# 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

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**EX.NO:** 1 **DATE:** 13-09-24

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

#### <u>**AIM**</u>:

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.



#### CODE:

```
def share_diagonal(x0, y0, x1, y1):
dx = abs(x0 - x1)
dy = abs(y0 - y1)
return dy == dx
def col_clashes(bs, c):
for i in range(c):
```

```
if share_diagonal(i, bs[i], c, bs[c]):
return True
return False
def has clashes(the board):
for col in range(1, len(the board)):
if col clashes(the board, col):
return True
return False
def main():
import random
n=int(input("Enter the number of queens: "))
rng = random.Random()
bd = list(range(n))
num_found = 0
tries = 0
result = []
while num found < 5:
rng.shuffle(bd)
tries += 1
if not has_clashes(bd) and bd not in result:
print("Found solution {0} in {1} tries.".format(bd, tries))
tries = 0
num found += 1
result.append(list(bd))
print(result)
main()
```

#### **OUTPUT:**

#### **RESULT:**

The code for N-Queens problem has been executed successfully.

#### **EX.NO:2**

#### **DATE:13-09-24**

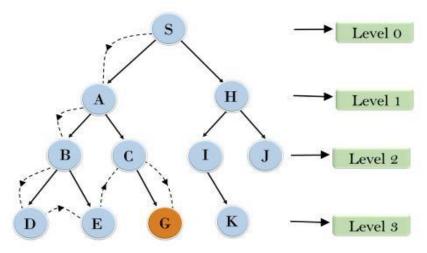
#### **DEPTH-FIRST SEARCH**

#### 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**



# CODE: def dfs(graph, start, visited=None): if visited is None: visited = set()visited.add(start) print(start, end=" ") for neighbour in graph.get(start, []): if neighbour not in visited: dfs(graph, neighbour, visited) num nodes = int(input("Enter the number of nodes: ")) $graph = \{\}$ for i in range(num nodes): node = input("Enter node " + str(i+1) + ": ").strip() neighbors = input("Enter neighbors of " + node + " (comma-separated): ").strip().split(',') neighbors = [n.strip() for n in neighbors] graph[node] = neighbors print("Graph:", graph) start node = input("Enter the starting node: ").strip() print("Starting node:", start node)

#### **OUTPUT:**

dfs(graph, start node)

```
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  Enter the number of nodes: 4
     Enter node 1: a
 Enter neighbors of a (comma-separated): b,c
      Enter node 2: b
      Enter neighbors of b (comma-separated): d
      Enter node 3: c
      Enter neighbors of c (comma-separated): d
      Enter node 4: d
      Enter neighbors of d (comma-separated): a
      Graph: {'a': ['b', 'c'], 'b': ['d'], 'c': ['d'], 'd': ['a']}
      Enter the starting node: a
      Starting node: a
      a b d c
```

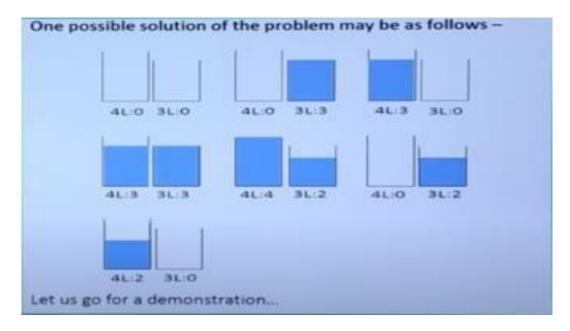
#### **RESULT:**

The code for DFS has been successfully executed.

