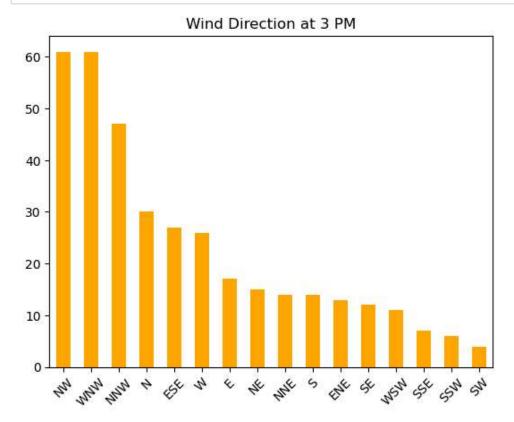
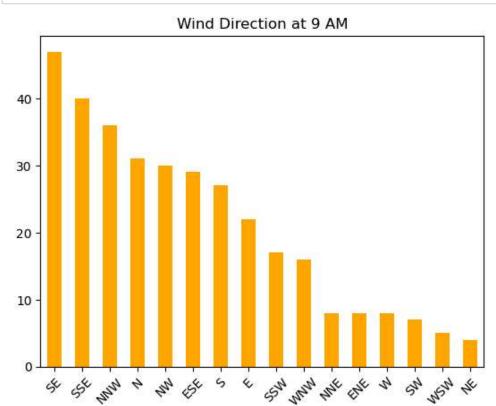
Logistric Regression Using Weather Data

```
In [1]: import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
In [2]: df = pd.read_csv("weather.csv")
In [3]: df.head()
Out[3]:
              MinTemp MaxTemp Rainfall Evaporation Sunshine WindGustDir WindGustSpeed WindDir9am WindDir3pm
                                                                                                                      NW
           0
                   8.0
                             24.3
                                       0.0
                                                    3.4
                                                              6.3
                                                                           NW
                                                                                           30.0
                                                                                                         SW
           1
                  14.0
                             26.9
                                       3.6
                                                    4.4
                                                              9.7
                                                                           ENE
                                                                                           39.0
                                                                                                           Ε
                                                                                                                       W
           2
                  13.7
                             23.4
                                                    5.8
                                                              3.3
                                                                           NW
                                                                                           85.0
                                                                                                                     NNE
                                       3.6
                                                                                                           Ν
                  13.3
                             15.5
                                      39.8
                                                    7.2
                                                              9.1
                                                                           NW
                                                                                           54.0
                                                                                                       WNW
                                                                                                                       W
           4
                   7.6
                             16.1
                                       2.8
                                                    5.6
                                                             10.6
                                                                           SSE
                                                                                           50.0
                                                                                                        SSE
                                                                                                                     ESE
          5 rows × 22 columns
In [4]: df.size
Out[4]: 8052
In [5]: df.shape
Out[5]: (366, 22)
In [6]: df.columns
Out[6]: Index(['MinTemp', 'MaxTemp', 'Rainfall', 'Evaporation', 'Sunshine',
                   'WindGustDir', 'WindGustSpeed', 'WindDir9am', 'WindDir3pm', 'WindSpeed9am', 'WindSpeed3pm', 'Humidity9am', 'Humidity3pm', 'Pressure9am', 'Cloud9am', 'Cloud3pm', 'Temp9am',
                   'Temp3pm', 'RainToday', 'RISK_MM', 'RainTomorrow'],
                 dtype='object')
```

```
In [7]: df['WindDir3pm'].value_counts().plot(kind="bar", color='orange')
    plt.title("Wind Direction at 3 PM")
    plt.xticks(rotation=45) # Rotate x-axis labels
    plt.show()
```



```
In [8]: df['WindDir9am'].value_counts().plot(kind="bar", color='orange')
plt.title("Wind Direction at 9 AM")
plt.xticks(rotation=45) # Rotate x-axis
plt.show()
```



Transform categorical data into numerical values in a DataFrame

```
In [9]: df['RainToday']=df['RainToday'].apply(lambda x:1 if x == "Yes" else 0)
In [10]: df['RainTomorrow']=df['RainTomorrow'].apply(lambda x:1 if x == "Yes" else 0)
In [11]: df.head()
```

