In [2]:	Import modules
	<pre>import numpy as np import pandas as pd import seaborn as sns</pre>
	<pre>import matplotlib.pyplot as plt from sklearn import preprocessing, svm from sklearn.preprocessing import StandardScaler from sklearn.model_selection import train_test_split from sklearn.linear model import LinearDegragation</pre>
	<pre>from sklearn.linear_model import LinearRegression import xgboost as xg from sklearn.metrics import mean_squared_error as MSE from sklearn.metrics import make_scorer from sklearn.neural network import MLPRegressor</pre>
	<pre>from sklearn.model_selection import KFold from sklearn.model_selection import cross_val_score from sklearn.pipeline import Pipeline</pre>
	<pre>from sklearn.pipeline import FeatureUnion from sklearn.decomposition import PCA from sklearn.feature_selection import SelectKBest from tabulate import tabulate</pre>
	/Users/kobr0v/opt/anaconda3/lib/python3.9/site-packages/xgboost/compat.py:36: FutureWarning: pandas.Int64Index is de precated and will be removed from pandas in a future version. Use pandas.Index with the appropriate dtype instead.
In [3]:	<pre>from pandas import MultiIndex, Int64Index  warnings.filterwarnings('ignore') def rmse(predict, actual):  graph = np graph(MCE(predict actual))</pre>
	<pre>score = np.sqrt(MSE(predict, actual)) return score  RMSE = make_scorer(rmse)</pre>
	Read the file and prepare the dataset
In [229	<pre>file="large_database.xlsx" database = pd.read_excel(file, usecols=[1,2,3,4,5,11, 9, 12]) X = pd.read_excel(file, usecols=[1,2,3,4,5])</pre>
	<pre>y = pd.read_excel(file, usecols=[9,11,12]) y1= pd.read_excel(file, usecols=[9]) y2= pd.read_excel(file, usecols=[12]) X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, shuffle=True, random_state=20)</pre>
	<pre>y1_train=y_train.iloc[:,0] y2_train=y_train.iloc[:,2] y1_test=y_test.iloc[:,0]</pre>
	p2_test=y_test.iloc[:,2]  Descriptive statistics
In [230	database.describe()
Out[230]:	count 48.000000 48.000000 48.000000 48.000000 48.000000 48.000000 48.000000 48.000000
	mean       1373.485571       971.240832       44.469771       0.038613       8.298153       40.348828       0.042395       16.659639         std       612.312042       177.198329       34.831949       0.035407       1.983831       34.406977       0.035242       5.506392         min       293.393004       665.105557       2.672223       0.001253       2.191926       0.215625       0.000143       7.771517
	25%       912.240196       850.297525       19.104834       0.010301       7.220885       18.377344       0.018851       12.799274         50%       1254.637830       944.967107       32.599803       0.028017       8.151168       28.523437       0.029917       15.544078
	<b>75</b> % 1747.808000 1076.448962 62.377415 0.057959 9.749421 55.094531 0.057217 20.164501 <b>max</b> 2911.111661 1503.124779 139.933084 0.182168 12.185881 172.640625 0.158789 32.765113
	Correlation matrix
In [231	<pre>corrmat = database.corr() plt.figure(figsize=(13, 6)) sns.heatmap(corrmat, vmax=1, annot=True, linewidths=.5) plt.wtickg(rotation=30, horizontalalignment="right")</pre>
	<pre>plt.xticks(rotation=30, horizontalalignment="right") plt.show()</pre>
	Vol dig - 1 0.4 0.032 -0.4 0.44 -0.05 -0.1 -0.52 -0.75  Vol PTV46 - 0.4 1 0.027 -0.13 0.11 0.015 -0.14 -0.087
	Overlap Vol - 0.032 0.027 1 0.81 -0.78 0.94 0.93 0.61 -0.50  OV/Vdig0.4 -0.13 0.81 1 -0.83 0.84 0.85 0.84 -0.25
	Distance - 0.44 0.11 -0.78 -0.83 1 -0.78 -0.8 -0.72 -0.00
	V450.05
	Dmoy - 0.52
	Replication of CFJ models
In [232	Linear Regression V45 = F(Overlap Vol)  regressor = LinearRegression() regressor fit(X train["Overlap Vol"], array reghano( 1 1) y1 train array)
	<pre>regressor.fit(X_train["Overlap Vol"].array.reshape(-1,1), y1_train.array) y1_pred = regressor.predict(X_test["Overlap Vol"].array.reshape(-1,1)) r_squared_train = regressor.score(X_train["Overlap Vol"].array.reshape(-1,1),y1_train) r_squared_test = regressor.score(X_test["Overlap Vol"].array.reshape(-1,1),y1_test)</pre>
	<pre>print("Training coefficient of determination R2: ",r_squared_train) print("Testing coefficient of determination R2: ",r_squared_test) plt.scatter(X_train["Overlap Vol"].array.reshape(-1,1), y1_train,color='g') plt.plot(X_test["Overlap Vol"].array.reshape(-1,1), y1_pred,color='k')</pre>
Out[232]:	Training coefficient of determination R2: 0.8754562517157654 Testing coefficient of determination R2: 0.9501134180471098
	175 -
	125 - 100 - 75 -
	75 - 50 - 25 -
	0 20 40 60 80 100 120 140
	Linear Regression V45 = F(OV/VoI)
In [233	<pre>regressor = LinearRegression() regressor.fit(X_train["OV/Vdig"].array.reshape(-1,1), y1_train.array) y1_pred = regressor.predict(X_test["OV/Vdig"].array.reshape(-1,1)) r_squared_test = regressor.score(X_test["OV/Vdig"].array.reshape(-1,1),y1_test)</pre>
	<pre>r_squared_train = regressor.score(X_train["OV/Vdig"].array.reshape(-1,1),y1_train) print("Training coefficient of determination R2: ",r_squared_train) print("Testing coefficient of determination R2: ",r_squared_test) plt.scatter(X_train["OV/Vdig"].array.reshape(-1,1), y1_train,color='g')</pre>
	<pre>plt.plot(X_test["OV/Vdig"].array.reshape(-1,1), y1_pred,color='k') Training coefficient of determination R2: 0.8086483932469174 Testing coefficient of determination R2: 0.3459264875385393</pre>
Out[233]:	[ <matplotlib.lines.line2d 0x7f98122b7dc0="" at="">]  175 -</matplotlib.lines.line2d>
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	0.000 0.025 0.050 0.075 0.100 0.125 0.150 0.175
	Linear Regression V45 = F(Distance)
In [234	<pre>regressor = LinearRegression() regressor.fit(X_train["Distance"].array.reshape(-1,1), y1_train.array) y1_pred = regressor.predict(X_test["Distance"].array.reshape(-1,1))</pre>
	<pre>r_squared_test = regressor.score(X_test["Distance"].array.reshape(-1,1),y1_test) r_squared_train = regressor.score(X_train["Distance"].array.reshape(-1,1),y1_train) print("Training coefficient of determination R2: ",r_squared_train) print("Testing coefficient of determination R2: ",r_squared_test)</pre>
	<pre>plt.scatter(X_train["Distance"].array.reshape(-1,1), y1_train,color='g') plt.plot(X_test["Distance"].array.reshape(-1,1), y1_pred,color='k')  Training coefficient of determination R2: 0.6743918380270656</pre> Training coefficient of determination R2: 0.6743918380270656
Out[234]:	Testing coefficient of determination R2: 0.3795024601681817 [ <matplotlib.lines.line2d 0x7f97f808bdf0="" at="">]</matplotlib.lines.line2d>
	150 -
	100 - 75 - 50 -
	25 - 0 -
	Multiple Linear Regression V45 = F(Overlap Vol, Distance)
In [235	<pre>regressor = LinearRegression() regressor.fit(X_train[["Overlap Vol","Distance"]], y1_train.array)</pre>
	<pre>y1_pred = regressor.predict(X_test[["Overlap Vol","Distance"]]) r_squared_test = regressor.score(X_test[["Overlap Vol","Distance"]],y1_test) r_squared_test = regressor.score(X_train[["Overlap Vol","Distance"]],y1_train) print("Training coefficient of determination R2: ",r_squared_train)</pre>
	<pre>print("Testing coefficient of determination R2: ",r_squared_test)</pre>
	Training coefficient of determination R2: 0.6743918380270656  Testing coefficient of determination R2: 0.883601176377753
	Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model
In [236	Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKBest(k=3)))
In [236	<pre>Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKBest(k=3))) feature_union = FeatureUnion(features) # create pipeline estimators = [] estimators.append(('standardize', StandardScaler()))</pre>
In [236	<pre>Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKBest(k=3))) feature_union = FeatureUnion(features) # create pipeline estimators = [] estimators.append(('standardize', StandardScaler())) estimators.append(('feature_union', feature_union)) estimators.append(('linear_regression', LinearRegression())) PCR = Pipeline(estimators) kfold = KFold(n_splits=3)</pre>
	<pre>Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKBest(k=3))) feature_union = FeatureUnion(features) # create pipeline estimators = [] estimators.append(('standardize', StandardScaler())) estimators.append(('feature_union', feature_union)) estimators.append(('linear_regression', LinearRegression())) PCR = Pipeline(estimators) kfold = KFold(n_splits=3) PCR_results1 = cross_val_score(PCR, X, y1, cv=kfold, scoring=RMSE) PCR_results2 = cross_val_score(PCR, X, y2, cv=kfold, scoring=RMSE)</pre> PCR_fit(X,y1)
	<pre>Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKBest(k=3))) feature_union = FeatureUnion(features) # create pipeline estimators = [] estimators.append(('standardize', standardScaler())) estimators.append(('feature_union', feature_union)) estimators.append(('linear_regression', LinearRegression())) PCR = Pipeline(estimators) kfold = KFold(n_splits=3) PCR_results1 = cross_val_score(PCR, X, y1, cv=kfold, scoring=RMSE) PCR_results2 = cross_val_score(PCR, X, y2, cv=kfold, scoring=RMSE)  PCR.fit(X,y1) out1 = PCR.predict(X) PCR.fit(X,y2) out2 = PCR.predict(X) print('RMSE Y1: ", rmse(out1,y1))</pre>
	<pre>Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKBest(k=3))) feature_union = FeatureUnion(features) # create pipeline estimators = [] estimators.append(('standardize', StandardScaler())) estimators.append(('feature_union', feature_union)) estimators.append(('linear_regression', LinearRegression())) PCR = Pipeline(estimators) kfold = KFold(n_splits=3) PCR_results1 = cross_val_score(PCR, X, y1, cv=kfold, scoring=RMSE) PCR_results2 = cross_val_score(PCR, X, y2, cv=kfold, scoring=RMSE)  PCR.fit(X,y1) out1 = PCR.predict(X) PCR.fit(X,y2) out2 = PCR.predict(X)</pre>
	<pre>Testing coefficient of determination R2: 0.883601176377753  Creating Principal Component Regression model  # create feature union features = [] features.append(('pca', PCA(n_components=3))) features.append(('select_best', SelectKRest(k=3))) feature union = FeatureUnion(features) # create pipeline estimators = [] estimators.append(('feature_union', feature_union)) estimators.append(('feature_union', feature_union))) PCR = Pipeline(estimators) Kfold = KFold(n_splits=3) PCR_results1 = cross_val_score(PCR, X, y1, cv=kfold, scoring=RMSE) PCR_results2 = cross_val_score(PCR, X, y2, cv=kfold, scoring=RMSE)  PCR.fit(X,y1) out1 = PCR.predict(X) PCR.fit(X,y2) out2 = PCR.predict(X) print("RMSE Y1: ", rmse(out1,y1)) print("RMSE Y2: ", rmse(out2,y2)) pd.DataFrame(out2).to_clipboard()</pre>
	Creating Principal Component Regression model  # create feature union features = [] features.append(('pea', PCA(n_components=3))) features.append(('select_best', SelectRBest(k=3))) feature union = PeatureUnion(features) # create pipeline estimators = [] estimators.append(('standardize', StandardScaler())) estimators.append(('feature_union', feature_union)) eStimators.append(('linear_regression', LinearRegression())) PCR = Pipeline(estimators) kfold = KFold(n_splits=3) PCR_results1 = cross_val_score(PCR, X, y1, cv=kfold, scoring=RMSE) PCR_results2 = cross_val_score(PCR, X, y2, cv=kfold, scoring=RMSE)  PCR.fit(X,y1) out1 = PCR.predict(X) PCR.fit(X,y2) out2 = PCR.predict(X) print("RMSE Y1: ", rmse(out1,y1)) print("RMSE Y1: ", rmse(out2,y2)) pd.DataFrame(out2).to_clipboard()  RMSE Y1: 9.97613459288511 RMSE Y2: 2.673158223016091
In [237	Creating Principal Component Regression model  # create feature union features = [] features.append(('poa', FCA(n_components=3))) features.append(('select_bost', SelectRest(k=3))) features.append(('select_bost', SelectRest(k=3))) feature_union = PeatureUnion(features) # create pipeline estimators.append(('standardize', StandardScaler())) estimators.append(('feature_union', feature_union)) estimators.append(('linear_regression', LinearRegression())) PCR: = Pipeline(estimators) kfold = Krold(n_splits=3) PCR_results1 = cross_val_score(PCR, X, y1, cv=kfold, scoring=RMSE) PCR_results2 = cross_val_score(PCR, X, y2, cv=kfold, scoring=RMSE) PCR_fit(X,y1) out1 = PCR.predict(X) PCR.fit(X,y2) out2 = PCR.predict(X) PCR.fit(X,y2) out2 = PCR.predict(X) print("RMSE Y1: *, rnse(out1,y1)) print("RMSE Y1: *, rnse(out2,y2)) pd.DataFrame(out2).to_clipboard() RMSE Y1: 9.97613459288511 RNSE Y2: 2.673158223016091  Select ML models to be cross validated  # prepare models models = [] models.append(('\sub ', LinearRegression())) models.append(('\sub ', K), XGBRegressor()))
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