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

In [12]: data = pd.read_csv("C:/Users/SANDEEP/OneDrive/Desktop/NASA datasers/kc1_csv.csv")

In [11]: for i in data.columns:
 data[i]=data[i].astype(int)

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	loc	v(g)	ev(g)	iv(g)	n	V	I	d	i	е	 IOCode	IOComment	IOBlank	lo
0	1	1	1	1	1	1	1	1	1	1	 2	2	2	
1	1	1	1	1	1	1	1	1	1	1	 1	1	1	
2	83	11	1	11	171	927	0	23	40	21378	 65	10	6	
3	46	8	6	8	141	769	0	14	51	11436	 37	2	5	
4	25	3	1	3	58	254	0	9	27	2381	 21	0	2	
2104	19	2	1	2	40	175	0	6	25	1197	 12	1	2	
2105	23	3	3	3	60	278	0	9	28	2700	 18	1	2	
2106	2	1	1	1	4	8	0	1	5	12	 0	0	0	
2107	13	1	1	1	17	60	0	4	15	243	 6	0	5	
2108	11	2	1	2	27	102	0	6	17	616	 9	0	0	

2109 rows × 22 columns

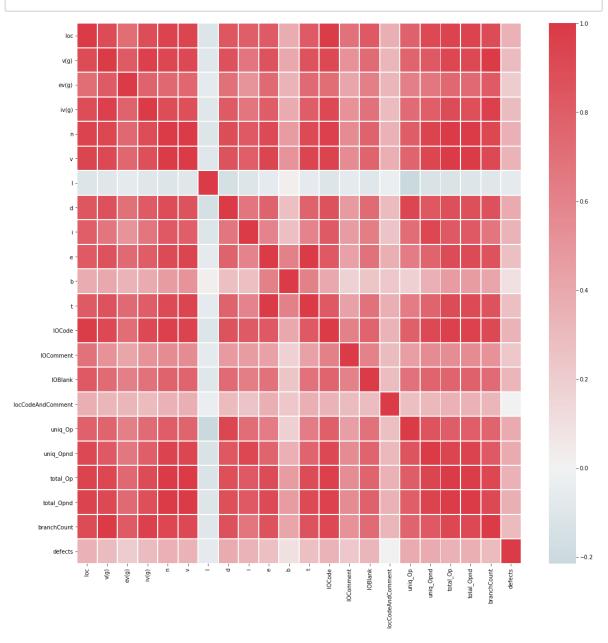
In [10]: corr=data.corr() corr

Out[10]:

	loc	v(g)	ev(g)	iv(g)	n	v	- 1
loc	1.000000	0.902632	0.718875	0.895183	0.948510	0.938509	-0.110648
v(g)	0.902632	1.000000	0.819979	0.965691	0.922520	0.915569	-0.096869
ev(g)	0.718875	0.819979	1.000000	0.775755	0.746289	0.753612	-0.062994
iv(g)	0.895183	0.965691	0.775755	1.000000	0.893780	0.884937	-0.094170
n	0.948510	0.922520	0.746289	0.893780	1.000000	0.994944	-0.113230
v	0.938509	0.915569	0.753612	0.884937	0.994944	1.000000	-0.100045
1	-0.110648	-0.096869	-0.062994	-0.094170	-0.113230	-0.100045	1.000000
d	0.846915	0.869069	0.699902	0.816050	0.891045	0.859948	-0.151379
i	0.803307	0.673686	0.511792	0.674865	0.827143	0.800298	-0.108742
е	0.821335	0.862408	0.738875	0.807284	0.908209	0.934954	-0.061712
b	0.375858	0.401335	0.336838	0.387409	0.463649	0.511154	0.023958
t	0.821337	0.862404	0.738865	0.807277	0.908202	0.934949	-0.061659
IOCode	0.985491	0.918526	0.723286	0.916928	0.961050	0.949673	-0.108407
lOComment	0.686717	0.517046	0.411720	0.519591	0.547562	0.545153	-0.061285
IOBlank	0.824426	0.731419	0.624956	0.689650	0.777578	0.772091	-0.091932
locCodeAndComment	0.365863	0.328269	0.321557	0.304393	0.347555	0.359965	-0.031088
uniq_Op	0.783487	0.768653	0.616156	0.729793	0.809491	0.771043	-0.215276
uniq_Opnd	0.915715	0.827922	0.666772	0.803426	0.945880	0.931039	-0.128103
total_Op	0.946597	0.925185	0.747860	0.901764	0.998146	0.992036	-0.115961
total_Opnd	0.944167	0.910997	0.737913	0.873896	0.995162	0.991849	-0.107831
branchCount	0.901113	0.998609	0.819810	0.964736	0.923873	0.917393	-0.096723
defects	0.348405	0.295585	0.205192	0.295712	0.355101	0.339487	-0.062956

22 rows × 22 columns

In [13]: f,ax=plt.subplots(figsize=(18,18)) cmap=sns.diverging_palette(220,10,as_cmap=True) heatmap=sns.heatmap(corr,cmap=cmap,center=0.0,vmax=1,linewidths=1,ax=ax) plt.show()



In [18]: dat=data.drop(['I','loc','ev(g)','b','t','lOComment','locCodeAndComment','branchCount'],axis=1) dat

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	v(g)	iv(g)	n	V	d	i	е	IOCode	IOBlank	uniq_Op	uniq_Opnd
0	1.4	1.4	1.3	1.30	1.30	1.30	1.30	2	2	1.2	1.2
1	1.0	1.0	1.0	1.00	1.00	1.00	1.00	1	1	1.0	1.0
2	11.0	11.0	171.0	927.89	23.04	40.27	21378.61	65	6	18.0	25.0
3	8.0	8.0	141.0	769.78	14.86	51.81	11436.73	37	5	16.0	28.0
4	3.0	3.0	58.0	254.75	9.35	27.25	2381.95	21	2	11.0	10.0
2104	2.0	2.0	40.0	175.69	6.82	25.77	1197.90	12	2	10.0	11.0
2105	3.0	3.0	60.0	278.63	9.69	28.75	2700.58	18	2	12.0	13.0
2106	1.0	1.0	4.0	8.00	1.50	5.33	12.00	0	0	3.0	1.0
2107	1.0	1.0	17.0	60.94	4.00	15.24	243.78	6	5	6.0	6.0
2108	2.0	2.0	27.0	102.80	6.00	17.13	616.79	9	0	8.0	6.0

2109 rows × 14 columns

In [19]: lis= ['v(g)','iv(g)','n','v','d','i','e','IOCode','IOBlank','uniq_Op','uniq_Opnd','total_Op','total_Opnd'] fea=data[lis] fea

Out[19]:

	v(g)	iv(g)	n	v	d	i	е	IOCode	IOBlank	uniq_Op	uniq_Opnd
0	1.4	1.4	1.3	1.30	1.30	1.30	1.30	2	2	1.2	1.2
1	1.0	1.0	1.0	1.00	1.00	1.00	1.00	1	1	1.0	1.0
2	11.0	11.0	171.0	927.89	23.04	40.27	21378.61	65	6	18.0	25.0
3	8.0	8.0	141.0	769.78	14.86	51.81	11436.73	37	5	16.0	28.0
4	3.0	3.0	58.0	254.75	9.35	27.25	2381.95	21	2	11.0	10.0
2104	2.0	2.0	40.0	175.69	6.82	25.77	1197.90	12	2	10.0	11.0
2105	3.0	3.0	60.0	278.63	9.69	28.75	2700.58	18	2	12.0	13.0
2106	1.0	1.0	4.0	8.00	1.50	5.33	12.00	0	0	3.0	1.0
2107	1.0	1.0	17.0	60.94	4.00	15.24	243.78	6	5	6.0	6.0
2108	2.0	2.0	27.0	102.80	6.00	17.13	616.79	9	0	8.0	6.0

2109 rows × 13 columns

```
In [20]:
         #KNN
In [21]: from imblearn.over sampling import SMOTE
         from sklearn.model selection import train test split
         from sklearn.neighbors import KNeighborsClassifier
         from imblearn.under sampling import RandomUnderSampler
         from sklearn.linear model import LogisticRegression
         from sklearn import metrics
In [23]: X = fea
         y = data['defects']
         X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=1)
         smote = SMOTE(random state=1)
In [25]:
         X train resampled,y train resampled = smote.fit resample(X train,y train)
In [26]: knn = KNeighborsClassifier(n neighbors=3)
In [27]: knn.fit(X_train, y_train)
Out[27]: 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 [28]: y pred = knn.predict(X test)
In [29]: a=metrics.accuracy score(y test,y pred)
         p=metrics.precision score(y test,y pred)
         r=metrics.recall_score(y_test,y_pred)
         f1=2*(p*r)/(p+r)
         print("Accuracy:",a," Precision:",p," Recall:",r," F1Score:",f1)
         Accuracy: 0.8262243285939969 Precision: 0.39655172413793105 Recall: 0.23469387755
         102042 F1Score: 0.2948717948717949
In [30]: #RUS
In [31]: | Xr = fea
         yr = data['defects']
         Xr train, Xr test, yr train, yr test = train test split(Xr, yr, test size=0.3, random state=1)
```

```
In [32]:
         smote = SMOTE(random state=42)
         X train resampled, y train resampled = smote.fit resample(Xr train, yr train)
In [33]: rus = RandomUnderSampler(random_state=42)
         X train rus, y train rus = rus.fit resample(Xr train, yr train)
In [34]: 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: Conv
         ergenceWarning: 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/mod
         ules/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-lea
         rn.org/stable/modules/linear model.html#logistic-regression)
          n iter i = check optimize result(
Out[34]: 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 [35]: yr pred = Ir.predict(Xr test)
         a=metrics.accuracy score(yr test,yr pred)
In [36]:
         p=metrics.precision score(yr test,yr pred)
         r=metrics.recall score(yr test,yr pred)
         f1=2*(p*r)/(p+r)
         print("Accuracy:",a," Precision:",p," Recall:",r," F1Score:",f1)
         Accuracy: 0.7472353870458136 Precision: 0.32967032967032966 Recall: 0.612244897959
         1837 F1Score: 0.4285714285714286
In [37]:
         #XGBOOST
In [38]: import xgboost as xgb
In [39]: | Xx = data[lis]
         yx = data["defects"]
         Xx train, Xx test, yx train, yx test = train test split(Xx, yx, test size=0.3, random state=1)
```

```
In [40]:
        smote = SMOTE(random state=1)
         X train resampled,y train resampled = smote.fit resample(Xx train,yx train)
        params = {'objective': 'binary:logistic', 'eval metric': 'logloss'}
In [41]:
         dtrain = xgb.DMatrix(Xx train, label=yx train)
In [42]:
         dtest = xgb.DMatrix(Xx test, label=yx test)
In [43]: model = xgb.train(params, dtrain, num boost round=100)
In [44]: | yx_pred = model.predict(dtest)
In [45]: |yx| \text{ pred} = [1 \text{ if p} >= 0.5 \text{ else } 0 \text{ for p in } yx \text{ pred}]
In [46]: a=metrics.accuracy_score(yx_test,yx_pred)
         p=metrics.precision score(yx test,yx pred)
         r=metrics.recall_score(yx_test,yx_pred)
         f1=2*(p*r)/(p+r)
         print("Accuracy:",a," Precision:",p," Recall:",r," F1Score:",f1)
         Accuracy: 0.8530805687203792    Precision: 0.545454545454    Recall: 0.306122448979
         59184 F1Score: 0.39215686274509803
         #Voting Ensemble
In [47]:
In [48]:
        from sklearn.ensemble import VotingClassifier
         from xgboost import XGBClassifier
In [50]: X = fea
         y = data['defects']
         X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=42)
In [51]: model1 = KNeighborsClassifier(n neighbors=5)
         model3 = XGBClassifier(n estimators=100, learning rate=0.1,objective='binary:logistic',eval me
         model2 = LogisticRegression(random state=42)
In [52]: voting clf = VotingClassifier(estimators=[('xgb', model1), ('rus', model2),('knn', model3)], voting=
```

```
In [53]: voting_clf.fit(X_train, y_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[53]: 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 [54]: y_predic = voting_clf.predict(X_test)
```

```
In [55]: a=metrics.accuracy_score(y_test,y_predic)
p=metrics.precision_score(y_test,y_predic)
r=metrics.recall_score(y_test,y_predic)
f1=metrics.f1_score(y_test,y_predic)
print("Accuracy:",a,"Precision: ",p,"Recall: ",r,"F1score: ",f1)

Accuracy: 0.8436018957345972 Precision: 0.5128205128205128 Recall: 0.2 F1score: 0.287
7697841726619
```