```
In [494]: import pandas as pd
    from collections import Counter
    import numpy as np
    from sklearn.linear_model import LogisticRegression
    from sklearn.model_selection import train_test_split
    from sklearn.svm import SVC
    from sklearn.preprocessing import StandardScaler
    from sklearn import metrics
    from array import *
```

Data Cleaning

```
In [541]: data = pd.read_csv('C:/Users/Frank Shi/Desktop/ADS Project 4/compas-scores-two-ye
           data.head()
In [542]:
Out[542]:
               id
                              first
                      name
                                         last compas_screening_date
                                                                               dob
                                                                                         age_cat
                                                                     sex
                                                                                    age
                      miguel
                                                                                         Greater
                             miguel hernandez
                                                                         18/04/1947
               1
                                                         14/08/2013
                                                                    Male
                                                                                                    0
                  hernandez
                                                                                         than 45
                      kevon
                                                                                                  Afric
               3
                                                         27/01/2013
                                                                         22/01/1982
                                                                                          25 - 45
                             kevon
                                        dixon
                                                                   Male
                                                                                                 Amer
                      dixon
                                                                                            Less
                                                                                                  Afric
                    ed philo
                                        philo
                                                         14/04/2013
                                                                   Male
                                                                         14/05/1991
                                                                                         than 25
                                                                                                 Amer
                                                                                           Less
                                                                                                  Afric
                      marcu
            3
               5
                                       brown
                                                         13/01/2013
                                                                   Male 21/01/1993
                             marcu
                                                                                         than 25
                                                                                                 Amer
                      brown
                     bouthy
                                                                                          25 - 45
                            bouthy pierrelouis
                                                         26/03/2013 Male 22/01/1973
                                                                                     43
                                                                                                    0
                   pierrelouis
           5 rows × 53 columns
In [543]:
           ### remove rows contains other races, update AA to be 1 and Cau to be 0
           data = data[data["race"].str.contains("Other")==False]
           data['race'] = data['race'].replace(['African-American', 'Caucasian'], [1, 0])
In [544]:
           #### update vr charge degree to be dummy
           data['vr charge degree']
           data['vr_charge_degree'] = data['vr_charge_degree'].fillna('0')
           data['vr charge degree'] = data['vr charge degree'].str.contains(pat = '0')
In [545]: | Counter(data['vr_charge_degree'])
Out[545]: Counter({False: 781, True: 6056})
```

```
In [546]:
           ### DROP the following columns
           df = data.drop(['type_of_assessment','id', 'name', 'first', 'last','compas_screer
           df.head()
Out[546]:
                           age_cat race juv_fel_count decile_score juv_misd_count juv_other_count prior:
                      age
                 sex
            1
                 Male
                       34
                            25 - 45
                             Less
            2
                 Male
                       24
                                     1
                                                  0
                                                              4
                                                                            0
                                                                                           1
                           than 25
                             Less
            3
                       23
                                                  0
                                                              8
                                                                                           0
                 Male
                                     1
                                                                             1
                           than 25
                           25 - 45
                                                              6
            6
                 Male
                       41
                           25 - 45
            8 Female
                       39
                                     0
                                                  0
                                                              1
                                                                             0
                                                                                           0
           5 rows × 24 columns
In [547]:
           ##fill na with 0
           df['days_b_screening_arrest'].fillna(0, inplace=True)
           df['c_days_from_compas'].fillna(0, inplace=True)
In [548]: ###Dummie transformation
           to_dummy = ['sex', 'age_cat' ,'c_charge_degree', 'vr_charge_degree', 'v_score_tex
           dummies = pd.get dummies(df[to dummy])
           df = pd.concat([df, dummies], axis=1)
```

df = df.drop(to dummy, axis=1)

```
In [549]: df.isna().sum()
Out[549]: age
                                      0
          race
                                      0
          juv_fel_count
                                      0
          decile_score
                                      0
          juv misd count
                                      0
          juv_other_count
                                      0
          priors_count
                                      0
          days_b_screening_arrest
                                      0
          c_days_from_compas
                                      0
          is_recid
                                      0
                                      0
          is violent recid
          decile_score.1
                                      0
          v_decile_score
                                      0
          priors_count.1
                                      0
                                      0
          start
          end
                                      0
          event
                                      0
          two_year_recid
                                      0
                                      0
          sex Female
          sex_Male
                                      0
          age_cat_25 - 45
                                      0
          age_cat_Greater than 45
          age cat Less than 25
                                      0
          c charge degree F
                                      0
          c_charge_degree_M
                                      0
          v score text High
                                      0
          v_score_text_Low
                                      0
          v_score_text_Medium
                                      0
          score text High
          score text Low
                                      0
          score_text_Medium
          dtype: int64
In [550]:
          df cau = df[df["race"] == 0]
          df aa = df[df["race"] == 1]
          print(df cau.shape[0])
          print(df_aa.shape[0])
          print('Number of Cau race Commit a Crime in 2 years', df_cau[df_cau["two_year_r
          print( 'Number of AA race Commit a Crime in 2 years', df_aa[df_aa["two_year_recic
          print('Percentage of Cau race Commit a Crime in 2 years' ,df_cau[df_cau["two_year
          print('Percentage of AA race Commit a Crime in 2 years', df_aa[df_aa["two_year_re
          2454
          3696
          Number of Cau race Commit a Crime in 2 years 966
          Number of AA race Commit a Crime in 2 years 1901
          Percentage of Cau race Commit a Crime in 2 years 0.39364303178484106
          Percentage of AA race Commit a Crime in 2 years 0.5143398268398268
```

```
In [551]: ## drop race
df_cau = df_cau.drop('race', axis=1)
df_aa = df_aa.drop('race', axis=1)
```

