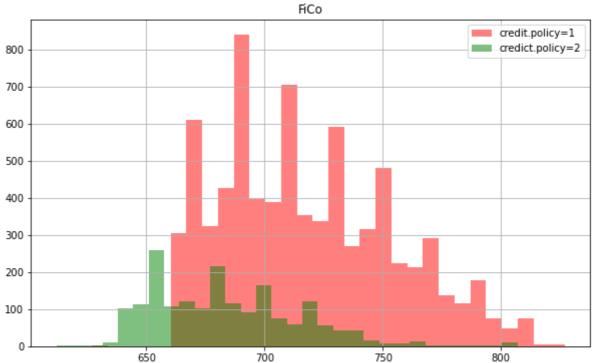
In [2]:

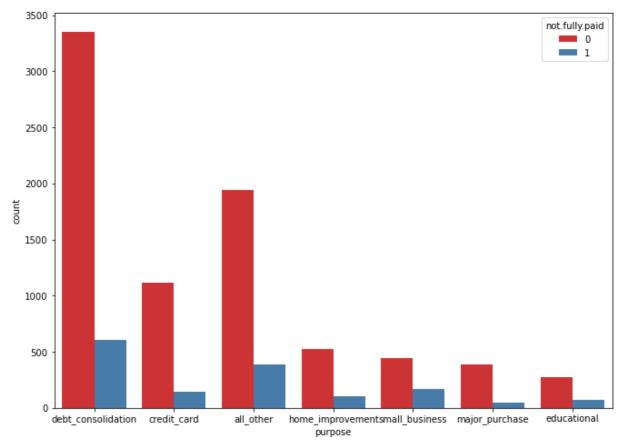
import os

```
import numpy as np
           import matplotlib.pyplot as plt
           import pandas as pd
           %matplotlib inline
           from sklearn.model selection import train test split
In [3]:
           os.chdir('C:\\Excel Class')
In [4]:
           df=pd.read_csv('loan_data.csv')
Out[4]:
                 credit.policy
                                                    int.rate
                                                             installment log.annual.inc
                                                                                            dti
                                                                                                fico
                                                                                                       days.with.cr
                                          purpose
              0
                            1
                                debt_consolidation
                                                     0.1189
                                                                  829.10
                                                                               11.350407
                                                                                          19.48
                                                                                                 737
                                                                                                           5639.958
              1
                            1
                                                                  228.22
                                                                                                 707
                                        credit_card
                                                     0.1071
                                                                               11.082143
                                                                                          14.29
                                                                                                           2760.000
              2
                            1
                                 debt_consolidation
                                                     0.1357
                                                                  366.86
                                                                               10.373491
                                                                                          11.63
                                                                                                 682
                                                                                                           4710.000
              3
                            1
                                 debt_consolidation
                                                                  162.34
                                                                               11.350407
                                                                                                 712
                                                                                                           2699.958
                                                     0.1008
                                                                                           8.10
              4
                            1
                                        credit_card
                                                     0.1426
                                                                  102.92
                                                                               11.299732
                                                                                          14.97
                                                                                                  667
                                                                                                           4066.000
                            0
                                                                               12.180755
                                                                                          10.39
                                                                                                 672
                                                                                                          10474.000
          9573
                                          all_other
                                                     0.1461
                                                                  344.76
                            0
                                                                                           0.21
                                                                                                 722
          9574
                                          all_other
                                                     0.1253
                                                                  257.70
                                                                               11.141862
                                                                                                           4380.000
          9575
                                debt_consolidation
                                                                   97.81
                                                                               10.596635
                                                                                          13.09
                                                                                                  687
                                                                                                           3450.041
                            0
                                                     0.1071
          9576
                            0
                               home_improvement
                                                     0.1600
                                                                  351.58
                                                                               10.819778
                                                                                         19.18
                                                                                                 692
                                                                                                           1800.000
          9577
                                debt_consolidation
                                                     0.1392
                                                                  853.43
                                                                               11.264464
                                                                                          16.28
                                                                                                 732
                                                                                                           4740.000
         9578 rows × 14 columns
                                                                                                                 >
In [5]:
           df.describe()
Out[5]:
                  credit.policy
                                               installment
                                                           log.annual.inc
                                                                                    dti
                                                                                                 fico
                                                                                                       days.with.cr
                                     int.rate
                  9578.000000
                                9578.000000
                                              9578.000000
                                                              9578.000000
                                                                           9578.000000
                                                                                         9578.000000
                                                                                                           9578.00
          count
           mean
                      0.804970
                                    0.122640
                                               319.089413
                                                                10.932117
                                                                              12.606679
                                                                                          710.846314
                                                                                                           4560.76
             std
                      0.396245
                                    0.026847
                                               207.071301
                                                                 0.614813
                                                                               6.883970
                                                                                            37.970537
                                                                                                           2496.93
            min
                      0.000000
                                    0.060000
                                                15.670000
                                                                 7.547502
                                                                               0.000000
                                                                                           612.000000
                                                                                                            178.95
            25%
                      1.000000
                                    0.103900
                                               163.770000
                                                                10.558414
                                                                               7.212500
                                                                                           682.000000
                                                                                                           2820.00
            50%
                      1.000000
                                    0.122100
                                               268.950000
                                                                10.928884
                                                                              12.665000
                                                                                           707.000000
                                                                                                           4139.95
                                               432.762500
            75%
                      1.000000
                                    0.140700
                                                                11.291293
                                                                              17.950000
                                                                                           737.000000
                                                                                                           5730.00
            max
                      1.000000
                                    0.216400
                                               940.140000
                                                                14.528354
                                                                              29.960000
                                                                                           827.000000
                                                                                                          17639.95
In [6]:
```

df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 9578 entries, 0 to 9577 Data columns (total 14 columns): # Column Non-Null Count Dtype -------------0 credit.policy 9578 non-null int64 1 purpose 9578 non-null object 2 int.rate 9578 non-null float64 3 installment 9578 non-null float64 4 log.annual.inc 9578 non-null float64 5 dti 9578 non-null float64 6 fico 9578 non-null int64 7 days.with.cr.line 9578 non-null float64 8 revol.bal 9578 non-null int64 revol.util 9578 non-null float64 10 inq.last.6mths 9578 non-null int64 11 delinq.2yrs 9578 non-null int64 12 pub.rec 9578 non-null int64 13 not.fully.paid 9578 non-null int64 dtypes: float64(6), int64(7), object(1) memory usage: 1.0+ MB In [7]: df.isnull().sum() Out[7]: credit.policy 0 purpose 0 int.rate 0 installment 0 log.annual.inc 0 dti 0 fico 0 days.with.cr.line 0 revol.bal 0 revol.util 0 inq.last.6mths 0 deling.2yrs 0 pub.rec 0 not.fully.paid 0 dtype: int64 In [10]: plt.figure(figsize=(10,6)) df[df['credit.policy']==1]['fico'].hist(alpha=0.5,color='red',bins=30,label='credit. df[df['credit.policy']==0]['fico'].hist(alpha=0.5,color='green',bins=30,label='credi plt.legend() plt.title('FiCo') Out[10]: Text(0.5, 1.0, 'FiCo')



```
650
                                              700
                                                               750
                                                                                800
In [12]:
          df['credit.policy']
                  1
Out[12]:
                  1
          2
                  1
          3
                  1
                  1
         9573
                 0
         9574
                  0
         9575
                  0
         9576
                  0
         9577
                  0
         Name: credit.policy, Length: 9578, dtype: int64
In [13]:
          import seaborn as sns
In [18]:
          plt.figure(figsize=(11,8))
          sns.countplot(x='purpose',hue='not.fully.paid',data=df,palette='Set1')
Out[18]: <AxesSubplot:xlabel='purpose', ylabel='count'>
```



```
In [20]:
           df.dtypes
          credit.policy
                                   int64
Out[20]:
          purpose
                                 object
          int.rate
                                 float64
                                 float64
          installment
                                 float64
          log.annual.inc
          dti
                                 float64
          fico
                                   int64
          days.with.cr.line
                                 float64
          revol.bal
                                   int64
          revol.util
                                 float64
          inq.last.6mths
                                   int64
          delinq.2yrs
                                   int64
          pub.rec
                                   int64
          not.fully.paid
                                   int64
          dtype: object
In [23]:
           cat = ['purpose']
           cat
          ['purpose']
Out[23]:
In [24]:
           final_data=pd.get_dummies(df,columns=cat,drop_first=True)
           final_data
Out[24]:
                credit.policy
                            int.rate installment log.annual.inc
                                                                    fico
                                                                         days.with.cr.line
                                                                                        revol.bal revo
                                                                dti
```