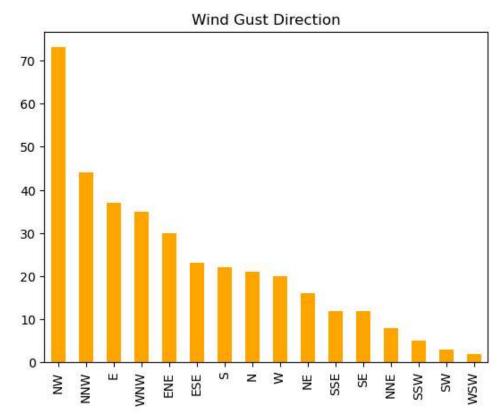
Out[11]:

	MinTemp	MaxTemp	Rainfall	Evaporation	Sunshine	WindGustDir	WindGustSpeed	WindDir9am	WindDir3pm
0	8.0	24.3	0.0	3.4	6.3	NW	30.0	SW	NW
1	14.0	26.9	3.6	4.4	9.7	ENE	39.0	E	W
2	13.7	23.4	3.6	5.8	3.3	NW	85.0	N	NNE
3	13.3	15.5	39.8	7.2	9.1	NW	54.0	WNW	W
4	7.6	16.1	2.8	5.6	10.6	SSE	50.0	SSE	ESE

5 rows × 22 columns

4

```
In [12]: df.WindGustDir.value_counts().plot(kind = "bar",color = 'orange')
    plt.title("Wind Gust Direction")
    plt.show()
```



```
In [13]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df=df.dropna()
```

```
In [14]: df.shape
Out[14]: (328, 22)

In [15]: df.WindGustDir = le.fit_transform(df.WindGustDir)
    df.WindDir3pm = le.fit_transform(df.WindDir3pm)
    df.WindDir9am = le.fit_transform(df.WindDir9am)
```

```
In [16]: df.describe()
Out[16]:
                    MinTemp
                              MaxTemp
                                           Rainfall Evaporation
                                                                 Sunshine WindGustDir WindGustSpeed WindDir9am V
           count 328.000000
                             328.000000 328.000000
                                                     328.000000
                                                               328.000000
                                                                            328.000000
                                                                                            328.000000
                                                                                                        328.000000
                                                                                             40.396341
                    7.742988
                              20.897561
                                          1.440854
                                                      4.702439
                                                                  8.014939
                                                                              6.192073
                                                                                                          7.067073
           mean
                    5.945199
                               6.707310
                                          4.289427
                                                       2.681183
                                                                  3.506646
                                                                              4.337765
                                                                                             13.132176
                                                                                                          3.897197
             std
                   -5.300000
                               7.600000
                                          0.000000
                                                      0.200000
                                                                  0.000000
                                                                              0.000000
                                                                                             13.000000
                                                                                                          0.000000
             min
                    2.850000
                              15.500000
                                          0.000000
                                                      2.550000
                                                                  6.000000
                                                                              2.000000
                                                                                             31.000000
                                                                                                          3.000000
            25%
                                          0.000000
                                                                              6.500000
                                                                                             39.000000
                                                                                                          7.500000
            50%
                    7.900000
                              20.400000
                                                      4.400000
                                                                  8.750000
            75%
                   12.800000
                              25.800000
                                          0.200000
                                                      6.600000
                                                                 10.700000
                                                                              8.000000
                                                                                             46.000000
                                                                                                         10.000000
                   20.900000
                              35.800000
                                         39.800000
                                                      13.800000
                                                                 13.600000
                                                                             15.000000
                                                                                             98.000000
                                                                                                         15.000000
            max
          8 rows × 22 columns
In [17]: #Training and Testing
          from sklearn.model_selection import train_test_split
          x = df.drop(['RainTomorrow'],axis=1)
          y = df['RainTomorrow']
In [18]: y
Out[18]: 0
                  1
                  1
          2
                  1
          3
                  1
          4
                  0
          361
                  0
          362
                  0
          363
                  0
          364
                  0
          365
          Name: RainTomorrow, Length: 328, dtype: int64
In [19]: | train_x , train_y ,test_x , test_y = train_test_split(x,y ,test_size = 0.3,random_state = 2)
In [20]: train_x.shape
Out[20]: (229, 21)
In [21]: | test_x.shape
Out[21]: (229,)
In [22]: train_y.shape
Out[22]: (99, 21)
```

```
In [23]: | from sklearn.linear_model import LogisticRegression
         model = LogisticRegression(max_iter=2000)
         model.fit(train_x , test_x)
Out[23]:
                 LogisticRegression
                                          (https://scikit-
                                           earn.org/1.4/modules/generated/sklearn.linear_model.LogisticRegression.h
          LogisticRegression(max_iter=2000)
In [24]: predict = model.predict(train_y)
In [25]: | from sklearn.metrics import accuracy_score, classification_report
         score = accuracy score(predict, test y)
Out[25]: 0.97979797979798
In [26]: report = classification_report(predict, test_y)
         print(report)
                        precision
                                     recall f1-score
                                                         support
                     0
                             1.00
                                       0.98
                                                 0.99
                                                              85
                     1
                             0.88
                                       1.00
                                                 0.93
                                                              14
                                                              99
                                                 0.98
             accuracy
                                                              99
                                       0.99
                                                 0.96
            macro avg
                             0.94
         weighted avg
                             0.98
                                       0.98
                                                 0.98
                                                              99
```

