## Logistic Regression Kaggle Data Set Processing

```
In [ ]:
In [4]:
         import numpy as np
         import matplotlib.pyplot as plt
         import pandas as pd
         # Used for Confusion Matrix
         from sklearn import metrics
         import seaborn as sns
         np.set printoptions(precision=2, suppress=True)
         from sklearn.datasets import fetch openml
         #dataset = fetch openml("mnist 784")
         # Used for Splitting Training and Test Sets
         from sklearn.model selection import train test split
         %matplotlib inline
         from sklearn.linear model import LogisticRegression
         from sklearn.linear model import LinearRegression
In [7]:
         s list = []
         intercept list = []
         weights_list = []
         df = pd.read csv("HR.csv")
         from sklearn.preprocessing import OneHotEncoder
         from sklearn.preprocessing import LabelEncoder
         labelencoder = LabelEncoder()
         df['n Gender'] = labelencoder.fit transform(df['Gender'])
         df['n JobRole']=labelencoder.fit transform(df['JobRole'])
         df['n Attrition'] = labelencoder.fit transform(df['Attrition'])
         df['n BusinessTravel'] = labelencoder.fit transform(df['BusinessTravel'])
         df['n Department'] = labelencoder.fit transform(df['Department'])
         df['n EducationField'] = labelencoder.fit transform(df['EducationField'])
         df.head()
```

```
df.drop(['Attrition','MaritalStatus','OverTime','Over18','BusinessTravel','JobRole','Gender','Department','Educ
        axis=1, inplace=True)
df.head()
p = df['n Attrition']
#df.drop(['n Attrition'],axis=1, inplace=True)
from sklearn.model selection import KFold
kf = KFold(n_splits=5, random state=None, shuffle=True)
train = df.to_numpy()
test = p.to numpy()
#.values.ravel()
dftemp = df
#p = df.from dict(p,orient='index',columns=['n Attrition'])
#p.shape()
for train index, test index in kf.split(df):
    #print("TRAIN:", train index, "TEST:", test index, "\n\n")
    #print("start TRAIN:", train_index, "TEST:", test_index,"end\n\n")
    X train, X test = df.iloc[train index], df.iloc[test index]
    y train, y test = X train.loc[:,['n Attrition']], X test.loc[:,['n Attrition']]
    #X train, X test = train index, test index
    #y train, y test = , p.values.ravel()
    X train.drop(['n Attrition'],axis=1, inplace=True)
    train_img = X_train
    #train img.drop(['n Attrition'],axis=1, inplace=True)
    X test.drop(['n Attrition'],axis=1, inplace=True)
    test img = X test
    #test img.drop(['n Attrition'],axis=1, inplace=True)
    train lbl = y train
    test lbl = y test
    # test size: what proportion of original data is used for test set
   train img, test img, train lbl, test lbl = train test split(df, p, test size=1/7.0, random state=0)
    #train lbl.head()
```

```
#train img.columns
#df.drop(['n Attrition'],axis=1, inplace=True)
X train = train img
X test = test img
Y_train = train_lbl
Y test = test lbl
logisticRegr = LogisticRegression( solver='lbfgs', max iter=9000,warm start=True)
#logisticRegr = LinearRegression()
logisticRegr
np.set_printoptions(precision=2, suppress=True)
logisticRegr.fit(train_img, train_lbl.values.ravel())
#p = logisticRegr.intercept
\#p = np.array2string(p)
\#p = str(round(p, 2))
print(f'intercept: {p}')
np.set printoptions(precision=2, suppress=True)
p = logisticRegr.coef_[0]
p = logisticRegr.coef_[0]
\#p = np.array2string(p)
\#p = round(p, 4)
\#p = np.array2string(p)
print(f" Weights {p}")
#logisticRegr.predict(test img)
#logisticRegr.predict(test img[0:10])
np.set printoptions(precision=2, suppress=True)
```

```
\#p = str(round(score, 2))
    #print(p)
    predictions = logisticRegr.predict(test img)
    print(f'\n Predict {predictions[:10]} \n Actual {test lbl[:10]}')
    \#std \ dev = [s, s]
    from sklearn.metrics import mean squared error
    s = mean squared error(test lbl, predictions)
    p = round(s, 4)
    print(f'mean squared error {p}')
    s_list.append(p)
/opt/anaconda3/lib/python3.8/site-packages/pandas/core/frame.py:4163: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#ret
urning-a-view-versus-a-copy
  return super().drop(
intercept: 0
                   1
1
        0
2
        1
        0
4
        0
1465
1466
1467
1468
1469
Name: n Attrition, Length: 1470, dtype: int64
 Weights [-0.03 - 0. 0.04 \ 0.04 \ 0. -0. -0.39 - 0. -0.35 -0.05 -0.35 - 0.
        0.14 - 0.02 \ 0.06 - 0.12 \ 0.05 - 0.35 - 0.05 - 0.19 - 0.18 \ 0.08 - 0.14
  0.14 -0.12 0.08 0.03 -0.05 0.13 0.08]
 Predict [0 0 0 0 0 0 0 0 0 0]
Actual
            n Attrition
2
              1
              0
3
6
7
              0
11
```

