

**AIM:** Identification and Installation of python environment towards the artificial intelligence and machine learning, installing python modules/Packages Import scikitlearn, keras etc.

**DESCRIPTION:** Artificial Intelligence and Machine Learning require a robust programming environment equipped with specialized libraries and tools. Python is the most widely used language for AI/ML due to its simplicity and extensive library support. This experiment focuses on:

- Installing Python and setting up an environment (Anaconda, Virtual Environment, or Google Colab).
- Installing essential AI/ML packages like numpy, pandas, matplotlib, scikit-learn, keras, and tensorflow.
- Verifying the successful installation of these packages.

**PROCEDURE:**

Step 1: Checking Python Installation

1. Open the terminal (Command Prompt or Anaconda Prompt).
2. Type the following command to check if Python is installed:  
`python --version`
3. If Python is not installed, download and install it from [Python's official website](https://www.python.org/).

Step 2: Setting Up a Virtual Environment (Optional but Recommended)

1. Create a new virtual environment:  
`python -m venv aiml_env`
2. Activate the virtual environment:
  - Windows:  
`aiml_env\Scripts\activate`
  - Mac/Linux:  
`source aiml_env/bin/activate`

Step 3: Installing Essential Python Modules

Use pip to install AI/ML packages:

```
pip install numpy pandas matplotlib scikit-learn keras tensorflow
```

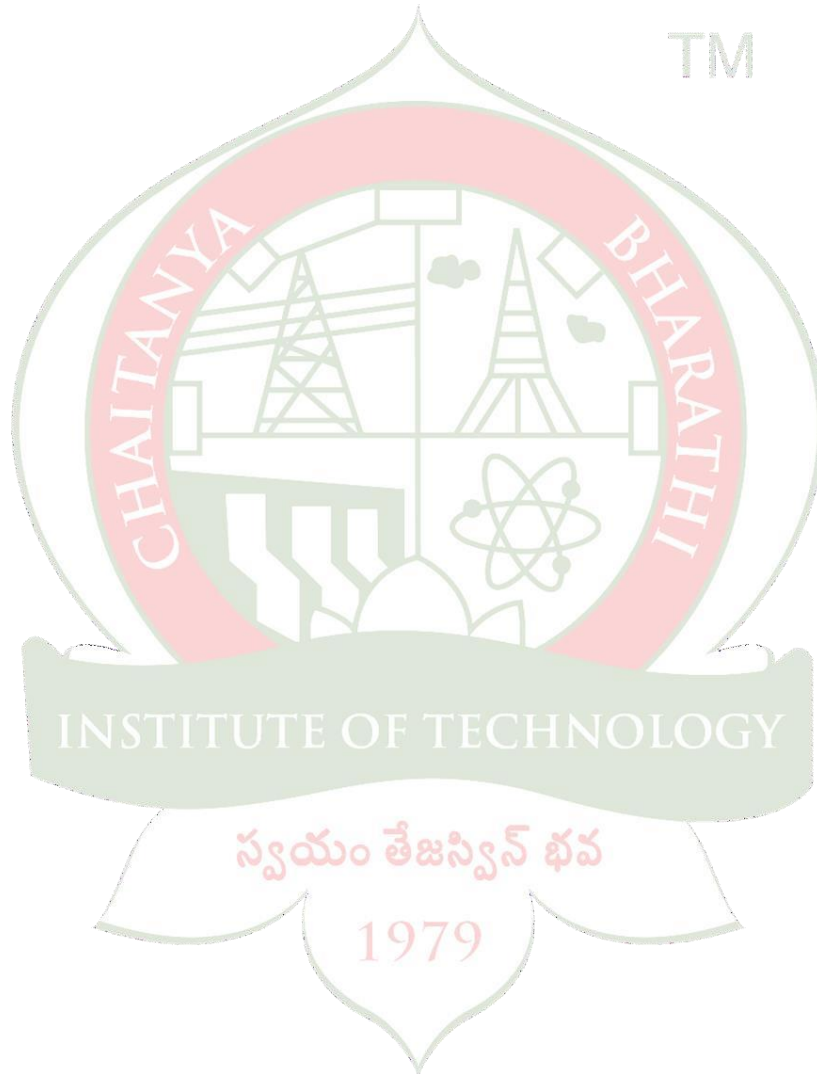
Step 4: Importing and Verifying Installed Packages

Create a Python script (verify\_installation.py) and run the following code:

```
import numpy as np
import pandas as pd
import sklearn
import keras
import tensorflow as tf
print("NumPy Version:", np.__version__)
print("Pandas Version:", pd.__version__)
print("Scikit-learn Version:", sklearn.__version__)
print("Keras Version:", keras.__version__)
print("TensorFlow Version:", tf.__version__)
```

**OUTPUT:**

NumPy Version: 1.26.4  
Pandas Version: 2.2.2  
Scikit-learn Version: 1.6.1  
Keras Version: 3.8.0  
TensorFlow Version: 2.18.0



## WEEK-2

**AIM:** Implement A\* algorithm on Graph Problem

**DISCRIPTION:** The A\* algorithm is a widely used graph traversal and pathfinding algorithm that finds the shortest path from a start node to a goal node. It combines the benefits of Dijkstra's algorithm and the Greedy Best-First Search by using a heuristic function to prioritize nodes. The algorithm maintains an open list of nodes to explore and a closed list of visited nodes, selecting the node with the lowest cost function  $f(n) = g(n) + h(n)$ , where  $g(n)$  is the actual cost from the start node and  $h(n)$  is the estimated cost to the goal. A\* is optimal and complete when using an admissible heuristic, making it efficient for applications like navigation, AI, and robotics.

### CODE:

```
import heapq

def a_star(graph, start, goal, heuristic):
    open_list = [] # Priority queue (min-heap)
    heapq.heappush(open_list, (heuristic[start], 0, start)) # (f, g, node)
    came_from = {} # To reconstruct path
    g_score = {start: 0} # Shortest path cost from start

    while open_list:
        _, g, current = heapq.heappop(open_list)
        if current == goal: # Goal reached, reconstruct path
            path = []
            while current in came_from:
                path.append(current)
                current = came_from[current]
            path.append(start)
            return path[::-1], g # Reverse path to get correct order

        for neighbor, cost in graph[current]:
            tentative_g = g + cost # Calculate new cost

            if neighbor not in g_score or tentative_g < g_score[neighbor]:
                g_score[neighbor] = tentative_g
                f_score = tentative_g + heuristic[neighbor]
                heapq.heappush(open_list, (f_score, tentative_g, neighbor))
                came_from[neighbor] = current # Track path
```

```
return None, None # No path found
```

```
graph = {
    'S': [('C', 3), ('B', 4)],
    'C': [('E', 10), ('D', 7)],
    'B': [('E', 12), ('F', 5)],
    'E': [('G', 5)],
    'D': [('E', 2)],
    'F': [('G', 16)],
    'G': []
}
```

```
# Define heuristic values for each node
```

```
heuristics = {
    'S': 14, 'C': 11, 'B': 12, 'E': 4, 'D': 6, 'F': 11, 'G': 0
}
```

```
# Example start and goal nodes
```

```
start = 'S'
```

```
goal = 'G'
```

```
path, total_cost = a_star(graph, start, goal, heuristic)
```

```
if path:
```

```
    print("Path:", path)
```

```
    print("Cost", total_cost)
```

```
else:
```

```
    print("No path found")
```

## OUTPUT:

```
PS C:\Users\HP> & C:/Users/HP/AppData/Local/Programs/Python/Python311/pyth
xe c:/Users/HP/Downloads/1b.py
Path: ['S', 'C', 'D', 'E', 'G']
Cost: 17
PS C:\Users\HP>
```

**AIM:** Implement A\* algorithm on Grid Problems

**DISCRIPTION:** The A\* algorithm is used in grid-based pathfinding problems to find the shortest path from a start position to a goal position. It works on a 2D grid where each cell represents a node and can be traversable or blocked. The algorithm prioritizes nodes using the cost function  $f(n) = g(n) + h(n)$ , where  $g(n)$  is the actual distance from the start node and  $h(n)$  is the estimated distance to the goal (often using Manhattan or Euclidean distance). A\* efficiently explores the most promising paths first, ensuring optimality when the heuristic is admissible. It is widely used in robotics, AI, and game development for navigation.

**CODE:**

```
import heapq
```

```
class Node:
```

```
    def __init__(self, position, parent=None):
```

```
        self.position = position
```

```
        self.parent = parent
```

```
        self.g = 0 # Cost from start to current node
```

```
        self.h = 0 # Heuristic cost estimate to goal
```

```
        self.f = 0 # Total cost (g + h)
```

```
    def __lt__(self, other):
```

```
        return self.f < other.f # Compare nodes based on total cost
```

```
def heuristic(a, b):
```

```
    """ Manhattan Distance Heuristic """
```

```
    return abs(a[0] - b[0]) + abs(a[1] - b[1])
```

```
def a_star_algorithm(grid, start, goal):
```

```
    open_list = []
```

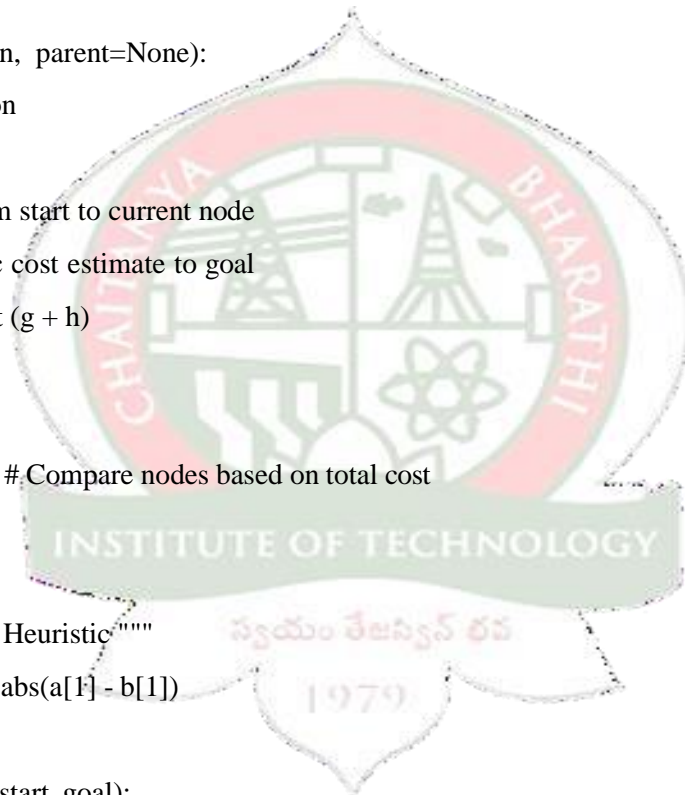
```
    closed_list = set()
```

```
    start_node = Node(start)
```

```
    goal_node = Node(goal)
```

```
    heapq.heappush(open_list, start_node) # Push start node to priority queue
```

```
    while open_list:
```



```

current_node = heapq.heappop(open_list) # Get node with lowest f-score
closed_list.add(current_node.position)

# Goal reached, reconstruct the path
if current_node.position == goal_node.position:
    path = []
    while current_node:
        path.append(current_node.position)
        current_node = current_node.parent
    return path[::-1] # Return reversed path

# Generate possible movements (up, down, left, right)
neighbors = [(0, -1), (0, 1), (-1, 0), (1, 0)]
for new_position in neighbors:
    node_position = (current_node.position[0] + new_position[0],
                    current_node.position[1] + new_position[1])

    # Check if within grid bounds
    if (node_position[0] < 0 or node_position[0] >= len(grid) or
        node_position[1] < 0 or node_position[1] >= len(grid[0])):
        continue # Skip invalid positions

    # Check if cell is walkable (0 = walkable, 1 = obstacle)
    if grid[node_position[0]][node_position[1]] != 0:
        continue # Skip obstacles

    neighbor = Node(node_position, current_node)

    if neighbor.position in closed_list:
        continue # Ignore nodes already processed

    # Calculate g, h, and f scores
    neighbor.g = current_node.g + 1
    neighbor.h = heuristic(neighbor.position, goal_node.position)
    neighbor.f = neighbor.g + neighbor.h

    # Add neighbor to open list if it's not already explored with a better g-score

```



```

if add_to_open(open_list, neighbor):
    heapq.heappush(open_list, neighbor)

```

```

return None # No path found

```

```

def add_to_open(open_list, neighbor):

```

```

    """ Check if neighbor is already in the open list with a lower g-score """

```

```

    for node in open_list:

```

```

        if neighbor.position == node.position and neighbor.g > node.g:

```

```

            return False

```

```

    return True

```

```

# Example grid (0 = open path, 1 = obstacle)

```

```

grid = [

```

```

    [0, 1, 0, 0, 0, 0],

```

```

    [0, 1, 0, 1, 1, 0],

```

```

    [0, 0, 0, 1, 0, 0],

```

```

    [0, 1, 1, 0, 0, 0],

```

```

    [0, 0, 0, 0, 1, 0],

```

```

    [0, 0, 1, 0, 0, 0]

```

```

]

```

```

# Define start and goal positions

```

```

start = (0, 0) # Top-left corner

```

```

goal = (5, 5) # Bottom-right corner

```

```

# Run A* algorithm

```

```

path = a_star_algorithm(grid, start, goal)

```

```

# Print results

```

```

print("Path:", path)

```

## OUTPUT:

```

PS C:\Users\HP> & C:/Users/HP/AppData/Local/Programs/Python/Python311/python.exe
xe c:/Users/HP/Downloads/1b.py
Path: [(0, 0), (1, 0), (2, 0), (3, 0), (4, 0), (4, 1), (4, 2), (4, 3), (5, 3),
(5, 4), (5, 5)]
PS C:\Users\HP>

```

## WEEK-3

**AIM:** Implement an 8-puzzle solver using Heuristic search technique (Misplaced Tiles)

**DISCRIPTION:** An 8-puzzle solver using a heuristic search technique, such as the A\* algorithm, finds the optimal sequence of moves to reach the goal state. The puzzle consists of a 3×3 grid with numbered tiles (1-8) and one empty space, where tiles can slide into the empty space. The solver uses heuristics like the **Manhattan Distance** (sum of tile distances from their goal positions) or **Misplaced Tiles** (count of tiles not in place) to guide the search efficiently. The A\* algorithm evaluates each state using  $f(n) = g(n) + h(n)$ , where  $g(n)$  is the cost so far and  $h(n)$  is the estimated cost to the goal. This approach ensures an optimal and efficient solution to the problem.

### CODE:

```
import heapq

def a_star(graph, start, goal, heuristic):
    open_list = [] # Priority queue (min-heap)
    heapq.heappush(open_list, (heuristic[start], 0, start)) # (f, g, node)
    came_from = {} # To reconstruct path
    g_score = {start: 0} # Shortest path cost from start
    while open_list:
        _, g, current = heapq.heappop(open_list)
        if current == goal: # Goal reached, reconstruct path
            path = []
            while current in came_from:
                path.append(current)
                current = came_from[current]
            path.append(start)
            return path[::-1], g # Reverse path to get correct order
        for neighbor, cost in graph[current]:
            tentative_g = g + cost # Calculate new cost
            if neighbor not in g_score or tentative_g < g_score[neighbor]:
                g_score[neighbor] = tentative_g
                f_score = tentative_g + heuristic[neighbor]
                heapq.heappush(open_list, (f_score, tentative_g, neighbor))
            came_from[neighbor] = current # Track path
```



```
return None, None # No path found
```

```
graph = { 'A': [('B', 6), ('F', 3)],
          'B': [('A', 6), ('C', 3), ('D', 2)],
          'C': [('B', 3), ('D', 1), ('E', 5)],
          'D': [('B', 2), ('C', 1), ('E', 8)],
          'E': [('C', 5), ('D', 8), ('I', 5), ('J', 5)],
          'F': [('A', 3), ('G', 1), ('H', 7)],
          'G': [('F', 1), ('I', 3)],
          'H': [('F', 7), ('I', 2)],
          'I': [('E', 5), ('G', 3), ('H', 2), ('J', 3)],
          'J': [('E', 5), ('I', 3)]}
```

```
heuristic = { 'A': 10, 'B': 8, 'C': 5, 'D': 7, 'E': 3, 'F': 6, 'G': 5, 'H': 3, 'I': 1, 'J': 0}
```

```
start, goal = 'A', 'J'
```

```
path, total_cost = a_star(graph, start, goal, heuristic)
```

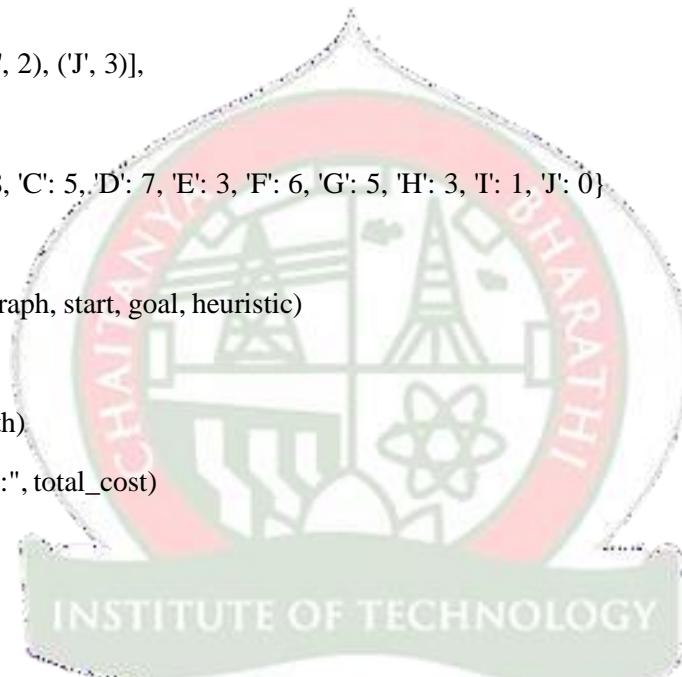
```
if path:
```

```
    print("Path found:", path)
```

```
    print("Shortest distance:", total_cost)
```

```
else:
```

```
    print("No path found")
```



### OUTPUT:

```
===== RESTART: /home/student/8puzzle.py
[2, 8, 3]
[1, 6, 4]
[7, 0, 5]
↓
[2, 8, 3]
[1, 0, 4]
[7, 6, 5]
↓
[2, 0, 3]
[1, 8, 4]
[7, 6, 5]
↓
[0, 2, 3]
[1, 8, 4]
[7, 6, 5]
↓
[1, 2, 3]
[0, 8, 4]
[7, 6, 5]
↓
[1, 2, 3]
[8, 0, 4]
[7, 6, 5]
```

**AIM:** Implement 8 puzzle problem Using Manhattan Distance

**DISCRIPTION:** The A\* algorithm using the Manhattan Distance heuristic is commonly applied to grid-based pathfinding problems. The Manhattan Distance heuristic  $h(n)$  calculates the sum of the absolute differences between the current node's coordinates and the goal node's coordinates ( $h(n) = |x1 - x2| + |y1 - y2|$ ). The algorithm prioritizes nodes based on the cost function  $f(n) = g(n) + h(n)$ , where  $g(n)$  is the actual cost from the start node. This heuristic is suitable for movement in a 4-directional grid (up, down, left, right) without diagonal movement. A\* ensures an optimal path when the heuristic does not overestimate the actual cost. It is widely used in AI, robotics, and game development for pathfinding.

