EXP.NO:1

BASIC PYTHON PROGRAMS

1) Factorial

```
def factorial (n):
if A==0
    return 1
else :
    return n * factorial (n-1)
num = int (input ("Enter a number : "))
print ("Factorial of ", num, " is ",
                    factorial (num))
```

Output

```
Enter a number : 5
Factorial of 5 is 12
```

2) Sum of digits

```
def sum_digits (n):
    sum = 0
    while (n != 0):
        sum = sum + int (n % 10)
        n = int (n / 10)
    return sum
num = int (input ("Enter a number : "))
print (sum_digits (num))
```

Output

Enter a number: 567

3) Palindrome

```
def is palindrome (s):
    return s == [::-1]
word = input ("Enter a string: ")
ans = is palindrome (word)
if ans:
    print ("Yes")
else
    print ("No")
```

4) Armstrong number

```
def is_armstrong_number (number):
    sum = 0
    while (number > 0):
        digit = number % 10
        sum += digit * 3
        number = 10
    return sum
number = int (input ())
b = number
if is_armstrong_number (number) =
    print ("Yes")
```

```
else:
    print ("NO")
```

Output

23

5) Swapping of two numbers

```
def swap(a, b)
    a = a - b
    b = b + a
    a = b - a
    return a, b
a = int (input())
b = int (input())
print (swap(a, b)
```

Output 10, 20

20, 10

6) Odd or even

```
num = int (input())
if num % 2 == 0:
    print ("Even")
else
    print ("odd")
```

output

7

odd

7) Area of a rectangle

```
l= int(input())
b= int(input())
print(l*b)
```

output

10

20

200

8) prime or not

```
def is_prime(n):
    if n<=1:
        print False
    for i in range(2, int(n**0.5)+1):
        if n%i==0:
            return False
    return True
n= int(input())
print(is_prime(n))
```

Output:

7

true

9) Fibonacci Series

```
def fib(n):
    a, b = 0, 1
    for i in range(n):
        print(a, end = ' ')
        a, b = b, a+b
n = int(input())
print(fib(n))
```

Output :

4

0 1 1 2

10) Greatest common Division

```
def gcd(a,b):
    while b:
        a, b = b, a % b
    return a
n = int(input())
m = int(input())
print(gcd(n, m))
```

31/01/2024

Output

16.

Predictive analysis of student final Performance.

~~Domain~~ Domain Type:

Educational Data Mining and Learning Analytics

Problem Statement

Predictive Modeling of Student's final performance

Algorithm:-

~~1. Statistical Models~~

* Decision trees:

Model decision making processes by splitting data based on feature value

* USE CASE: understanding which features (eg attendance, participation

most influence final grades

1) Criteria for splitting

* For classification
    1) Gini impurity
    2) Entropy
* For Regression

2) Splitting Process:

1) Choose a Feature:

Select the feature that provides the best split according to the criterion

2) Split Data

Divide the data into subset based on the chosen features

3) Repeat.

Objective:

Develop and implement a decision tree model to predict Students' final performance in a specific course based on historical academic data, behavioral metrics and personal Charecteristics. The aim is to accurately forecast students' final grade or performance outcomes.

People benefit by this project

1) students
2) Educators
3) Academic Institutions.
4) Parents and Guardians.

Ex.no: 3

Date. 4/9/24

N QUEENS

Aim:

To implement a N queens problem.

Program:

```
def Solve nqueens(n):

    def is_safe(row, col, board):
        for i in range(row):
            if board[i] == col:
                return false
        for r, c in enumerate(board[:row]):
            if abs(row-i) == abs(col-c):
                Return False
        Return true

    def Solve_recursive(row, board):
        if row == n:
            return [board[:]]
        solution = []
```

```
for col in range(n):
    if is_safe(row, col, board):
        board.append(col)
        solution.extend(solve_recursive
                        (row+1, board))
        board.pop()
    return solution
solution = solve_recursive(0, [])
return solution
n=8
solution = solve_nqueens(n)
if solution:
    print(f"Found {len(solution)}
          solution for the {n}-
          Queens problem:")
    for solution in solution:
        print(solution)
else:
    print(f"No solution found
          for the {n}-queens
          problem.")
```

Output:

```
0 0 1 0
1 0 0 0
0 0 0 1
0 1 0 0
```

True

Result:

Thus the program was executed successfully & out is verified

Ex no:

Date.

Depth First Search

Ex

Aim:

To Implement a depth-first search problem using python

Program:

```
def dfs(graph, start_node):

    visited = set()
    stack = [start_node]
    traversal_order = []

    while stack:
        node = stack.pop()
        if node not in visited:
            visited.add(node)
            traversal_order.append(node)
            neighbors = graph.get
                (node, [])
            stack.extend([neighbor for
                neighbor in reversed(neighbors)
                if neighbor not in
                visited])
```

```
return traversal_order

graph = {
    'A' : ['B', 'C'],
    'B' : ['D', 'E']
    'C' : ['F']
    'D' : [],
    'E' : ['F'],
    'F' : []

start_node = 'A'
dfs_order = dfs (graph, start_node)
print ("DFS transversal order :",
                    dfs_order)
```

Output:

1 2 0 3 4

Result:

Thus the program was executed sucessfully & output was verified

Date
Exno: 11/9/24
Exno: 5

A* algorithm

Aim: To implement a A* algorithm using python

Program:

```
import heapq

def __init__(self, state, parent=None, cost=0,
             heuristic=0):
    self.state = state
    self.parent = parent
    self.cost = cost
    self.heuristic = heuristic
    self.f_value = cost + heuristic

def __lt__(self, other):
    return self.f_value < other.f_value

def a_star(graph, start_node, goal_node,
           heuristic_func):
```

```
open_list = [Node(start_node, cost=0,
                  heuristic = heuristic_func
                              (start_node))]

closed_list = set()

while open_list:
    current_node = heapq.heappop
                   (open_list)
    if current_node.state == goal_node:
        path = []
        while current_node:
            path.insert(0, current_node.state)
            current_node = current_node.
                           parent
        return path
    closed_list.add(current_node.
                    state)
    for neighbor, cost in graph[current_
                                node.state]
                                .items():
        if neighbor not in closed_list:
            new_cost = current_node.cost +
                       cost
```

```
new_heuristic = heuristic_func
                (neighbor),
                current_node, new_cost, new_
                heuristic)

in_open_list = False

for open_node in open_list:
    if OPEN_node.state == neighbor
        and open_cost <= new_
        cost

        in_open_list = True
        break

if not in_open_list:
    heapq.heappush(open_list,
                   new_node)

return none
```

```
graph = {
    'A' : {'B' : 1, 'C' : 3},
    'B' : {'D' : 2, 'E' : 4},
    'C' : {'F' : 5},
    'D' : {'G' : 1}
```

```
'E' : {'G': 5},
'F' : {'G': 2},
'G' : {}
}

start_node = 'A'
goal_node = 'G'

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

if shortest_path
    print("shortest path",
          shortest_path)
else
    print("no path found")
```

output:

Path found: ['A', 'E', 'D', 'G']

Result:

Thus the program was executed successfully & output was verified

Ex no: 18/9/24 Date: 18/9/24

10

Aim: To Implement a A*O* algorithm using Python

```
class node
    def -int- (self -name value)
        Self . name = name
        Self. value = value
        Self = child = end()
    def add - child (self node)
        Self. children (node)
    def - ao - sba (root, gaal)
        queue = ((root, (root)))
    while queue:
    node. path = queue pop (0)
        if node = = goal
    for child in node. children
        new - path = path + (node)
            queue . append ((child))
    return new path
```

Input

A = Node ("A", 0)

B = Node ('B', 1)

C = Node ('C', 2)

D = Node ('D', 3)

E = Node ('E', 4)

Output

solution path ('A', 'B', 'D', 'E')

RESULT:

Thus the above program has been executed succesfully

1.

Ex no 7

Date 25/9/24

DECISION TREE classification

Aim: To Implement a Decision tree.

Program

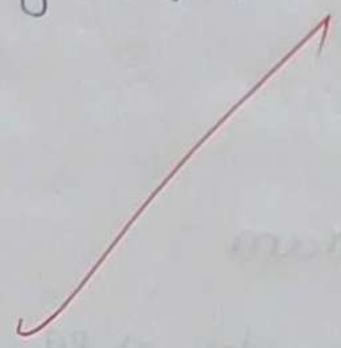
```
import pandas as pd
from sklearn.tree import DecisionTreeClassifier

data = {
    'Height': [152, 155, 172, 185, 167, 180, 157, 180, 164, 177]
    'Weight': [45, 57, 72, 85, 68, 78, 22, 90, 66, 88]
    'Gender': ['Female', 'Female', 'Male', 'Male', 'Female',
               'Male', 'Female', 'Male', 'Female', 'Male']
}
df = pd.DataFrame(data)
x = df[['Height', 'Weight']]
y = df['Gender']

classifier = DecisionTreeClassifier()
classifier.fit(x, y)

height = float(input("Enter height (in cm) for prediction: "))
weight = float(input("Enter weight (in kg) for prediction: "))
```

```
random_values = pd.DataFrame([[height, weight]],
                columns = ["Height", "Weight"])
predicted_gender = classifier.predict(random_values)
print(f"predicted gender for height {height} cm
      and weight} kg: {predicted_gender[0]}")
```



Output

Enter height for prediction : 169
Enter weight for prediction : 61
Predicted gender for height 169.0 cm and weight 61.0 kg: Female.

Result:

The above program was executed and implement successfully.

Ex: no. 9

Date 16/10/24

Implementing Artificial Neural network for an application using python - classification.

Aim:

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

Program.

```
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import standard scaler
from keras.models import sequential
from keras.layers import Dense
from keras.optimizers import Adam
import matplotlib.pyplot as plt

# Generating synthetic dataset for demonstration
# Replace this with your own dataset
np.random.seed(42)
X = np.random.rand(1000, 3) # 1000 samples, 3 features
Y = 5 * X[:, 0] + 2 * X[:, 1] ** 2 + 1.5 * np.sin(X[:, 2] *
    np.pi) + np.random.randn(
    (0, 0.1, 1000) # Non-linear relationship.
```

Split the 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)

Feature Scaling

Scaler = Standard Scaler()

X train = scale.fit transform (X train)

Y test = scaler. transform (X_test)

Building the ANN Model

input layer and first hidden layer with 10 neurons and ReLu activation model. add (Dense (10, input dim = X train. Shape [1], activation = relu))

Second hidden layer with 10 neurons and ReLu activation model. add (Dense (10, activation = relu))

output layer with linear activation for regression
model. add (Dense (1., activation = 'linear'))

Compiling the model

model. Compile (optimizer = Adam (learning_rate = 0.01)
loss = mean_squared_error)

Training the model

history = model. fit (X train, Y train, epochs = 100
batch_size = 32 validation_split = 0.2, verbose = 1)

Making predictions

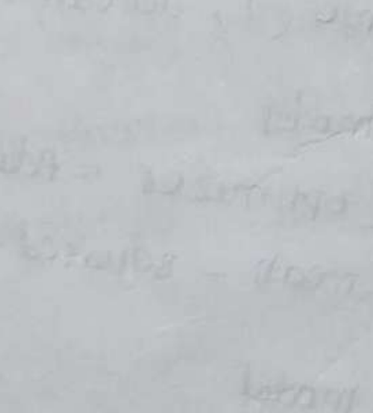
Y_prod = model. predict (X_test)

Evaluating the model

```
mse = np.mean((y_test - y_pred.flatten())**2)
print(f"mean Squared Error: {mse:.4f}")
```

Plotting training history

```
plt.figure(figsize=(12,6))
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



Out put

Epoch 1/100

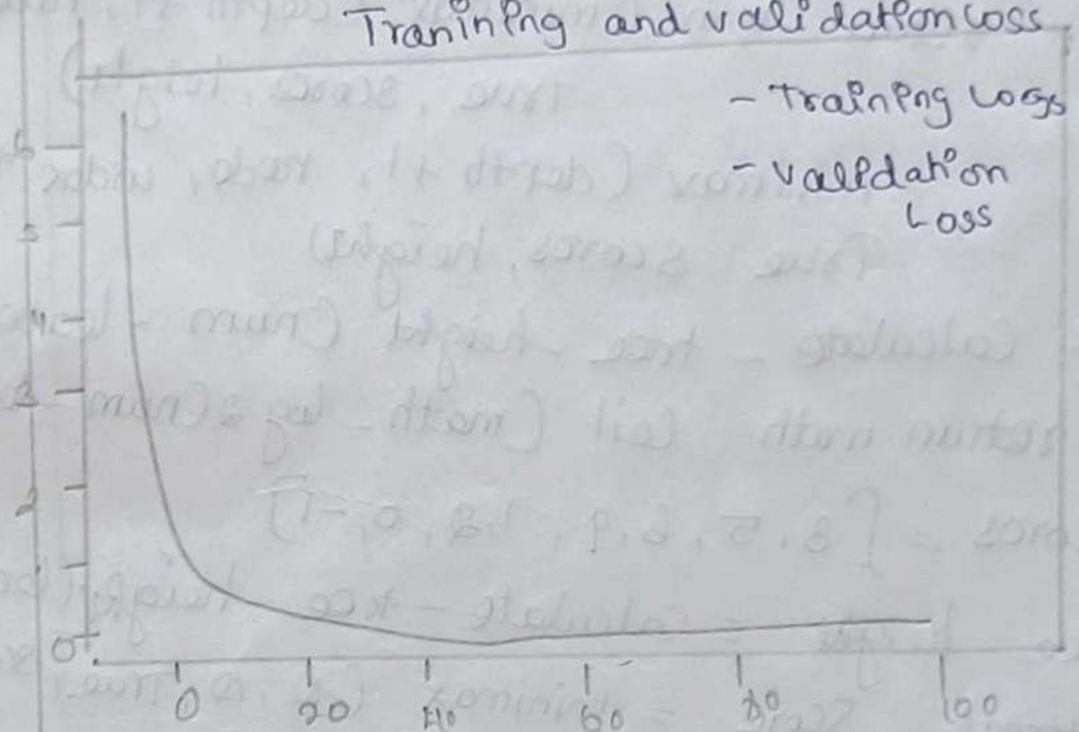
20/20 ———— 20 19ms/st - loss: 8.847

- val_loss : 1.9373

Epoch 2/100

20/20 ———— 0s 4ms/step - loss 1.5651

- val_loss : 0.6565



Output:

The above ANN program was successfully executed and Implemented.

Aim: To write a program to implementation of Alpha-Beta pruning technique in minimax algorithm

```
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)
               minimax (depth+1, node_index*2+1,
                True, scores, height)
def calculate_tree_height (num_leaves)
    return math.ceil (math.log2 (num_leaves)
scores = [3, 5, 6, 9, 1, 2, 0, -1]
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:

The optimal score is = 5

Output: The optimal score is: 5

Result:

The above program was successfully implement and executed.

Ex no: 11

Date: 9/10

Implementation of Clustering Technique K-Means

Aim:

To implement a k-Means clustering technique using Python language.

```
Import numpy as np
import mat plotlib.py plot as pit
from skiearn. cluster import k means
from sklern. data sets import make_blobs
```

Step 1: import K means from sklearn.cluster (already done above)

Step 2: Generate synthetic data and assign x and y

Creating a data set with 3 clusters

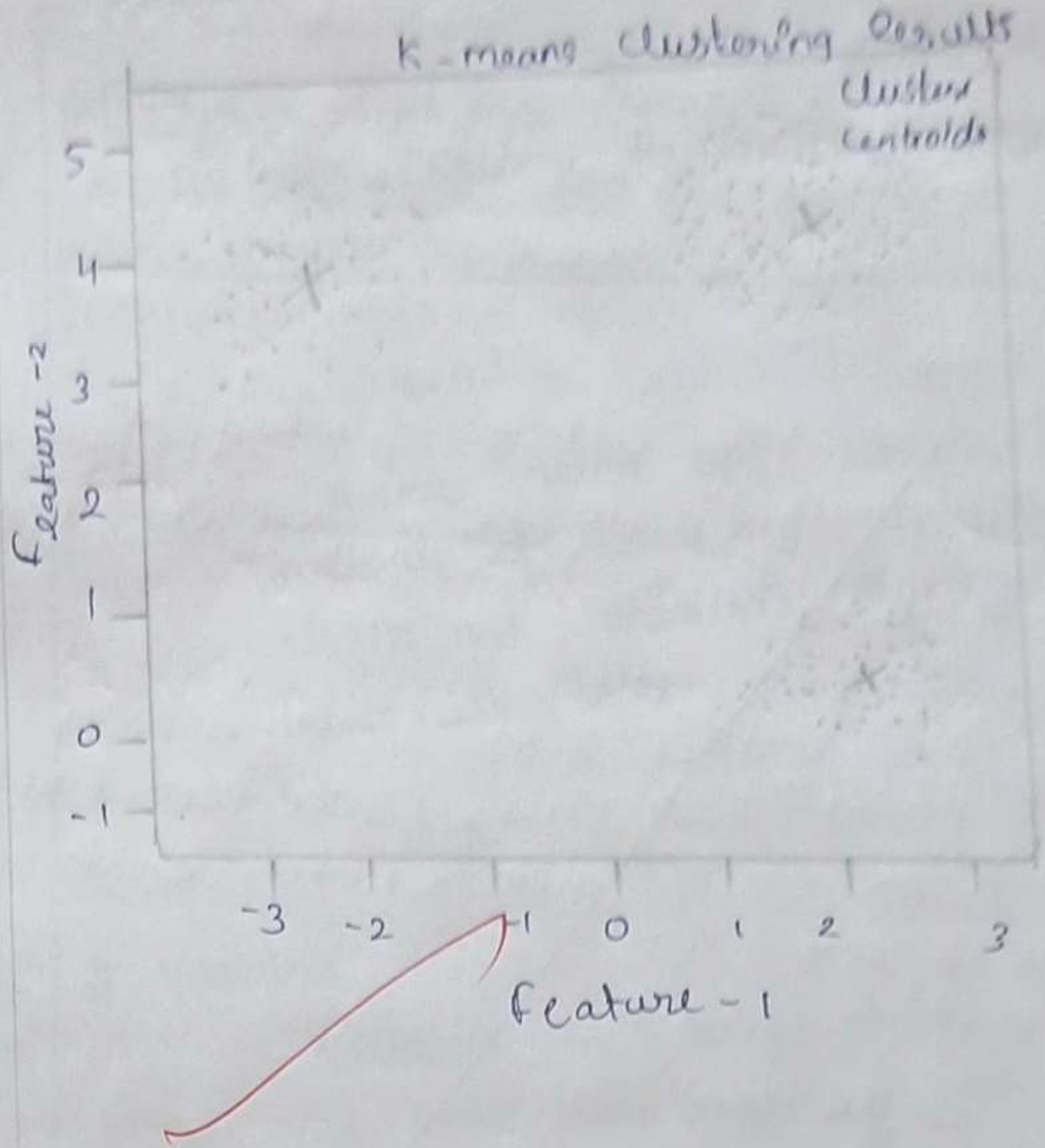
```
x, y true = make_blobs (n snaples = 300 centres
                        cluster_std = 0.60 random_state = 0)
```

Step3: Call the function k-means()
fit k-means with the chosen number of cluster (k=3)

```
K=3
k means = k means (n_cluster = k, random_State...
y-k means = k means - fit predict (x)
```

Step 4: perform scatter operation and display the output

```
plt.figure (fig size = (8,6))
ptt.scatter (x[:0], x[:,1], c=y_k means, s=30
    c=map = 'virdis', label = ('clusters')
Centers = k means. cluster_Centers
pts.scatter. (Centers[:,0], Centers[:,1]
    c='red', s=200, alpha=0.75, marker='x'
label = ("centroids')
plt.title (k-means clustering Results)
plt.xlabel ('feature')
plt.ylabel ('Feature 2')
plt.legend ()
plt.show ()
```



Result:

The above program was executed and implement successfully.

Ex 14 Data Scientist

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 underscore, starting with a lower case letter.

Ex

dog
ab_c_391

2. Variables:

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

Ex:

Dog

Apple

3. Compound Terms:

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

Ex. is bigger (elephant, x)

f(g(x)).

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 small (X, Y) :- is_bigger (Y, X)
aunt (Aunt, child) :- sister (Aunt, Parent)
parent (Parent, child)

SOURCE CODE:

KB1

woman (mia)
woman (jody)
woman (yolanda)

plays Air Guitar (jody)
Party
Query 1 : ? - woman (mia)
Query 2 : ? - plays Air Guitar (mia)
Query 3 : ? - party
Query 4 : ? - Concert

OUTPUT:

? = woman (mia)
true

? - Plays Air Guitar (mia)
false
? party
true
? - Concert
ERROR : Unknown procedure: Concert/0 (DWM, Could not find Correct goal)
? -

KB2:

happy (yolanda)
Listen 2 music (mia)
Listens 2 music (yolanda) - happy (yolanda)
Plays Air Guitar (mia) :- Listens 2 music (mia)
plays Air Guitar (yolanda) : Listens 2 Music (yolanda)

OUTPUT:

? - Plays Air Guitar (mia)
true
? - plays Air Guitar (yolanda)
true.
? -

KB3:

Likes (dan, sally)
Likes (sally, dan)
Likes (John, brittney)
married (X, X) :- Likes (X, X), Likes (Y, X)
friends (X, X) :- Likes (X, X) : Likes (Y, X)

OUTPUT :-

? - likes (dan . x)

x = Sally

? - married (dan, sally)

true

? - married (john, brithay)

false

KB4

food (burger)
food (sandwich)
food (pizza)
lunch (sandwich)
dinner (pizza)
meal (x) :- food (x)

OUTPUT

? -
1 food (pizza)
true
? - real (x) lunch (x)
x - sandwich
? - dinner (sandwich)
false

KBS

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 :

To learn prolog terminologies and write basic programs.

Exno: 12 Date: 6/11/24.

PROLOG - FAMILY TREE

AIM:

To develop a family tree program using PROLOG with all possib 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)

Parent of (chris , peter)
Parent of (chris , betty)
parent of (helen , peter)
parent of (helen , betty)
parent of (kevin , chris)
parent of (kevin , lisa)
Parent of (jeny , john)
Parent of (jeny , helen)
```

/RULES::

/ son . parent

son, grandparent/

father (x,y) :- male (y), parent of (x,y)

mother (x,y) :- female (y), parent of (x,y)

grand father (x,y) :- male y, parent of (x,z),
parent of (z,y)

grandmother (x,y) :- female (y), parent of (x,z)
parent of (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)
=w

Result:

The program was executed and output was verified.

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Date
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NAME: Akilesh. B STD: CSE SEC: A ROLL NO: 021 SUB POAI

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completed