Import Packages

```
import numpy as np
import pandas as pd
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import make column transformer
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import make pipeline
from sklearn.model selection import StratifiedKFold
from sklearn.model selection import cross val score
from sklearn.linear model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from keras.wrappers.scikit learn import KerasClassifier
from keras.models import Sequential
from keras.layers import Dense, Dropout
from imblearn.over sampling import SMOTE
from statistics import mode
import warnings
warnings.filterwarnings('ignore')
```

Feature Builder Utility

def feature builder(inp data):

```
# Remove Numeric Features outliers based on the EDA
inp_data["age"] = np.where(inp_data["age"] > 71, 71, inp_data['age'])
inp_data["duration"] = np.where(inp_data["duration"] > 1260, 1260, inp_data['duration"]
inp_data["campaign"] = np.where(inp_data["campaign"] > 14, 14, inp_data['campaign'])
inp_data["previous"] = np.where(inp_data["previous"] > 3, 3, inp_data['previous'])
# Column Transformers
 # Scaling for Numeric Columns
 # One Hot Encoding for Categoriccal Features
numerical_features = ['age', 'duration', 'campaign', 'previous']
categorical features = [ 'job', 'marital', 'education', 'default', 'housing', 'loan
feature_transformer = make_column_transformer((StandardScaler(), numerical_features
# Create Feature Data
feature array = feature transformer.fit transform(inp data)
feature cols = list(feature_transformer.get_feature_names_out())
feature_cols = [col.replace('standardscaler__', '').replace('onehotencoder ', '').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').replace(').r
feature data = pd.DataFrame.sparse.from_spmatrix(feature_array,columns=feature_cols)
return feature_transformer, feature_data
```

train data = pd.read excel('https://github.com/rsdevanathan/Customer Subscription/blok

Model Training

```
# Remove Duplicates
train_data = train_data.drop_duplicates()
#Encode Target Variable
train data = train data.replace({'y': {'yes': 1,
                                 'no': 0}})
X_train = train_data.drop(columns = 'y')
y train = train data[['y']]
cw = {0:1,1:np.count_nonzero(y_train==0)/np.count_nonzero(y_train==1)}
feature transformer pipeline, X train = feature builder(X train)
smote_sampler = SMOTE(random_state=14)
columns = X train.columns
arr X, arr y = smote sampler.fit resample(X train.values, y train)
sampled X = pd.DataFrame(data=arr X,columns=X train.columns)
```

sampled_y = pd.DataFrame(data=arr_y,columns=['y'])

print("Training AUC:", np.mean(cv_auc))

Training AUC: 0.9266264073790428

ml lr = LogisticRegression(max iter=1000) ml lr.fit(sampled X, sampled y)

Logistic Regression

```
kfold = StratifiedKFold(n_splits=3, shuffle=True)
 cv_auc = cross_val_score(ml_lr, X_train.values, y_train.values, scoring='roc_auc',cv=l
 print("Training AUC:", np.mean(cv auc))
Training AUC: 0.9126389285216318
Random Forest
```

ml_rf=RandomForestClassifier(random_state=14,class_weight = cw,max_depth=100,max_feath

cv_auc = cross_val_score(ml_rf, X_train.values, y_train.values, scoring='roc_auc',cv=l

ml_rf.fit(X_train, y_train) kfold = StratifiedKFold(n_splits=3, shuffle=True)

In [74]:

Neural Network

```
def create nn():
  model = Sequential()
  model.add(Dense(32, input dim=57, activation='relu'))
  model.add(Dense(16, activation='relu'))
  model.add(Dense(1, activation='sigmoid'))
  model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['binary_accurae
  return model
dl cfr cw = KerasClassifier(build fn=create nn, epochs=100, batch size=1028, verbose=0
dl cfr cw.fit(X train, y train, class weight = cw)
kfold = StratifiedKFold(n splits=3, shuffle=True)
cv auc = cross val score(dl cfr cw, X train.values, y train.values, scoring='roc auc',
print("Training AUC:", np.mean(cv auc))
Training AUC: 0.9211359915936056
```

Name: Final Prediction, dtype: int64

Model Prediction

```
# Read Input Files
test data = pd.read excel('https://github.com/rsdevanathan/Customer Subscription/blob,
test data["age"] = np.where(test data["age"] > 71, 71, test data['age'])
test data["duration"] = np.where(test data["duration"] > 1260, 1260, test data['duration']
test data["campaign"] = np.where(test_data["campaign"] > 14, 14, test_data['campaign'
test_data["previous"] = np.where(test_data["previous"] > 3, 3, test_data['previous'])
test_features = feature_transformer_pipeline.transform(test_data)
lr_pred = ml_lr.predict(test_features)
rf_pred = ml_rf.predict(test_features)
xgb pred = dl cfr cw.predict(test features)
```

```
prediction_data = test_data.copy()
prediction_data['LR_Prediction'] = lr_pred
prediction data['RF Prediction'] = rf pred
prediction data['XGB Prediction'] = xgb pred
```

```
prediction_data['Final_Prediction'] = np.where((prediction_data['LR_Prediction']+prediction')
prediction data['Final Prediction'].value counts()
```

prediction_data.to_csv('Final_Prediction.csv')

End

Out[90]: 0