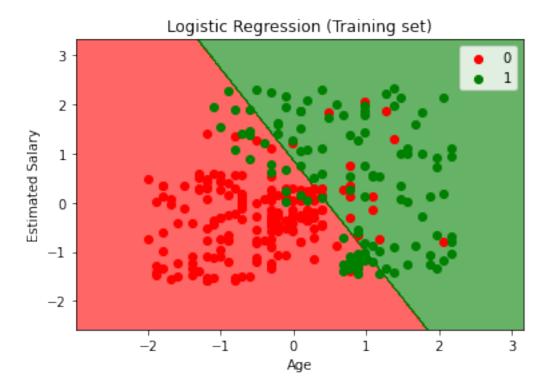
April 15, 2024

```
[1]: import numpy as np
      import matplotlib.pyplot as plt
      import pandas as pd
 [3]: pwd
 [3]: 'C:\\Users\\Tej'
 [6]: cd E:\
     E:\
 [8]: dataset = pd.read_csv('Social Network Ads.csv')
 [9]: dataset.head()
 [9]:
         User ID Gender
                           Age EstimatedSalary
                                                 Purchased
      0 15624510
                                          19000
                                                         0
                     Male
                            19
      1 15810944
                     Male
                            35
                                          20000
                                                         0
      2 15668575
                  Female
                            26
                                          43000
                                                         0
      3 15603246
                  Female
                            27
                                          57000
                                                         0
      4 15804002
                     Male
                            19
                                          76000
[10]: X = dataset.iloc[:, [2, 3]].values
      y = dataset.iloc[:, 4].values
      print(X[:3, :])
      print('-'*15)
      print(y[:3])
     19 19000]
          35 20000]
          26 43000]]
     [0 \ 0 \ 0]
[11]: from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25, __
       →random_state = 0)
      print(X_train[:3])
      print('-'*15)
      print(y train[:3])
      print('-'*15)
      print(X_test[:3])
      print('-'*15)
      print(y_test[:3])
           44 39000]
     32 120000]
           38 50000]]
     [0 1 0]
     [[ 30 87000]
          38 50000]
          35 75000]]
     [0 0 0]
[12]: from sklearn.preprocessing import StandardScaler
      sc_X = StandardScaler()
      X_train = sc_X.fit_transform(X_train)
      X_test = sc_X.transform(X_test)
      print(X_train[:3])
      print('-'*15)
      print(X_test[:3])
     [[ 0.58164944 -0.88670699]
      [-0.60673761 1.46173768]
      [-0.01254409 -0.5677824 ]]
     [[-0.80480212 0.50496393]
      [-0.01254409 -0.5677824 ]
      [-0.30964085 0.1570462 ]]
[13]: from sklearn.linear_model import LogisticRegression
      classifier = LogisticRegression(random_state = 0, solver='lbfgs')
      classifier.fit(X_train, y_train)
      y_pred = classifier.predict(X_test)
      print(X_test[:10])
      print('-'*15)
      print(y_pred[:10])
```

```
print(y_pred[:20])
      print(y_test[:20])
      [[-0.80480212 0.50496393]
      [-0.01254409 -0.5677824 ]
      [-0.30964085 0.1570462]
      [-0.80480212 0.27301877]
      [-0.30964085 -0.5677824 ]
      [-1.10189888 -1.43757673]
      [-0.70576986 -1.58254245]
      [-0.21060859 2.15757314]
      [-1.99318916 -0.04590581]
      [ 0.8787462 -0.77073441]]
      [0 0 0 0 0 0 0 1 0 1]
      [0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0]
     [0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0]
[14]: from sklearn.metrics import classification_report
      classification_report(y_test,y_pred)
[14]: '
                                   recall f1-score
                                                       support\n\n
                     precision
                                                                              0
      0.89
                0.96
                           0.92
                                       68\n
                                                       1
                                                               0.89
                                                                          0.75
                                                                                    0.81
                                                     0.89
      32\n\n
                                                                100\n
                accuracy
                                                                         macro avg
                                      100\nweighted avg
      0.89
                0.85
                           0.87
                                                               0.89
                                                                          0.89
                                                                                     0.89
      100\n'
[15]: from sklearn.metrics import confusion_matrix
      cm = confusion_matrix(y_test, y_pred)
      print(cm)
     [[65 3]
      [ 8 24]]
[16]: from sklearn.metrics import accuracy_score
      accuracy_score(y_test,y_pred)
[16]: 0.89
[17]: # Visualizing the Training set results
      from matplotlib.colors import ListedColormap
      X_set, y_set = X_train, y_train
      X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, __
       0].max() + 1, step = 0.01),
                            np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:,__
       \rightarrow 1].max() + 1, step = 0.01))
```

c argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points. *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.



```
[18]: # Visualizing the Test set results
      from matplotlib.colors import ListedColormap
      X_set, y_set = X_test, y_test
      X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, __
       0].max() + 1, step = 0.01),
                           np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, __
       41].max() + 1, step = 0.01))
      plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).
       ⇒reshape(X1.shape),
                   alpha = 0.6, cmap = ListedColormap(('red', 'green')))
      plt.xlim(X1.min(), X1.max())
      plt.ylim(X2.min(), X2.max())
      for i, j in enumerate(np.unique(y_set)):
          plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
                      c = ListedColormap(('red', 'green'))(i), label = j)
      plt.title('Logistic Regression (Test set)')
      plt.xlabel('Age')
      plt.ylabel('Estimated Salary')
      plt.legend()
      plt.show()
```

c argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points. *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.

