TUGUME JAMES 2023-U-MMU-BCS-01680

	Feature 1	Feature 2	Label
0	1.764052	4.764052	0
1	0.978738	-2.021262	1
2	1.867558	4.867558	0
3	0.950088	-2.049912	1
4	-0.103219	2.896781	0
95	-1.292857	-4.292857	1
96	-0.039283	-3.039283	1
97	0.523277	-2.476723	1
98	0.771791	3.771791	0
99	2.163236	5.163236	0

100 rows × 3 columns

```
In [66]: X = np.array(data[["Feature 1","Feature 2"]])
y = np.array(data["Label"])
```

In [67]: X

```
Out[67]: array([[ 1.76405235, 4.76405235],
                [0.97873798, -2.02126202],
                [ 1.86755799, 4.86755799],
                [0.95008842, -2.04991158],
                [-0.10321885, 2.89678115],
                [0.14404357, -2.85595643],
                [0.76103773, 3.76103773],
                [0.44386323, -2.55613677],
                [ 1.49407907, -1.50592093],
                [ 0.3130677 , 3.3130677 ],
                [-2.55298982, 0.44701018],
                [ 0.8644362 , -2.1355638 ],
                [ 2.26975462, 5.26975462],
                [ 0.04575852, -2.95424148],
                [1.53277921, 4.53277921],
                [0.15494743, -2.84505257],
                [-0.88778575, 2.11221425],
                [-0.34791215, 2.65208785],
                [ 1.23029068, 4.23029068],
                [-0.38732682, 2.61267318],
                [-1.04855297, -4.04855297],
                [-1.70627019, -4.70627019],
                [-0.50965218, 2.49034782],
                [-1.25279536, -4.25279536],
                [-1.61389785, 1.38610215],
                [-0.89546656, -3.89546656],
                [-0.51080514, -3.51080514],
                [-0.02818223, 2.97181777],
                [0.06651722, -2.93348278],
                [-0.63432209, 2.36567791],
                [-0.67246045, 2.32753955],
                [-0.81314628, 2.18685372],
                [ 0.17742614, 3.17742614],
                [-1.63019835, 1.36980165],
                [-0.90729836, -3.90729836],
                [ 0.72909056, -2.27090944],
                [ 1.13940068, 4.13940068],
                [ 0.40234164, 3.40234164],
                [-0.87079715, 2.12920285],
                [-0.31155253, 2.68844747],
                [-1.16514984, -4.16514984],
                [0.46566244, -2.53433756],
                [ 1.48825219, 4.48825219],
```

```
[ 1.17877957, 4.17877957],
[-1.07075262, -4.07075262],
[-0.40317695, 2.59682305],
[0.20827498, -2.79172502],
[0.3563664, -2.6436336],
[ 0.01050002, -2.98949998],
[0.12691209, -2.87308791],
[ 1.8831507 , -1.1168493 ],
[-1.270485, -4.270485],
[-1.17312341, -4.17312341],
[-0.41361898, -3.41361898],
[ 1.92294203, -1.07705797],
[ 1.86755896, -1.13244104],
[-0.86122569, -3.86122569],
[-0.26800337, 2.73199663],
[0.94725197, 3.94725197],
[ 0.61407937, -2.38592063],
[ 0.37642553, -2.62357447],
[0.29823817, -2.70176183],
[-0.69456786, -3.69456786],
[-0.43515355, 2.56484645],
[ 0.67229476, 3.67229476],
[-0.76991607, -3.76991607],
[-0.67433266, 2.32566734],
[-0.63584608, 2.36415392],
[0.57659082, -2.42340918],
[ 0.39600671, 3.39600671],
[-1.49125759, -4.49125759],
[ 0.1666735 , 3.1666735 ],
[ 2.38314477, -0.61685523],
[-0.91282223, 2.08717777],
[-1.31590741, 1.68409259],
[-0.06824161, -3.06824161],
[-0.74475482, 2.25524518],
[-0.09845252, 2.90154748],
[1.12663592, 4.12663592],
[-1.14746865, 1.85253135],
[-0.49803245, 2.50196755],
[0.94942081, -2.05057919],
[-1.22543552, 1.77456448],
[-1.00021535, -4.00021535],
[ 1.18802979, 4.18802979],
[ 0.92085882, 3.92085882],
```

```
[ 0.85683061, -2.14316939],
               [-1.03424284, -4.03424284],
               [-0.80340966, -3.80340966],
               [-0.4555325, -3.4555325],
               [-0.35399391, 2.64600609],
               [-0.6436184, 2.3563816],
               [ 0.62523145, -2.37476855],
               [-1.10438334, 1.89561666],
               [-0.739563, -3.739563],
               [-1.29285691, -4.29285691],
               [-0.03928282, -3.03928282],
               [ 0.52327666, -2.47672334],
               [ 0.77179055, 3.77179055],
               [ 2.16323595, 5.16323595]])
In [68]: y
Out[68]: array([0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1,
               0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0,
               1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1,
               0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1,
               1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0], dtype=int64)
In [69]: # spliting data int train and test
        X train, X test, y train, y test = train test split(X,y ,test size=0.2, random state = 42)
In [70]: # building the model
        model = LogisticRegression()
        model.fit(X_train,y_train)
Out[70]:
         ▼ LogisticRegression
         LogisticRegression()
In [71]: y pred = model.predict(X test)
In [72]: y pred
dtype=int64)
```

```
In [73]: | accuracy = accuracy score(y test, y pred)
In [74]: accuracy
Out[74]: 1.0
In [75]: coefficient = model.coef_
         intercept = model.intercept
In [76]: | coefficient
Out[76]: array([[ 0.66645107, -1.67786247]])
In [77]: intercept
Out[77]: array([0.28951949])
        # model optimaization on logostic regression
In [78]: |# importing the necessary libraries
         import numpy as np
         import pandas as pd
         from sklearn.model_selection import train_test_split, GridSearchCV
         from sklearn.linear_model import LogisticRegression
         from sklearn.metrics import accuracy score
In [88]: # Loading the datasets
         data =pd.read csv("C:\\Users\\tugumejames\\Desktop\\Logistic dataset.csv")
```

```
In [89]: data
```

Out[89]:

	Feature 1	Feature 2	Label
0	1.764052	4.764052	0
1	0.978738	-2.021262	1
2	1.867558	4.867558	0
3	0.950088	-2.049912	1
4	-0.103219	2.896781	0
95	-1.292857	-4.292857	1
96	-0.039283	-3.039283	1
97	0.523277	-2.476723	1
98	0.771791	3.771791	0
99	2.163236	5.163236	0

100 rows × 3 columns

```
In [80]: X = np.array(data[["Feature 1","Feature 2"]])
y = np.array(data["Label"])
```

In [81]: X

```
Out[81]: array([[ 1.76405235, 4.76405235],
                [0.97873798, -2.02126202],
                [ 1.86755799, 4.86755799],
                [0.95008842, -2.04991158],
                [-0.10321885, 2.89678115],
                [0.14404357, -2.85595643],
                [0.76103773, 3.76103773],
                [0.44386323, -2.55613677],
                [ 1.49407907, -1.50592093],
                [ 0.3130677 , 3.3130677 ],
                [-2.55298982, 0.44701018],
                [ 0.8644362 , -2.1355638 ],
                [ 2.26975462, 5.26975462],
                [0.04575852, -2.95424148],
                [1.53277921, 4.53277921],
                [0.15494743, -2.84505257],
                [-0.88778575, 2.11221425],
                [-0.34791215, 2.65208785],
                [ 1.23029068, 4.23029068],
                [-0.38732682, 2.61267318],
                [-1.04855297, -4.04855297],
                [-1.70627019, -4.70627019],
                [-0.50965218, 2.49034782],
                [-1.25279536, -4.25279536],
                [-1.61389785, 1.38610215],
                [-0.89546656, -3.89546656],
                [-0.51080514, -3.51080514],
                [-0.02818223, 2.97181777],
                [0.06651722, -2.93348278],
                [-0.63432209, 2.36567791],
                [-0.67246045, 2.32753955],
                [-0.81314628, 2.18685372],
                [ 0.17742614, 3.17742614],
                [-1.63019835, 1.36980165],
                [-0.90729836, -3.90729836],
                [ 0.72909056, -2.27090944],
                [ 1.13940068, 4.13940068],
                [ 0.40234164, 3.40234164],
                [-0.87079715, 2.12920285],
                [-0.31155253, 2.68844747],
                [-1.16514984, -4.16514984],
                [0.46566244, -2.53433756],
                [ 1.48825219, 4.48825219],
```

```
[ 1.17877957, 4.17877957],
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[-0.40317695, 2.59682305],
[0.20827498, -2.79172502],
[0.3563664, -2.6436336],
[ 0.01050002, -2.98949998],
[ 0.12691209, -2.87308791],
[ 1.8831507 , -1.1168493 ],
[-1.270485, -4.270485],
[-1.17312341, -4.17312341],
[-0.41361898, -3.41361898],
[ 1.92294203, -1.07705797],
[ 1.86755896, -1.13244104],
[-0.86122569, -3.86122569],
[-0.26800337, 2.73199663],
[0.94725197, 3.94725197],
[ 0.61407937, -2.38592063],
[ 0.37642553, -2.62357447],
[0.29823817, -2.70176183],
[-0.69456786, -3.69456786],
[-0.43515355, 2.56484645],
[ 0.67229476, 3.67229476],
[-0.76991607, -3.76991607],
[-0.67433266, 2.32566734],
[-0.63584608, 2.36415392],
[0.57659082, -2.42340918],
[ 0.39600671, 3.39600671],
[-1.49125759, -4.49125759],
[ 0.1666735 , 3.1666735 ],
[2.38314477, -0.61685523],
[-0.91282223, 2.08717777],
[-1.31590741, 1.68409259],
[-0.06824161, -3.06824161],
[-0.74475482, 2.25524518],
[-0.09845252, 2.90154748],
[ 1.12663592, 4.12663592],
[-1.14746865, 1.85253135],
[-0.49803245, 2.50196755],
[0.94942081, -2.05057919],
[-1.22543552, 1.77456448],
[-1.00021535, -4.00021535],
[ 1.18802979, 4.18802979],
[ 0.92085882, 3.92085882],
```

```
[ 0.85683061, -2.14316939],
                [-1.03424284, -4.03424284],
                [-0.80340966, -3.80340966],
                [-0.4555325, -3.4555325],
                [-0.35399391, 2.64600609],
                [-0.6436184 , 2.3563816 ],
