

```
In [2]: import joblib
import pandas as pd
from sklearn.metrics import f1_score
```

Function-1

```
In [3]: def final_fun_1( x ):
        """
        This function performs data preprocessing
        and makes final predictions and
        returns the class labels
        """

        # Load preprocessing models
        median_imputer = joblib.load("../median_imputer.pkl")
        mice_imputer = joblib.load("../mice_imputer.pkl")

        # Drop features with max null values
        x = x.drop(['br_000', 'bq_000', 'bp_000', 'bo_000', 'ab_000', 'cr_000', 'bn_000', 'cd_000'], axis=1)

        # Specify features whose missing values are imputed using Median Imputer
        median_features = ['ak_000', 'ca_000', 'dm_000', 'df_000', 'dg_000', \
                           'dh_000', 'dl_000', 'dj_000', 'dk_000', 'eb_000', \
                           'di_000', 'ac_000', 'bx_000', 'cc_000']

        # Median Imputation
        x[median_features] = median_imputer.transform(x[median_features])

        # MICE Imputation
        x = pd.DataFrame(data = mice_imputer.transform(x), columns = x.columns)

        # Load our Best Model
        model = joblib.load("../gbdt_model.pkl")

        # Predict class label
        y = model.predict(x)

        return y
```

Function-2

```
In [11]: def final_fun_2( x , y ):
        """
        This function performs data preprocessing,
        makes final predictions and returns the computed
        performance metric and class labels
        """

        # Load preprocessing models
        median_imputer = joblib.load("../median_imputer.pkl")
        mice_imputer = joblib.load("../mice_imputer.pkl")

        # Drop features with max null values
        x = x.drop(['br_000', 'bq_000', 'bp_000', 'bo_000', 'ab_000', 'cr_000', 'bn_000', 'cd_000'], axis=1)

        # Specify features whose missing values are imputed using Median Imputer
        median_features = ['ak_000', 'ca_000', 'dm_000', 'df_000', 'dg_000', \
                           'dh_000', 'dl_000', 'dj_000', 'dk_000', 'eb_000', \
                           'di_000', 'ac_000', 'bx_000', 'cc_000']

        # Median Imputation
        x[median_features] = median_imputer.transform(x[median_features])

        # MICE Imputation
        x = pd.DataFrame(data = mice_imputer.transform(x), columns = x.columns)

        # Load our Best Model
        model = joblib.load("../gbdt_model.pkl")

        # Predict Class Labels
        y_pred = model.predict(x)

        # Return Performance Metric
        return f1_score(y, y_pred, average='macro'), y_pred
```

Testing both functions on test dataset

```
In [12]: data = pd.read_csv("../aps_failure_test_set.csv", skiprows=20, na_values=["na"])

def get_correct_label(y):
    """
    This function converts the class labels
    from 'neg' and 'pos' to 0 and 1 respectively
    """
    return y.replace(['neg', 'pos'], [0, 1])

print(data['class'].unique())
data['class'] = get_correct_label(data['class'])
print(data['class'].unique())

y_test = data['class']
x_test = data.drop('class', axis=1)

['neg' 'pos']
[0 1]
```

```
In [13]: F1_test , y_pred = final_fun_2(x_test, y_test)
print("Macro-F1 Score: ", F1_test)

Macro-F1 Score:  0.8817895552269138
```