In [2]: import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns

In [3]: data = pd.read_csv("C:/Users/SANDEEP/OneDrive/Desktop/Eclipse/Eclipsedata2.csv")

In [5]: for i in data.columns:
 data[i]=data[i].astype(int)
 data.shape

Out[5]: (7888, 200)

In [6]: zero_cols = (data == 0).all() data = data.loc[:, ~zero_cols] data

Out[6]:

	pre	ACD	FOUT_avg	FOUT_max	FOUT_sum	MLOC_avg	MLOC_max	MLOC_sum	NBD
0	0	0	3	15	35	7	35	79	
1	0	0	2	12	25	10	16	92	
2	0	0	0	0	0	0	0	0	
3	0	0	4	9	32	7	15	49	
4	0	0	4	11	24	8	39	53	
7883	0	0	3	10	31	5	15	52	
7884	0	0	0	2	2	2	6	8	
7885	1	6	10	46	120	17	81	204	
7886	0	0	6	25	106	8	34	134	
7887	0	0	0	2	4	1	6	13	

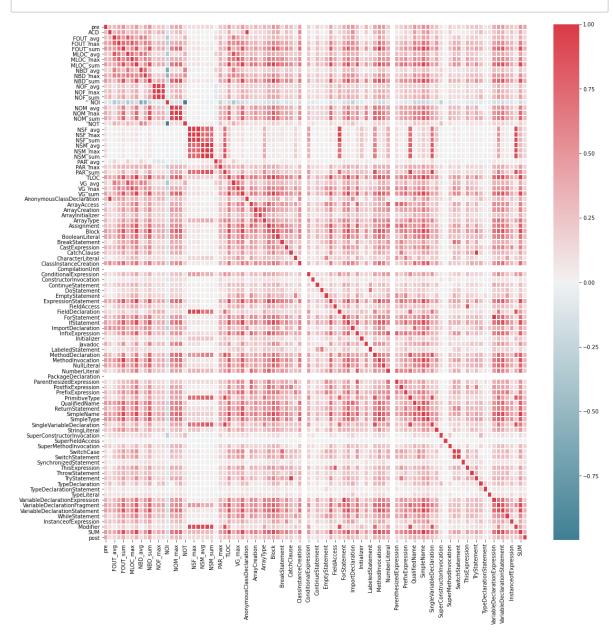
7888 rows × 96 columns

In [7]: corr=data.corr() corr

Out[7]:

	pre	ACD	FOUT_avg	FOUT_max	FOUT_sum	MLOC_avg	MLO
pre	1.000000	0.313830	0.283869	0.407779	0.501487	0.251237	0.0
ACD	0.313830	1.000000	0.308763	0.448432	0.454992	0.248542	0.0
FOUT_avg	0.283869	0.308763	1.000000	0.733544	0.471107	0.836103	0.
FOUT_max	0.407779	0.448432	0.733544	1.000000	0.697394	0.660235	0.8
FOUT_sum	0.501487	0.454992	0.471107	0.697394	1.000000	0.398068	0.
WhileStatement	0.278649	0.140794	0.181262	0.351975	0.472553	0.308678	0.4
InstanceofExpression	0.262850	0.263854	0.191771	0.322750	0.418362	0.204886	0.2
Modifier	0.189775	0.102344	0.045596	0.122223	0.195417	0.052720	0.
SUM	0.507290	0.396681	0.378159	0.629589	0.876762	0.432298	0.0
post	0.429497	0.144605	0.157955	0.236446	0.357571	0.142240	0.
00							

96 rows × 96 columns



In [9]: #Identify the features with highest correlation to the target variable target_corr = corr['post'] top_features = target_corr.abs().sort_values(ascending=False)[:100].index

In [10]: selected_data = data[top_features] selected_data

Out[10]:

	post	pre	SUM	SimpleName	NBD_sum	Block	QualifiedName	TLOC	VariableDeclaratio
0	0	0	506	181	18	25	12	115	
1	0	0	547	240	16	24	11	145	
2	0	0	86	33	0	0	7	14	
3	0	0	429	184	12	12	13	72	
4	0	0	298	129	9	14	17	73	
7883	0	0	438	182	13	14	23	80	
7884	0	0	102	40	3	3	15	20	
7885	0	1	1378	590	27	34	84	265	
7886	0	0	911	399	29	38	35	177	
7887	0	0	181	67	7	7	15	34	

7888 rows × 96 columns

In [11]: selected_data = selected_data.drop(['post'],axis=1) selected_data

Out[11]:

	pre	SUM	SimpleName	NBD_sum	Block	QualifiedName	TLOC	VariableDeclarationState
0	0	506	181	18	25	12	115	
1	0	547	240	16	24	11	145	
2	0	86	33	0	0	7	14	
3	0	429	184	12	12	13	72	
4	0	298	129	9	14	17	73	
7883	0	438	182	13	14	23	80	
7884	0	102	40	3	3	15	20	
7885	1	1378	590	27	34	84	265	
7886	0	911	399	29	38	35	177	
7887	0	181	67	7	7	15	34	