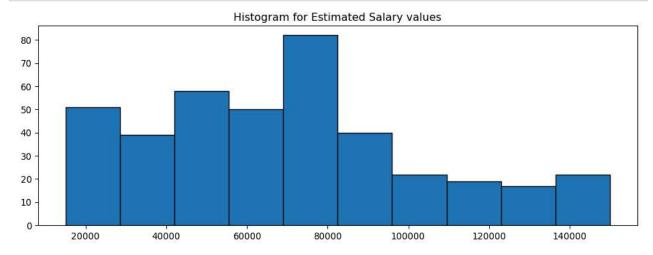
Logistric Regression using Social Network Ads

```
In [27]: from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
```

```
In [28]: | df = pd.read_csv('Social_Network_Ads.csv')
Out[28]:
                 User ID Gender Age EstimatedSalary Purchased
             0 15624510
                           Male
                                  19
                                              19000
                                                            0
             1 15810944
                                              20000
                                                            0
                           Male
                                  35
                                              43000
                                                            0
             2 15668575 Female
                                 26
               15603246 Female
                                  27
                                              57000
                                                            0
               15804002
                                              76000
                                                            0
                           Male
                                  19
           395 15691863 Female
                                  46
                                              41000
                                                            1
           396 15706071
                           Male
                                  51
                                              23000
                                                            1
                                              20000
           397 15654296 Female
                                  50
           398 15755018
                                              33000
                                                            0
                           Male
                                  36
           399 15594041 Female
                                  49
                                              36000
                                                            1
          400 rows × 5 columns
In [29]: df.head()
Out[29]:
               User ID Gender Age EstimatedSalary Purchased
           0 15624510
                                                          0
                         Male
                                            19000
           1 15810944
                         Male
                                35
                                            20000
                                                          0
           2 15668575 Female
                                26
                                            43000
           3 15603246 Female
                                27
                                            57000
                                                          0
           4 15804002
                                            76000
                                                          0
                         Male
                                19
In [30]: df.shape
Out[30]: (400, 5)
In [31]: df.columns
Out[31]: Index(['User ID', 'Gender', 'Age', 'EstimatedSalary', 'Purchased'], dtype='object')
```

In [32]: | df = df.drop(columns=['Gender'])

```
In [33]: plt.figure(figsize = (12,4), dpi = 96)
    plt.title("Histogram for Estimated Salary values")
    plt.hist(df['EstimatedSalary'], bins = 'sturges', edgecolor = 'black')
    plt. show()
```



```
In [34]: from sklearn.model_selection import train_test_split

X = df.drop(columns = 'Purchased')
y = df['Purchased']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3, random_state = 42
```

```
In [35]: lr = LogisticRegression()
    lr.fit(X_train, y_train)
```

```
Out[35]: 

LogisticRegression (1) (https://scikit-
LogisticRegression()

LogisticRegression()
```

```
In [36]: print(lr.score(X_train, y_train))
    y_train_pred = lr.predict(X_train)
```

0.8357142857142857

```
In [37]: from sklearn.metrics import confusion_matrix
    from sklearn.metrics import classification_report
    print("\n Confusion Matrix \n")
    print(confusion_matrix(y_train, y_train_pred))

print("\n Classification Report\n")
    print(classification_report(y_train, y_train_pred))
```

Confusion Matrix

[[170 14] [32 64]]

Classification Report

	precision	recall	f1-score	support
0 1	0.84 0.82	0.92 0.67	0.88 0.74	184 96
accuracy macro avg weighted avg	0.83 0.83	0.80 0.84	0.84 0.81 0.83	280 280 280

In []: