

```

import numpy # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import io
# Models:
from sklearn import svm
from sklearn.model_selection import cross_validate
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import KFold, cross_val_score
# Python utilities:
import time
import os
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.metrics import roc_auc_score
from sklearn.metrics import roc_curve
import matplotlib.pyplot as plt
import datetime
%tensorflow_version 2.x
import tensorflow as tf

```

## Google Collabrate Import Data

```

from google.colab import drive
drive.mount('/content/gdrive')

```

☞ Drive already mounted at /content/gdrive; to attempt to forcibly remount, call

## Google Collabrate files uploading

```

from google.colab import files
uploaded = files.upload()
for fn in uploaded.keys():
    print('User uploaded file "{name}" with length {length} bytes'.format(
        name=fn, length=len(uploaded[fn])))

```

☞

## Train data and test data imported and test data listed

```

train_data=pd.read_csv('gdrive/My Drive/GiveMeSomeCredit/cs-training.csv')
test_data=pd.read_csv('gdrive/My Drive/GiveMeSomeCredit/cs-test.csv')
cv = KFold(n_splits=2,random_state=None, shuffle=False)

```

## First 6 data information

```
train_data.head()
```

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	Unnamed: 0	SeriousDlqin2yrs	RevolvingUtilizationOfUnsecuredLines	age	59Days
0	1	1	0.766127	45	
1	2	0	0.957151	40	
2	3	0	0.658180	38	
3	4	0	0.233810	30	
4	5	0	0.907239	49	

```
train_data.shape
```

```
(150000, 12)
```

Get overall information more statistical

```
train_data.describe()
```

```
[>
```

	Unnamed: 0	SeriousDlqin2yrs	RevolvingUtilizationOfUnsecuredLines	
<b>count</b>	150000.000000	150000.000000	150000.000000	15
<b>mean</b>	75000.500000	0.066840	6.048438	
<b>std</b>	43301.414527	0.249746	249.755371	
<b>min</b>	1.000000	0.000000	0.000000	
<b>25%</b>	37500.750000	0.000000	0.029867	
<b>50%</b>	75000.500000	0.000000	0.154181	
<b>75%</b>	112500.250000	0.000000	0.559046	
<b>max</b>	150000.000000	1.000000	50708.000000	

The ratio of debt to assets

```
train_data.DebtRatio.describe()
```

```
[> count      150000.000000
mean         353.005076
std          2037.818523
min           0.000000
25%           0.175074
50%           0.366508
75%           0.868254
max          329664.000000
Name: DebtRatio, dtype: float64
```

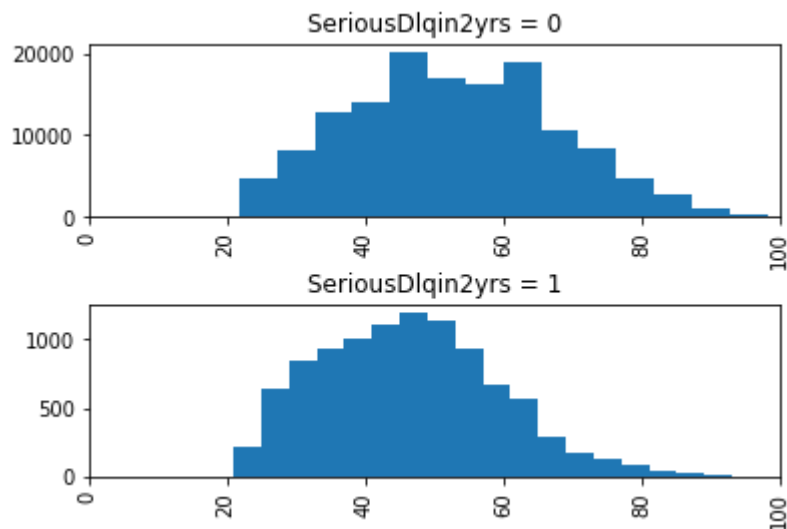
How balanced the data set is

```
train_data.SeriousDlqin2yrs.mean()
```

```
↳ 0.06684
```

```
age_hist = train_data['age'].hist(by=train_data['SeriousDlqin2yrs'], bins=20, layout=(2,1))
age_hist[0].set_xlim((0,100))
age_hist[0].set_title('SeriousDlqin2yrs = 0')
age_hist[1].set_xlim((0,100))
age_hist[1].set_title('SeriousDlqin2yrs = 1')
```

```
↳ Text(0.5, 1.0, 'SeriousDlqin2yrs = 1')
```



Investigate further to see if these are outliers or not

```
train_data.DebtRatio.quantile([.975])
```

```
↳ 0.975    3489.025
   Name: DebtRatio, dtype: float64
```

```
train_data[train_data['DebtRatio'] > 3489.025][['SeriousDlqin2yrs', 'MonthlyIncome']]
```

```
↳
```

```
train_data.groupby('NumberOfTimes90DaysLate').NumberOfTimes90DaysLate.count()
```

```

↳ NumberOfTimes90DaysLate
0      141662
1       5243
2      1555
3       667
4       291
5       131
6        80
7        38
8        21
9        19
10       8
11       5
12       2
13       4
14       2
15       2
17       1
96       5
98      264
Name: NumberOfTimes90DaysLate, dtype: int64

```

```

train_data[train_data['NumberOfTimes90DaysLate'] > 95][['SeriousDlqin2yrs',
                                                         'NumberOfTime60-89DaysPastI
'NumberOfTime30-59DaysPastDueNotWorse',
'NumberOfTimes90DaysLate']].describe()

```

```

↳

```

	SeriousDlqin2yrs	NumberOfTime60-89DaysPastDueNotWorse	NumberOfTime30-59DaysPastDueNotWorse	NumberOfTimes90DaysLate
count	269.000000	269.000000	269.000000	269.000000
mean	0.546468	97.962825	97.962825	97.962825
std	0.498764	0.270628	0.270628	0.270628
min	0.000000	96.000000	96.000000	96.000000
25%	0.000000	98.000000	98.000000	98.000000
50%	1.000000	98.000000	98.000000	98.000000
75%	1.000000	98.000000	98.000000	98.000000
max	1.000000	98.000000	98.000000	98.000000

choose specific data and get the arg minumum of them

```

def rmDataAndPutArgmin(train_data, first = 20, second = 40):
    New = []
    argminx = train_data.argmin()
    for val in train_data:
        if ((val == first) | (val == second)):
            New.append(argminx)
        else:

```

```

        New.append(val)

    return New

```

```

not_missing = train_data.dropna()
target = 'MonthlyIncome'
predictors = [c for c in list(not_missing) if c not in [target, 'Unnamed: 0', 'Series']]
X_data = not_missing[predictors]
y_data = not_missing[target]
regr = LinearRegression().fit(X_data, y_data)
regr.score(X_data, y_data)

```

```

☞ 0.02201505632577072

```

```

train_data=train_data.fillna(round(train_data.median()))
train_data.loc[train_data.DebtRatio > 1, 'DebtRatio'] = train_data['DebtRatio'].median()

```

## Training Data

```

trainnew = rmDataAndPutArgmin(train_data.SeriousDlqin2yrs)
train_data.SeriousDlqin2yrs=trainnew

x = train_data.drop(['SeriousDlqin2yrs'], axis=1)
y = trainnew

gb_scores=[]
for n in [150, 160, 170, 180, 190, 200, 210]:
    gb_scores.append(numpy.mean(cross_val_score(GradientBoostingClassifier(learning_rate=0.1, n_estimators=n), x, y)))
print(gb_scores)
start_time = datetime.datetime.now()
print (numpy.mean(cross_val_score(GradientBoostingClassifier(learning_rate=0.1, n_estimators=210), x, y)))
print ('Time elapsed:', datetime.datetime.now() - start_time)

```

```

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```

Iter	Train Loss	Remaining Time
1	0.4575	23.65s
2	0.4392	23.24s
3	0.4256	22.91s
4	0.4154	23.01s
5	0.4074	22.83s
6	0.4009	22.79s
7	0.3954	22.54s
8	0.3904	22.33s
9	0.3863	22.11s
10	0.3828	22.05s
20	0.3647	20.35s
30	0.3581	18.69s
40	0.3546	17.18s
50	0.3521	15.57s
60	0.3502	14.01s
70	0.3486	12.43s
80	0.3474	10.88s
90	0.3465	9.32s
100	0.3455	7.77s

Iter	Train Loss	Remaining Time
1	0.4571	22.71s
2	0.4388	22.87s
3	0.4263	22.71s
4	0.4162	22.50s
5	0.4081	22.56s
6	0.4023	22.34s
7	0.3970	22.13s
8	0.3926	21.93s
9	0.3886	21.78s
10	0.3860	21.66s
20	0.3689	20.30s
30	0.3621	18.60s
40	0.3583	17.03s
50	0.3556	15.46s
60	0.3537	13.89s
70	0.3523	12.32s
80	0.3510	10.78s
90	0.3500	9.24s
100	0.3492	7.71s

Iter	Train Loss	Remaining Time
1	0.4575	24.49s
2	0.4392	24.57s
3	0.4256	24.23s
4	0.4154	24.00s
5	0.4074	23.93s
6	0.4009	23.77s
7	0.3954	23.57s
8	0.3904	23.42s
9	0.3863	23.26s
10	0.3828	23.27s
20	0.3647	21.68s
30	0.3581	20.13s
40	0.3546	18.58s
50	0.3521	17.01s
60	0.3502	15.46s
70	0.3486	13.92s
80	0.3474	12.37s
90	0.3465	10.82s
100	0.3455	9.27s

Iter	Train Loss	Remaining Time
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1	0.4571	24.37s
2	0.4388	24.85s
3	0.4263	24.52s
4	0.4162	24.34s
5	0.4081	24.18s
6	0.4023	24.08s
7	0.3970	23.82s
8	0.3926	23.74s
9	0.3886	23.55s
10	0.3860	23.45s
20	0.3689	21.84s
30	0.3621	20.21s
40	0.3583	18.60s
50	0.3556	17.09s
60	0.3537	15.50s
70	0.3523	13.94s
80	0.3510	12.38s
90	0.3500	10.83s
100	0.3492	9.29s
Iter	Train Loss	Remaining Time
1	0.4575	25.95s
2	0.4392	26.02s
3	0.4256	25.62s
4	0.4154	25.87s
5	0.4074	25.65s
6	0.4009	25.46s
7	0.3954	25.25s
8	0.3904	25.06s
9	0.3863	24.87s
10	0.3828	24.81s
20	0.3647	23.23s
30	0.3581	21.65s
40	0.3546	20.12s
50	0.3521	18.58s
60	0.3502	17.01s
70	0.3486	15.44s
80	0.3474	13.89s
90	0.3465	12.34s
100	0.3455	10.81s
Iter	Train Loss	Remaining Time
1	0.4571	26.41s
2	0.4388	26.62s
3	0.4263	26.36s
4	0.4162	26.33s
5	0.4081	26.13s
6	0.4023	26.13s
7	0.3970	25.86s
8	0.3926	25.68s
9	0.3886	25.49s