1

1

1

0.1189

0.1071

0.1357

829.10

228.22

366.86

11.350407 19.48

10.373491 11.63

14.29

11.082143

737

707

682

5639.958333

2760.000000

4710.000000

28854

33623

3511

0

1

2

| | credit.policy | int.rate | installment | log.annual.inc | dti | fico | days.with.cr.line | revol.bal | revo |
|------|---------------|----------|-------------|----------------|-------|------|-------------------|-----------|------|
| 3 | 1 | 0.1008 | 162.34 | 11.350407 | 8.10 | 712 | 2699.958333 | 33667 | |
| 4 | 1 | 0.1426 | 102.92 | 11.299732 | 14.97 | 667 | 4066.000000 | 4740 | |
| ••• | | | | | | | | | |
| 9573 | 0 | 0.1461 | 344.76 | 12.180755 | 10.39 | 672 | 10474.000000 | 215372 | |
| 9574 | 0 | 0.1253 | 257.70 | 11.141862 | 0.21 | 722 | 4380.000000 | 184 | |
| 9575 | 0 | 0.1071 | 97.81 | 10.596635 | 13.09 | 687 | 3450.041667 | 10036 | |
| 9576 | 0 | 0.1600 | 351.58 | 10.819778 | 19.18 | 692 | 1800.000000 | 0 | |
| 9577 | 0 | 0.1392 | 853.43 | 11.264464 | 16.28 | 732 | 4740.000000 | 37879 | |
| | | | | | | | | | |

9578 rows × 19 columns

```
In [25]: final_data.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9578 entries, 0 to 9577
Data columns (total 19 columns):

Column # Non-Null Count Dtype credit.policy 0 9578 non-null int64 1 int.rate 9578 non-null float64 2 installment 9578 non-null float64 3 log.annual.inc 9578 non-null float64 4 9578 non-null float64 dti 5 fico 9578 non-null int64 6 9578 non-null days.with.cr.line float64 7 revol.bal 9578 non-null int64 8 revol.util 9578 non-null float64 inq.last.6mths 9578 non-null 9 int64 10 deling.2yrs 9578 non-null int64 11 pub.rec 9578 non-null int64 12 not.fully.paid 9578 non-null int64 13 purpose_credit_card 9578 non-null uint8 14 purpose_debt_consolidation 9578 non-null uint8 15 purpose educational 9578 non-null uint8 16 purpose home improvement 9578 non-null uint8 17 purpose major purchase 9578 non-null uint8 18 purpose small business 9578 non-null uint8

dtypes: float64(6), int64(7), uint8(6)

memory usage: 1.0 MB

