1. Loading and Preprocessing

Load the breast cancer dataset from sklearn.

```
In [1]: from sklearn.datasets import load_breast_cancer
       data = load_breast_cancer()
       x = data.data
       y = data.target
       х,у
Out[1]: (array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
                1.189e-01],
               [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
                8.902e-02],
               [1.969e+01, 2.125e+01, 1.300e+02, ..., 2.430e-01, 3.613e-01,
                8.758e-02],
               [1.660e+01, 2.808e+01, 1.083e+02, ..., 1.418e-01, 2.218e-01,
                7.820e-02],
               [2.060e+01, 2.933e+01, 1.401e+02, ..., 2.650e-01, 4.087e-01,
                1.240e-01],
               [7.760e+00, 2.454e+01, 4.792e+01, ..., 0.000e+00, 2.871e-01,
                7.039e-02]]),
        0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0,
               1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0,
               1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
               1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
               0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1,
               1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
               1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0,
               0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
               1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
               1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
               0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0,
               0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0,
               0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0,
               1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1,
               1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1,
               1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
               1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1,
               1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1,
               1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1,
               1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1))
```

Preprocess the data to handle any missing values and perform necessary feature scaling.

```
In [2]: import numpy as np
print(np.isnan(x).sum())
```

Explain the preprocessing steps you performed and justify why they are necessary for this dataset.

```
In [4]: from sklearn.preprocessing import StandardScaler
        scaler = StandardScaler()
        x_scaled = scaler.fit_transform(x)
In [5]: x_scaled
Out[5]: array([[ 1.09706398, -2.07333501,
                                          1.26993369, ..., 2.29607613,
                 2.75062224, 1.93701461],
               [ 1.82982061, -0.35363241,
                                          1.68595471, ..., 1.0870843 ,
                -0.24388967, 0.28118999],
               [1.57988811, 0.45618695, 1.56650313, ..., 1.95500035,
                 1.152255 , 0.20139121],
               [ 0.70228425, 2.0455738 , 0.67267578, ..., 0.41406869,
                -1.10454895, -0.31840916],
               [ 1.83834103, 2.33645719, 1.98252415, ..., 2.28998549,
                 1.91908301, 2.21963528],
               [-1.80840125, 1.22179204, -1.81438851, ..., -1.74506282,
                -0.04813821, -0.75120669]])
```

2. Classification Algorithm Implementation

1. Logistic Regression

2. Decision Tree Classifier

3. Random Forest Classifier

4. Support Vector Machine (SVM)

5. k-Nearest Neighbors (k-NN)

3. Model Comparison

```
In [15]: from sklearn.model_selection import cross_val_score
   models = [log_reg, tree_clf, rf_clf, svm_clf, knn_clf]
   model_names = ['Logistic Regression', 'Decision Tree', 'Random Forest', 'SV
   for model, name in zip(models, model_names):
        scores = cross_val_score(model, x_scaled, y, cv=5, scoring='accuracy')
        print(f"{name}: {scores.mean():.4f}")
```

Logistic Regression: 0.9807

Decision Tree: 0.9138 Random Forest: 0.9631

SVM: 0.9736 k-NN: 0.9649

Which algorithm performed the best and which one performed the worst?

Best Algorithm: The one with the highest average accuracy. Worst Algorithm: The one with the lowest average accuracy.

In []:	