



Experiment No :- 1

Practical Aim:- Implement Depth First Search algorithm and Breadth First Search algorithm using an Undirected graph and develop a recursive algorithm for searching all the vertices of a graph by tree data structures.

Aim:- To implement Depth First Search (DFS) and Breadth First Search (BFS) algorithm using an Undirected graph and develop a recursive algorithm to traverse all vertices of a graph by tree data structures.

Objectives

1. To understand and implement graph traversal technique.
2. To differentiate Both DFS & BFS approaches.
3. To write recursive and iterative solution for traversal
4. To explore all vertices in a connected Undirected graph.



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Software Requirement

- Anaconda with python 3.7
- Python 3.6 Q4 above.
- Text Editor
- Cups
- Exporta Python package

Hardware Requirement:-

PIN 2GB RAM, 500GB HDD

processor :- Intel i3 Q4 above

RAM - 4GB minimum

Hard disk :- 500MB free space

Input / Output device - Keyboard

Monitor

Theory:-

Graph Basics

A graph is a set of vertices and edges.

A graph can be undirected or directed.

Graph and Commonly Represented Using

1. Adjacency Matrix
2. Adjacency List

BFS (Breadth first search)

- Explores Vertices level by level
- Uses a queue (FIFO)
- Starts from a source node and explores its neighbour first.
- Useful for finding the shortest path ~~so in unweighted graph~~.

DFS - (Depth first Search)

- Explores vertices depth wise before Backtracking
- Uses Recursion On a Stack
- Travels a path fully Before moving to next path
- Useful in topological sorting, cycle detection and maze solving.

Algorithm

~~DFS Algorithm~~ Algorithm (Recursive)

DFS(G, v , visited)

mark ' v ' as visited.

for all neighbours of v in G

if u is not visited

DFS (G, u , visited).



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Lab experiment No.: 2

Practical Title:-

Implementation of A* search algorithm for solving the 8-puzzle game

AIM:-

To implement the A* Search algorithm for solving the 8-puzzle game and demonstrate finding the optimal solution path.

* Objective

① To understand the working principle of the A* search algorithm.

② To apply A* to a game search problem (8-puzzle)

③ To compare A* to a game search algorithm of possible moves using



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

④ To produce the shortest sequence of mouse from the initial state of goal state.

⑤ To produce the shortest sequence of mouse from the initial state of goal state.

* Software requirements

① Operating system : windows

② Programming language : python

③ IDE : python pycharm / JS code /

* Hardware requirement

① Processor : Intel i5 or above

② RAM 4GB minimum

③ input devices : keyboard

④ Output devices : monitor

* Formula

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

$g(n)$:- Cost from start node to the current

$h(n)$:- estimate cost from current node to the goal.

$f(n)$:- estimate total cost from Start to goal via node.

For the 8-puzzle problem, the game consists of 3×3 board containing file no. 18 and one blank space the goal is to reach the target arrangement by sliding files.

Common heuristic

- ① Misplaced :- Count of tiles not in the Position.
- ② Manhattan distance:- Sum of the distance of file from there correct position.

* Algorithm

- ① Start with an initial board state.



- ② Initialize priority queue with start node
- ③ While queue is not empty,
- 1) Remove the node with lowest $f(n)$
 - 2) If the node is the goal state return the path.
 - 3) Generate all valid neighbour
 - 4) For each neighbour
 - Calculate $g(n)$ & $h(n)$
 - path into the priority queue if not already visited.
 - 5) If no solution is found return

Conclusion :-

The A* algorithm successfully finds the shortest path for solving the 8 puzzle game correctly combining the 8 puzzle game the actual cost and heuristic estimate. It is efficient for problems with well defined heavily on the occurring of the heuristic function.

Lab experiment : 03

Title : Solving the N-Queen problems using constraint satisfaction with back tracking and Branch & Bound.

Solving the N-Queen problems using constraint satisfaction with back tracking and Branch & Bound.

AIM :-

To implement and compare the back tracking and branch & Bound approaches for solving the N-Queens problem as a constraint satisfaction problem (CSP)

Objectives

① To understand the concept of constraint satisfaction problem in Artificial intelligence.

② To solve the N-Queen problems using the back tracking method.



③ To optimize the solution and using the branch and bound approach.

④ To compare performance in term of search space and execution time.

⑤ To demonstrate the efficiency of connecting in reducing unnecessary computation.

* Software requirement

- Operating System

Windows / Linux / program language : Python 3x

I-DF : PyCharm / VS code / Jupiter Notebook.

libraries: No external libraries required.



* Hardware Requirement

- ① Processor :- intel core i8 or higher.
- ② RAM :- 4GB or more.
- ③ Storage :- 500 MB free space.
- ④ Input Devices: keyboard.
- ⑤ Output device:- monitor.

* Theory

The N-queens problem is a classic constraint satisfaction problem in which we place N queens on $N \times N$ chess board so that

No two queens share the same row.

No two queens share the same column.



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Back Tracking approach.

A depth first search method places queen one by one in different columns.

If conflict is found back track to the previous try another position.

Branch and Bound approach.

Improves over back tracking by more.

Mount wins theme array.

- Columns [] : marks occupied
- Left diagonal [] : marks occupied left diagonal.
- Right diagonal [] : marks occupied diagonals.

*Algorithm

- ① Start from left most column
- ② Place a queen in a row when it not attacked.
- ③ If safe place the queen & move to the next column.



④ If placing leads to a conflict, remove the queen and try next row

⑤ Repeat until all queens one placed or no valid placement exists.

* Branch and Bound.

1) Initialize column [] , left diagonal [] right diagonal [] one / false .

2) Straight from column 0 .

3) For each row in the column 0 and right diagonal .

4) Back track by marking them false if no further placement possible .



Conclusion

Both Back tracking & Branch Bound can solve the Ns problem but branch bound is generally faster because it prunes ~~infeasible solution~~ earlier reducing the search space. This almost demonstrates the advantages of combining with optimization strategies.

Timely Submission	Performance	Understanding
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Aim :- Implement Greedy Search algorithm for any of the following application.

1. Selection Sort
2. Minimum Spanning tree
3. Single Scheduling problem
4. Job - Source Shortest path problem
5. prim's minimal Spanning tree algorithm
6. Kruskal's minimum Spanning tree

Also implement Dijkstra's Algorithm

Objective :-

To Study & Implement the Greedy search strategy

To understand how local Optimal choice can lead to a globally Optimal solution

Software Requirement :-
programming language / python / C++ / Java

IDE - pycharm / vs code / Turbo C++ / Eclipse
Compiler : Gcc for C++ , python 3x

IDE for java : Netbeans
Operating System : Window / Linux



Hardware Requirement: Intel i5
OS above

- RAM 4 GB minimum
- Hard disk 1250 MB free space
- I/O Device Keyboard, Monitor, Mouse.

Theory:-

The Greedy search algorithm works by making the locally optimal choice of each step with the hope of finding a global optimum. Greedy algorithms build up solutions piece by piece, always choosing the next piece that offers the most immediate benefit.

They are simple, fast & well work well for optimization problem.

Example:- Dijkstra's Algorithm

To find the shortest path from a single source node to all other nodes in a weighted path.

At each steps it picks the nodes with the minimum distance.

Algorithm

Steps:

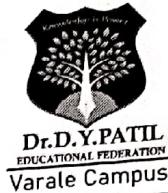
1. Initialize distance of all vertices at infinity & distance of source as 0.
2. Mark all vertices unvisited.
3. Select the unvisited vertex with the smallest distance.
4. Update the distance value of its neighbouring vertices.
5. Mark the current vertex as visited.
6. Repeat steps 3-5 until all vertices are visited.
7. Display the shortest path distance from the source.

Conclusion :- Greedy algorithm are useful for solving optimization problems efficiently. Implementing Dijkstra's algorithm using the greedy approach helps us understand how local decision can lead to global optimization.



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Experiment NO:05

* Problem Title: Develop an elementary chatbot power to morai interaction

* Objective

- To design by implement the simple chatbot per customer interaction.
- To understand the working of rule based retrieval - based chatbots
- To demonstrate natural language interaction between a user or the system.

* software Requirement

Programming language
IDE - pycharm / vscode / Jupiter
Library : NLTK, REOR simple conditional statement.

OS: windows/linux

Hardware Requirement

Program: Intel i3 or above

RAM: 4GB minimum

Hard disk: 200MB free space

I/O devices: Keyboard monitor

Theory

- A chatbot is an AI-based program that simulates human conversation.
- It can interact with users through text.
- Two main types
 - Rule-based chatbot: Predefined processing of AI to understand user queries.

In this experiment, we developed a rule-based chatbot. It matches user queries with predefined responses.

- If the I/P do not match, it provides a default "I don't understand" type response.
- If the I/P responses

This approach is simple and suitable for elementary applications.

Algorithm:-

1. Start the program
2. Define a set of predefined questions and their responses.

3. Accept IP from the user.
4. Compare user IP with stored keyboards using string matching.
5. If a match is found, display the corresponding response.
6. If no match is found display a default message.
7. Continue the conversion until the user types "exit".
8. End the program.

Conclusion:

An elementary chatbot can simulate simple human-like conversation using rule-based responses. This experiment gives a foundation for building advanced chatbots using NLP & ML.

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