**CODE:**

```
import heapq

class Puzzle:
    def __init__(self, board, g, parent=None):
        self.board = board
        self.g = g # Depth (cost so far)
        self.h = self.manhattan_distance()
        self.f = self.g + self.h # A* function: f(n) = g(n) + h(n)
        self.parent = parent # To track the path

    def __lt__(self, other):
        return self.f < other.f # Priority queue comparison

    def manhattan_distance(self):
        goal_pos = {1: (0, 0), 2: (0, 1), 3: (0, 2),
                    4: (1, 0), 5: (1, 1), 6: (1, 2),
                    7: (2, 0), 8: (2, 1), 0: (2, 2)}

        distance = 0

        for i in range(3):
            for j in range(3):
                value = self.board[i][j]

                if value != 0: # Ignore blank tile
                    goal_x, goal_y = goal_pos[value]

                    distance += abs(i - goal_x) + abs(j - goal_y)

        return distance

    def get_blank_pos(self):
        """Find blank space (0)"""

        for i in range(3):
```

```

    for j in range(3):
        if self.board[i][j] == 0:
            return i, j
def generate_moves(self):
    """Generate possible moves"""
    x, y = self.get_blank_pos()
    moves = []
    directions = [(-1, 0), (1, 0), (0, -1), (0, 1)] # Up, Down, Left, Right
    for dx, dy in directions:
        nx, ny = x + dx, y + dy
        if 0 <= nx < 3 and 0 <= ny < 3:
            new_board = [row[:] for row in self.board]
            new_board[x][y], new_board[nx][ny] = new_board[nx][ny], new_board[x][y]
            moves.append(Puzzle(new_board, self.g + 1, self))
    return moves
def solve_puzzle(initial_state):
    """A* Search to solve the 8-puzzle"""
    start = Puzzle(initial_state, 0)
    goal = [[1, 2, 3], [4, 5, 6], [7, 8, 0]]
    open_list = []
    closed_set = set()
    heapq.heappush(open_list, start)
    while open_list:
        current = heapq.heappop(open_list)
        if current.board == goal:
            path = []
            while current:
                path.append(current)
                current = current.parent
            return path[::-1] # Reverse to get the correct order
        closed_set.add(tuple(map(tuple, current.board)))
    for move in current.generate_moves():
        if tuple(map(tuple, move.board)) not in closed_set:

```

```

    heapq.heappush(open_list, move)

return None # No solution found

initial_state = [[1, 2, 3], [4, 0, 6], [7, 5, 8]]

solution = solve_puzzle(initial_state)

if solution:

    for step in solution:

        print("\nStep:", step.g)

        for row in step.board:

            print(row)

        print(f"h(n) = {step.h}, g(n) = {step.g}, f(n) = {step.f}")

        print("↓" if step.board != [[1, 2, 3], [4, 5, 6], [7, 8, 0]] else "Goal Reached 🚩")

else:

    print("No solution found.")

```

**OUTPUT:**

```

===== RESTART: /home/student/manhatted distance.py =====

Step: 0
[1, 2, 3]
[4, 0, 6]
[7, 5, 8]
h(n) = 2, g(n) = 0, f(n) = 2
↓

Step: 1
[1, 2, 3]
[4, 5, 6]
[7, 0, 8]
h(n) = 1, g(n) = 1, f(n) = 2
↓

Step: 2
[1, 2, 3]
[4, 5, 6]
[7, 8, 0]
h(n) = 0, g(n) = 2, f(n) = 2
Goal Reached

```