# RAJALAKSHMI ENGINEERING COLLEGE (Autonomous)

# RAJALAKSHMI NAGAR, THANDALAM, CHENNAI-602105

# DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING



#### AI19341

#### PRINCIPLES OF ARTIFICIAL INTELLIGENCE LAB

THIRD YEAR

FIFTH SEMESTER

# **INDEX**

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#### **Working Tools and Language**

#### **PYTHON LANGUAGE**

- Interpreter Languages
- Interpreted
- Not compiled like Java
- Code is written and then directly executed by an interpreter
- Type commands into interpreter and see immediate results

Java: Code Compiler Runtime Environment Computer

Python: Code Interpreter Computer

# **Python Installation Steps**

#### Windows:

- Download Python from http://www.python.org
- Install Python.
- · Run Idle from the Start Menu.

#### Mac OS X:

- Python is already installed.
- Open a terminal and run python or run Idle from Finder.

#### Linux:

- Chances are you already have Python installed. To check, run python from the terminal.
- If not, install from your distribution's package system.

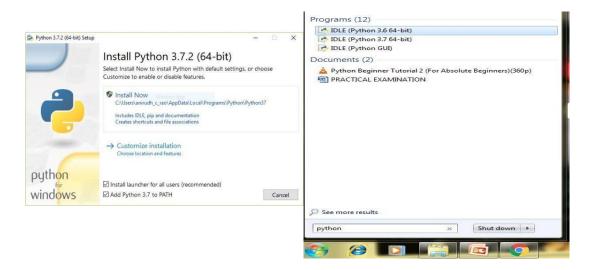
**Note:** For step by step installation instructions, see the course web site.

#### Step 1: Download the Python 3 Installer

- Open a browser window and navigate to the Download page for Windows at python.org.
- Underneath the heading at the top that says Python Releases for Windows, click on the link for the Latest Python 3 Release Python 3.x.x. (As of this writing, the latest version is Python 3.7.2.)
- Scroll to the bottom and select either Windows x86-64 executable installer for 64-bit or Windows x86 executable installer for 32-bit.

#### Step 2: Run the Installer

• Once you have chosen and downloaded an installer, simply run it by double-clicking on the downloaded file. A dialog should appear that looks something like this:



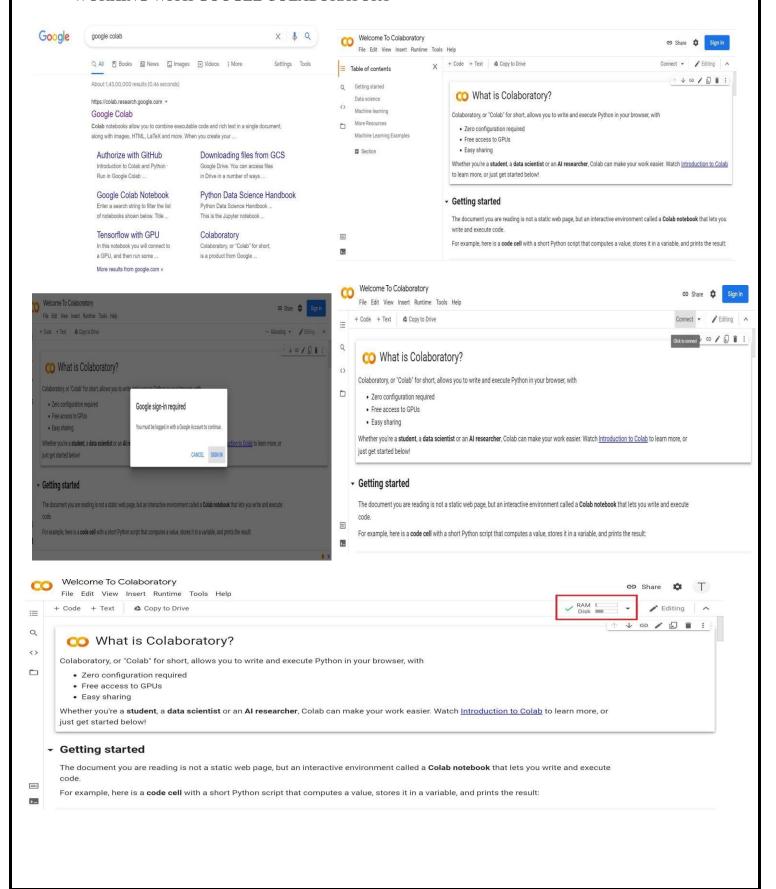
# INSTALLATION STEPS ON PYTHON

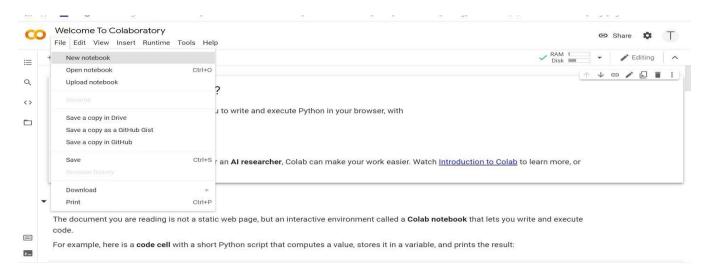


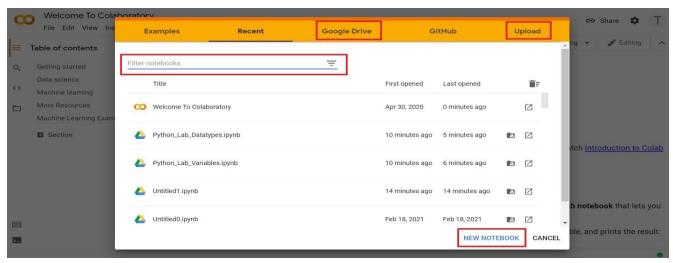


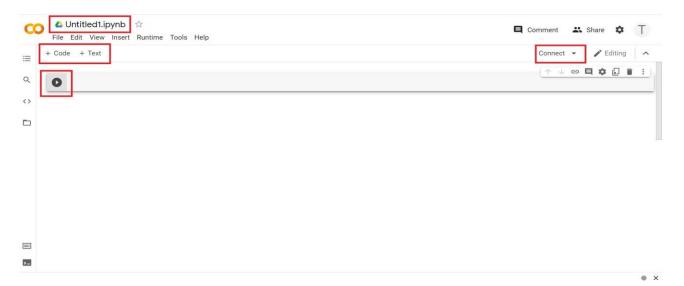
Version	Operating System	Description	MD5 Sum	File Size	GPG
Gzipped source tarball	Source release		e1a36bfffdd1d3a780b1825daf16e56c	22580749	SIG
XZ compressed source tarball	Source release		2c68846471994897278364fc18730dd9	16907204	SIG
Mac OS X 64-bit/32-bit installer	Mac OS X	for Mac OS X 10.6 and later	86e6193fd56b1e757fc8a5a2bb6c52ba	27561006	SIG
Windows help file	Windows		e520a5c1c3e3f02f68e3db23f74a7a90	8010498	SIG
Windows x86-64 embeddable zip file	Windows	for AMD64/EM64T/x64	0fdfe9f79e0991815d6fc1712871c17f	7047535	SIG
Windows x86-64 executable installer	Windows	for AMD64/EM64T/x64	4377e7d4e6877c248446f7cd6a1430cf	31434856	SIG
Windows x86-64 web-based installer	Windows	for AMD64/EM64T/x64	58ffad3d92a590a463908dfedbc68c18	1312496	SIG
Windows x86 embeddable zip file	Windows		2ca4768fdbadf6e670e97857bfab83e8	6332409	SIG
Windows x86 executable installer	Windows		8d8e1711ef9a4b3d3d0ce21e4155c0f5	30507592	SIG
Windows x86 web-based installer	Windows		ccb7d66e3465eaf40ade05b76715b56c	1287040	SIG

#### WORKING WITH GOOGLE COLABORATORY









EX.NO:

#### **8- QUEENS PROBLEM**

#### AIM:

To implement an 8-Queesns problem using Python.

You are given an 8x8 board; find a way to place 8 queens such that no queen can attack any other queen on the chessboard. A queen can only be attacked if it lies on the same row, same column, or the same diagonal as any other queen. Print all the possible configurations.

To solve this problem, we will make use of the Backtracking algorithm. The backtracking algorithm, in general checks all possible configurations and test whether the required result is obtained or not. For the given problem, we will explore all possible positions the queens can be relatively placed at. The solution will be correct when the number of placed queens = 8.





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```
# Function to print the board configuration
def print board(board):
     for row in board:
         print(" ".join("Q" if col else "." for col in row))
     print("\n")
# Function to check if a queen can be placed at board[row][col]
def is_safe(board, row, col):
     # Check the current row on left side
     for i in range(col):
         if board[row][i]:
             return False
     # Check upper diagonal on left side
     for i, j in zip(range(row, -1, -1), range(col, -1, -1)):
         if board[i][j]:
             return False
     # Check lower diagonal on left side
     for i, j in zip(range(row, len(board)), range(col, -1, -1)):
         if board[i][j]:
             return False
     return True
# Recursive function to solve the 8-Queens problem
def solve 8 queens(board, col):
     # Base case: If all queens are placed, print the board
     if col >= len(board):
         print_board(board)
         return True
     # Consider this column and try placing queens in all rows one by one
     res = False
     for i in range(len(board)):
         if is_safe(board, i, col):
             # Place this queen in board[i][col]
             hound[f1][col1 - 1
```

# **RESULT:**

Thus ,To implement an 8-Queesns problem using Python has been executed and verified.

**EX.NO:** 

**DATE:** 

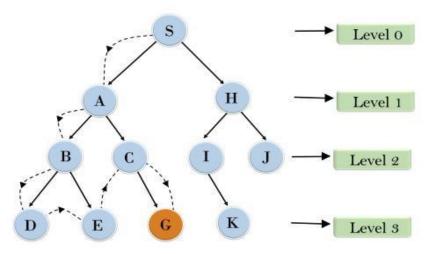
### **DEPTH-FIRST SEARCH**

#### AIM:

To implement a depth-first search problem using Python.

- Depth-first search (DFS) algorithm or searching technique starts with the root node of graph G, and then travel deeper and deeper until we find the goal node or the node which has no children by visiting different node of the tree.
- The algorithm, then backtracks or returns back from the dead end or last node towards the most recent node that is yet to be completely unexplored.
- The data structure (DS) which is being used in DFS Depth-first search is stack. The process is quite similar to the BFS algorithm.
- In DFS, the edges that go to an unvisited node are called discovery edges while the edges that go to an already visited node are called block edges.

# **Depth First Search**



**PROGRAM:** 

```
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 → True
 # Define the graph as an adjacency list
     graph = {
         'A': ['B', 'C'],
         'B': ['D', 'E'],
         'C': ['F'],
         'D': [],
         'E': ['F'],
         'F': []
     # DFS function using recursion
     def dfs_recursive(graph, node, visited):
         if node not in visited:
             print(node, end=" ")
             visited.add(node)
             # Recursively visit all the unvisited neighbors
             for neighbor in graph[node]:
                dfs_recursive(graph, neighbor, visited)
     # DFS function using an explicit stack
     def dfs_stack(graph, start):
         visited = set()
         stack = [start]
         while stack:
             node = stack.pop()
             if node not in visited:
                 print(node, end=" ")
                 visited.add(node)
```

# Add neighbors to the stack in reverse order to maintain DFS order

```
DFS:
A
B
D
E
F
C
```

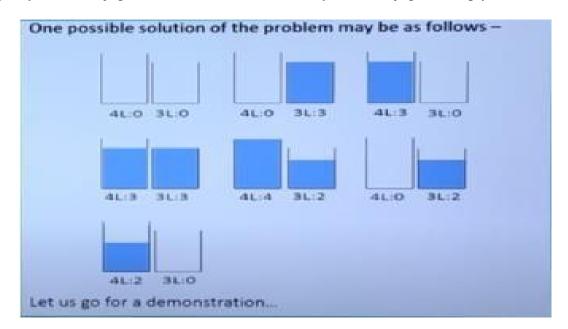
# **RESULT:**

Therefore, To implement a depth-first search problem using Python has been executed and verified.

EX.NO:

#### **DEPTH-FIRST SEARCH – WATER JUG PROBLEM**

In the water jug problem in Artificial Intelligence, we are provided with two jugs: one having the capacity to hold 3 gallons of water and the other has the capacity to hold 4 gallons of water. There is no other measuring equipment available and the jugs also do not have any kind of marking on them. So, the agent's task here is to fill the 4-gallon jug with 2 gallons of water by using only these two jugs and no other material. Initially, both our jugs are empty.



**PROGRAM:** 







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```
from collections import deque
    # Function to solve the water jug problem
    def water_jug_problem():
        # Initialize the queue for BFS
        queue = deque()
        queue.append((0, 0)) # Starting state with both jugs empty
        visited = set()
                             # Set to track visited states
        visited.add((0, 0))
        # BFS Loop
        while queue:
            x, y = queue.popleft()
            # If we achieve the goal, print the state and return
            if y == 2:
                print(f"Solution found: (3-gallon jug: {x}, 4-gallon jug: {y})")
                return
            # Print the current state
            print(f"Current state: (3-gallon jug: {x}, 4-gallon jug: {y})")
            # Possible operations
            next_states = [
                (3, y), # Fill the 3-gallon jug
                (x, 4), # Fill the 4-gallon jug
                (0, y), # Empty the 3-gallon jug
                (x, 0), # Empty the 4-gallon jug
                (x - min(x, 4 - y), y + min(x, 4 - y)), # Pour water from 3-gallon jug to 4-gallon jug
                (x + min(y, 3 - x), y - min(y, 3 - x)), # Pour water from 4-gallon jug to 3-gallon jug
            # Process each next possible state
            for next_state in next_states:
                if next_state not in visited:
                    queue.append(next_state)
                    visited.add(next_state)
        nmint("No solution found ")
```

```
    [('fill', 2), ('pour', 2, 1), ('fill', 2), ('pour', 2, 1)]
```

#### **RESULT:**

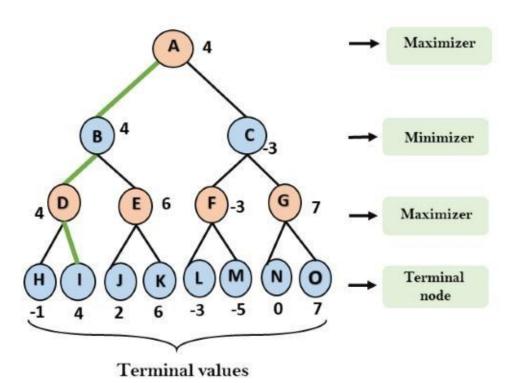
Thus, DEPTH-FIRST SEARCH – WATER JUG PROBLEM has been executed and verified.

EX.NO:

#### **MINIMAX ALGORITHM**

• A simple example can be used to explain how the minimax algorithm works. We've included an example of a game-tree below, which represents a two-player game.

- There are two players in this scenario, one named Maximizer and the other named Minimizer.
- Maximizer will strive for the highest possible score, while Minimizer will strive for the lowest possible score.
- Because this algorithm uses DFS, we must go all the way through the leaves to reach the terminal nodes in this game-tree.
- The terminal values are given at the terminal node, so we'll compare them and retrace the tree till we reach the original state.

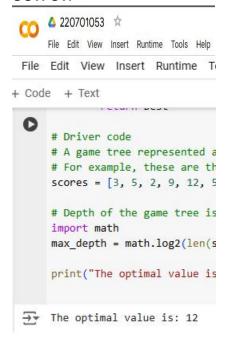


#### **PROGRAM:**



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```
# Minimax function to evaluate a game tree
def minimax(depth, node_index, is_maximizer, scores, max_depth):
    # Base case: If we've reached the terminal node
    if depth == max depth:
        return scores[node_index]
    # If it's Maximizer's turn, choose the maximum value
    if is maximizer:
         best = float('-inf')
         # Recur for left and right children
        best = max(best, minimax(depth + 1, node_index * 2, False, scores, max_depth))
        best = max(best, minimax(depth + 1, node_index * 2 + 1, False, scores, max_depth))
         return best
    # If it's Minimizer's turn, choose the minimum value
    else:
        best = float('inf')
         # Recur for left and right children
        best = min(best, minimax(depth + 1, node_index * 2, True, scores, max_depth))
         best = min(best, minimax(depth + 1, node_index * 2 + 1, True, scores, max_depth))
         return best
# Driver code
# A game tree represented as leaf nodes
# For example, these are the terminal nodes and their scores
scores = [3, 5, 2, 9, 12, 5, 23, 23]
# Depth of the game tree is log2(len(scores))
import math
max_depth = math.log2(len(scores))
print("The optimal value is:", minimax(0, 0, True, scores, int(max_depth)))
```



#### **RESULT:**

Thus, the program to implement **MINIMAX ALGORITHM** has been successfully executed and verified

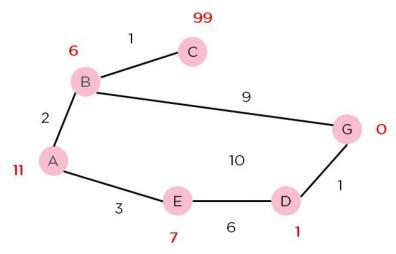
EX.No:

#### **A\* SEARCH ALGORITHM**

A heuristic algorithm sacrifices optimality, with precision and accuracy for speed, to solve problems faster and more efficiently.

All graphs have different nodes or points which the algorithm has to take, to reach the final node. The paths between these nodes all have a numerical value, which is considered as the weight of the path. The total of all paths transverse gives you the cost of that route.

Initially, the Algorithm calculates the cost to all its immediate neighboring nodes,n, and chooses the one incurring the least cost. This process repeats until no new nodes can be chosen and all paths have been traversed. Then, you should consider the best path among them. If f(n) represents the final cost, then it can be denoted as : f(n) = g(n) + h(n), where :  $g(n) = \cos t$  of traversing from one node to another. This will vary from node to node h(n) = heuristic approximation of the node's value. This is not a real value but an approximation cost.



#### **PROGRAM:**



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```
import heapq
    # A class to represent the graph
    class Graph:
        def __init__(self):
            self.edges = {}
                                  # Stores edges as adjacency lists
            self.h = {}
                                    # Stores heuristic values
        # Add an edge to the graph
        def add_edge(self, from_node, to_node, cost):
            if from_node not in self.edges:
                self.edges[from node] = []
            self.edges[from_node].append((to_node, cost))
        # Set heuristic value for each node
        def set_heuristic(self, node, heuristic_value):
            self.h[node] = heuristic_value
        # A* algorithm implementation
        def a_star(self, start, goal):
            # Priority queue to store nodes with their f(n) values
            open_list = []
            heapq.heappush(open_list, (0, start))
            # Dictionaries to store g(n) values and the optimal path
            g = {start: 0}
            came_from = {start: None}
            while open_list:
                # Pop the node with the lowest f(n) value
                _, current = heapq.heappop(open_list)
                # If the goal is reached, construct the path
                if current == goal:
                    path = []
                    while current:
                        path.append(current)
                        current - came from[current]
```

```
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graph.set_heuristic("G", 0) f

# Find the shortest path from start_node = "A"
goal_node = "G"
path, cost = graph.a_star(star

# Print the results
print(f"Path: {path}")
print(f"Total cost: {cost}")

Path: ['A', 'B', 'E', 'G']
Total cost: 4
```

#### **RESULT:**

Thus, the program implementing A\* SEARCH ALGORITHM has been executed and verified

EX.NO:

#### **INTRODUCTION TO PROLOG**

#### AIM:

To learn PROLOG terminologies and write basic programs.

#### **TERMINOLOGIES**

1. Atomic Terms: -

Atomic terms are usually strings made up of lower- and uppercase letters, digits, and the underscore, starting with a lowercase letter. Ex:

dog ab c 321

2. Variables: -

Variables are strings of letters, digits, and the underscore, starting with a capital letter or an underscore.

Ex:

Dog Apple 420

3. Compound Terms: -

Compound terms are made up of a PROLOG atom and a number of arguments (PROLOG terms, i.e., atoms, numbers, variables, or other compound terms) enclosed in parentheses and separated by commas.

Ex:

is\_bigger(elephant,X)

f(g(X,\_),7) 4. Facts: 
A fact is a predicate followed by a dot.

Ex:

bigger\_animal(whale).

life\_is\_beautiful.

5. Rules: -

A rule consists of a head (a predicate) and a body (a sequence of predicates separated by commas).

Ex:

is\_smaller(X,Y):-is\_bigger(Y,X). aunt(Aunt,Child):-sister(Aunt,Parent),parent(Parent,Child).

#### **SOURCE CODE:**

```
KB1:
```

#### **KB2**:

happy(yolanda). listens2music(mia).

Listens2music(yolanda):-happy(yolanda). playsAirGuitar(mia):-listens2music(mia). playsAirGuitar(Yolanda):-listens2music(yolanda).

ERROR: Unknown procedure: concert/0 (DWIM could not correct goal)

#### **OUTPUT: -**

```
?- playsAirGuitar(mia).
true .
?- playsAirGuitar(yolanda).
true.
?- ■
KB3: likes(dan,sally). likes(sally,dan).
likes(john,brittney). married(X,Y):-
likes(X,Y), likes(Y,X). friends(X,Y):-
likes(X,Y); likes(Y,X).
```

```
OUTPUT: -
?- likes(dan, X).
X = sally.
?- married(dan, sally).
true.
?- married(john, brittney).
false.
KB4: food(burger).
food(sandwich).
food(pizza).
lunch(sandwich).
dinner(pizza).
meal(X):-food(X).
```

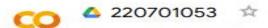
```
?-
| food(pizza).
true.
?- meal(X).lunch(X).
X = sandwich ,
?- dinner(sandwich).
false.
?-
```

#### **KB5**:

```
owns(jack,car(bmw)).
owns(john,car(chevy)).
owns(olivia,car(civic)).
owns(jane,car(chevy)).
sedan(car(bmw)). sedan(car(civic)).
truck(car(chevy)).
```

```
?-
| owns(john,X).
X = car(chevy).
?- owns(john,_).
true.
?- owns(Who,car(chevy)).
Who = john ,
?- owns(jane,X),sedan(X).
false.
?- owns(jane,X),truck(X).
X = car(chevy).
```

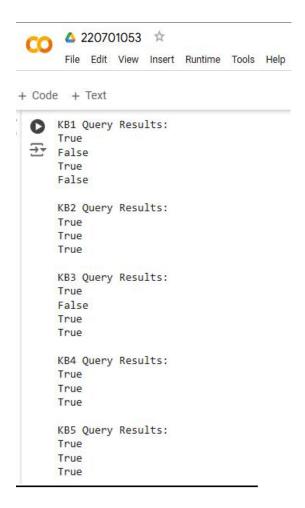
#### **PROGRAM:**



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```
KB1 = {
    "woman": ["mia", "jody", "yolanda"],
"playsAirGuitar": ["jody"],
    "party": True
def query_kb1(query):
    if query[0] == "woman" and query[1] in KB1["woman"]:
        return True
    elif query[0] == "playsAirGuitar" and query[1] in KB1["playsAirGuitar"]:
        return True
    elif query[0] == "party" and KB1["party"]:
        return True
    else:
        return False
# Queries for KB1
print("KB1 Query Results:")
print(query_kb1(["woman", "mia"]))  # Expected True
print(query_kb1(["playsAirGuitar", "mia"])) # Expected False
                                  # Expected True
print(query_kb1(["party"]))
print(query_kb1(["concert"]))
                                            # Expected False
# Define KB2 with facts and rules
KB2 = {
    "happy": ["yolanda"],
    "listens2music": ["mia"],
def listens2music(person):
    return person in KB2["listens2music"] or person in KB2["happy"]
def playsAirGuitar(person):
    return listens2music(person)
# Queries for KB2
print("\nKB2 Query Results:")
print(listens2music("mia"))
                                             # Expected True
                                          # Expected True
nrint(nlavsAirGuitar("mia"))
```



# **RESULT:**

Thus ,the program implementing Prolog has been executed successfully.

EX.NO DATE:

#### PROLOG- FAMILY TREE

#### **AIM:**

To develop a family tree program using PROLOG with all possible facts, rules, and queries.

# **SOURCE CODE: KNOWLEDGE BASE:**

