```
Step-1
```

import os

import numpy as np

import pandas as pd

import shap

shap.initjs()

import time

import warnings

warnings.filterwarnings('ignore')

from sklearn.model_selection import train_test_split

from sklearn import preprocessing

from sklearn.neighbors import KNeighborsClassifier

from sklearn.linear_model import LogisticRegression

from sklearn.naive_bayes import MultinomialNB

from sklearn.ensemble import ExtraTreesClassifier

from sklearn.svm import SVC

from sklearn.ensemble import RandomForestClassifier

from sklearn.ensemble import GradientBoostingClassifier

from sklearn.feature_selection import f_classif

from sklearn.feature_selection import VarianceThreshold

from sklearn.feature_selection import mutual_info_classif

from sklearn.feature_selection import SelectKBest

from mlxtend.feature_selection import ExhaustiveFeatureSelector as EFS

from mlxtend.feature_selection import SequentialFeatureSelector as SFS

from sklearn import metrics

from sklearn.metrics import confusion_matrix, accuracy_score, classification_report svc=SVC(probability=True, kernel='linear')

start = time.time()

Step-2

d_1=pd.read_excel(r"training_file.xlsx")

d_2=pd.read_excel(r"testing_file.xlsx")

Step-3

 $x_{tr} = d_{1.iloc[:,:-1]}$

 $y_{tr1} = d_{1.iloc[:,-1:]}$

x_ts = d_2.iloc[:,:-1]

y_ts1 = d_2.iloc[:,-1:]

Step-4

 $x_t = d_1[selected features from fusion]$

y_tr1= d_1.iloc[:,-1:]

 $x_ts = d_2[selected features from fusion]$

y_ts1= d_1.iloc[:,-1:]

Sten-5

x_tr=d_1[selected feature from SHAP]

y_tr1= d_1.iloc[:,-1:]

x_ts= d_2 [selected feature from SHAP]

y_ts1= d_2.iloc[:,-1:]

Step-6

x_tr= d_1[fusion and then SHAP to selected feature]

y_tr1= d_1.iloc[:,-1:]

 $x_ts = d_2$ [fusion and then SHAP to selected feature]

y_ts1= d_1.iloc[:,-1:]

```
Step-7
(x_{train}, y_{train}), (x_{test}, y_{test}) = (x_{tr}, y_{tr1}), (x_{ts}, y_{ts1})
model = SVC()
model = KNeighborsClassifier()
model = MultinomialNB()
model = LogisticRegression()
model = ExtraTreesClassifier()
history = model.fit(x_train, y_train)
y_pred = model.predict(x_test)
classification_report = classification_report(y_test, y_pred, output_dict=True)
end = time.time()
print(end - start, "seconds")
Step-8
explainer = shap.TreeExplainer(model)
shap_values = explainer.shap_values(x_train)
shap.summary_plot(shap_values, x_train.values, max_display=10, plot_type="bar", class_names= label,
feature_names = x_train.columns)
Step-9
feature_names = x_train.columns)
f_value = f_classif(x_train, y_train)
feature_data = [(feature, value) for feature, value in zip(feature_names, f_value[0])]
sorted_feature_data = sorted(feature_data, key=lambda x: x[1], reverse=True)
df = pd.DataFrame(sorted_feature_data, columns=['Feature', 'Value'])
selector = VarianceThreshold()
selector.fit_transform(x_train)
feature_data = [(feature, variance) for feature, variance in zip(feature_names, selector.variances_)]
sorted_feature_data = sorted(feature_data, key=lambda x: x[1], reverse=True)
df = pd.DataFrame(feature_data, columns=['Feature', 'Variance'])
MI_score = mutual_info_classif(X_data, y_data, random_state=0)
feature_data = list(zip(feature_names, MI_score))
df = pd.DataFrame(feature_data, columns=['Feature', 'MI Score'])
skb = SelectKBest(score_func=f_classif, k=10)
x_train_new = skb.fit_transform(x_train, y_train)
selected_features = [feature_names[index] for index in skb.get_support(indices=True)]
df = pd.DataFrame({'Selected Features': selected_features})
sfs = SFS(estimator=lr, k_features=(1, 10), forward=True, scoring='accuracy', cv=5)
sfs = sfs.fit(x_train, y_train)
X_data_new = sfs.transform(X_data)
sbs_results = pd.DataFrame.from_dict(sbs.subsets_).T
sbs = SFS(estimator=lr, k_features=(1, 10),forward=False, scoring='accuracy', cv=5)
sbs = sbs.fit(x_train, y_train)
X_data_new = sbs.transform(X_data)
sbs_results = pd.DataFrame.from_dict(sbs.subsets_).T
model = RandomForestClassifier()
history = model.fit(x_train, y_train)
y_pred = model.predict(x_test)
feat_importances = pd.Series(model.feature_importances_, index= x_train.columns)
rf=feat_importances.nlargest(10)
model1= GradientBoostingClassifier()
history = model1.fit(x_train, y_train)
y_pred = model1.predict(x_test)
feature_importances = pd.Series(model1.feature_importances_, index= x_train.columns)
gb=feature_importances.nlargest(10)
```

Follow the instructions: Here 4 Baseline model scenarios

- B1: Performance Analysis of Machine Learning Classifiers
- B2: Performance Analysis of Fusion-based Feature Selection
- B3: Analysis of XAI-based Feature Selection
- B4: Performance Analysis of sXFS

Follow the steps for B1:

- 1. Run the step 1
- 2. Run the step 2
- 3. Run the step 3
- 4. Run the step 7

Here choose the best model for the further case on the basis of highest accuracy.

Follow the steps for B2:

- 1. Run the step 1
- 2. Run the step 2
- 3. Run the step 3
- 4. Run the step 9

Make fusion of 10 most influential feature

- 5. Run the step 4
- 6. Run the step 7

Follow the steps for B3:

- 1. Run the step 1
- 2. Run the step 2
- 3. Run the step 3
- 4. Run the step 7
- 5. Run the step 8

Select most 50% feature based on model interpretability.

- 6. Run the step 5
- 7. Run the step 7

Follow the steps for B4:

- 1. Run the step 1
- 2. Run the step 2
- 3. Run the step 3
- 4. Run the step 9

Make fusion of 10 most influential feature

- 5. Run the step 4
- 6. Run the step 7
- 7. Run the step 8

Select most 50% feature based on model interpretability.

- 8. Run the step 6
- 9. Run the step 9