## **#Red Wine Quality Dataset**

from google.colab import files

```
uploaded = files.upload()
import pandas as pd
wn = pd.read csv('winequality2.csv')
wn.head()
wn.isnull().values.any()
wn.describe()
x = wn.drop('quality', axis = 1)
y = wn['quality']
from sklearn.model selection import train test split
from sklearn.pipeline import Pipeline
from sklearn.feature selection import VarianceThreshold
from imblearn.over sampling import SMOTE
import numpy as np
from sklearn.datasets import make classification
rng = np.random.RandomState(42)
x, y = make_classification(random_state=rng)
oversample=SMOTE()
wn=oversample.fit resample(x,y)
import sklearn.model selection
!pip install scikit-optimize
from skopt import BayesSearchCV
x train, x test, y train, y test = sklearn.model selection.train test split(x, y,
random state=42, stratify=y)
from sklearn.ensemble import RandomForestClassifier
pipe = Pipeline([
  ('selector', VarianceThreshold()),
  ('model', RandomForestClassifier())
1)
param grid = {
  'selector threshold': (0,0.1),
  'model': [RandomForestClassifier()],
  'model max depth': (1,500),
  'model max leaf nodes': (10,100),
```

```
'model max samples': (0.1,0.8),
  'model min samples leaf': (1,6),
  'model min samples split': (2,8),
  'model n estimators': (50,700),
  'model__n_jobs': (1,400)
}
opt = BayesSearchCV(
  pipe,
  # (parameter space, # of evaluations)
  [(param grid, 500)],
  cv=2)
opt.fit(x train, y train)
print("test score: %s" % opt.score(x test, y test))
clf = RandomForestClassifier(random_state=42)
clf = clf.fit(x train, y train)
y hat = clf.predict(x test)
print("RF Accuracy", sklearn.metrics.accuracy_score(y_test, y_hat))
# Brain Stroke Dataset
from google.colab import files
uploaded = files.upload()
import pandas as pd
st = pd.read csv('brain stroke.csv')
st.head()
print(st['gender'].unique())
st['gender'].value counts()
print(st['ever married'].unique())
st['ever married'].value counts()
print(st['work type'].unique())
st['work type'].value counts()
print(st['Residence type'].unique())
st['Residence type'].value counts()
print(st['smoking status'].unique())
st['smoking status'].value counts()
ohed = pd.get dummies(st, columns = ['gender', 'ever married', 'work type',
'Residence type', 'smoking status'])
ohed.head()
```

```
ohed.describe()
from sklearn.model selection import train test split
x = ohed.drop('stroke', axis = 1)
y = ohed['stroke']
import numpy as np
from sklearn.datasets import make classification
from imblearn.over sampling import SMOTE
oversample=SMOTE()
ohed=oversample.fit resample(x,y)
import sklearn.model selection
x train, x test, y train, y test = sklearn.model selection.train test split(x, y,
random state=42, stratify=y)
from sklearn.ensemble import RandomForestClassifier
!pip install scikit-optimize
from skopt import BayesSearchCV
from sklearn.pipeline import Pipeline
from sklearn.feature selection import VarianceThreshold
x train, x test, y train, y test = sklearn.model selection.train test split(x, y,
random state=42, stratify=y)
pipe = Pipeline([
  ('selector', VarianceThreshold()),
  ('model', RandomForestClassifier())
])
rfc search = {
  'selector threshold': (0,0.1),
  'model':([RandomForestClassifier()]),
  'model max depth': (1,500),
  'model max leaf nodes': (10,100),
  'model max samples': (0.1,0.8),
  'model min samples leaf': (1,6),
   'model min samples split': (2,8),
  'model n estimators': (50,700),
  'model n jobs': (-1,400)
}
opt = BayesSearchCV(
  pipe.
  # (parameter space, # of evaluations)
  [(rfc search, 50)],
  cv=3)
```

opt.fit(x\_train, y\_train)
print("test score: %s" % opt.score(x\_test, y\_test))
clf = RandomForestClassifier(random\_state=42)
clf = clf.fit(x\_train, y\_train)
y\_hat = clf.predict(x\_test)
print("RF Accuracy", sklearn.metrics.accuracy\_score(y\_test, y\_hat))

Table 1: Red Wine Quality Pipeline	
Mean Accuracy Score Before Bayesian Optimization	0.88
Mean Accuracy Score After Bayesian Optimization	0.88

Table 2: Brain Stroke Pipeline	
Mean Accuracy Score Before Bayesian Optimization	0.9478
Mean Accuracy Score After Bayesian Optimization	0.9502