#### **EX.NO:** 3 **DATE:** 20-09-24

#### **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.



```
CODE:
```

```
def fill_4_gallon(x, y, x_max, y_max):
return (x_max, y)
def fill_3_gallon(x, y, x_max, y_max):
return (x, y_max)
def empty_4_gallon(x, y, x_max, y_max):
return (0, y)
def empty_3_gallon(x, y, x_max, y_max):
return (x, 0)
def pour_4_to_3(x, y, x_max, y_max):
```

```
transfer = min(x, y max - y)
return (x - transfer, y + transfer)
def pour 3 to 4(x, y, x max, y max):
transfer = min(y, x max - x)
return (x + transfer, y - transfer)
def dfs water jug(x max, y max, goal x, visited=None, start=(0, 0)):
if visited is None:
visited = set()
stack = [start]
while stack:
state = stack.pop()
x, y = state
if state in visited:
continue
visited.add(state)
print(f"Visiting state: {state}")
if x == goal x:
print(f"Goal reached: {state}")
return state
next states = [
fill 4 gallon(x, y, x max, y max),
fill 3 gallon(x, y, x max, y max),
empty_4_gallon(x, y, x_max, y_max),
empty 3 gallon(x, y, x max, y max),
pour 4 to 3(x, y, x max, y max),
pour 3 to 4(x, y, x max, y max)
for new state in next states:
if new state not in visited:
stack.append(new state)
return None
x max = 4
y max = 3
goal x = 2
```

```
dfs_water_jug(x_max, y_max, goal_x)
```

#### **OUTPUT:**

```
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 Visiting state: (0, 0)
     Visiting state: (0, 3)
 Yisiting state: (3, 0)
     Visiting state: (3, 3)
     Visiting state: (4, 2)
     Visiting state: (4, 0)
     Visiting state: (1, 3)
     Visiting state: (1, 0)
     Visiting state: (0, 1)
     Visiting state: (4, 1)
     Visiting state: (2, 3)
     Goal reached: (2, 3)
     (2, 3)
```

#### **RESULT:**

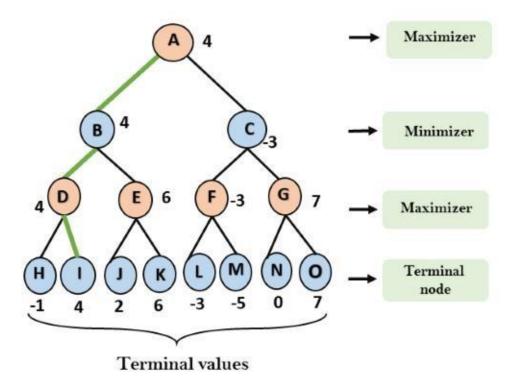
The code for Water jug problem has been executed successfully.

#### **EX.NO: 4**

#### **DATE:20-09-24**

#### **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.



#### **CODE:**

import math

```
def minimax(depth, node_index, is_maximizer, scores, height):
if depth == height:
return scores[node_index]
if is_maximizer:
return max(minimax(depth + 1, node_index * 2, False, scores, height),
minimax(depth + 1, node_index * 2 + 1, False, scores, height))
else:
return min(minimax(depth + 1, node_index * 2, True, scores, height))
else:
return min(minimax(depth + 1, node_index * 2 + 1, True, scores, height))
def calculate_tree_height(num_leaves):
return math.ceil(math.log2(num_leaves))
scores = list(map(int, input("Enter the scores separated by spaces: ").split()))
tree_height = calculate_tree_height(len(scores))
optimal_score = minimax(0, 0, True, scores, tree_height)
print(f"The optimal score is: {optimal_score}")
```

#### **OUTPUT:**

```
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                                                           0
    def minimax(depth, node_index, is_maximizer, scores, neignt):
        if depth == height:
            return scores[node_index]
        if is_maximizer:
            return max(minimax(depth + 1, node_index * 2, False, scores, height),
                      minimax(depth + 1, node_index * 2 + 1, False, scores, height)
            return min(minimax(depth + 1, node_index * 2, True, scores, height),
                      minimax(depth + 1, node_index * 2 + 1, True, scores, height))
    def calculate_tree_height(num_leaves):
        return math.ceil(math.log2(num_leaves))
    scores = list(map(int, input("Enter the scores separated by spaces: ").split()))
    tree_height = calculate_tree_height(len(scores))
    optimal_score = minimax(0, 0, True, scores, tree_height)
    print(f"The optimal score is: {optimal_score}")
→ Enter the scores separated by spaces: -1 4 2 6 -3 -5 0 7
    The optimal score is: 4
                                                      27s completed at 11:04 AM
```

#### **RESULT:**

The code for DFS has successfully been executed.

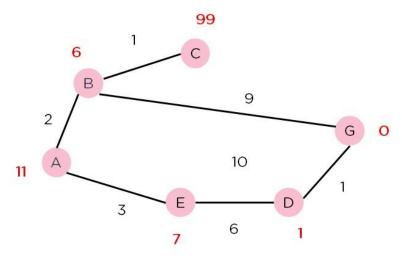
#### EX.No: 5 DATE: 27-09-24

#### **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  $h(n) = \frac{1}{2} \ln t$  heuristic approximation of the node's value. This is not a real value but an approximation cost.



CODE:

import heapq

class Node:

def \_\_init\_\_(self, name, parent=None, g=0, h=0):
 self.name = name
 self.parent = parent

```
self.g = g
    self.h = h
    self.f = g + h
  def lt (self, other):
    return self.f < other.f
def a_star(graph, start, goal, h_func):
  open_list = []
  heapq.heappush(open list, Node(start, None, 0, h func(start, goal)))
  closed list = set()
  while open list:
    current node = heapq.heappop(open list)
    if current_node.name == goal:
       path = []
       while current node:
         path.append(current node.name)
         current node = current node.parent
       return path[::-1]
    closed list.add(current node.name)
```

```
for neighbor, cost in graph[current node.name]:
       if neighbor in closed list:
          continue
       g_new = current_node.g + cost
       h_new = h_func(neighbor, goal)
       f_{new} = g_{new} + h_{new}
       neighbor node = Node(neighbor, current node, g new, h new)
       heapq.heappush(open list, neighbor node)
  return None
graph = {
  'A': [('D', 1), ('C', 3)],
  'B': [('A', 1), ('D', 1), ('E', 3)],
  'C': [('A', 3), ('F', 2)],
  'D': [('B', 1)],
  'E': [('B', 3), ('F', 1)],
  'F': [('C', 2), ('E', 1)]
}
def heuristic(node, goal):
```

```
return 0

start_node = 'D'
goal_node = 'F'

path = a_star(graph, start_node, goal_node, heuristic)

if path:
    print(f'Path found: {path}")

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

```
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path = a_star(graph, start_node, goal_node, heuristic)

if path:
    print(f"Path found: {path}")

else:
    print("No path found")

Path found: ['D', 'B', 'E', 'F']
```

#### **RESULT:**

The code for A\* algorithm has successfully executed.

EX.NO :6

#### **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: -

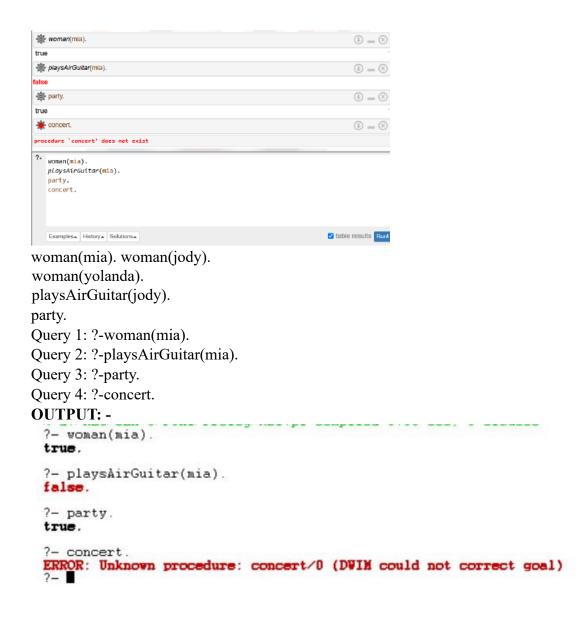
A rule consists of a head (a predicate) and a body (a sequence of predicates separated by commas).

Ex:

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

#### **SOURCE CODE:**

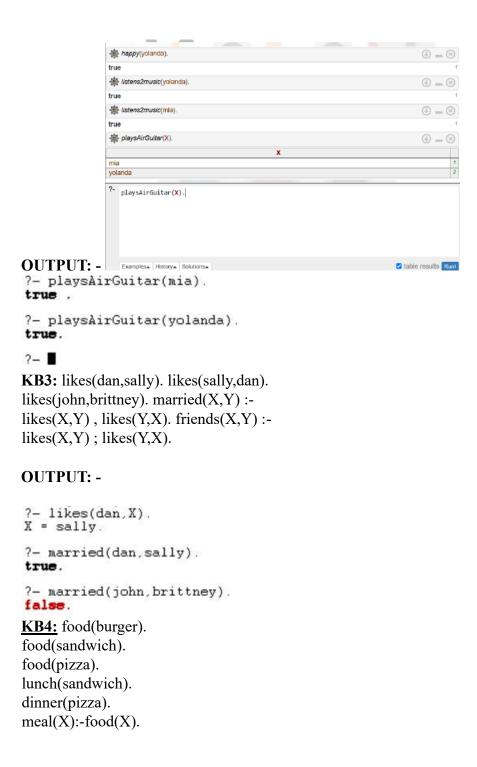
**KB1**:



#### **KB2**:

happy(yolanda). listens2music(mia).

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

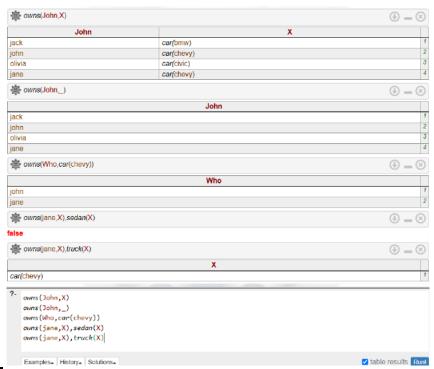


#### **OUTPUT:**

```
food(pizza).
?- 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)).



**OUTPUT:** 

```
?-
| 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).
```

#### **RESULT:**

The prolog has been successfully executed and verified.

#### EX.NO :7 DATE: <u>07-10-24</u>

#### **PROLOG- FAMILY TREE**

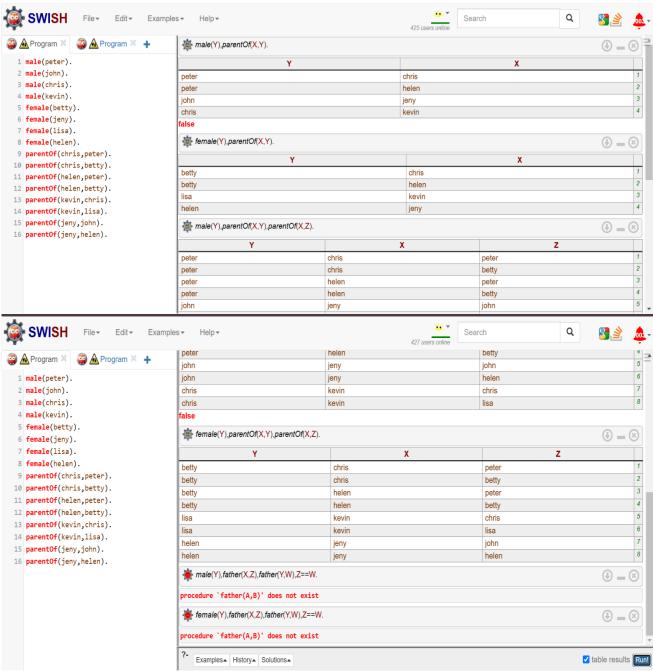
#### AIM:

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.
```

#### **OUTPUT:**



#### **RESULT:**

The prolog family-tree has been implemented successfully and verified.

**EX.NO**: 8 **DATE**: 18-11-24

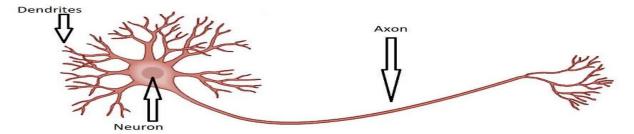
# IMPLEMENTING ARTIFICIAL NEURAL NETWORKS FOR AN APPLICATION USING PYTHON - CLASSIFICATION AIM

<u>:</u>

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

#### What is an Artificial Neural Network?

Artificial Neural Network is much similar to the human brain. The human Brain consist of **neurons**. These neurons are connected. In the human brain, neuron looks something like this...



As you can see in this image, There are **neurons**, **Dendrites**, **and axons**.

#### What do you think?

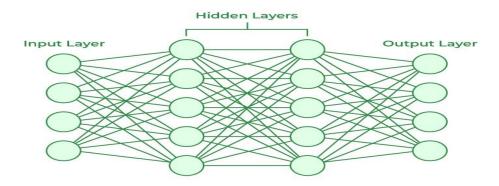
When you touch the hot surface, how you suddenly remove your hand? This is the procedure that happens inside you. When you touch some hot surface. Then automatically your skin sends a signal to the neuron. And then the neuron takes a decision, "Remove your hand". So that's all about the Human Brain. In the same way, Artificial Neural Network works.

#### **Artificial Neural Networks**

Artificial Neural Networks contain artificial neurons which are called **units**. These units are arranged in a series of layers that together constitute the whole Artificial Neural Network in a system. A layer can have only a dozen units or millions of units as this depends on how the complex neural networks will be required to learn the hidden patterns in the dataset. Commonly, Artificial Neural Network has an input layer, an output layer as well as hidden layers. The input layer receives data from the outside world which the neural network needs to analyze or learn about. Then this data passes through one or multiple hidden layers that transform the input into data that is valuable for the output layer. Finally, the output layer provides an output in the form of a response of the Artificial Neural Networks to input data provided.

The structures and operations of human neurons serve as the basis for artificial neural networks. It is also known as neural networks or neural nets. The input layer of an artificial neural network is the first layer, and it receives input from external sources and releases it to the hidden layer, which is the second layer. In the hidden layer, each neuron receives input from the previous layer neurons, computes the weighted sum, and sends it to the neurons in the next layer. These connections are weighted means effects of the inputs

from the previous layer are optimized more or less by assigning different-different weights to each input and it is adjusted during the training process by optimizing these weights for improved model performance.

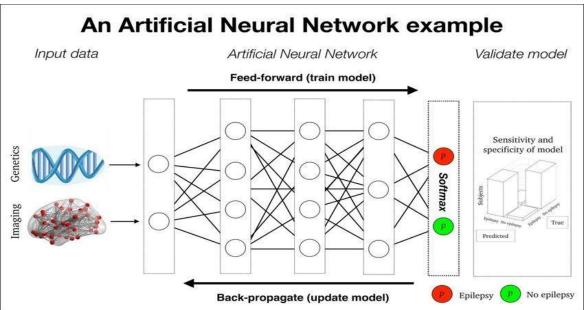


#### What are the types of Artificial Neural Networks?

- 1. Feedforward Neural Network
- 2. Convolutional Neural Network
- 3. Modular Neural Network
- 4. Radial basis function Neural Network
- 5. Recurrent Neural Network

#### **Applications of Artificial Neural Networks**

- 1. Social Media
- 2. Marketing and Sales
- 3. Healthcare
- 4. Personal Assistants



**CODE:** 

```
import numpy as np
from keras.models import Sequential
from keras.layers import Dense
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
from sklearn.datasets import make regression
X, y = \text{make regression}(n \text{ samples}=1000, n \text{ features}=5, \text{noise}=0.1)
X train, X test, y train, y test = train test split(X, y, test size=0.2,
random state=42)
scaler = StandardScaler()
X train = scaler.fit transform(X train)
X \text{ test} = \text{scaler.transform}(X \text{ test})
model = Sequential()
model.add(Dense(units=64, activation='relu', input dim=X train.shape[1]))
model.add(Dense(units=32, activation='relu'))
model.add(Dense(units=1, activation='linear'))
model.compile(optimizer='adam', loss='mean squared error')
model.fit(X train, y train, epochs=5, batch size=32, verbose=1)
loss = model.evaluate(X test, y test)
print(f"Model Loss: {loss}")
y pred = model.predict(X test)
print("Predictions for the first 5 test samples:", y pred[:5])
OUTPUT
 Epoch 1/5
                         - 6s 10ms/step - loss: 0.5633 - mae: 0.5609 - val loss: 0.0651 - val mae: 0.1992
160/160
 Epoch 2/5
                        - 1s 4ms/step - loss: 0.0412 - mae: 0.1588 - val loss: 0.0217 - val mae: 0.1176
 160/160
 Epoch 3/5
                         - 1s 3ms/step - loss: 0.0146 - mae: 0.0964 - val loss: 0.0175 - val mae: 0.1027
160/160 -
 Epoch 4/5
                        - 0s 2ms/step - loss: 0.0093 - mae: 0.0788 - val_loss: 0.0153 - val_mae: 0.0945
 160/160
 Epoch 5/5
                        - 1s 2ms/step - loss: 0.0068 - mae: 0.0660 - val loss: 0.0101 - val mae: 0.0759
 Mean Absolute Error on Test Set: 0.08500416576862335
                    — 0s 8ms/step
Predicted values: [ 35.185875 72.94722 16.496237 -307.4313
                                                            29.497248]
Actual values: [ 42.67137813 75.01408257 -4.05539077 -295.72163432 44.43243324]
```

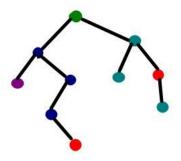
#### **RESULT:**

The Code for ANN is successfully executed and verified.

EX.NO :9 DATE : 25-10-24

#### 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.

#### **CODE:**

```
from sklearn import tree X = [[150, 50, 37], [160, 60, 38], [170, 70, 39], [180, 80, 40], [165, 55, 36]] Y = [0, 0, 1, 1, 0] clf = tree.DecisionTreeClassifier() clf = clf.fit(X, Y) prediction = clf.predict([[175, 75, 41]]) print("Predicted Gender (0 = Female, 1 = Male):", prediction[0])
```

#### **OUTPUT:**

The predicted gender for the input [170, 63, 41] is: Male

#### **RESULT:**

The decision tree classification technique for gender classification using python has been executed successfully.

<u>EX NO :10</u> <u>DATE :</u> 1-11-24

#### <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:

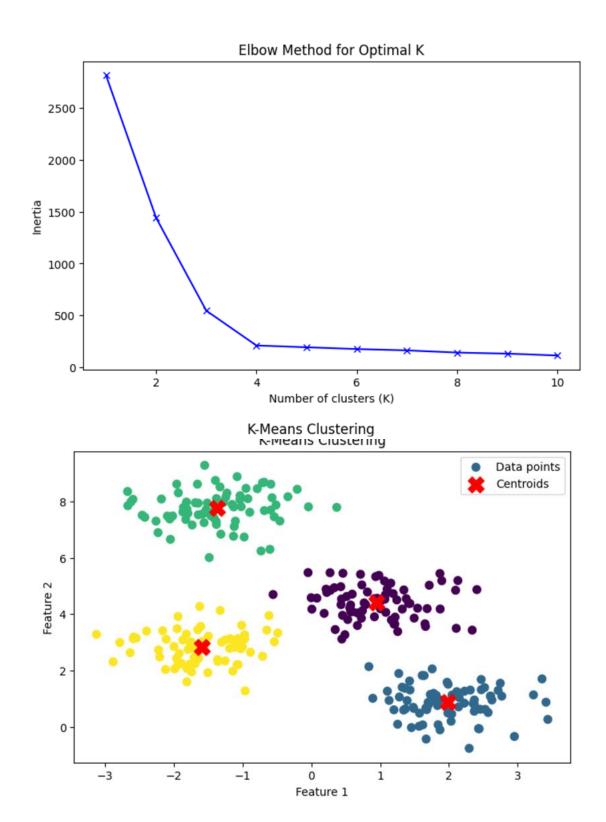
To implement a K - Means clustering technique using python language.

#### **EXPLANATION:**

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

#### **CODE:**

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.datasets import make blobs
X, = make blobs(n samples=300, centers=4, random state=42)
plt.scatter(X[:, 0], X[:, 1], s=30)
plt.title("Generated Data Points")
plt.show()
kmeans = KMeans(n clusters=4, random state=42)
kmeans.fit(X)
centroids = kmeans.cluster centers
labels = kmeans.labels
plt.scatter(X[:, 0], X[:, 1], c=labels, cmap='viridis', s=30)
plt.scatter(centroids[:, 0], centroids[:, 1], c='red', marker='X', s=200,
label='Centroids')
plt.title("K-Means Clustering")
plt.legend()
plt.show()
print(end="\n")
print("Centroids of the clusters:")
print(centroids,end="\n\n")
inertia = []
for k in range(1, 11):
kmeans = KMeans(n clusters=k, random state=42)
kmeans.fit(X)
inertia.append(kmeans.inertia)
plt.plot(range(1, 11), inertia, marker='o')
plt.title("Elbow Method to find optimal K")
plt.xlabel("Number of clusters (K)")
plt.ylabel("Inertia")
plt.show()
OUTPUT:
```



**RESULT:** Thus, we have successfully implemented a K-Means clustering technique using python language.