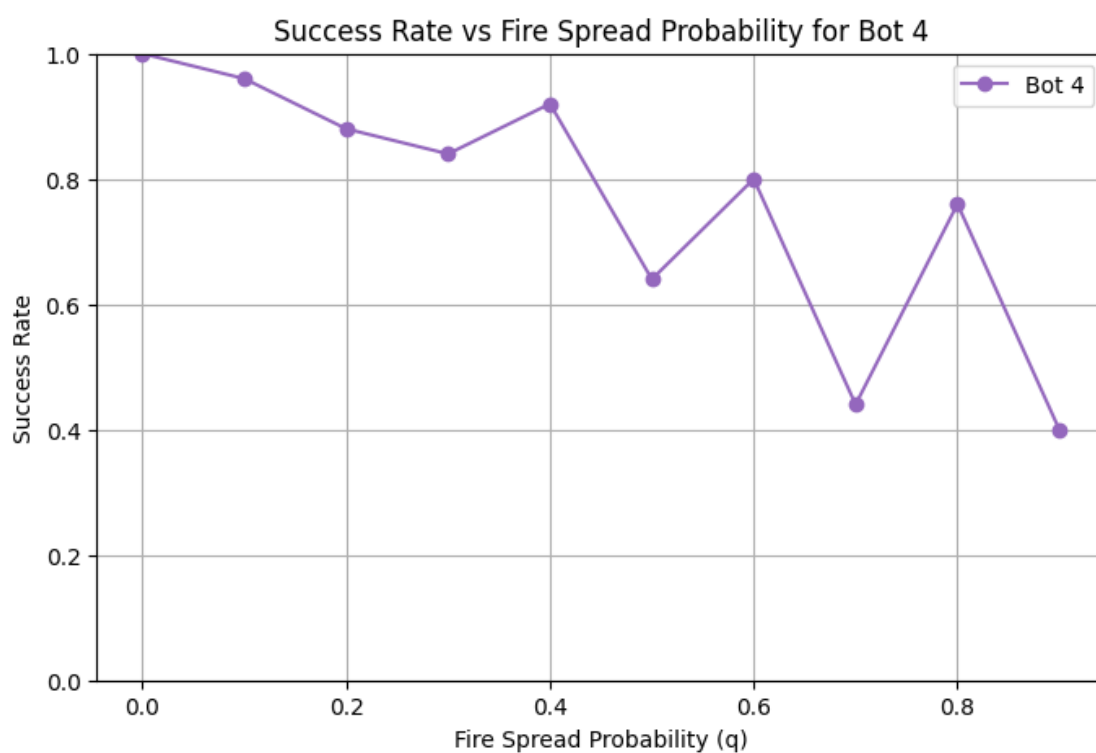
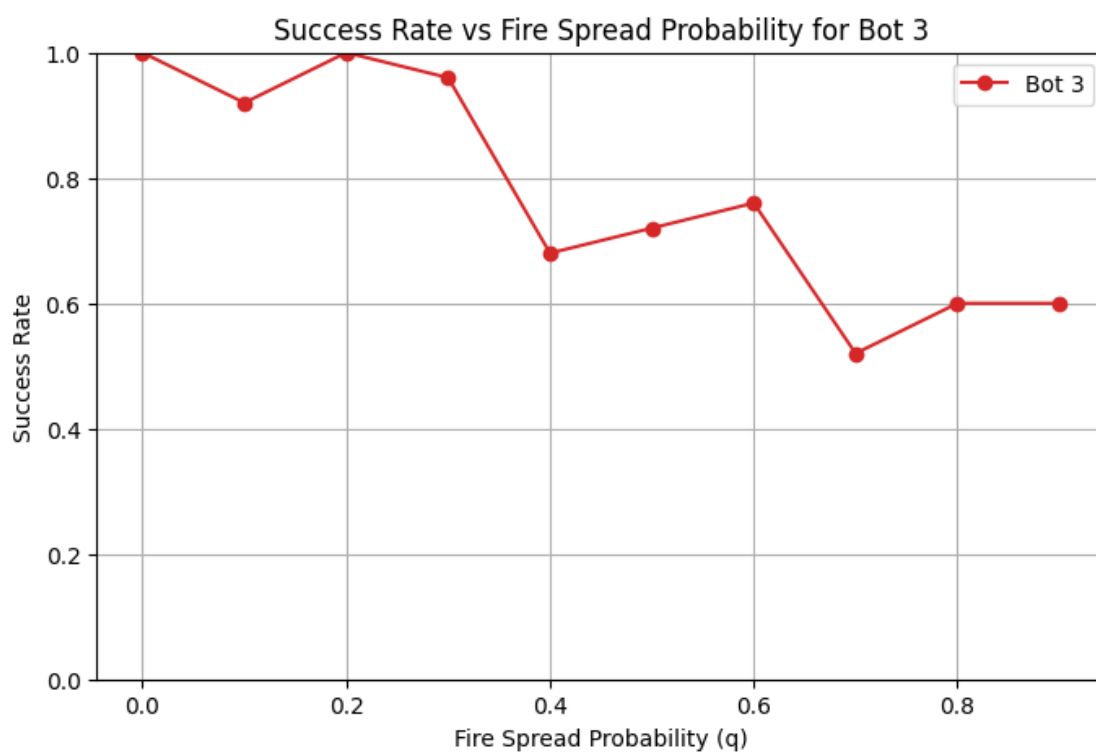
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[ ]: success_rates
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[ ]: {1: [1.0, 0.92, 0.84, 0.8, 0.72, 0.72, 0.68, 0.56, 0.6, 0.52],
      2: [1.0, 1.0, 0.88, 0.92, 0.8, 0.8, 0.64, 0.72, 0.64, 0.76],
      3: [1.0, 0.92, 1.0, 0.96, 0.68, 0.72, 0.76, 0.52, 0.6, 0.6],
      4: [1.0, 0.96, 0.88, 0.84, 0.92, 0.64, 0.8, 0.44, 0.76, 0.4]}
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```
[ ]: # Function to plot each bot's graph separately
def plot_individual_bots(fire_spread_values: List[float], success_rates: Dict[int, List[float]]):
    for bot_type in range(1, 5):
        plt.figure(figsize=(8, 5))
        plt.plot(fire_spread_values, success_rates[bot_type], label=f'Bot {bot_type}', marker='o', color=f'C{bot_type}')
        plt.xlabel('Fire Spread Probability (q)')
        plt.ylabel('Success Rate')
        plt.title(f'Success Rate vs Fire Spread Probability for Bot {bot_type}')
        plt.legend()
        plt.grid(True)
        plt.ylim(0, 1)
        plt.show()
```

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[ ]: # Run the individual bot plots
plot_individual_bots(fire_spread_values, success_rates)
```