```
In [26]: x=final_data.drop('not.fully.paid',axis=1)
x
```

| Out[26]: | | credit.policy | int.rate | installment | log.annual.inc | dti | fico | days.with.cr.line | revol.bal | revo |
|----------|-----|---------------|----------|-------------|----------------|-------|------|-------------------|-----------|------|
| | 0 | 1 | 0.1189 | 829.10 | 11.350407 | 19.48 | 737 | 5639.958333 | 28854 | |
| | 1 | 1 | 0.1071 | 228.22 | 11.082143 | 14.29 | 707 | 2760.000000 | 33623 | |
| | 2 | 1 | 0.1357 | 366.86 | 10.373491 | 11.63 | 682 | 4710.000000 | 3511 | |
| | 3 | 1 | 0.1008 | 162.34 | 11.350407 | 8.10 | 712 | 2699.958333 | 33667 | |
| | 4 | 1 | 0.1426 | 102.92 | 11.299732 | 14.97 | 667 | 4066.000000 | 4740 | |
| | ••• | | | | | | | | ••• | |

0.1461

9573

credit.policy int.rate installment log.annual.inc

344.76

12.180755 10.39 672

| | 9574 | 0 | 0.1253 | 257.70 | 11.141862 | 0.21 | 722 | 4380.000000 | 184 | | |
|----------|--|-----------|--------------|--------------|--------------|--------------|-----|-------------|-------|---|--|
| | 9575 | 0 | 0.1071 | 97.81 | 10.596635 | 13.09 | 687 | 3450.041667 | 10036 | | |
| | 9576 | 0 | 0.1600 | 351.58 | 10.819778 | 19.18 | 692 | 1800.000000 | 0 | | |
| | 9577 | 0 | 0.1392 | 853.43 | 11.264464 | 16.28 | 732 | 4740.000000 | 37879 | | |
| | 9578 rows × | : 18 colu | ımns | | | | | | | | |
| | < | | | | | | | | | > | |
| In [27]: | <pre>y=final_data['not.fully.paid'] y</pre> | | | | | | | | | | |
| Out[27]: | 0 0 1 0 2 0 3 0 4 0 | | | | | | | | | | |
| | 9573 1 9574 1 9575 1 9576 1 9577 1 Name: not. | .fully.ŗ | paid, Len | gth: 9578, (| dtype: int64 | 4 | | | | | |
| In [28]: | x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=1) | | | | | | | | | | |
| In [29]: | <pre>from sklearn.tree import DecisionTreeClassifier</pre> | | | | | | | | | | |
| In [30]: | <pre>dt=DecisionTreeClassifier() dt.fit(x_train,y_train)</pre> | | | | | | | | | | |
| Out[30]: | DecisionTreeClassifier() | | | | | | | | | | |
| In [31]: | <pre>pred=dt.predict(x_test) pred</pre> | | | | | | | | | | |
| Out[31]: | array([1, 1, 0,, 0, 0], dtype=int64) | | | | | | | | | | |
| In [40]: | <pre>from sklearn.metrics import accuracy_score,classification_report print(accuracy_score(y_test,pred)) print(classification_report(y_test,pred))</pre> | | | | | | | | | | |
| | 0.7458246346555324 | | | | | | | | | | |
| | precision recall f1-score support | | | | | | | | | | |
| | | 0 1 | 0.85 0.25 | 0.84 0.26 | 0.85 0.25 | 1593 323 | | | | | |
| | accura macro a | - | 0.55 | 0.55 | 0.75 0.55 | 1916 1916 | | | | | |

dti fico days.with.cr.line revol.bal revo

215372

10474.000000

weighted avg 0.75 0.75 0.75 1916 In [34]: from sklearn.ensemble import RandomForestClassifier In [35]: rf=RandomForestClassifier() rf.fit(x_train,y_train) RandomForestClassifier() In [36]: predy=rf.predict(x_test) predy Out[36]: array([0, 1, 0, ..., 0, 0, 0], dtype=int64) In [41]: print(accuracy_score(y_test,predy)) print(classification_report(y_test,predy)) 0.8335073068893528 precision recall f1-score support 0 0.84 1.00 0.91 1593 0.61 0.03 0.06 323 accuracy 0.83 1916 macro avg 0.72 0.51 0.49 1916 weighted avg 0.80 0.83 0.77 1916 In []: In []: