

8_B

Importing libraries

```
In [52]: import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import SGDClassifier
from sklearn.linear_model import LogisticRegression
import pandas as pd
import numpy as np
from sklearn.preprocessing import StandardScaler, Normalizer
import matplotlib.pyplot as plt
from sklearn.svm import SVC
import warnings
warnings.filterwarnings("ignore")
```

Reading the dataset

```
In [53]: data = pd.read_csv('task_b.csv')
data=data.iloc[:,1:]
```

```
In [54]: data.std() # CALCULATING THE STANDARD DEVIATION
```

```
Out[54]: f1      488.195035
f2     10403.417325
f3        2.926662
y         0.501255
dtype: float64
```

Task 1

```
In [55]: X=data[['f1','f2','f3']].values
        Y=data['y'].values
        print(X.shape)
        print(Y.shape)
```

```
(200, 3)
(200,)
```

```
In [56]: clf = SGDClassifier(loss = 'log',random_state = 42)
```

```
In [57]: clf.fit(X,Y)
```

```
Out[57]: SGDClassifier(alpha=0.0001, average=False, class_weight=None, epsilon=
0.1,
          eta0=0.0, fit_intercept=True, l1_ratio=0.15,
          learning_rate='optimal', loss='log', max_iter=None, n_iter=None,
          n_jobs=1, penalty='l2', power_t=0.5, random_state=42, shuffle=True,
          tol=None, verbose=0, warm_start=False)
```

```
In [58]: coef_dict = {}
        for coef, feat in zip(clf.coef_[0,:],['f1','f2','f3']): # PRINTING THE
            WEIGHT COEFFICIENT LOGISTIC REGRESSION
            coef_dict[feat] = coef
```

```
In [59]: coef_dict
```

```
Out[59]: {'f1': 13617.93477406387, 'f2': -15636.943966575614, 'f3': 5704.9422734
29228}
```

```
In [60]: clf = SGDClassifier(loss = 'hinge',random_state = 42)
```

```
In [61]: clf.fit(X,Y)
```

```
Out[61]: SGDClassifier(alpha=0.0001, average=False, class_weight=None, epsilon=
```

```
0.1,
    eta0=0.0, fit_intercept=True, l1_ratio=0.15,
    learning_rate='optimal', loss='hinge', max_iter=None, n_iter=Non
e,
    n_jobs=1, penalty='l2', power_t=0.5, random_state=42, shuffle=Tr
ue,
    tol=None, verbose=0, warm_start=False)
```

```
In [62]: coef_dict = {}
        for coef, feat in zip(clf.coef_[0,:],['f1','f2','f3']): # # PRINTING TH
        E WEIGHT COEFFICIENT SVM
            coef_dict[feat] = coef
```

```
In [63]: coef_dict
```

```
Out[63]: {'f1': -5752.778874720262, 'f2': -37467.51703073541, 'f3': 5284.7826362
11242}
```

Observations

- ACCORDING TO ABOVE RESULTS IMPORTANT FEATURES ARE $f_2 > f_1 > f_3$ IT IS PROPORTIONAL TO VARIANCE OF FEATURES($\text{var}(F_2) > \text{var}(F_1) > \text{Var}(F_3)$)
- BOTH IN LR AND SVM FEATURES WHICH HAVE HIGH VARIANCE IS THE MOST IMPORTANT FEATURE
- SINCE f_2 HAS HIGH NEGATIVE VALUE DOES NOT MEAN IT HAVE IMPACT ON CLASSIFYING NEGATIVE CLASS BECAUSE OUR DATA NOT STANDARDIZED

Task 2

```
In [64]: df = StandardScaler().fit_transform(data[['f1','f2','f3']]) # STANDARDI
        ZING THE DATA
```

```
In [65]: clf = SGDClassifier(loss = 'log',random_state = 42)
```

```
In [66]: clf.fit(df,Y)
```

```
Out[66]: SGDClassifier(alpha=0.0001, average=False, class_weight=None, epsilon=0.1,
eta0=0.0, fit_intercept=True, l1_ratio=0.15,
learning_rate='optimal', loss='log', max_iter=None, n_iter=None,
n_jobs=1, penalty='l2', power_t=0.5, random_state=42, shuffle=True,
tol=None, verbose=0, warm_start=False)
```

```
In [67]: coef_dict = {}
for coef, feat in zip(clf.coef_[0,:],['f1','f2','f3']):
    coef_dict[feat] = coef
```

```
In [68]: coef_dict
```

```
Out[68]: {'f1': -7.27392033428641, 'f2': 3.1539523793187896, 'f3': 38.87585461795804}
```

```
In [69]: clf = SGDClassifier(loss = 'hinge',random_state = 42)
```

```
In [70]: clf.fit(df,Y)
```

```
Out[70]: SGDClassifier(alpha=0.0001, average=False, class_weight=None, epsilon=0.1,
eta0=0.0, fit_intercept=True, l1_ratio=0.15,
learning_rate='optimal', loss='hinge', max_iter=None, n_iter=None,
n_jobs=1, penalty='l2', power_t=0.5, random_state=42, shuffle=True,
tol=None, verbose=0, warm_start=False)
```

```
In [71]: coef_dict = {}
for coef, feat in zip(clf.coef_[0,:],['f1','f2','f3']):
    coef_dict[feat] = coef
```

```
In [72]: coef_dict
```

```
Out[72]: {'f1': -8.608946430950358, 'f2': 2.8645127265880586, 'f3': 41.29014342543963}
```

Observations

- AFTER STANDARDIZATION MOST IMPORTANT FEATURE IS f3
- f3>f1>f2 IS THE NEW FEATURE IMPORTANCE"
- AFTER STANDARDIZATION EFFECT OF VARIANCE IS REMOVED.