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[ ]: # Function to calculate and plot average accuracy for each bot in percentage
def plot_average_accuracy(success_rates: Dict[int, List[float]]):
    bot_averages = {}

    # Calculate the average success rate for each bot and convert it to
    ↪percentage
    for bot_type in range(1, 5):
        avg_success_rate = np.mean(success_rates[bot_type]) * 100 # Convert to
        ↪percentage
        bot_averages[bot_type] = avg_success_rate

    # Plot the average success rates as a bar graph
    plt.figure(figsize=(8, 6))
    bars = plt.bar(bot_averages.keys(), bot_averages.values(), color=['C1',
    ↪'C2', 'C3', 'C4'])
    plt.xlabel('Bot Type')
    plt.ylabel('Average Success Rate (%)')
    plt.title('Average Success Rate for Each Bot (in %)')
    plt.ylim(0, 100)
    plt.xticks([1, 2, 3, 4], ['Bot 1', 'Bot 2', 'Bot 3', 'Bot 4'])

    # Add percentage labels above the bars
    for bar in bars:
        yval = bar.get_height()
        plt.text(bar.get_x() + bar.get_width()/2, yval, f'{yval:.2f}%',
        ↪ha='center', va='bottom')

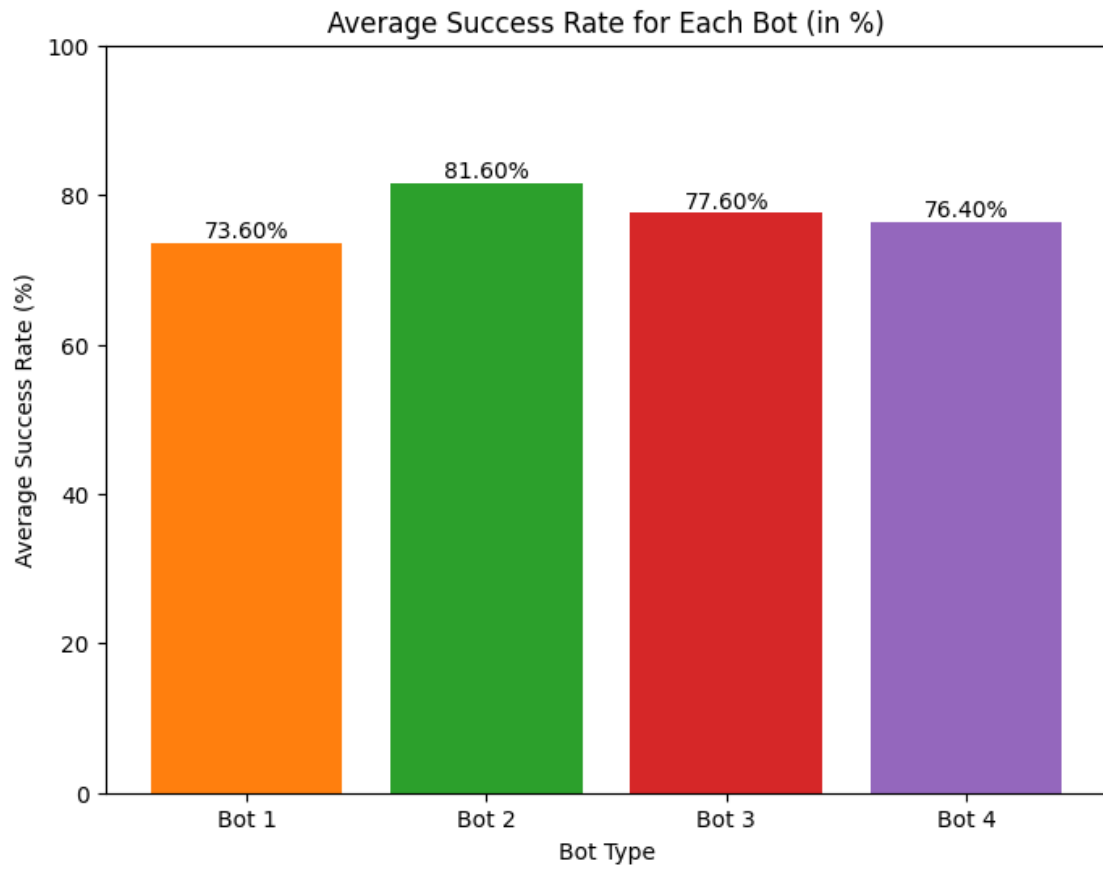
    plt.show()

    return bot_averages
```

```
[ ]: # Calculate and plot the average accuracy in percentage
average_success_rates = plot_average_accuracy(success_rates)

# Display the average success rates in percentage for reference
for bot, avg_rate in average_success_rates.items():
    print(f"Bot {bot} Average Success Rate: {avg_rate:.2f}%") # Print in
    ↪percentage form
```





Bot 1 Average Success Rate: 73.60%

Bot 2 Average Success Rate: 81.60%

Bot 3 Average Success Rate: 77.60%

Bot 4 Average Success Rate: 76.40%

[ ]: