

1. Identifying Agents and Environments – List five real-world AI agents with their environment, sensors, actuators and goals.

List of 5 real-world AI agents, along with their environments, sensors, actuators, and goals

1. Autonomous Car (e.g., Tesla Autopilot)

- **Environment:** Roads, traffic, pedestrians, weather conditions
- **Sensors:** Cameras, LiDAR, radar, GPS, ultrasonic sensors
- **Actuators:** Steering, throttle, brake, signal indicators
- **Goal:** Safely navigate to a destination while obeying traffic rules and avoiding collisions

2. Smart Home Thermostat (e.g., Nest)

- **Environment:** Indoor home environment (temperature, humidity, occupancy)
- **Sensors:** Temperature sensor, humidity sensor, motion detector, Wi-Fi module
- **Actuators:** HVAC controls (heater/cooler on/off)
- **Goal:** Maintain comfortable temperature efficiently, save energy based on occupancy

3. Voice Assistant (e.g., Amazon Alexa)

- **Environment:** Human-inhabited space (home, office)
- **Sensors:** Microphones, sometimes cameras
- **Actuators:** Audio speakers, smart device signals (e.g., lights, alarms)
- **Goal:** Understand and fulfill user voice commands (e.g., play music, control devices, answer questions)

4. Warehouse Robot (e.g., Amazon Kiva System)

- **Environment:** Warehouse floor with shelves and other robots
- **Sensors:** Cameras, RFID readers, proximity sensors, wheel encoders
- **Actuators:** Wheels, lifting arms, directional motors
- **Goal:** Pick and deliver inventory items to specific locations efficiently and safely

5. Email Spam Filter

- **Environment:** Incoming email stream on a mail server
- **Sensors:** Email text, metadata (sender, subject, links), user feedback
- **Actuators:** Classifier label (spam/ham), email routing (inbox or spam folder)
- **Goal:** Classify and filter out unwanted/spam messages with high accuracy

6. Self-Checkout Kiosk (e.g., in supermarkets)

- **Environment:** Store checkout area, interacting with products and customers
- **Sensors:** Barcode scanner, weight sensors, cameras, touchscreen input
- **Actuators:** Display screen, speaker, payment terminal, receipt printer
- **Goal:** Accurately scan items, calculate total, handle payment, and reduce checkout time without human cashiers

7. Drone Delivery System (e.g., Zipline or Amazon Prime Air)

- **Environment:** Outdoor airspace, wind/weather conditions, GPS-based location data
- **Sensors:** GPS, altimeter, gyroscope, cameras, proximity sensors
- **Actuators:** Propellers, navigation control surfaces, package drop mechanism
- **Goal:** Deliver packages autonomously to precise locations safely and efficiently

8. Autonomous Vacuum Cleaner (e.g., Roomba)

- **Environment:** Home floors with furniture, pets, and humans
- **Sensors:** Infrared sensors, bump sensors, cliff sensors, gyroscope, cameras
- **Actuators:** Wheels, vacuum motor, rotating brushes, sound system
- **Goal:** Navigate and clean floors thoroughly while avoiding obstacles and stairs

9. Facial Recognition System (e.g., airport security systems)

- **Environment:** Public access points like airports, office buildings, or phones
- **Sensors:** Camera input (static or video), sometimes infrared sensors
- **Actuators:** Access control system (e.g., gates), alerting systems
- **Goal:** Identify or verify individuals based on facial features for security or personalization

10. Stock Trading Bot (e.g., algorithmic trading systems)

- **Environment:** Digital financial markets and trading platforms
- **Sensors:** Real-time financial data, market news feeds, economic indicators
- **Actuators:** Trade execution APIs (buy/sell orders), portfolio rebalancing tools
- **Goal:** Maximize profit or minimize risk through rapid and data-driven trading decisions

3. Demonstrate basic problem-solving using Breadth-First Search on a simple grid.

Program/Code:

```
from collections import deque

grid = [

    ['.', '.', '.', '#', '.', '.'],

    ['#', '#', '.', '#', '.', '#'],

    ['.', '.', '.', '.', '.', '.'],

    ['.', '#', '#', '#', '#', '.'],

    ['.', '.', '.', '.', '#', '.']

]

start = (0, 0)
goal = (4, 5)
directions = [(-1, 0), (1, 0), (0, -1), (0, 1)]

def bfs(grid, start, goal):
    rows, cols = len(grid), len(grid[0])
    visited = set()
    queue = deque()
    queue.append((start, [start]))

    while queue:
        (x, y), path = queue.popleft()

        if (x, y) == goal:
            return path

        for dx, dy in directions:
            nx, ny = x + dx, y + dy
```

```

        if (0 <= nx < rows and 0 <= ny < cols and
            grid[nx][ny] != '#' and (nx, ny) not in visited):
            visited.add((nx, ny))
            queue.append(((nx, ny), path + [(nx, ny)]))

    return None

path = bfs(grid, start, goal)

if path:
    print("Shortest Path:")
    for step in path:
        print(step)
else:
    print("No path found.")

```

Output:

Shortest Path:

```

(0, 0)
(0, 1)
(0, 2)
(1, 2)
(2, 2)
(2, 3)
(2, 4)
(2, 5)
(3, 5)
(4, 5)

```

4. Implement Depth-First Search (DFS) on a small graph.

Program/Code:

```
graph = {  
    'A': ['B', 'C'],  
    'B': ['D', 'E'],  
    'C': ['F'],  
    'D': [],  
    'E': ['F'],  
    'F': []  
}  
  
def dfs(graph, start, visited=None):  
    if visited is None:  
        visited = set()  
    visited.add(start)  
    print(start)  
  
    for neighbor in graph[start]:  
        if neighbor not in visited:  
            dfs(graph, neighbor, visited)  
  
# Start DFS from node 'A'  
print("DFS Traversal Starting from Node A:")  
dfs(graph, 'A')
```

Output:

DFS Traversal Starting from Node A:

A

B

D

E

F

C

7. Apply the A* Search algorithm to find the shortest path in a 4x4 grid.

Program/Code:

```
import heapq

grid = [
    ['.', '.', '.', '.'],
    ['.', '#', '#', '.'],
    ['.', '.', '.', '.'],
    ['#', '.', '#', '.']
]

start = (0, 0)
goal = (3, 3)

# Directions: Up, Down, Left, Right
directions = [(-1,0), (1,0), (0,-1), (0,1)]

# Heuristic: Manhattan distance
def heuristic(a, b):
    return abs(a[0] - b[0]) + abs(a[1] - b[1])

def a_star(grid, start, goal):
    rows, cols = len(grid), len(grid[0])
    open_set = []
    heapq.heappush(open_set, (0 + heuristic(start, goal), 0, start, [start]))
    visited = set()

    while open_set:
        f, g, current, path = heapq.heappop(open_set)
        if current == goal:
            return path
        if current in visited:
            continue
```

```

visited.add(current)

for dx, dy in directions:
    nx, ny = current[0] + dx, current[1] + dy
    neighbor = (nx, ny)

    if 0 <= nx < rows and 0 <= ny < cols and grid[nx][ny] != '#' and neighbor not
in visited:
        heapq.heappush(open_set, (
            g + 1 + heuristic(neighbor, goal),
            g + 1,
            neighbor,
            path + [neighbor]
        ))
return None

shortest_path = a_star(grid, start, goal)
if shortest_path:
    print("Shortest path from start to goal:")
    for step in shortest_path:
        print(f"Step: {step}")
else:
    print("No path found.")

```

Output:

Shortest path from start to goal:

Step: (0, 0)

Step: (0, 1)

Step: (0, 2)

Step: (0, 3)

Step: (1, 3)

Step: (2, 3)

Step: (3, 3)

5. Solve the Water Jug Problem using Breadth First Search (BFS).

Program/Code:

```
from collections import deque

# Define jug capacities and goal

jug1_capacity = 4
jug2_capacity = 3
goal = 2

def is_goal(state):
    x, y = state
    return x == goal or y == goal

def get_next_states(x, y):
    states = []

    # Fill either jug
    states.append((jug1_capacity, y)) # Fill jug1
    states.append((x, jug2_capacity)) # Fill jug2

    # Empty either jug
    states.append((0, y)) # Empty jug1
    states.append((x, 0)) # Empty jug2

    # Pour jug1 -> jug2
    pour = min(x, jug2_capacity - y)
    states.append((x - pour, y + pour))

    # Pour jug2 -> jug1
    pour = min(y, jug1_capacity - x)
    states.append((x + pour, y - pour))
    return states

def bfs():
    visited = set()
    queue = deque()
```

```

start_state = (0, 0)
queue.append((start_state, [start_state]))

while queue:

    (x, y), path = queue.popleft()
    if (x, y) in visited:
        continue
    visited.add((x, y))
    if is_goal((x, y)):
        return path

    for next_state in get_next_states(x, y):
        if next_state not in visited:
            queue.append((next_state, path + [next_state]))
    return None

solution_path = bfs()
if solution_path:
    print("Steps to reach the goal (2 liters in one jug):")
    for step in solution_path:
        print(f"Jug1: {step[0]}L, Jug2: {step[1]}L")
else:
    print("No solution found.")

```

Output:

Steps to reach the goal (2 liters in one jug):

Jug1: 0L, Jug2: 0L

Jug1: 0L, Jug2: 3L

Jug1: 3L, Jug2: 0L

Jug1: 3L, Jug2: 3L

Jug1: 4L, Jug2: 2L

10. Use constraint propagation to solve a Magic Square puzzle.

Program/Code:

```
from itertools import permutations

def is_magic(square):
    return (
        sum(square[0:3]) == 15 and
        sum(square[3:6]) == 15 and
        sum(square[6:9]) == 15 and
        sum(square[0::3]) == 15 and
        sum(square[1::3]) == 15 and
        sum(square[2::3]) == 15 and
        sum(square[0::4]) == 15 and
        sum(square[2:8:2]) == 15
    )

def solve_magic_square():
    digits = [1, 2, 3, 4, 5, 6, 7, 8, 9]

    for square in permutations(digits):
        if is_magic(square):
            return square

    return None
```

```
solution = solve_magic_square()

if solution:
    print("Magic Square Solution:")
    for i in range(0, 9, 3):
        print(solution[i], solution[i+1], solution[i+2])
else:
    print(" No solution found.")
```

Output:

Magic Square Solution:

2 7 6

9 5 1

4 3 8

9. Solve the 4 – Queens Problem as a CSP backtracking problem

Program/Code:

```
def is_safe(queen_positions, row, col):
    for r in range(row):
        c = queen_positions[r]
        if c == col or abs(c - col) == abs(r - row):
            return False
    return True

def solve_n_queens(n):
    solutions = []
    queen_positions = [-1] * n

    def backtrack(row):
        if row == n:
            solutions.append(queen_positions[:])
            return

        for col in range(n):
            if is_safe(queen_positions, row, col):
                queen_positions[row] = col
                backtrack(row + 1)
                queen_positions[row] = -1

    backtrack(0)
    return solutions

solutions = solve_n_queens(4)

def print_boards(solutions):
    for sol in solutions:
        print(" Solution:")
```

```

for i in sol:
    row = ['.'] * len(sol)
    row[i] = 'Q'
    print(" ".join(row))
    print()

print_boards(solutions)

```

Output:

Solution:

. Q ..

... Q

Q ...

.. Q .

Solution:

.. Q .

Q ...

... Q

. Q ..

11. Apply optimization techniques to find the maximum value in a list.

Simple optimization technique to find the **maximum value in a list** — using a **linear scan** (also called **brute force search**).

Program/Code:

```
def find_max_value(data):  
    max_val = data[0]  
  
    for val in data:  
        if val > max_val:  
            max_val = val  
  
    return max_val  
  
data = [12, 5, 23, 7, 34, 9, 30]  
print("Max value found using Brute Force:", find_max_value(data))
```

Output:

Max value found using Brute Force: 34

Another simple technique to find the **maximum value in a list** — this time using **Python's built-in max() function**, which is the most concise and efficient method.

```
# Example list
```

```
data = [12, 5, 23, 7, 34, 9, 30]
```

```
# Use Python's built-in max() function
```

```
max_value = max(data)
```

```
print("Max value found using built-in max():", max_value)
```

Output:

```
Max value found using built-in max(): 34
```


8. Implement the Minimax search algorithm for 2-player games. You may use a game tree with 3 plies.

Program/Code:

```
# Leaf node values (3 plies: root -> children -> grandchildren)
# Structure: MAX -> MIN -> Leaf values
```

```
tree = [
    [3, 5],
    [2, 9],
    [0, -1]
]
```

```
def minimax(tree):
    min_values = []
```

```
    for branch in tree:
        min_val = min(branch)
        min_values.append(min_val)
```

```
    return max(min_values)
```

```
best_value = minimax(tree)
print(f"Best value for MAX: {best_value}")
```

Output:

Best value for MAX: 3