#score = logisticRegr.score(test img, test lbl)

```
26
               1
34
               1
45
               1
50
               1
mean squared error 0.1463
/opt/anaconda3/lib/python3.8/site-packages/pandas/core/frame.py:4163: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#ret
urning-a-view-versus-a-copy
  return super().drop(
intercept: 0.1463
 Weights [-0.03 -0.
                        0.03 - 0.05 \ 0. -0. -0.32 - 0. -0.25 - 0.01 - 0.35 - 0.
        0.17 - 0.03 \quad 0.01 - 0.26 \quad 0.05 - 0.39 - 0.04 - 0.18 - 0.13 \quad 0.08 - 0.14
  0.16 -0.16 0.07 0.04 0.04 0.14 0.03]
Predict [0 0 0 0 0 0 0 0 1 0]
Actual
            n Attrition
10
               0
13
               0
14
               1
17
20
               0
27
28
               0
29
               0
42
               1
53
mean squared error 0.1701
/opt/anaconda3/lib/python3.8/site-packages/pandas/core/frame.py:4163: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#ret
urning-a-view-versus-a-copy
  return super().drop(
intercept: 0.1701
Weights [-0.03 -0.
                        0.04 \quad 0.09 \quad 0. \quad -0. \quad -0.32 \quad -0. \quad -0.43 \quad -0.08 \quad -0.35 \quad -0.
        0.13 - 0.02 \quad 0.03 - 0.17 \quad 0.05 - 0.46 - 0.05 - 0.27 - 0.28 \quad 0.08 - 0.13
  0.11 -0.11 0.12 0.06 0.01 0.14 0.09]
Predict [0 0 0 0 0 0 0 0 0 1]
Actual
            n Attrition
5
               0
               0
19
25
               0
30
               0
31
```

21

1

```
35
               0
40
               0
46
               0
               0
47
mean squared error 0.1531
/opt/anaconda3/lib/python3.8/site-packages/pandas/core/frame.py:4163: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#ret
urning-a-view-versus-a-copy
  return super().drop(
intercept: 0.1531
 Weights [-0.04 - 0.
                         0.03 \quad 0.03 \quad 0. \quad -0. \quad -0.35 \quad -0. \quad -0.4 \quad -0.03 \quad -0.32 \quad -0.
        0.11 - 0.02 \quad 0.01 - 0.23 \quad 0.05 - 0.49 - 0.03 - 0.15 - 0.11 \quad 0.07 - 0.12
  0.12 -0.11 0.08 0.05 0.04 0.15 0.02]
Predict [0 0 0 0 0 0 0 0 0]
Actual
             n Attrition
1
               0
               0
8
               0
15
18
23
24
               1
               0
32
36
               1
37
mean squared error 0.1429
/opt/anaconda3/lib/python3.8/site-packages/pandas/core/frame.py:4163: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#ret
urning-a-view-versus-a-copy
  return super().drop(
intercept: 0.1429
Weights [-0.03 -0.
                         0.03 - 0.02 \quad 0. \quad -0. \quad -0.42 - 0. \quad -0.46 - 0.06 - 0.4 \quad -0.
        0.15 - 0.02 \quad 0.07 - 0.09 \quad 0.06 - 0.52 - 0.04 - 0.14 - 0.27 \quad 0.1 \quad -0.15
  0.14 -0.14 0.02 0.02 0.05 0.12 0.06]
 Predict [0 0 1 0 0 0 0 0 0 0]
Actual
             n Attrition
0
               1
               0
               0
12
16
               0
22
```