7888 rows × 95 columns

```
import xgboost as xgb
        from sklearn.model_selection import train_test_split
        from sklearn import metrics
In [15]: X = selected data
        y = data['post']
        X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.2, random_state=1)
        params = {'objective': 'binary:hinge', 'eval metric': 'logloss'}
In [16]:
In [17]:
        dtrain = xgb.DMatrix(X train, label=y train)
        dtest = xgb.DMatrix(X test, label=y test)
In [18]:
        model = xgb.train(params, dtrain, num boost round=100)
In [19]: y pred = model.predict(dtest)
In [20]: y pred = [1 if p >= 0.5 else 0 for p in y pred]
        a=metrics.accuracy score(y test,y pred)
        p=metrics.precision score(y test,y pred,average='weighted',zero division=0)
        r=metrics.recall score(y test,y pred,average='weighted',zero division=0)
        f1=metrics.f1 score(y test,y pred,average='weighted',zero division=0)
        print("Accuracy:",a,"Precision: ",p,"Recall: ",r,"F1score: ",f1)
        Accuracy: 0.8605830164765526 Precision: 0.8340547447855596 Recall: 0.86058301647655
        26 F1score: 0.8470135537988538
  In [ ]:
        #RUS
In [22]: from imblearn.under sampling import RandomUnderSampler
        from sklearn.linear model import LogisticRegression
        from sklearn.multiclass import OneVsRestClassifier
In [23]: Xr = selected data
        yr = data['post']
        Xr train, Xr test, yr train, yr test = train test split(Xr, yr, test size=0.3, random state=1)
In [24]: rus = RandomUnderSampler(random_state=42)
        X train rus, y train rus = rus.fit resample(Xr train, yr train)
```

```
In [25]: Ir = LogisticRegression(random_state=42)
Ir.fit(X_train_rus, y_train_rus)
```

C:\Users\SANDEEP\anaconda3\lib\site-packages\sklearn\linear_model_logistic.py:458: ConvergenceWarning: lbfgs failed to converge (status=1):

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max iter) or scale the data as shown in:

https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear model.html#logistic-regression)

```
n iter i = check optimize result(
```

Out[25]: LogisticRegression(random_state=42)

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [26]: yr_pred = Ir.predict(Xr_test)
```

```
In [27]: a=metrics.accuracy_score(yr_test,yr_pred)
p=metrics.precision_score(yr_test,yr_pred,average='weighted',zero_division=0)
r=metrics.recall_score(yr_test,yr_pred,average='weighted',zero_division=0)
f1=2*(p*r)/(p+r)
print("Accuracy:",a,"Precision: ",p,"Recall: ",r,"F1score: ",f1)
```

Accuracy: 0.5369666244190959 Precision: 0.826294355946815 Recall: 0.536966624419095 9 F1score: 0.6509281751322792

In [28]: from sklearn.neighbors import KNeighborsClassifier

```
In [30]: Xn= selected_data yn = data['post'] Xn_train, Xn_test, yn_train, yn_test = train_test_split(Xn, yn, test_size=0.3, random_state=1)
```

In [31]: knn = KNeighborsClassifier(n_neighbors=3)

```
In [32]: knn.fit(Xn_train, yn_train)
```

Out[32]: KNeighborsClassifier(n_neighbors=3)

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [33]: yn_pred = knn.predict(Xn_test)
```

In [35]: a=metrics.accuracy_score(yn_test,yn_pred)
p=metrics.precision_score(yn_test,yn_pred,average='weighted',zero_division=0)
r=metrics.recall_score(yn_test,yn_pred,average='weighted',zero_division=0)
f1=metrics.f1_score(y_test,y_pred,average='weighted',zero_division=0)
print("Accuracy:",a,"Precision: ",p,"Recall: ",r,"F1score: ",f1)

Accuracy: 0.8719898605830165 Precision: 0.8213355403725865 Recall: 0.87198986058301 65 F1score: 0.8470135537988538

- In [37]: from sklearn.ensemble import VotingClassifier from xgboost import XGBClassifier
- In [38]: Xv= selected_data
 yv = data['post']
 Xv_train, Xv_test, yv_train, yv_test = train_test_split(Xv, yv, test_size=0.3, random_state=1)
- In [39]: model1 = KNeighborsClassifier(n_neighbors=5)
 model3 = XGBClassifier(n_estimators=100, learning_rate=0.1,objective='binary:logistic',eval_memodel2 = LogisticRegression(random_state=42)
- In [40]: voting_clf = VotingClassifier(estimators=[('xgb', model1), ('rus', model2),('knn', model3)], voting=

```
In [41]: voting_clf.fit(Xv_train, yv_train)
```

C:\Users\SANDEEP\anaconda3\lib\site-packages\sklearn\linear_model_logistic.py:458: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max iter) or scale the data as shown in:

https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear model.html#logistic-regression)

n iter i = check optimize result(

```
Out[41]: VotingClassifier(estimators=[('xgb', KNeighborsClassifier()),
```

('rus', LogisticRegression(random_state=42)), ('knn',

XGBClassifier(base_score=None, booster=None,

callbacks=None,

colsample_bylevel=None,

colsample_bynode=None,

colsample bytree=None,

early_stopping_rounds=None,

enable_categorical=False,

eval metric='logloss',

feature_types=None, gamma=None,

gpu id=None, grow policy=None,

importance_type=None,

interaction constraints=None,

learning rate=0.1, max bin=None,

max cat threshold=None,

max_cat_to_onehot=None,

max_delta_step=None, max_depth=None,

max leaves=None,

min child weight=None, missing=nan,

monotone constraints=None,

n estimators=100, n jobs=None,

num parallel tree=None,

predictor=None, random_state=None, ...))])

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [43]: yv_pred = voting_clf.predict(Xv_test)
```

In [44]: a=metrics.accuracy_score(yv_test,yv_pred)
p=metrics.precision_score(yv_test,yv_pred,average='weighted',zero_division=0)
r=metrics.recall_score(yv_test,yv_pred,average='weighted',zero_division=0)
f1=metrics.f1_score(y_test,y_pred,average='weighted',zero_division=0)
print("Accuracy:",a,"Precision: ",p,"Recall: ",r,"F1score: ",f1)

Accuracy: 0.8859315589353612 Precision: 0.8205093489640337 Recall: 0.88593155893536 12 F1score: 0.8470135537988538