```
In [1]:
import pyodbc
In [2]:
conn = pyodbc.connect(r'DSN=tekkredi;UID=yavuzs;PWD=18651438-155E-4450-859D-803181407D18')
In [3]:
import pandas as pd
import numpy as np
In [4]:
#database connection and import table
df = pd.read_sql("select * from dbo.Sample5", conn)
In [5]:
df.head()
Out[5]:
  RatioCons RatioCred
                      RatioOD RatioMorg RatioCar
                                                 RatioFullDebtLimit RatioConsDebtLimit RatioCredDebtLimit Ratio
0 0.307692
                      0.307692 NaN
             0.384615
                                                  0.440091
                                                                   0.260075
                                                                                     0.730099
                                                                                                       0.68
                                         None
```

#### 1 0.777778 0.037037 0.185185 NaN None 0.257212 0.201451 0.993843 NaN 0.538462 0.230769 0.230769 NaN None 0.249186 0.258512 -0.000192 0.36 3 0.428571 0.250000 0.321429 NaN 0.179675 0.225764 0.076379 0.21 None 0.333333 0.208333 0.333333 0.125 None 0.425787 0.246722 0.136672 0.00

5 rows × 184 columns

```
In [6]:
```

```
In [7]:
```

### In [ ]:

```
df_2['Score_Inv'] = np.where(df_2.Score == 0, 1 / min(df_2.Score[df_2.Score > 0]), 1 / df_2.Score)
df_2['RatioConsDebtLimit_Inv'] = np.where(df_2.RatioConsDebtLimit == 0, 1 / min(df_2.RatioConsDebtL
imit[df_2.RatioConsDebtLimit > 0]), 1 / df_2.RatioConsDebtLimit)
df_2['Ratio90of3to12_Inv'] = np.where(df_2.Ratio90of3to12 == 0, 1 / min(df_2.Ratio90of3to12[df_2.Ratio90of3to12 > 0]), 1 / df_2.Ratio90of3to12)
df_2['Ratio90of12to18_Inv'] = np.where(df_2.Ratio90of12to18 == 0, 1 / min(df_2.Ratio90of12to18[df_2.Ratio90of12to18 > 0]), 1 / df_2.Ratio90of12to18)
df_2['FullAvg30Mon3_Inv'] = np.where(df_2.FullAvg30Mon3 == 0, 1 / min(df_2.FullAvg30Mon3[df_2.FullA
```

```
vg30Mon3 > 0]), 1 / df 2.FullAvg30Mon3)
\label{eq:df_2['FullAvg60Mon3_Inv']} $$ df_2['FullAvg60Mon3_Inv'] = np.where(df_2.FullAvg60Mon3 == 0, 1 / min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) $$ df_2['FullAvg60Mon3_Inv'] = np.where(df_2.FullAvg60Mon3 == 0, 1 / min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) $$ df_2['FullAvg60Mon3_Inv'] = np.where(df_2.FullAvg60Mon3 == 0, 1 / min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) $$ df_2['FullAvg60Mon3_Inv'] = np.where(df_2.FullAvg60Mon3 == 0, 1 / min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) $$ df_2['FullAvg60Mon3_Inv'] = np.where(df_2.FullAvg60Mon3 == 0, 1 / min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) $$ df_2['FullAvg60Mon3_Inv'] = np.where(df_2.FullAvg60Mon3 == 0, 1 / min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) $$ df_2['FullAvg60Mon3] = np.where(df_2.FullAvg60Mon3) $$ df_2['FullAvg60Mon3] = np.where(df_2.FullAvg60Mon3) $$ df_2['FullAvg60Mon3] = np.where(df_2.FullAvg60Mon3) $$ df_2['FullAvg60Mon3] = np.where(df_2.FullAvg60Mon3) = np.where(df_2.FullAvg6
vg60Mon3 > 0), 1 / df 2.FullAvg60Mon3)
df 2['ConsAvg90Mon3_Inv'] = np.where(df_2.ConsAvg90Mon3 == 0, 1 / min(df_2.ConsAvg90Mon3[df_2.ConsA
vg90Mon3 > 0), 1 / df 2.ConsAvg90Mon3)
df 2['CredAvg90Mon3 Inv'] = np.where(df 2.CredAvg90Mon3 == 0, 1 / min(df 2.CredAvg90Mon3[df 2.CredA
vg90Mon3 > 0]), 1 / df_2.CredAvg90Mon3)
df 2['ODAvg90Mon3 Inv'] = np.where(df 2.ODAvg90Mon3 == 0, 1 / min(df 2.ODAvg90Mon3[df 2.ODAvg90Mon3
> 0]), 1 / df 2.0DAvg90Mon3)
 df 2['FullAvg90Mon12 Inv'] = np.where (df 2.FullAvg90Mon12 == 0, 1 / min(df 2.FullAvg90Mon12[df 2.FullAvg90Mon12] = 0, 1 / min(df 2.FullAvg90Mon12[df 2.FullAvg90Mon12] = 0, 1 / min(df 2.FullAvg90Mon12] = 0, 1 / min(df 2.FullAvg90Mon12[df 2.FullAvg90Mon12] = 0, 1 / min(df 2.FullAvg90Mon12] = 0, 1 / min(df 2.FullAvg90Mon12[df 2.FullAvg90Mon12] = 0, 1 / min(df 2.FullAvg90Mon12] = 0, 1 / min(df
11Avg90Mon12 > 0]), 1 / df 2.FullAvg90Mon12)
df 2['Score Log'] = np.where(df 2.Score == 0, np.log(min(df 2.Score[df 2.Score > 0])), np.log(df 2.
Score))
df_2['RatioConsDebtLimit_Log'] = np.where(df_2.RatioConsDebtLimit == 0, np.log(min(df_2.RatioConsDe
btLimit[df 2.RatioConsDebtLimit > 0])), np.log(df 2.RatioConsDebtLimit))
df 2['Ratio90of3to12 Log'] = np.where(df 2.Ratio90of3to12 == 0, np.log(min(df 2.Ratio90of3to12[df 2
.Ratio90of3to12 > 0])), np.log(df_2.Ratio90of3to12))
df 2['Ratio90of12to18 Log'] = np.where(df 2.Ratio90of12to18 == 0, np.log(min(df 2.Ratio90of12to18[d
f 2.Ratio90of12to18 > 0])), np.log(df 2.Ratio90of12to18))
df 2['FullAvg30Mon3 Log'] = np.where(df 2.FullAvg30Mon3 == 0, np.log(min(df 2.FullAvg30Mon3[df 2.Fu
11Avg30Mon3 > 0])), np.log(df 2.FullAvg30Mon3))
 df_2['FullAvg60Mon3\_Log'] = np.where \\ (df_2.FullAvg60Mon3 == 0, np.log(min(df_2.FullAvg60Mon3[df_2.FullAvg60Mon3]) \\ (df_2.FullAvg60Mon3_Log') \\ (df_2.FullAvg60Mon3_L
11Avg60Mon3 > 0])), np.log(df_2.FullAvg60Mon3))
df 2['ConsAvg90Mon3 Log'] = np.where(df 2.ConsAvg90Mon3 == 0, np.log(min(df 2.ConsAvg90Mon3[df 2.Co
nsAvg90Mon3 > 0])), np.log(df_2.ConsAvg90Mon3))
df 2['CredAvg90Mon3 Log'] = np.where(df 2.CredAvg90Mon3 == 0, np.log(min(df 2.CredAvg90Mon3[df 2.Cr
edAvg90Mon3 > 0])), np.log(df_2.CredAvg90Mon3))
df 2['ODAvg90Mon3 Log'] = np.where(df 2.ODAvg90Mon3 == 0, np.log(min(df 2.ODAvg90Mon3[df 2.ODAvg90M
on3 > 0])), np.log(df 2.ODAvg90Mon3))
df 2['FullAvg90Mon12 Log'] = np.where(df 2.FullAvg90Mon12 == 0, np.log(min(df 2.FullAvg90Mon12[df 2
.FullAvg90Mon12 > 0])), np.log(df_2.FullAvg90Mon12))
```

#### In [ ]:

df\_2.describe()

### In [43]:

 $corr_1 = df_2.corr()$ 

# In [44]:

```
import matplotlib.pyplot as plt
import seaborn as sns
plt.subplots(figsize=(20, 10))
sns.heatmap(corr_1, annot=True, cmap="RdYlGn")
plt.show()
```



0.9

0.6

0.3

0.0

-0.3

main

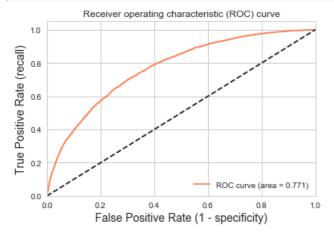
```
In [8]:
```

```
df 2['RatioConsDebtLimit'] = df 2['RatioConsDebtLimit'].fillna(df 2['RatioConsDebtLimit'].mean())
df 2['Ratio90of3to12'] = df 2['Ratio90of3to12'].fillna(df_2['Ratio90of3to12'].mean())
df 2['Ratio90of12to18'] = df 2['Ratio90of12to18'].fillna(df 2['Ratio90of12to18'].mean())
df 2['FullAvg30Mon3'] = df 2['FullAvg30Mon3'].fillna(df 2['FullAvg30Mon3'].mean())
\label{eq:df2['FullAvg60Mon3'] = df2['FullAvg60Mon3'].fillna(df2['FullAvg60Mon3'].mean())} \\
\label{eq:df2['ConsAvg90Mon3']} $$ df_2['ConsAvg90Mon3'].fillna(df_2['ConsAvg90Mon3'].mean()) $$ $$ df_2['ConsAvg90Mon3'].$$
df 2['FullAvg90Mon12'] = df 2['FullAvg90Mon12'].fillna(df_2['FullAvg90Mon12'].mean())
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel launcher.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer, col indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
  """Entry point for launching an IPython kernel.
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel launcher.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
 This is separate from the ipykernel package so we can avoid doing imports until
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel launcher.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
 after removing the cwd from sys.path.
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel launcher.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel launcher.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel launcher.py:7: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer, col indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
 import sys
C:\Users\Tekkredi\Anaconda3\lib\site-packages\ipykernel_launcher.py:8: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer, col indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer, col indexer] = value instead
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
```

```
In [9]:
df_2[df_2.isnull().any(axis=1)]
Out[9]:
 Def | Score | RatioConsDebtLimit | Ratio90of3to12 | Ratio90of12to18 | FullAvg30Mon3 | FullAvg60Mon3 | ConsAvg90Mon3
                                                                                             Cre
In [11]:
df_2_vars=df_2.columns.values.tolist()
In [35]:
y=df 2[['Def']]
X=df 2[[i for i in df 2 vars if i not in y]]
In [36]:
from sklearn.model_selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_state=42)
In [37]:
df 2 test= pd.concat([X test, y test], axis=1, join='inner')
In [45]:
#standardized xtrain
X train std = X train.std(0)
X_train_mean = X_train.mean(0)
X train edit=(X train-X train mean)/X train std
In [46]:
#standardized xtest
X \text{ test\_std} = X_{\text{test.std}}(0)
X test mean = X test.mean(0)
X_test_edit=(X_test-X_test_mean)/X_test_std
In [47]:
# Parameters initialization
weights = np.random.normal(0, 0.1, 10)
bias = np.random.normal(0, 0.1)
lr = 0.05
n = X train edit.shape[0]
In [48]:
for epoch in range(300):
    # Logistic function
    Z = np.dot(X train edit, weights) + bias
   A = 1 / (1 + np.exp(-Z))
    # Negative log likelihood -- loss function
    # Gradient computation
    dZ = A - y_train['Def']
    dw = np.dot(dZ, X_train_edit) / n
    db = np.sum(dZ) / n
    # Update weights
```

```
weights = weights - lr * dw
    bias = bias - lr * db
    if epoch % 10 == 0:
       print("epoch %s - loss %s" % (epoch, J))
epoch 0 - loss 0.6731099431713983
epoch 10 - loss 0.6367877695717422
epoch 20 - loss 0.616139112812915
epoch 30 - loss 0.6036143608295776
epoch 40 - loss 0.5956123119537848
epoch 50 - loss 0.5902864743232956
epoch 60 - loss 0.5866222149437182
epoch 70 - loss 0.5840299688584913
epoch 80 - loss 0.5821515005539408
epoch 90 - loss 0.580761053907593
epoch 100 - loss 0.5797119674610682
epoch 110 - loss 0.5789064648608515
epoch 120 - loss 0.5782778822504294
epoch 130 - loss 0.5777798634271202
epoch 140 - loss 0.577379603374761
epoch 150 - loss 0.5770535184529614
epoch 160 - loss 0.5767844110263288
epoch 170 - loss 0.5765595765369712
epoch 180 - loss 0.5763695175681626
epoch 190 - loss 0.5762070563103725
epoch 200 - loss 0.5760667130422781
epoch 210 - loss 0.575944265037956
epoch 220 - loss 0.5758364296303956
epoch 230 - loss 0.5757406338630138
epoch 240 - loss 0.5756548452870904
epoch 250 - loss 0.575577446445119
epoch 260 - loss 0.5755071409080369
epoch 270 - loss 0.5754428823372971
epoch 280 - loss 0.5753838205091868
epoch 290 - loss 0.5753292599468717
In [62]:
weights, bias
Out[62]:
(array([-0.4608249 , 0.21033903, -0.14422634, -0.06532443, 0.27781995,
         0.40399846, 0.2029573, 0.16374308, 0.09746318, 0.07586096]),
 -0.16477774283800228)
In [50]:
#logreg prediction score for test dataset
pred_score = 1 / (1 + np.exp(-(np.dot(X_test_edit, weights) + bias)))
In [51]:
from sklearn.metrics import roc_curve, auc, log_loss, accuracy_score, confusion_matrix
[fpr, tpr, thr] = roc curve(y test, pred score)
In [52]:
import matplotlib.pyplot as plt
import seaborn as sns
sns.set(style="white") #white background style for seaborn plots
sns.set(style="whitegrid", color codes=True)
In [53]:
#draw the roc curve
plt.plot(fpr, tpr, color='coral', label='ROC curve (area = %0.3f)' % auc(fpr, tpr))
plt.plot([0, 1], [0, 1], 'k--')
plt.xlim([0.0, 1.0])
```

```
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate (1 - specificity)', fontsize=14)
plt.ylabel('True Positive Rate (recall)', fontsize=14)
plt.title('Receiver operating characteristic (ROC) curve')
plt.legend(loc="lower right")
plt.show()
```



#### In [54]:

```
#calculate gini
2*0.771-1
```

## Out[54]:

0.542

#### In [55]:

```
#adding predicted probability results to test data
df_test_result = df_2_test
df_test_result = df_test_result.assign(pred_prob=pred_score)
```

### In [57]:

```
#creating deciles for ranking
df_test_result['decile_pred'] = pd.qcut(df_test_result['pred_prob'], 10, labels=np.arange(10, 0, -1))
df_test_result['decile_Bureau'] = pd.qcut(df_test_result['Score'].rank(method='first'), 10, labels=
False)+1

#making deciles integers
df_test_result['decile_pred']=df_test_result['decile_pred'].astype(int)
df_test_result['decile_Bureau']=df_test_result['decile_Bureau'].astype(int)
```

### In [58]:

### In [59]:

```
rank_merge
```

# Out[59]:

	decile_pred	0_x	1_x	decile	bad_rate_x	decile_Bureau	0_y	1_y	bad_rate_y
0	1	271	1391	1	0.836943	1	592	1070	0.643803
1	2	444	1218	2	0.732852	2	534	1128	0.678700
2	3	641	1020	3	0.614088	3	502	1159	0.697772
3	4	737	925	4	0.556558	4	613	1049	0.631167
4	5	857	804	5	0.484046	5	820	842	0.506619
5	6	968	694	6	0.417569	6	956	705	0.424443
6	7	1076	586	7	0.352587	7	1074	588	0.353791
7	8	1251	410	8	0.246839	8	1206	455	0.273931
8	9	1368	294	9	0.176895	9	1352	310	0.186522
9	10	1529	133	10	0.080024	10	1493	169	0.101685

### In [60]:

```
plt.plot(rank_merge['decile'], rank_merge['bad_rate_x'], color='blue')
plt.plot(rank_merge['decile'], rank_merge['bad_rate_y'], color='red')
plt.xlim((1, 10))
plt.ylim((0.0, 1.0))
plt.xticks(np.arange(1, 11, 1))
plt.show()
```

