Load data

```
In [2]:
import numpy as np
import pandas as pd
from sklearn.linear_model import SGDClassifier
from sklearn.model selection import GridSearchCV
from sklearn.model selection import RandomizedSearchCV
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
import pickle
from sklearn.linear_model import LogisticRegression
import xgboost as xgb
from scipy import stats
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import LinearSVC
from scipy.stats import uniform
from sklearn.calibration import CalibratedClassifierCV
from sklearn.svm import SVC
import scipy
from catboost import CatBoostClassifier
In [18]:
# loading train data
train = pd.read_csv('/content/drive/MyDrive/train.csv')
# loading test data
test = pd.read_csv('/content/drive/MyDrive/test.csv')
In [19]:
Y = train['ACTION']
X = train[train.columns.difference(['ACTION'])]
In [20]:
X test=test.drop(columns=['id'],axis=1)
X test.shape
Out[20]:
(58921, 9)
In [21]:
# loading train/test one hot encoded data
f = open('/content/drive/MyDrive/1_hot_enc.pckl','rb')
X train ohe, X test ohe = pickle.load(f)
f.close()
def generating samples(input data, target data):
    '''In this function, we will write code for generating samples with replacement of size 15k ''
    rows = np.random.choice(len(input data), size=15000, replace=True)
    x = input_data.iloc[rows]
    y = target data.iloc[rows]
    return x, y
```

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In [93]:
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class Classifier (object):
    '''class that trains & predict sklearn classifiers '''
    def __init__ (self, clf, seed=0, params=None):
        params['random_state'] = seed
        self.clf = clf(**params)

def train(self, x_train, y_train):
        self.clf.fit(x_train, y_train)

def predict(self, x):
        return self.clf.predict_proba(x)[:,1]

def feature_names(self):
    # for XGBoost classifier only
    return self.clf.get_booster().feature_names
```

In [103]:

```
class StackingClassifier(object):
    '''Custom stacking-classifier '''
   def __init__(self, models, params):
       self.models = models
       self.params = params
   def train(self, X, Y, test, k):
        # splitting data into D1,D2 (50-50)
       X D1, X D2, Y D1, Y D2 = train test split(X, Y, test size=0.5, stratify=Y, random state=42)
       D2 pred = \{\}
       test pred = {}
       for i in range(k):
           model = self.models[i % len(self.models)]
           param = self.params[i % len(self.params)]
           print(" training classifier "+str(i+1))
            # sampling data
           x1,y1 = generating_samples(X_D1,Y_D1)
           base model = Classifier(model, params=param)
                                                                 # training on sample data
           base model.train(x1,y1)
           D2_pred[i] = base_model.predict(X_D2) # base_model prediction on D2 data
            # base model prediction on test data
           if model==xgb.XGBClassifier:
               t = test[base_model.feature_names()]
               test_pred[i] = base_model.predict(t)
           else:
                test_pred[i] = base_model.predict(test)
        # storing k predictions in a dataframe
       train_data = pd.DataFrame(D2_pred)
       test_data = pd.DataFrame(test_pred)
        # training meta classifier
       lr = LogisticRegression(random state=0, class weight='auto')
       parameter = {'C':[10**i for i in range(-5,5)]}
       clf = GridSearchCV(lr, parameter, scoring='roc auc', return train score=True, n jobs=-1)
hyperparameter tuning using gridsearch
       grid result = clf.fit(train data,Y D2)
       print("Best auc: %f using %s" % (grid_result.best_score_, grid_result.best_params_)) #prin
ting best auc score for best alpha
       c = grid result.best params ['C']
       meta = LogisticRegression(random state=0, class weight='auto', C=c)
       meta.fit(train data,Y D2)
       # final prediction on test data
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predictions = meta.predict proba(test data)[:,1]
                                # returning final classifier and predictions
       return meta, predictions
In [42]:
features = list(range(X.shape[1]))
print(features)
[0, 1, 2, 3, 4, 5, 6, 7, 8]
In [104]:
# list of models and there parameters
models = [CatBoostClassifier,xgb.XGBClassifier,RandomForestClassifier,LogisticRegression]
LR_param = {'C':1,
           'max iter':1000,
           'class_weight':'balanced'}
RF param = { 'max depth': 25,
           'max features': 3,
           'min samples split': 3,
           'n estimators': 437}
XGB param = {'colsample bytree': 0.3732907507832548,
           'learning rate': 0.2620452218743664,
           'max depth': 10,
           'min child weight': 1,
           'n estimators': 309,
           'subsample': 0.9755215050028858}
CAT_param = {'loss_function':'Logloss',
         'eval metric':'AUC',
         'cat_features':features,
          'verbose':False,
         'early stopping rounds': 200}
params = [CAT param, XGB param, RF param, LR param]
In [105]:
stc = StackingClassifier(models,params)
In [106]:
# checking score for different no of base_models
for k in range (2,12):
   print("\n no. of base_learners = ",k)
   meta,predictions = stc.train(X, Y, X_test, k=k)
   print("======="")
   print("======"")
no. of base learners = 2
training classifier 1
{\tt training\ classifier\ 2}
Best auc: 0.848565 using {'C': 0.01}
_____
no. of base_learners = 3
training classifier 1
training classifier 2
training classifier 3
Best auc: 0.856634 using {'C': 10000}
_____
_____
no. of base_learners = 4
training classifier 1
```

```
cramming cracement
training classifier 2
 training classifier 3
 training classifier 4
Best auc: 0.856930 using {'C': 10}
 ______
_____
no. of base learners = 5
training classifier 1
 training classifier 2
 training classifier 3
 training classifier 4
 training classifier 5
Best auc: 0.863643 using {'C': 1}
_____
-----
no. of base learners = 6
 training classifier 1
 training classifier 2
 training classifier 3
 training classifier 4
 training classifier 5
 training classifier 6
Best auc: 0.864614 using {'C': 0.1}
_____
_____
no. of base learners = 7
 {\tt training\ classifier\ 1}
 training classifier 2
 training classifier 3
 training classifier 4
 training classifier 5
 training classifier 6
 training classifier 7
Best auc: 0.864899 using {'C': 1}
_____
_____
 no. of base learners = 8
 training classifier 1
 training classifier 2
 training classifier 3
 training classifier 4
 training classifier 5
 training classifier 6
 training classifier 7
 training classifier 8
Best auc: 0.868887 using {'C': 1}
_____
_____
no. of base learners = 9
 training classifier 1
 training classifier 2
 training classifier 3
 training classifier 4
 training classifier 5
 training classifier 6
 training classifier 7
 training classifier 8
 training classifier 9
Best auc: 0.871383 using {'C': 1}
_____
_____
no. of base learners = 10
  \  \, \text{training classifier 1} \\
 training classifier 2
 training classifier 3
 training classifier 4
 training classifier 5
 training classifier 6
 training classifier 7
 training classifier 8
 training classifier 9
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craining crassifier >
training classifier 10
Best auc: 0.873051 using {'C': 1}
_____
_____
no. of base_learners = 11
training classifier 1
training classifier 2
training classifier 3
training classifier 4
training classifier 5
training classifier 6
training classifier 7
training classifier 8
training classifier 9
training classifier 10
training classifier 11
Best auc: 0.871822 using {'C': 1}
_____
_____
```

In []:

Submission and Description	Private Score	Public Score
stacking_classifier.csv	0.85816	0.85651
just now by ankit chandrakar		

Conclusion

- Got the best private_score of 0.85 with 10 base_models.
- Not that satisfactory result as compared to the single CatBoostClassifier(private_score = 0.908)