```
/*FACTS :: */
male(peter). male(john).
male(chris).
male(kevin).
female(betty).
female(jeny). female(lisa).
female(helen).
parentOf(chris,peter).
parentOf(chris,betty).
parentOf(helen,peter).
parentOf(helen,betty).
parentOf(kevin,chris).
parentOf(kevin,lisa). parentOf(jeny,john).
parentOf(jeny,helen).
/*RULES :: */
/* son,parent
* son,grandparent*/
father(X,Y):-male(Y), parentOf(X,Y).
mother(X,Y):- female(Y), parentOf(X,Y).
grandfather(X,Y):-male(Y),parentOf(X,Z),parentOf(Z,Y).
grandmother(X,Y):- female(Y), parentOf(X,Z), parentOf(Z,Y).
brother(X,Y):- male(Y), father(X,Z), father(Y,W),Z==W.
```

sister(X,Y):- female(Y), father(X,Z),father(Y,W),Z==W. **PROGRAM:** 

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+ Code + Text
 # Facts
     male = {"peter", "john", "chris", "kevin"}
     female = {"betty", "jeny", "lisa", "helen"}
     # Parent relationships
     parent_of = {
         "chris": ["peter", "betty"],
         "helen": ["peter", "betty"],
         "kevin": ["chris", "lisa"],
         "jeny": ["john", "helen"]
     # Functions for family relationships
     def father(child, person):
         return person in male and person in parent_of and child in parent_of[person]
     def mother(child, person):
         return person in female and person in parent_of and child in parent_of[person]
     def grandfather(grandchild, person):
         if person in male:
             for parent in parent_of.get(person, []):
                 if grandchild in parent_of.get(parent, []):
                     return True
         return False
     def grandmother(grandchild, person):
         if person in female:
             for parent in parent_of.get(person, []):
                 if grandchild in parent_of.get(parent, []):
                     return True
         return False
     def brother(sibling1, sibling2):
         for parent, children in parent of.items():
             if sibling1 in children and sibling2 in children and sibling1 != sibling2:
                 return sibling1 in male and sibling2 in male
          noturn Falco
```



#### **RESULT:**

Thus, the program implementing prolog family tree has been executed successfully.

EX.NO:

# IMPLEMENTING ARTIFICIAL NEURAL NETWORKS FOR AN APPLICATION USING PYTHON - REGRESSION

#### **Regression using Artificial Neural Networks**

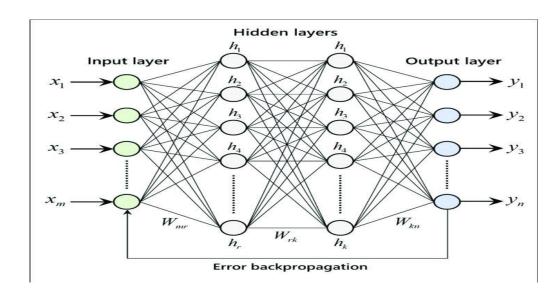
Why do we need to use Artificial Neural Networks for Regression instead of simply using Linear Regression?

The purpose of using Artificial Neural Networks for Regression over Linear Regression is that the linear regression can only learn the linear relationship between the features and target and therefore cannot learn the complex non-linear relationship. In order to learn the complex non-linear relationship between the features and target, we are in need of other techniques. One of those techniques is to use Artificial Neural Networks. Artificial Neural Networks have the ability to learn the complex relationship between the features and target due to the presence of activation function in each layer. Let's look at what are Artificial Neural Networks and how do they work.

#### **Artificial Neural Networks**

Artificial Neural Networks are one of the deep learning algorithms that simulate the workings of neurons in the human brain. There are many types of Artificial Neural Networks, Vanilla Neural Networks, Recurrent Neural Networks, and Convolutional Neural Networks. The Vanilla Neural Networks have the ability to handle structured data only, whereas the Recurrent Neural Networks and Convolutional Neural Networks have the ability to handle unstructured data very well. In this post, we are going to use Vanilla Neural Networks to perform the Regression Analysis.

#### **Structure of Artificial Neural Networks**



The Artificial Neural Networks consists of the Input layer, Hidden layers, Output layer. The hidden layer can be more than one in number. Each layer consists of n number of neurons. Each layer will be having an Activation Function associated with each of the neurons. The activation function is the function that is responsible for introducing non-linearity in the relationship. In our case, the output layer must contain a linear activation function. Each layer can also have regularizers associated with it. Regularizers are responsible for preventing overfitting.

Artificial Neural Networks consists of two phases,

- Forward Propagation
- Backward Propagation

Forward propagation is the process of multiplying weights with each feature and adding them. The bias is also added to the result. Backward propagation is the process of updating the weights in the model.

#### AIM:

To implementing artificial neural networks for an application in Regression using python.

Backward propagation requires an optimization function and a loss function.

#### **PROGRAM**



- Code + Text

```
# Import necessary libraries
    import numpy as np
    import pandas as pd
    from sklearn.model selection import train test split
    from sklearn.preprocessing import StandardScaler
    from sklearn.metrics import mean_squared_error
    from tensorflow.keras.models import Sequential
    from tensorflow.keras.layers import Dense
    from tensorflow.keras.optimizers import Adam
    # Generate synthetic dataset (you can replace this with your own data)
    np.random.seed(42)
    data size = 1000
    X = np.random.rand(data_size, 3) # 3 input features
    y = X[:, 0] * 5 + X[:, 1] * 3 + X[:, 2] * 2 + np.random.randn(data_size) * 0.5 # Target with some noise
    # Split dataset into training and testing sets
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
    # Standardize the features for better performance
    scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
    X_test = scaler.transform(X_test)
    # Define the neural network model
    model = Sequential()
    model.add(Dense(64, input_dim=X_train.shape[1], activation='relu'))
    model.add(Dense(32, activation='relu'))
    model.add(Dense(1)) # Output layer for regression (1 neuron)
    # Compile the model with appropriate loss and optimizer for regression
    model.compile(optimizer=Adam(learning_rate=0.01), loss='mean_squared_error')
    # Train the model
    history = model.fit(X_train, y_train, epochs=100, batch_size=32, validation_split=0.2, verbose=1)
    # Evaluate the model on the test set
    y_pred = model.predict(X_test)
```

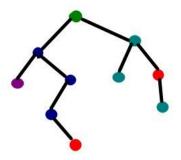


**RESULT:** Thus, the program implementing artificial neural networks for an AN application suing python regression has been executed successfully.

EX.NO: DATE:

#### IMPLEMENTATION OF DECISION TREE CLASSIFICATION TECHNIQUES

<u>Decision Tree</u> is one of the most powerful and popular algorithm. Decision-tree algorithm falls under the category of supervised learning algorithms. It works for both continuous as well as categorical output variables.



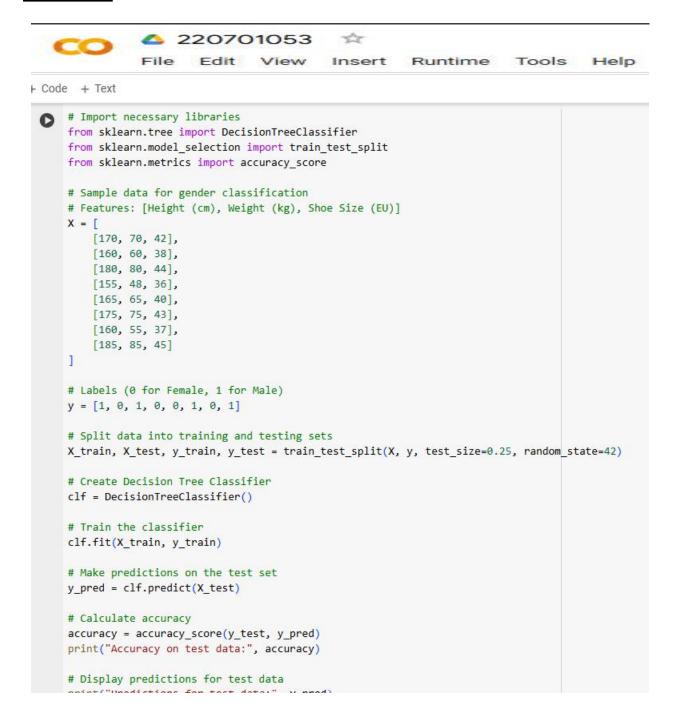
#### AIM:

To implement a decision tree classification technique for gender classification using python.

#### **EXPLANATION:**

- Import tree from sklearn.
- Call the function DecisionTreeClassifier() from tree
- Assign values for X and Y.
- Call the function predict for Predicting on the basis of given random values for each given feature.
- Display the output.

#### **PROGRAM:**





# **RESULT:**

Thus, the program implementing decision tree classification has has been executed successfully.

EX NO:

#### <u>IMPLEMENTATION OF CLUSTERING TECHNIQUES K - MEANS</u>

The *k*-means clustering method is an <u>unsupervised machine learning</u> technique used to identify clusters of data objects in a dataset. There are many different types of clustering methods, but *k*means is one of the oldest and most approachable. These traits make implementing *k*-means clustering in Python reasonably straightforward, even for novice programmers and data scientists.

If you're interested in learning how and when to implement *k*-means clustering in Python, then this is the right place. You'll walk through an end-to-end example of *k*-means clustering using Python, from preprocessing the data to evaluating results.

#### How does it work?

First, each data point is randomly assigned to one of the K clusters. Then, we compute the centroid (functionally the center) of each cluster, and reassign each data point to the cluster with the closest centroid. We repeat this process until the cluster assignments for each data point are no longer changing.

K-means clustering requires us to select K, the number of clusters we want to group the data into. The elbow method lets us graph the inertia (a distance-based metric) and visualize the point at which it starts decreasing linearly. This point is referred to as the "eblow" and is a good estimate for the best value for K based on our data.

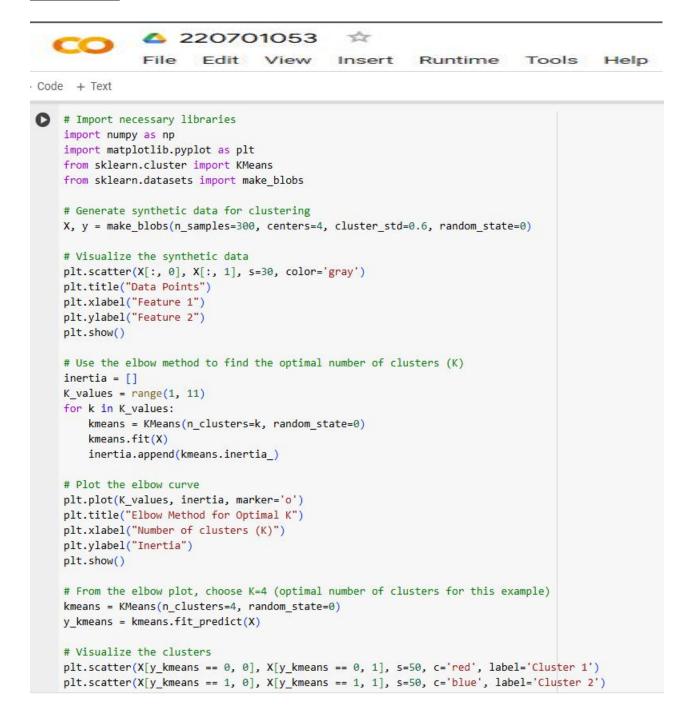
#### AIM:

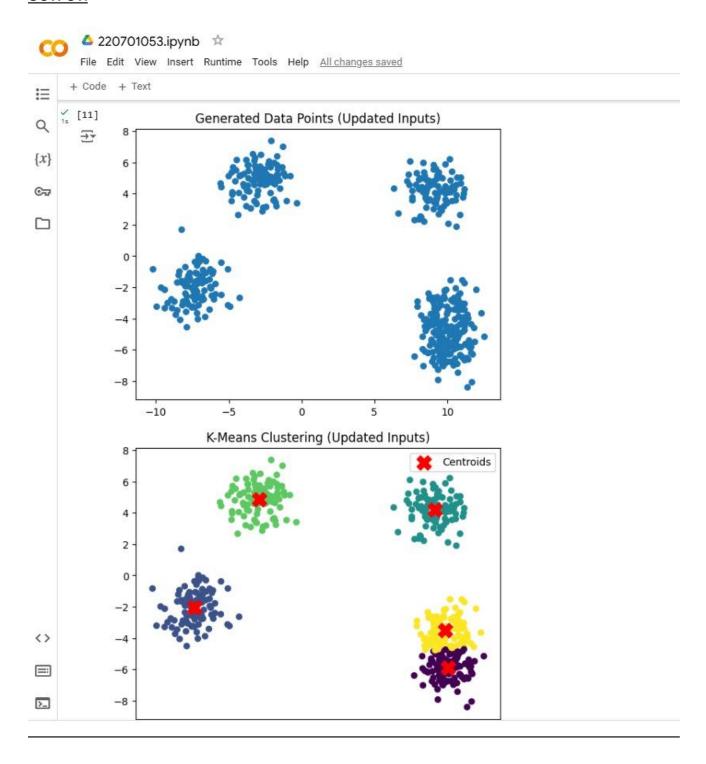
To implement a K - Means clustering technique using python language.

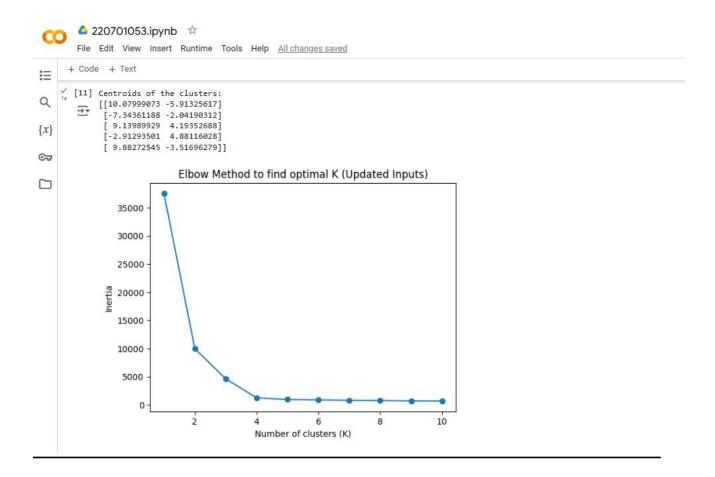
#### **EXPLANATION:**

- Import KMeans from sklearn.cluster
- Assign X and Y.
- Call the function KMeans().
- Perform scatter operation and display the output.

#### **PROGRAM:**







# **RESULT:**

Thus, the program implementing Cluster tree K-means has been executed successfully.