# Lab 6 Dimension Reduction – WBCD

Dataset:

```
In [1]:
import from sklearn.datasets load_breast_cancer
# Load dataset
data = load_breast_cancer()
X = data.data
y = data.target
print("Original Shape:", X.shape)
```

Original Shape: (569, 30)

```
import matplotlib.pyplot as plt
importfrom sklearn.model_selection train_test_split
importfrom sklearn.linear_model LogisticRegression
importfrom sklearn.metrics accuracy_score, precision_score, recall_score,
f1_score
```

```
In [3]:
# define function for train test split and logistic regrression
def LogisticRegression_clasify_eval (X, y, method_name="Original"):
    # Split the data 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)
    # logistic regression
    clf = LogisticRegression(max iter=10000)
    clf.fit(X_train, y_train)
   y pred = clf.predict(X test)
    accuracy = accuracy_score(y_test, y_pred)
    precision = precision_score(y_test, y_pred)
    recall = recall_score(y_test, y_pred)
   f1 = f1_score(y_test, y_pred)
    # print
    print(f"==== Method: {method_name} Evaluation ====")
    print(f"Accuracy: {accuracy}")
    print(f"Precision: {precision}")
    print(f"Recall: {recall}")
    print(f"F1 Score: {f1}")
    print()
    #visualize
    plt.figure(figsize=(8, 6))
    plt.scatter(range(len(y_test)), y_test, label="Actual", alpha=0.7,
color="blue")
    plt.scatter(range(len(y_pred)), y_pred, label="Predicted", alpha=0.7,
color="red", marker = "x")
    plt.title(f"{method_name} Data Visualization")
    plt.xlabel("Sample Index ")
```

```
plt.ylabel("Class")
plt.legend()
plt.show

return {"Method" : method_name, "Accuracy": accuracy, "Precison": precision,
"Recall": recall, "F1": f1}
```

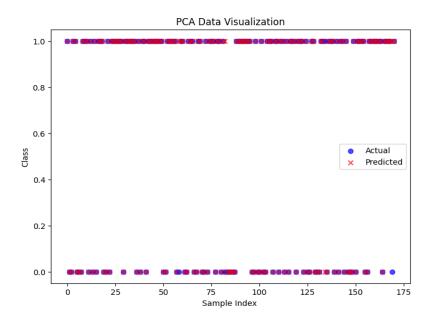
## **Exercise**

#### **PCA**

```
In [4]:
importfrom sklearn.decomposition PCA
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X)

pca_results = LogisticRegression_clasify_eval(X_pca, y, method_name="PCA")
```

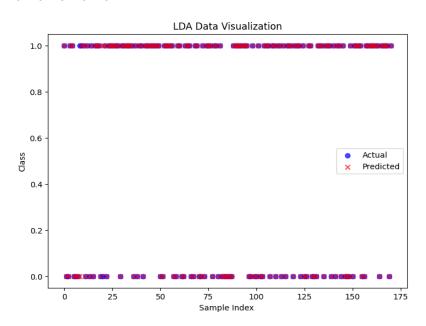
==== Method: PCA Evaluation ==== Accuracy: 0.9766081871345029 Precision: 0.97272727272728 Recall: 0.9907407407407407 F1 Score: 0.981651376146789



## LDA

```
In [5]:
importfrom sklearn.discriminant_analysis LinearDiscriminantAnalysis as LDA
lda = LDA(n_components=1)
X_lda = lda.fit_transform(X, y)
lda_results = LogisticRegression_clasify_eval(X_lda, y, method_name="LDA")
```

==== Method: LDA Evaluation ==== Accuracy: 0.9883040935672515 Precision: 0.9907407407407407 Recall: 0.9907407407407407 F1 Score: 0.9907407407407407



## **T-SNE**

```
In [6]:
importfrom sklearn.manifold TSNE
tsne = TSNE(n_components=2, random_state=42)
X_tsne = tsne.fit_transform(X)

tsne_results = LogisticRegression_clasify_eval(X_tsne, y, method_name="T-SNE")
```

==== Method: T-SNE Evaluation ==== Accuracy: 0.9298245614035088 Precision: 0.9285714285714286 Recall: 0.9629629629629629 F1 Score: 0.945454545454545454

T-SNE Data Visualization 1.0 0.8 0.6 Actual Predicted 0.4 0.2 0.0 25 50 75 100 125 150 175 Sample Index