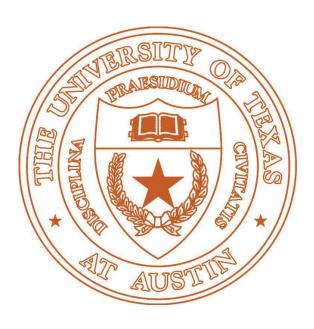
University of Texas at Austin, Cockrell School of Engineering Data Mining – EE 380L



Problem Set # 3
April 11, 2016

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Discussed Homework with Following Students:

- 1. Mudra Gandhi
- 2. Rayo Landeros

```
In [1]: # Name: Gabe Eapen
        # UT EID: eapengp
        # PS3 - Q1
In [2]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        %matplotlib inline
        from sklearn import datasets, linear model
        from pandas import DataFrame, Series
        import seaborn as sns
        sns.set(style='ticks', palette='Set2')
In [3]: def extract_int(some_string):
            int_as_string = (str(some_string)).split('.')[0]
            return int(int_as_string)
In [4]: df=pd.read stata("nes5200 processed voters realideo.dta")
Out[4]: (41498, 62)
In [5]: print(df.columns)
        u'union', u'religion', u'educ2', u'educ3', u'martial_status', u'occup2',
               u'icpsr_cty', u'fips_cty', u'partyid7', u'partyid3', u'partyid3_b',
               u'str partyid', u'father party', u'mother party', u'dlikes', u'rlikes',
               u'dem_therm', u'rep_therm', u'regis', u'vote', u'regisvote',
               u'presvote', u'presvote_2party', u'presvote_intent', u'ideo_feel',
               u'ideo7', u'ideo', u'cd', u'state', u'inter_pre', u'inter_post',
               u'black', u'female', u'age_sq', u'rep_presvote', u'rep_pres_intent',
               u'south', u'real_ideo', u'presapprov', u'perfin1', u'perfin2', u'perfin', u'presadm', u'age_10', u'age_sq_10', u'newfathe', u'newmoth',
               u'parent_party', u'white'],
              dtype='object')
In [6]: df_1992_raw = df[(df['year'] == 1992.0) & ((df['presvote'] == "1. democrat") | (df['presvote'] =
        print df 1992 raw.shape
        df_ed1992 = df.loc[(df['year'] == 1992.0) & ((df['presvote'] == "1. democrat") | (df['presvote']
        print df_ed1992.shape
         (1304, 62)
        (1304, 62)
```

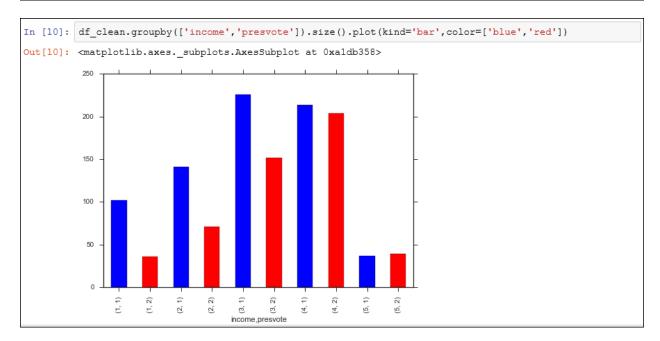
Part a]

```
In [7]: df_vote_inc = pd.DataFrame(df_ed1992,columns=['presvote','income'])
        print df_vote_inc.shape
print df_vote_inc.head()
        df_clean = df_vote_inc.dropna(how='any')
        print df_clean.shape
        #print df_clean.dtypes
        print df_clean.head()
        (1304, 2)
                    presvote
                                               income
        32092 2. republican 4. 68 to 95 percentile
        32093 2. republican 2. 17 to 33 percentile
                              1. 0 to 16 percentile
        32095

    democrat

              2. republican 2. 17 to 33 percentile
        32096
        32097
                1. democrat 3. 34 to 67 percentile
        (1222, 2)
        presvote
                    category
        income
                    category
        dtype: object
                    presvote
                                               income
        32092 2. republican 4. 68 to 95 percentile
        32093 2. republican 2. 17 to 33 percentile
        32095
                1. democrat
                              1. 0 to 16 percentile
        32096
               2. republican 2. 17 to 33 percentile
        32097
                1. democrat 3. 34 to 67 percentile
```

```
In [8]: cat_columns = df_clean.select_dtypes(['category']).columns
    #cat_columns
In [9]: df_clean[cat_columns] = df_clean[cat_columns].apply(lambda x: x.cat.codes + 1)
    #print vote_D.head()
    #df_clean.head()
```



Part b]

```
In [14]: X = df_clean['income'].values
         Y = df_clean['presvote'].values
         # Reshape as X has only single feature
         #X.reshape(-1, 1)
         print X.shape
         print Y.shape
         # create the logistic regression object
         logreg = linear model.LogisticRegression(fit intercept=True, solver='liblinear', penalty='l1')
         logreg.fit(X.reshape(-1, 1), Y)
         (1222L,)
         (1222L,)
Out[14]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                   intercept scaling=1, max iter=100, multi class='ovr', n jobs=1,
                   penalty='11', random_state=None, solver='liblinear', tol=0.0001,
                   verbose=0, warm_start=False)
In [16]: print "Coeff:",logreg.coef
         print "Intercept (B0)", logreg.intercept
         Coeff: [[ 0.29072854]]
         Intercept (B0) [-1.25875036]
```

Part c]

```
In [56]: # Part C #
In [57]: print "Probability for Republican Inc level 2", logreg.predict_proba(2)[0][1]
         print "Probability for Republican Inc level 3", logreg.predict proba(3)[0][1]
         Probability for Republican Inc level 2 0.336865682301
         Probability for Republican Inc level 3 0.404544542996
In [58]: XRepInc_2 = vote_R.loc[(vote_R['income'] == 2) ].values
         XRepInc_3 = vote_R.loc[(vote_R['income'] == 3) ].values
In [63]: print logreg.predict_proba(XRepInc_2.reshape(-1,1))
          [[ 0.66313432  0.33686568]
             0.66313432 0.33686568]
0.66313432 0.33686568]
             0.66313432 0.336865681
             0.66313432
                        0.33686568]
             0.66313432
                        0.336865681
             0.66313432 0.33686568]
             0.66313432
                        0.33686568]
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                        0.336865681
             0.66313432 0.33686568]
             0.66313432 0.33686568]
             0.66313432
                        0.336865681
             0.66313432
                        0.33686568]
             0.66313432
                        0.336865681
                        0.33686568]
             0.66313432
             0.66313432
                         0.33686568]
             0.66313432
                        0.33686568]
                         0.33686568]
             0.66313432
             0.66313432
                        0.33686568]
```

```
In [64]: print logreg.predict_proba(XRepInc_3.reshape(-1,1))
        [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         [ 0.59545546  0.40454454]
         [ 0.66313432  0.33686568]
         I O 505/55/6 O 40/5//5/1
```