                [0.62523145, -2.37476855],
                [-1.10438334, 1.89561666],
                [-0.739563, -3.739563],
                [-1.29285691, -4.29285691],
                [-0.03928282, -3.03928282],
                [0.52327666, -2.47672334],
                [ 0.77179055, 3.77179055],
                [ 2.16323595, 5.16323595]])
In [82]: y
Out[82]: array([0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1,
                0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0,
                1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1,
                0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1,
                1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0], dtype=int64)
In [83]: # splitting the data into training and testing
         X train, X test, y train, y test = train test split(X,y,test size=0.2, random state=42)
In [84]: # building the logistic regression model
         model = LogisticRegression()
```

```
In [85]: #Implementing the hyperparamter tuning using gridsearchCV
         param grid = {
                      "C":[0.001,0.01,0.1,1,10,100],
                       "penalty":["l1","l2",'none'],
                       "solver":[ 'lbfgs', 'liblinear', 'newton-cg', 'newton-cholesky', 'sag', 'saga'],
                       "max iter":[100,1000,]
In [86]: # perform a gridsearch (cross validation)
         grid_search= GridSearchCV(model, param_grid, cv=5)
         grid search.fit(X_train, y_train )
           warnings.warn(
         C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear model\ logistic.py:1182: FutureWarning: `penalty
         ='none'`has been deprecated in 1.2 and will be removed in 1.4. To keep the past behaviour, set `penalty=Non
         e`.
           warnings.warn(
         C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear_model\_logistic.py:1192: UserWarning: Setting pen
         alty=None will ignore the C and l1_ratio parameters
           warnings.warn(
         C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear_model\_sag.py:350: ConvergenceWarning: The max it
         er was reached which means the coef_ did not converge
           warnings.warn(
         C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear_model\_logistic.py:1182: FutureWarning: `penalty
         ='none'`has been deprecated in 1.2 and will be removed in 1.4. To keep the past behaviour, set `penalty=Non
         e`.
           warnings.warn(
         C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear model\ logistic.py:1192: UserWarning: Setting pen
         alty=None will ignore the C and l1 ratio parameters
           warnings.warn(
         C:\ProgramData\anaconda3\Lib\site-packages\sklearn\linear model\ sag.py:350: ConvergenceWarning: The max it
         er was reached which means the coef did not converge
```

```
In [87]: #Geting the best paramters found from Grid Search
         best_params = grid_search.best_params_
         print("Best parameters: ", best params)
         # use the best parameters to train the model
         best_model = LogisticRegression(**best_params)
         best model.fit(X train,y train)
         y pred = best model.predict(X test)
         #Evaluating the performance of the model
         accuracy = accuracy score(y test,y pred)
         print("accuracy: ", accuracy)
         coefficient = best model.coef
         intercept = best model.intercept
         print("coefficient:", coefficient)
         print("intercept:", intercept)
         Best parameters: {'C': 0.001, 'max iter': 100, 'penalty': '12', 'solver': 'liblinear'}
         accuracy: 1.0
         coefficient: [[ 0.00112375 -0.10074539]]
         intercept: [0.0052003]
```

In []: