13. Write a program to construct a Bayesian network considering medical data. Use this model to demonstrate the diagnosis of heart patients using standard Heart Disease Data Set. import numpy as np import pandas as pd from pgmpy.models import BayesianNetwork from pgmpy.models import BayesianModel from pgmpy.estimators import MaximumLikelihoodEstimator from pgmpy.inference import VariableElimination # Read Cleveland Heart Disease data heartDisease = pd.read_csv('C:/Users/kamle/Downloads/ML/prac8/heart.csv') heartDisease = heartDisease.replace('?', np.nan) # Display the data print('Few examples from the dataset are given below') print(heartDisease.head()) # Model Bayesian Network model = BayesianModel([('age', 'trestbps'), ('age', 'fbs'),('sex', 'trestbps'), ('exang', 'trestbps'), ('trestbps', 'heartdisease'),('fbs', 'heartdisease'), ('heartdisease', 'restecg'),('heartdisease', 'thalach'), ('heartdisease', 'chol')]) #estimator = MaximumLikelihoodEstimator(model, data) # Learning CPDs using Maximum Likelihood Estimators print('\nLearning CPD using Maximum likelihood estimators') model.fit(heartDisease, estimator=MaximumLikelihoodEstimator) # Inferencing with Bayesian Network print('\nInferencing with Bayesian Network:') HeartDisease infer = VariableElimination(model) # Computing the Probability of HeartDisease given Age print('\n1. Probability of HeartDisease given Age=28')

```
q = HeartDisease_infer.query(variables=['heartdisease'], evidence={'age': 28})
print(q['heartdisease'])
# Computing the Probability of HeartDisease given cholesterol
print('\n2. Probability of HeartDisease given cholesterol=100')
q = HeartDisease_infer.query(variables=['heartdisease'], evidence={'chol': 100})
print(q['heartdisease'])
```

```
Analysis (PCA)
import numpy as np
import matplotlib.pyplot as plt
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.decomposition import PCA
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
# Load the Iris dataset
iris = datasets.load_iris()
X = iris.data
y = iris.target
# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
# Apply PCA to reduce the dimensionality of the dataset
pca = PCA(n_components=2)
X_train_pca = pca.fit_transform(X_train)
X_test_pca = pca.transform(X_test)
# Train an SVM classifier on the reduced dataset
svm = SVC(kernel='linear')
svm.fit(X_train_pca, y_train)
# Make predictions on the test set
y_pred = svm.predict(X_t
```

14. Write a program to implement Support Vector Machines (SVM) and Principal Component

```
15. Write a Program to implement Principle Component Analysis.
import numpy as np
from sklearn.decomposition import PCA
import matplotlib.pyplot as plt
# Sample data: 5 samples with 3 features
data = np.array([
  [2.5, 2.4, 1.5],
  [0.5, 0.7, 0.8],
  [2.2, 2.9, 2.1],
  [1.9, 2.2, 1.8],
  [3.1, 3.0, 2.9]
])
# Standardize the data (mean = 0, variance = 1)
data_mean = np.mean(data, axis=0)
data_std = np.std(data, axis=0)
standardized_data = (data - data_mean) / data_std
# Apply PCA
pca = PCA(n_components=2) # Reduce to 2 dimensions
principal_components = pca.fit_transform(standardized_data)
# Print the principal components
print("Principal Components:\n", principal_components)
# Explained variance
print("Explained Variance Ratio:\n", pca.explained_variance_ratio_)
# Plotting the principal components
plt.scatter(principal_components[:, 0], principal_components[:, 1], c='blue', marker='o')
```

