## homework 3

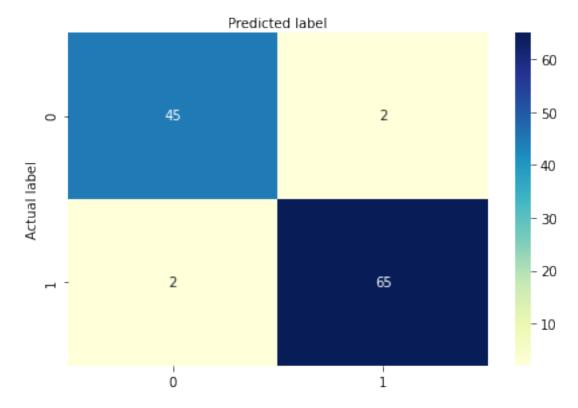
## November 14, 2021

```
[14]: import numpy as np
      import matplotlib.pyplot as plt
      import pandas as pd
      import seaborn as sns
      import warnings
      import matplotlib.pyplot as plt
      from sklearn.datasets import load breast cancer
      from sklearn.model_selection import train_test_split
      from sklearn.preprocessing import StandardScaler
      from sklearn.linear_model import LogisticRegression
      from sklearn.metrics import confusion_matrix
      from sklearn import metrics
      from matplotlib.colors import ListedColormap
      from sklearn.naive_bayes import GaussianNB
      from sklearn.decomposition import PCA
      from sklearn.discriminant_analysis import LinearDiscriminantAnalysis as LDA
[15]: # Importing and cleaning the data
      breast = load_breast_cancer()
      breast_data = breast.data
      breast_input = pd.DataFrame(breast_data)
      breast_labels = breast.target
      labels = np.reshape(breast labels,(569,1))
      final_breast_data = np.concatenate([breast_data,labels],axis=1)
      final breast data.shape
      breast_dataset = pd.DataFrame(final_breast_data)
      features = breast.feature_names
      features_labels = np.append(features, 'label')
      breast_dataset.columns = features_labels
[16]: X = breast_dataset.iloc[:,:30].values
      Y = breast_dataset.iloc[:, 30].values
      # Scaling the features
      sc_X = StandardScaler()
      std_X = sc_X.fit_transform(X)
```

Accuracy: 0.9649122807017544 Precision: 0.9701492537313433 Recall: 0.9701492537313433

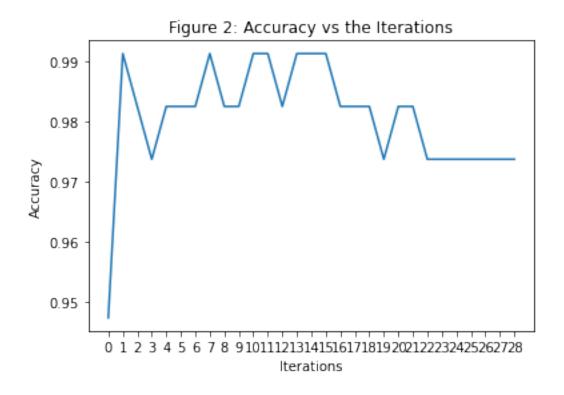
```
[18]: # Confusion matrix
    class_names=[0,1] # name of classes
    fig, ax = plt.subplots()
    tick_marks = np.arange(len(class_names))
    plt.xticks(tick_marks, class_names)
    plt.yticks(tick_marks, class_names)
# create heatmap
sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
ax.xaxis.set_label_position("top")
plt.tight_layout()
plt.title('Figure 1: Confusion matrix for Logistic Regression', y=1.1)
plt.ylabel('Actual label')
plt.xlabel('Predicted label');
```

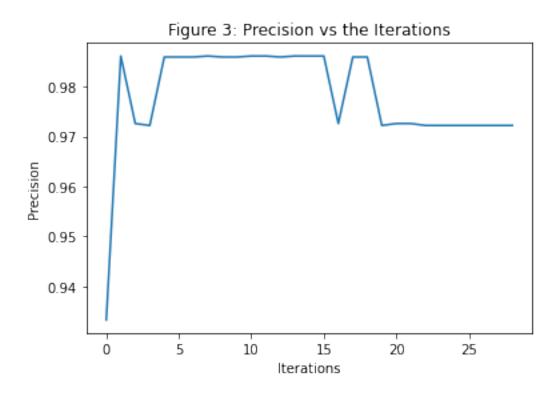


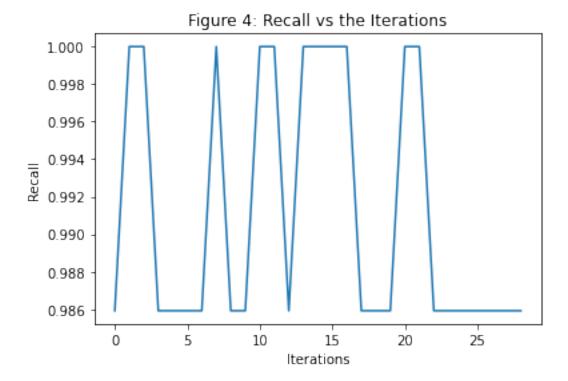


```
[19]: # Problem 2 - Feature extraction using PCA
      # Scaling the features
      accuracy = []
      precision = []
      recall = []
      for i in range(29):
          pca = PCA(n_components=i+1)
          principalComponents = pca.fit_transform(std_X)
          principalDf = pd.DataFrame(data = principalComponents)
          # Creating the logistic model and fitting it
          X_train, X_test, Y_train, Y_test = train_test_split(principalDf, Y,__
       →test_size = 0.20, random_state=42)
          classifier = classifier = LogisticRegression(random_state=42)
          classifier.fit(X_train, Y_train)
          # Creating predictions with the test data
          Y_pred = classifier.predict(X_test)
```

```
accuracy_append(metrics_accuracy_score(Y_test, Y_pred))
    precision.append(metrics.precision_score(Y_test, Y_pred))
    recall.append(metrics.recall_score(Y_test, Y_pred))
# Plotting the accuracy, precision, and recall against the iterations
plt.figure()
plt.plot(range(29),accuracy)
plt.xlabel('Iterations')
plt.ylabel('Accuracy')
plt.title('Figure 2: Accuracy vs the Iterations')
plt.xticks(range(29))
plt.show()
plt.figure()
plt.plot(range(29),precision)
plt.xlabel('Iterations')
plt.ylabel('Precision')
plt.title('Figure 3: Precision vs the Iterations')
plt.show()
plt.figure()
plt.plot(range(29),recall)
plt.xlabel('Iterations')
plt.ylabel('Recall')
plt.title('Figure 4: Recall vs the Iterations')
plt.show()
\#Explain that there is no O value in here and so the graph starts at iteration \sqcup
→ 0 and that is why the the graph does not show 0 on the y axis. You can more ⊔
→easily see the results by not showing iteration 0
```





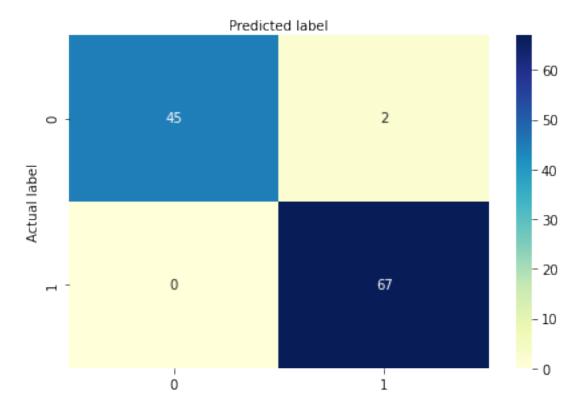


Accuracy: 0.9824561403508771 Precision: 0.9710144927536232

Recall: 1.0

```
[22]: # Confusion matrix
    class_names=[0,1] # name of classes
    fig, ax = plt.subplots()
    tick_marks = np.arange(len(class_names))
    plt.xticks(tick_marks, class_names)
    plt.yticks(tick_marks, class_names)
# create heatmap
sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
ax.xaxis.set_label_position("top")
plt.tight_layout()
plt.title('Figure 5: Confusion matrix for Logistic Regression', y=1.1)
plt.ylabel('Actual label')
plt.xlabel('Predicted label');
```

Figure 5: Confusion matrix for Logistic Regression



```
[23]: # Problem 4 - Feature extraction using LDA and Logistics Regression
lda = LDA(n_components=1)
lda_t = lda.fit_transform(X,Y)
```

[24]: # Splitting the data and fitting the model

Accuracy: 0.9736842105263158 Precision: 0.9705882352941176 Recall: 0.9850746268656716

```
[25]: # Confusion matrix
    class_names=[0,1] # name of classes
    fig, ax = plt.subplots()
    tick_marks = np.arange(len(class_names))
    plt.xticks(tick_marks, class_names)
    plt.yticks(tick_marks, class_names)
    # create heatmap
    sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu",fmt='g')
    ax.xaxis.set_label_position("top")
    plt.tight_layout()
    plt.title('Figure 6: Confusion matrix for Logistic Regression', y=1.1)
    plt.ylabel('Actual label')
    plt.xlabel('Predicted label');
```