10	0.3860	25.45s
20	0.3689	23.68s
30	0.3621	21.94s
40	0.3583	20.31s
50	0.3556	18.74s
60	0.3537	17.14s
70	0.3523	15.54s
80	0.3510	13.96s
90	0.3500	12.41s
100	0.3492	10.84s
Iter	Train Loss	Remaining Time
1	0.4575	27.78s
2	0.4392	27.61s

3	0.4256	27.44s
4	0.4154	27.19s
5	0.4074	27.01s
6	0.4009	27.05s
7	0.3954	26.86s
8	0.3904	26.66s
9	0.3863	26.48s
10	0.3828	26.52s
20	0.3647	24.84s
30	0.3581	23.24s
40	0.3546	21.67s
50	0.3521	20.13s
60	0.3502	18.54s
70	0.3486	16.98s
80	0.3474	15.45s
90	0.3465	13.91s
100	0.3455	12.36s
Iter	Train Loss	Remaining Time
1	0.4571	27.23s
2	0.4388	27.26s
3	0.4263	27.07s
4	0.4162	26.91s
5	0.4081	26.80s
6	0.4023	26.82s
7	0.3970	26.59s
8	0.3926	26.41s
9	0.3886	26.26s
10	0.3860	26.22s
20	0.3689	24.71s
30	0.3621	23.19s
40	0.3583	21.63s
50	0.3556	20.08s
60	0.3537	18.53s
70	0.3523	16.96s
80	0.3510	15.41s
90	0.3500	13.86s
100	0.3492	12.32s
Iter	Train Loss	Remaining Time
1	0.4575	28.79s
2	0.4392	28.96s
3	0.4256	28.86s
4	0.4154	28.60s
5	0.4074	28.40s
6	0.4009	28.30s
7	0.3954	28.10s
8	0.3904	27.96s
9	0.3863	27.97s
10	0.3828	28.04s
20	0.3647	26.22s
30	0.3581	24.63s
40	0.3546	23.08s
50	0.3521	21.57s
60	0.3502	20.00s
70	0.3486	18.52s
80	0.3474	17.08s
90	0.3465	15.51s
100	0.3455	13.95s
Iter	Train Loss	Remaining Time
1	0.4571	28.50s
2	0.4388	28.97s
3	0.4263	28.72s
4	0.4162	28.44s



4	0.4102	20.44s
5	0.4081	28.57s
6	0.4023	28.33s
7	0.3970	28.06s
8	0.3926	27.88s
9	0.3886	27.80s
10	0.3860	27.75s
20	0.3689	26.23s
30	0.3621	24.60s
40	0.3583	23.01s
50	0.3556	21.47s
60	0.3537	19.92s
70	0.3523	18.37s
80	0.3510	16.84s
90	0.3500	15.31s
100	0.3492	13.77s
Iter	Train Loss	Remaining Time
1	0.4575	30.43s
2	0.4392	30.48s
3	0.4256	30.44s
4	0.4154	30.25s
5	0.4074	30.02s
6	0.4009	29.91s
7	0.3954	29.76s
8	0.3904	29.71s
9	0.3863	29.64s
10	0.3828	29.63s
20	0.3647	28.04s
30	0.3581	26.41s
40	0.3546	24.82s
50	0.3521	23.24s
60	0.3502	21.65s
70	0.3486	20.11s
80	0.3474	18.56s
90	0.3465	17.00s
100	0.3455	15.44s
200	0.3391	0.00s
Iter	Train Loss	Remaining Time
1	0.4571	30.39s
2	0.4388	30.48s
3	0.4263	30.19s
4	0.4162	29.97s
5	0.4081	29.92s
6	0.4023	29.89s
7	0.3970	29.62s
8	0.3926	29.49s
9	0.3886	29.40s
10	0.3860	29.34s
20	0.3689	27.76s
30	0.3621	26.09s
40	0.3583	24.55s
50	0.3556	23.01s
60	0.3537	21.46s
70	0.3523	19.95s
80	0.3510	18.42s
90	0.3500	16.88s
100	0.3492	15.41s
200	0.3431	0.00s
Iter	Train Loss	Remaining Time
1	0.4575	32.20s
2	0.4392	32.57s
3	0.4256	32.06s

4	0.4154	31.95s
5	0.4074	31.72s
6	0.4009	31.87s
7	0.3954	31.61s
8	0.3904	31.40s
9	0.3863	31.25s
10	0.3828	31.16s
20	0.3647	29.38s
30	0.3581	27.77s
40	0.3546	26.21s
50	0.3521	24.64s
60	0.3502	23.07s
70	0.3486	21.58s
80	0.3474	20.03s
90	0.3465	18.49s
100	0.3455	16.95s
200	0.3391	1.54s

Iter	Train Loss	Remaining Time
1	0.4571	32.85s
2	0.4388	32.43s
3	0.4263	32.09s
4	0.4162	31.80s
5	0.4081	31.77s
6	0.4023	31.54s
7	0.3970	31.27s
8	0.3926	31.18s
9	0.3886	31.00s
10	0.3860	30.95s
20	0.3689	29.34s
30	0.3621	27.73s
40	0.3583	26.12s
50	0.3556	24.62s
60	0.3537	23.02s
70	0.3523	21.48s
80	0.3510	19.94s
90	0.3500	18.41s
100	0.3492	16.87s
200	0.3431	1.53s

[0.7701272042736258, 0.7702553260139671, 0.7691237412549932, 0.769091447101051

Iter	Train Loss	Remaining Time
1	0.4575	48.09s
2	0.4392	46.83s
3	0.4256	46.41s
4	0.4154	45.77s
5	0.4074	45.94s
6	0.4009	45.61s
7	0.3954	45.52s
8	0.3904	45.34s
9	0.3863	45.28s
10	0.3828	45.17s
20	0.3647	43.37s
30	0.3581	41.71s
40	0.3546	40.28s
50	0.3521	38.62s
60	0.3502	37.02s
70	0.3486	35.42s
80	0.3474	33.87s
90	0.3465	32.29s
100	0.3455	30.75s
200	0.3391	15.35s
300	0.3331	0.00s

Iter	Train Loss	Remaining Time
------	------------	----------------

1	0.4571	46.16s
2	0.4388	46.69s
3	0.4263	46.81s
4	0.4162	47.51s
5	0.4081	47.04s
6	0.4023	46.70s
7	0.3970	46.23s
8	0.3926	45.97s
9	0.3886	45.84s
10	0.3860	45.83s
20	0.3689	44.00s
30	0.3621	42.19s
40	0.3583	40.46s
50	0.3556	38.79s
60	0.3537	37.11s
70	0.3523	35.56s
80	0.3510	33.96s
90	0.3500	32.38s
100	0.3492	30.80s
200	0.3431	15.33s
300	0.3380	0.00s

0.7657463539285931  
Time elapsed: 0:01:32.755747

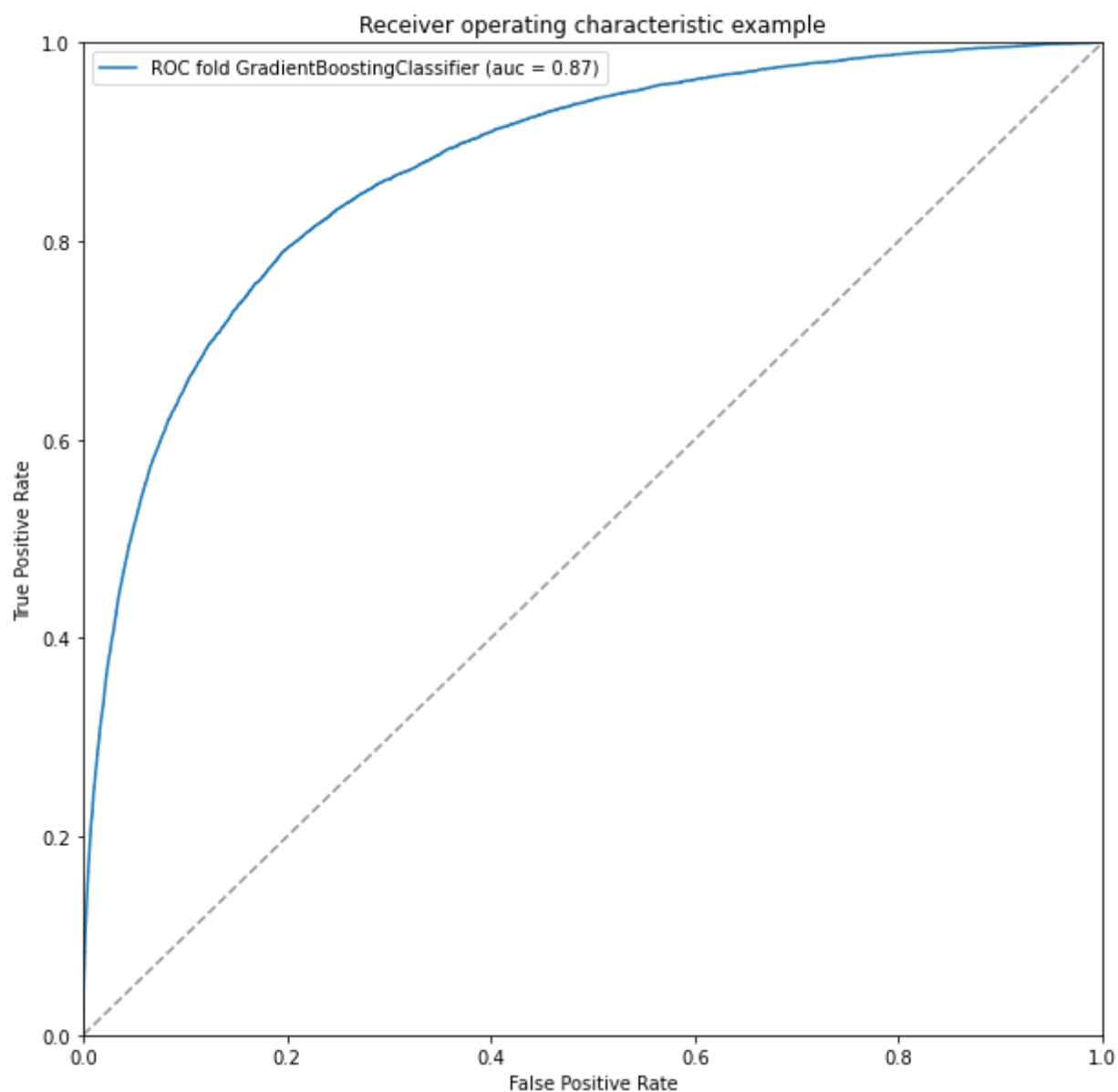
## Roc curve of gradient boosting classifier

```
plt.figure(figsize=(10, 10))
models=[]
models.append(GradientBoostingClassifier(learning_rate=0.1, n_estimators=300, verbose=0))
for model in models:
    model.fit(x,y)
    testscore=model.predict_proba(x)[:, 1]
    fpr, tpr, thresholds = roc_curve(y, testscore)
    roc_auc = roc_auc_score(y, testscore)
    md = str(model)
    md = md[:md.find('(')]
    plt.plot(fpr, tpr, label='ROC fold %s (auc = %0.2f)' % (md, roc_auc))

plt.plot([0, 1], [0, 1], '--', color=(0.6, 0.6, 0.6))
plt.xlim([0, 1])
plt.ylim([0, 1])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Receiver operating characteristic example')
plt.legend(loc="best")
plt.show()
```



Iter	Train Loss	Remaining Time
1	0.4578	1.66m
2	0.4391	1.64m
3	0.4260	1.64m
4	0.4164	1.62m
5	0.4082	1.62m
6	0.4024	1.61m
7	0.3966	1.60m
8	0.3922	1.59m
9	0.3881	1.59m
10	0.3854	1.58m
20	0.3680	1.52m
30	0.3618	1.46m
40	0.3585	1.40m
50	0.3562	1.35m
60	0.3548	1.29m
70	0.3536	1.24m
80	0.3528	1.18m
90	0.3520	1.13m
100	0.3514	1.07m
200	0.3469	32.17s
300	0.3435	0.00s



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