```
In [1]: import numpy as np
         import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         %matplotlib inline
 In [2]: from google.colab import drive
         drive.mount('/content/drive')
         Mounted at /content/drive
 In [3]: train = pd.read csv("/content/drive/My Drive/DSN AI Bootcamp Qualification Hackat
         test = pd.read csv("/content/drive/My Drive/DSN AI Bootcamp Qualification Hackath
         sample = pd.read csv("/content/drive/My Drive/DSN AI Bootcamp Qualification Hackage
 In [4]: print(train.shape)
         print(test.shape)
         print(sample.shape)
         (56000, 52)
         (24000, 51)
         (24000, 2)
 In [5]: | submit = test[['Applicant_ID']]
 In [6]: train.fillna(-1, inplace = True)
         test.fillna(-1, inplace = True)
 In [7]: train.replace(np.inf, -1, inplace = True)
         test.replace(np.inf, -1, inplace = True)
 In [8]: |train.drop('Applicant_ID', axis = 1, inplace= True)
         test.drop('Applicant_ID', axis = 1, inplace= True)
 In [9]: train[['form_field47','default_status']] = train[['form_field47','default_status']]
         test['form field47'] = test['form field47'].astype('category')
In [10]: train['form_field47'] = train['form_field47'].cat.codes
         train['default_status'] = train['default_status'].cat.codes
         test['form field47'] = test['form field47'].cat.codes
In [11]: | X = train.drop('default_status', axis = 1)
         y = train['default_status']
```

In [12]: !pip install catboost !pip install lightgbm

Collecting catboost

Downloading https://files.pythonhosted.org/packages/90/86/c3dcb600b4f9e7584ed 90ea9d30a717fb5c0111574675f442c3e7bc19535/catboost-0.24.1-cp36-none-manylinux1 x86_64.whl (https://files.pythonhosted.org/packages/90/86/c3dcb600b4f9e7584ed90 ea9d30a717fb5c0111574675f442c3e7bc19535/catboost-0.24.1-cp36-none-manylinux1 x8 6 64.whl) (66.1MB)

66.1MB 54kB/s

Requirement already satisfied: numpy>=1.16.0 in /usr/local/lib/python3.6/dist-p ackages (from catboost) (1.18.5)

Requirement already satisfied: scipy in /usr/local/lib/python3.6/dist-packages (from catboost) (1.4.1)

Requirement already satisfied: pandas>=0.24.0 in /usr/local/lib/python3.6/distpackages (from catboost) (1.1.2)

Requirement already satisfied: graphviz in /usr/local/lib/python3.6/dist-packag es (from catboost) (0.10.1)

Requirement already satisfied: matplotlib in /usr/local/lib/python3.6/dist-pack ages (from catboost) (3.2.2)

Requirement already satisfied: six in /usr/local/lib/python3.6/dist-packages (f rom catboost) (1.15.0)

Requirement already satisfied: plotly in /usr/local/lib/python3.6/dist-packages (from catboost) (4.4.1)

Requirement already satisfied: python-dateutil>=2.7.3 in /usr/local/lib/python 3.6/dist-packages (from pandas>=0.24.0->catboost) (2.8.1)

Requirement already satisfied: pytz>=2017.2 in /usr/local/lib/python3.6/dist-pa ckages (from pandas>=0.24.0->catboost) (2018.9)

Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.6/di st-packages (from matplotlib->catboost) (1.2.0)

Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /us r/local/lib/python3.6/dist-packages (from matplotlib->catboost) (2.4.7)

Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.6/dist-pa ckages (from matplotlib->catboost) (0.10.0)

Requirement already satisfied: retrying>=1.3.3 in /usr/local/lib/python3.6/dist -packages (from plotly->cathoost) (1.3.3)

Installing collected packages: catboost

Successfully installed catboost-0.24.1

Requirement already satisfied: lightgbm in /usr/local/lib/python3.6/dist-packag es (2.2.3)

Requirement already satisfied: numpy in /usr/local/lib/python3.6/dist-packages (from lightgbm) (1.18.5)

Requirement already satisfied: scikit-learn in /usr/local/lib/python3.6/dist-pa ckages (from lightgbm) (0.22.2.post1)

Requirement already satisfied: scipy in /usr/local/lib/python3.6/dist-packages (from lightgbm) (1.4.1)

Requirement already satisfied: joblib>=0.11 in /usr/local/lib/python3.6/dist-pa ckages (from scikit-learn->lightgbm) (0.16.0)

from sklearn.model selection import StratifiedKFold In [13]:

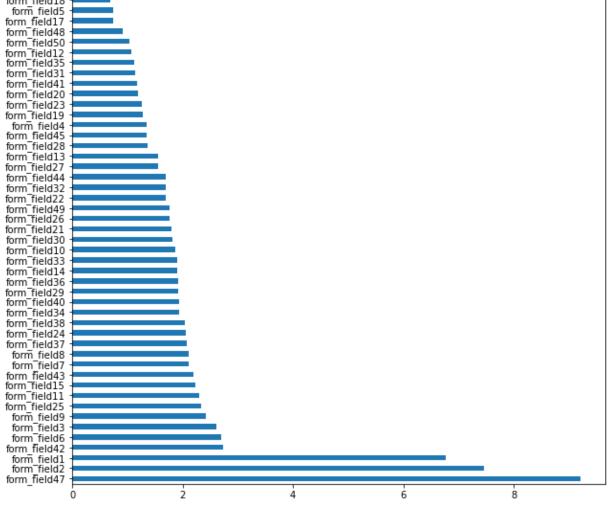
from sklearn.metrics import accuracy score, roc auc score

from catboost import CatBoostClassifier

from lightgbm import LGBMClassifier

```
In [14]: kf = StratifiedKFold(n_splits = 20, random_state = 1)
         /usr/local/lib/python3.6/dist-packages/sklearn/model selection/ split.py:296: F
         utureWarning: Setting a random state has no effect since shuffle is False. This
         will raise an error in 0.24. You should leave random state to its default (Non
         e), or set shuffle=True.
           FutureWarning
In [ ]: # # INITIALIZE THE MODEL
         cat4 = CatBoostClassifier(learning rate = 0.01, max depth = 8, n estimators = 50€
                                 eval metric = 'AUC', bootstrap type = 'Bayesian', random s
In [ ]: | scores40 = []
         scores41 = []
         for fold,(tr in,te in) in enumerate(kf.split(X, y)):
            X_train,X_test = X.iloc[tr_in],X.iloc[te_in]
            y train,y test = y.iloc[tr in],y.iloc[te in]
            cat4.fit(X_train,y_train,eval_set=[(X_train,y_train),(X_test,y_test)],early_s
            y pred = cat4.predict(X test)
            scores40.append(accuracy score(y test,y pred))
            scores41.append(roc_auc_score(y_test,y_pred))
           0:
                test: 0.7684223 test1: 0.7533164
                                                      best: 0.7533164 (0)
                                                                             tota
                        remaining: 7m 16s
         1: 87.4ms
         200:
                test: 0.8429056 test1: 0.8130717
                                                      best: 0.8130717 (200)
                                                                             tota
         1: 12.4s
                        remaining: 4m 56s
                test: 0.8512820 test1: 0.8168956
         400:
                                                      best: 0.8169654 (396)
                                                                             tota
                        remaining: 4m 38s
         1: 24.3s
         600:
                test: 0.8580685 test1: 0.8188417
                                                      best: 0.8188417 (600)
                                                                             tota
                        remaining: 4m 28s
         1: 36.6s
         800:
                test: 0.8641671 test1: 0.8200149
                                                      best: 0.8200382 (792)
                                                                             tota
         1: 48.7s
                        remaining: 4m 15s
         1000:
                test: 0.8697181 test1: 0.8210826
                                                      best: 0.8211261 (993)
                                                                             tota
         1: 1m 1s
                        remaining: 4m 3s
         1200:
                test: 0.8753295 test1: 0.8217794
                                                      best: 0.8218322 (1188)
                                                                             tota
         1: 1m 13s
                        remaining: 3m 53s
         1400:
                test: 0.8809316 test1: 0.8226919
                                                      best: 0.8227121 (1399)
                                                                             tota
         1: 1m 27s
                        remaining: 3m 45s
                test: 0.8863681 test1: 0.8231203
                                                      best: 0.8231777 (1543)
         1600:
                                                                             tota
         1: 1m 41s
                        remaining: 3m 35s
In [ ]: |print(np.mean(scores40))
         print(np.mean(scores41))
         0.8096611422742035
         0.6828921839980325
In [ ]: | pred4 = cat4.predict proba(test)[:,1]
```

```
bootcamp_final - Jupyter Notebook
 In [ ]: | print(cat4.feature importances )
          [6.77136385 7.46620841 2.61815811 1.34200953 0.73694863 2.69707917
            2.11824181 2.10233648 2.42828359 1.86620978 2.29982856 1.06825823
            1.55724121 1.90030774 2.22595604 0.54425601 0.74708346 0.68797583
            1.27392072 1.19588189 1.8014224 1.69031352 1.25768361 2.05734229
            2.34242887 1.76860867 1.56197984 1.3692587 1.92297207 1.82258266
            1.13557115 1.69000183 1.8976538 1.93243687 1.11762714 1.91949144
            2.07746344 2.03871629 0.40825328 1.93047426 1.17280155 2.73952482
            2.19505261 1.68765745 1.35205541 0.55699034 9.20095781 0.91409259
            1.76036672 1.03066951]
 In [ ]: feat_imp = pd.Series(cat4.feature_importances_, index = X.columns)
 In [ ]: |plt.figure(figsize = (10,10))
          feat imp.nlargest(50).plot(kind= 'barh')
Out[23]: <matplotlib.axes. subplots.AxesSubplot at 0x7f6b5e0bf9e8>
            form_field39
            form_field16
form_field46
            form_field18
            form field5
            form_field17
form_field48
            form_field50
form_field12
            form_field35
form_field31
            form_field41
            form_field20
            form_field23
            form field 19
```



```
In [ ]: X.drop('form_field39', axis = 1, inplace = True)
  test.drop('form_field39', axis = 1, inplace = True)
```

```
In [ ]: | scores10 = []
       scores11 = []
       for fold,(tr in,te in) in enumerate(kf.split(X, y)):
           X_train,X_test = X.iloc[tr_in],X.iloc[te_in]
           y_train,y_test = y.iloc[tr_in],y.iloc[te_in]
           cat4.fit(X_train,y_train,eval_set=[(X_train,y_train),(X_test,y_test)],early_s
           v pred = cat4.predict(X test)
           scores10.append(accuracy score(y test,y pred))
           scores11.append(roc_auc_score(y_test,y_pred))
        test: 0.8072782 test1: 0.7821756
                                                     best: 0.7821756 (0)
       0:
                                                                            tota
       1: 62.6ms
                       remaining: 5m 12s
       200:
               test: 0.8429309 test1: 0.8126232
                                                     best: 0.8126480 (197)
                                                                            tota
       1: 12.7s
                       remaining: 5m 3s
       400:
               test: 0.8512761 test1: 0.8160451
                                                     best: 0.8160560 (396)
                                                                            tota
       1: 24.8s
                       remaining: 4m 43s
       600:
               test: 0.8583740 test1: 0.8181868
                                                     best: 0.8182023 (599)
                                                                            tota
       1: 37.2s
                       remaining: 4m 32s
       800:
               test: 0.8642650 test1: 0.8197542
                                                     best: 0.8197666 (799)
                                                                            tota
       1: 49.7s
                       remaining: 4m 20s
       1000:
               test: 0.8699764 test1: 0.8207474
                                                     best: 0.8207691 (990)
                                                                            tota
       1: 1m 2s
                       remaining: 4m 8s
               test: 0.8754421 test1: 0.8215684
       1200:
                                                     best: 0.8215854 (1199)
                                                                            tota
       1: 1m 14s
                       remaining: 3m 56s
       1400:
               test: 0.8812183 test1: 0.8221410
                                                     best: 0.8222465 (1361)
                                                                            tota
       1: 1m 27s
                       remaining: 3m 43s
       1600:
               test: 0.8864300 test1: 0.8228611
                                                     best: 0.8229216 (1588)
                                                                            tota
       1: 1m 39s
                       remaining: 3m 31s
                                                           0 000000 (4760)
In [ ]: |print(np.mean(scores10))
       print(np.mean(scores11))
       0.8097681703217519
       0.6843677444985984
In [ ]: | pred1 = cat4.predict proba(test)[:,1]
In [ ]: |lgbm = LGBMClassifier(boosting type='gbdt', objective='binary', num leaves=50,
                                      learning rate=0.01, n estimators=2000, max depth=
```

```
In [ ]: | scores20 = []
       scores21 = []
       for fold,(tr in,te in) in enumerate(kf.split(X, y)):
           X_train,X_test = X.iloc[tr_in],X.iloc[te_in]
           y_train,y_test = y.iloc[tr_in],y.iloc[te_in]
           lgbm.fit(X_train,y_train,eval_set=[(X_train,y_train),(X_test,y_test)],early_s
           y pred = lgbm.predict(X test)
           scores20.append(accuracy score(y test,y pred))
           scores21.append(roc_auc_score(y_test,y_pred))
       Early stopping, best iteration is:
                                                  valid 1's binary logloss: 0.4
       [1094] training's binary logloss: 0.336182
       04563
       Training until validation scores don't improve for 100 rounds.
                                                  valid 1's binary logloss: 0.4
       [200]
              training's binary logloss: 0.406455
       12874
              training's binary logloss: 0.380678
                                                  valid 1's binary logloss: 0.3
       [400]
       97644
       [600]
              training's binary_logloss: 0.365
                                                  valid_1's binary_logloss: 0.3
       9467
              training's binary logloss: 0.352323
       [800]
                                                  valid 1's binary logloss: 0.3
       93908
       Early stopping, best iteration is:
       [801]
              training's binary logloss: 0.352269
                                                  valid 1's binary logloss: 0.3
       93904
       Training until validation scores don't improve for 100 rounds.
       [200]
              training's binary logloss: 0.405855
                                                  valid 1's binary logloss: 0.4
In [ ]: |print(np.mean(scores20))
       print(np.mean(scores21))
       0.8081785714285715
       0.6853363349877205
In [ ]: | pred2 = lgbm.predict proba(test)[:,1]
In [ ]: | submit['2'] = pred2
       submit['4'] = pred4
       submit['1'] = pred1
In [ ]: submit['default status'] = ((submit['2']* 0.2)+ ((submit['4'] * 0.7)+ submit['1'
In [ ]: | submit.drop(['2', '4', '1'], axis =1, inplace = True)
In [ ]: | submit.to_csv('submit2d5c.csv', index = False)
```