```
In [ ]:
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-pytho
# For example, here's several helpful packages to load
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
from imblearn.under sampling import NearMiss
from sklearn.feature selection import SelectKBest
from sklearn.feature selection import f classif
from numpy import set printoptions
from sklearn.model selection import train test split
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import GradientBoostingClassifier
import xgboost as xgb
from sklearn import svm
from sklearn import tree
from sklearn.feature selection import SelectPercentile, f classif
import matplotlib.pyplot as plt
from sklearn.neural network import MLPClassifier
from sklearn.decomposition import PCA
from sklearn.metrics import accuracy score
from sklearn import metrics
import lime.lime tabular as lm#the type of LIIME analysis we'll do
import shap #SHAP package
from sklearn.inspection import permutation importance
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files
under the input directory
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
       print(os.path.join(dirname, filename))
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserve
d as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of
the current session
In [ ]:
data antibody = pd.read csv("/kaggle/input/data-antibody/data antibody.csv")
data pcr= pd.read csv("/kaggle/input/pcr-data/pcr data.csv")
data both= pd.read csv("/kaggle/input/both-covid-data/both covid data.csv")
In [ ]:
def RandomForest classif(x_train,y_train):
    #Classification
    clf= RandomForestClassifier()
    clf=clf.fit(x train, y train)
    return clf
def DecisionTree classif(x train, y train):
   #Classification
   clf = tree.DecisionTreeClassifier()
   clf = clf.fit(x train, y train)
   return clf
def mpl classif(x train, y train):
   clf = MLPClassifier(max iter=300,solver='lbfgs', alpha=1e-5, random state=42)
    clf=clf.fit(x train, y train)
   return clf
```

def gb classif(x train,y train):

```
param dist = {'n estimators':500, 'max depth':5}
    clf=GradientBoostingClassifier(**param dist)
    clf=clf.fit(x train, y train)
    return clf
def xgb classif(x train, y train):
    param_dist = {'n_estimators':300,'max_depth':9,'min_child_weight': 2}
    clf = xgb.XGBClassifier(**param dist)
    return clf.fit(x train, y train)
In [ ]:
def calculate feature importance(model) :
    feat_importances = pd.Series(model.feature_importances , index=x.columns)
    # determine 6 most important features
    return feat importances.nlargest(len(x.columns))
In [ ]:
def plot feature importance (importance, color importace, classif name):
    indices =importance.values
    features = importance.index
    plt.title(classif name+'\nTop Feature Importances')
    plt.barh(range(len(indices)), indices, color=color importace, align='center')
    plt.yticks(range(len(indices)), features)
    plt.xlabel('Relative Importance')
    plt.show()
data_antibody
In [ ]:
x=data antibody.iloc[:,:10]
y=data antibody['Class']
In [ ]:
model rf= RandomForest classif(x,y)
importance rf=calculate feature importance(model rf)
plot feature importance(importance rf,'pink','Random Forest')
In [ ]:
importance rf
In [ ]:
model dt= DecisionTree classif(x,y)
importance dt=calculate feature importance (model dt)
plot feature importance (importance dt, 'purple', 'Decision Tree')
In [ ]:
importance_dt
In [ ]:
model gb=gb classif(x,y)
importance gb=calculate feature importance(model gb)
plot feature importance(importance gb,'green','Gradient Boosting')
In [ ]:
importance gb
```

In []:

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model xgb=xgb classif(x,y)
importance xgb=calculate feature importance(model xgb)
plot feature importance(importance xgb, 'red', 'Xgboos')
In [ ]:
importance xgb
In [ ]:
def mlp feature importance(x,y):
    X_train, X_val, y_train, y_val = train_test_split(
            x, y, test_size=0.3, random_state=0, stratify=y)
    model=mpl_classif(X_train,y_train)
    r = permutation importance(model, X val, y val, n repeats=50, random state=0)
    for i in r.importances mean.argsort()[::-1]:
            print(f"{X train.columns[i]:<8}" f"{r.importances mean[i]:.3f}"</pre>
                  f" +/- {r.importances std[i]:.3f}")
    sorted idx = r.importances mean.argsort()
    fig, ax = plt.subplots()
    ax.boxplot(r.importances[sorted idx].T,
               vert=False, labels=X train.columns[sorted idx])
    ax.set title("Permutation Importances (train set)")
    fig.tight layout()
    plt.show()
mlp feature importance (x, y)
pcr
In [ ]:
x pcr=data pcr.iloc[:,:10]
y_pcr=data_pcr['Class']
In [ ]:
model rf= RandomForest classif(x pcr,y pcr)
importance rf=calculate feature importance (model rf)
plot feature importance(importance rf,'pink','Random Forest')
In [ ]:
importance rf
In [ ]:
model dt= DecisionTree classif(x_pcr,y_pcr)
importance dt=calculate feature importance (model dt)
plot feature importance(importance dt, 'purple', 'Decision Tree')
In [ ]:
importance dt
In [ ]:
model gb=gb classif(x_pcr,y_pcr)
importance gb=calculate feature importance (model gb)
plot feature importance(importance gb, 'green', 'Gradient Boosting')
In [ ]:
importance_gb
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In []:

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model xgb=xgb classif(x pcr,y pcr)
importance xgb=calculate feature importance(model xgb)
plot feature importance(importance xgb, 'red', 'Xgboos')
In [ ]:
importance xgb
In [ ]:
mlp feature importance(x pcr, y pcr)
both data
In [ ]:
x_both=data_both.iloc[:,:10]
y both=data both['Class']
In [ ]:
model rf= RandomForest classif(x both, y both)
importance rf=calculate feature importance(model rf)
plot feature importance(importance rf, 'pink', 'Random Forest')
In [ ]:
importance rf
In [ ]:
model dt= DecisionTree classif(x both, y both)
importance dt=calculate feature importance(model dt)
plot feature importance(importance dt, 'purple', 'Decision Tree')
In [ ]:
importance dt
In [ ]:
model gb=gb classif(x_both,y_both)
importance gb=calculate feature importance (model gb)
plot feature importance(importance gb, 'green', 'Gradient Boosting')
In [ ]:
importance gb
In [ ]:
model xgb=xgb classif(x both, y both)
importance xgb=calculate feature importance(model xgb)
plot feature importance(importance xgb, 'red', 'Xgboos')
In [ ]:
importance xgb
In [ ]:
mlp_feature_importance(x_both,y_both)
```