33

## **Splitting Data into Training and Test Sets**

```
train img.columns
In [10]:
Out[10]: Index(['Age', 'DailyRate', 'DistanceFromHome', 'Education', 'EmployeeCount',
                 'EmployeeNumber', 'EnvironmentSatisfaction', 'HourlyRate',
                 'JobInvolvement', 'JobLevel', 'JobSatisfaction', 'MonthlyIncome',
                 'MonthlyRate', 'NumCompaniesWorked', 'PercentSalaryHike',
                 'PerformanceRating', 'RelationshipSatisfaction', 'StandardHours',
                 'StockOptionLevel', 'TotalWorkingYears', 'TrainingTimesLastYear',
                 'WorkLifeBalance', 'YearsAtCompany', 'YearsInCurrentRole',
                 'YearsSinceLastPromotion', 'YearsWithCurrManager', 'n Gender',
                 'n JobRole', 'n BusinessTravel', 'n Department', 'n EducationField'],
               dtype='object')
          X train = train_img
In [11]:
          X test = test img
          Y train = train lbl
          Y test = test lbl
```

```
#print(f'Xtrain {X_train} X_test {X_test}')
         print(f'x_train {X_train[0:1]}')
In [12]:
         print(f'y_train {Y_train[0:1]}')
         print(f'x_train {X_test[0:1]}')
                   Age DailyRate DistanceFromHome Education EmployeeCount EmployeeNumber \
         x train
         1 49
                                                    1
           EnvironmentSatisfaction HourlyRate JobInvolvement JobLevel ... \
         1
                                           61
           WorkLifeBalance YearsAtCompany YearsInCurrentRole \
         1
                                       10
           YearsSinceLastPromotion YearsWithCurrManager n Gender n JobRole \
         1
           n_BusinessTravel n_Department n_EducationField
         1
                          1
                                       1
         [1 rows x 31 columns]
         y train
                   n Attrition
         1
                   Age DailyRate DistanceFromHome Education EmployeeCount EmployeeNumber \
         x train
         0 41
                   1102
                                         1
                                                    2
                                                                  1
                                                                                  1
           EnvironmentSatisfaction HourlyRate JobInvolvement JobLevel ... \
                                           94
           WorkLifeBalance YearsAtCompany YearsInCurrentRole \
           YearsSinceLastPromotion YearsWithCurrManager n Gender n JobRole \
           n_BusinessTravel n_Department n_EducationField
         [1 rows x 31 columns]
In [ ]:
In [13]: | print(train img.shape)
         (1176, 31)
```

## **Using Logistic Regression on Entire Dataset**

```
In [ ]:
 In [ ]:
In [17]:
          logisticRegr
Out[17]: LogisticRegression(max_iter=9000, warm_start=True)
          np.set_printoptions(precision=2, suppress=True)
In [18]:
          logisticRegr.fit(train_img, train_lbl)
          #logisticRegr = LogisticRegression(solver = 'lbfqs', max iter=1200)
         /opt/anaconda3/lib/python3.8/site-packages/sklearn/utils/validation.py:72: DataConversionWarning: A column-vect
         or y was passed when a 1d array was expected. Please change the shape of y to (n samples, ), for example using
         ravel().
           return f(**kwargs)
Out[18]: LogisticRegression(max_iter=9000, warm start=True)
          print('intercept:', logisticRegr.intercept_)
In [19]:
         intercept: [0.]
In [20]:
          np.set printoptions(precision=2, suppress=True)
          p = logisticRegr.coef [0]
          \#p = np.array2string(p)
          print(f"{p}")
```

```
 \begin{bmatrix} -0.03 & -0. & 0.03 & -0.02 & 0. & -0. & -0.42 & -0. & -0.46 & -0.06 & -0.4 & -0. \\ 0. & 0.15 & -0.02 & 0.07 & -0.09 & 0.06 & -0.52 & -0.04 & -0.14 & -0.27 & 0.1 & -0.15 \\ 0.14 & -0.14 & 0.02 & 0.02 & 0.05 & 0.12 & 0.06 \end{bmatrix}
```

Uses the information the model learned during the model training process

```
In [21]: | # Returns a NumPy Array
    # Predict for One Observation (image)
    logisticRegr.predict(test img)
0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
      0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
      1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
      0, 0, 0, 0, 0, 0, 0, 0]
In [22]: # Predict for Multiple Observations (images) at Once
    logisticRegr.predict(test img[0:10])
Out[22]: array([0, 0, 1, 0, 0, 0, 0, 0, 0])
```

## **Measuring Model Performance**

accuracy (fraction of correct predictions): correct predictions / total number of data points

Basically, how the model performs on new data (test set)