```
In [1]: import pandas as pd
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
import numpy as np
```

```
In [2]:  #1. Loading and Preprocessing
    #loading dataset
    data = load_breast_cancer()
    X = pd.DataFrame(data.data, columns=data.feature_names)
    y = pd.Series(data.target)
```

```
In [3]:  #Preprocessing steps
# Check for missing values
print(X.isnull().sum())
```

```
mean radius
                            0
mean texture
mean perimeter
                            0
                            0
mean area
                            0
mean smoothness
mean compactness
                            0
mean concavity
                            0
mean concave points
                            0
mean symmetry
                            0
mean fractal dimension
                            0
radius error
                            0
texture error
                            0
perimeter error
                            0
area error
                            0
smoothness error
                            0
                            0
compactness error
                            0
concavity error
concave points error
                            0
symmetry error
                            0
fractal dimension error
                            0
worst radius
                            0
                            0
worst texture
worst perimeter
                            0
worst area
                            0
                            0
worst smoothness
worst compactness
                            0
                            0
worst concavity
worst concave points
                            0
worst symmetry
                            0
worst fractal dimension
                            0
dtype: int64
```

```
MLBreastCancer - Jupyter Notebook
In [4]: ▶ # Feature Scaling
           scaler = StandardScaler()
           X_scaled = scaler.fit_transform(X)
           X train, X test, y train, y test = train test split(X scaled, y, test size
In [5]: | from sklearn.linear model import LogisticRegression
           from sklearn.metrics import accuracy score
           # Logistic Regression
           log reg = LogisticRegression()
           log_reg.fit(X_train, y_train)
           y pred log reg = log reg.predict(X test)
           accuracy_log_reg = accuracy_score(y_test, y_pred_log_reg)
           print(f'Logistic Regression Accuracy: {accuracy_log_reg:.2f}')
           Logistic Regression Accuracy: 0.97
In [6]:
       # Decision Tree Classifier
```

Decision Tree Accuracy: 0.92

Random Forest Accuracy: 0.96

SVM Accuracy: 0.97

k-NN Accuracy: 0.95

```
In [16]:
           accuracies = {
                  'Logistic Regression': accuracy_log_reg,
                  'Decision Tree': accuracy dt,
                  'Random Forest': accuracy_rf,
                  'SVM': accuracy_svm,
                  'k-NN': accuracy_knn
              }
              # Identify the best and worst performing models
              max accuracy = max(accuracies.values())
              min_accuracy = min(accuracies.values())
              best models = [model for model, accuracy in accuracies.items() if accuracy
              worst models = [model for model, accuracy in accuracies.items() if accuracy
              # Print model accuracies
              print("\nModel Accuracies:")
              for model, accuracy in accuracies.items():
                  print(f"{model}: {accuracy:.2f}")
              # Print best and worst models
             print(f"\nBest Model(s): {', '.join(best_models)} with accuracy {max_accur
print(f"Worst Model(s): {', '.join(worst_models)} with accuracy {min_accur
              Model Accuracies:
              Logistic Regression: 0.97
              Decision Tree: 0.92
              Random Forest: 0.96
              SVM: 0.97
              k-NN: 0.95
              Best Model(s): Logistic Regression, SVM with accuracy 0.97
             Worst Model(s): Decision Tree with accuracy 0.92
```

```
In [ ]: ▶
```