In [552]: df

Out[552]:

	age	race	juv_fel_count	decile_score	juv_misd_count	juv_other_count	priors_count	da
1	34	1	0	3	0	0	0	
2	24	1	0	4	0	1	4	
3	23	1	0	8	1	0	1	
6	41	0	0	6	0	0	14	
8	39	0	0	1	0	0	0	
7208	20	1	0	9	0	0	0	
7209	23	1	0	7	0	0	0	
7210	23	1	0	3	0	0	0	
7212	33	1	0	2	0	0	3	
7213	23	Hispanic	0	4	0	0	2	

6837 rows × 31 columns

```
In [553]: ###base model combined every thing
    df = df.drop('race', axis=1)
    X = df.drop("two_year_recid", axis=1)
    y = df["two_year_recid"]
```

```
In [554]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
          log = LogisticRegression()
          log.fit(X train, y train)
          y_pred = log.predict(X_test)
          accuracy = metrics.accuracy_score(y_test, y_pred)
          accuracy
          C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.p
          y:814: ConvergenceWarning: lbfgs failed to converge (status=1):
          STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
          Increase the number of iterations (max_iter) or scale the data as shown in:
              https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-
          learn.org/stable/modules/preprocessing.html)
          Please also refer to the documentation for alternative solver options:
              https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
          on (https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
          on)
            n_iter_i = _check_optimize_result(
Out[554]: 0.97953216374269
```

Local Sampling method (Logitic regression)

Race: African American. Local Sampling

```
In [555]: ###data split for 2 races
X_cau = df_cau.drop("two_year_recid", axis=1)
y_cau = df_cau["two_year_recid"]
X_train_cau, X_test_cau, y_train_cau, y_test_cau = train_test_split(X_cau, y_cau,

X_aa = df_aa.drop("two_year_recid", axis=1)
y_aa = df_aa["two_year_recid"]
X_train_aa, X_test_aa, y_train_aa, y_test_aa = train_test_split(X_aa, y_aa, test_aa)
```

```
In [556]: ### Initial Logistic regression on training data for African American
          log aa = LogisticRegression()
          log_aa.fit(X_train_aa, y_train_aa)
          y pred aa = log_aa.predict(X_test_aa)
          accuracy = metrics.accuracy_score(y_test_aa, y_pred_aa)
          print('ACC for AA without resampling: ', accuracy)
          ACC for AA without resampling: 0.9648648648648649
          C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear model\ logistic.p
          y:814: ConvergenceWarning: lbfgs failed to converge (status=1):
          STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
          Increase the number of iterations (max_iter) or scale the data as shown in:
              https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-
          learn.org/stable/modules/preprocessing.html)
          Please also refer to the documentation for alternative solver options:
              https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressi
          on (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressi
          on)
            n_iter_i = _check_optimize_result(
In [557]: | ## probability table construction
          dd = log aa.predict proba(X train aa)
          (abs(dd[1,0] - dd[1,1]))
Out[557]: 0.9987336295627811
In [558]: ## calculate the logit differences
          logit diff = []
          for i in range(len(dd)):
              logit diff.append(abs(dd[i,0] - dd[i,1]))
In [559]:
          np.array(logit diff)[np.array(logit diff) <= 0.4] ###max Logit = 0.65
          print(len(np.array(logit_diff)[np.array(logit_diff) <= 0.4])) ### number of logit</pre>
          ### contains True False, length equals to number of rows in African American trai
          position = np.array(logit diff) <= 0.4</pre>
          136
In [560]:
          ##label resample for df aa
          selected rows = [] #### index with True
          not_selected_rows = [] ### index with False
          for i in range(len(position)):
              if position[i] == True :
                  selected rows.append(i)
              else:
                  not_selected_rows.append(i)
```

```
In [561]: ### X and y with distance below threshold
          selected_X = X_train_aa.iloc[selected_rows, ] ### with true in positions
          selected_y = y_train_aa.iloc[selected_rows, ]
          ### X and y with distance above threshold
          unselected_X = X_train_aa.iloc[not_selected_rows, ] ### with true in positions
          unselected_y = y_train_aa.iloc[not_selected_rows, ]
          ### merge X and y
          selected = pd.concat([selected_X, selected_y], axis=1)
          unselected = pd.concat([unselected_X, unselected_y], axis=1)
          ### Only keep rows from selected that has two_year_recid == 0
          ### Duplicate kept rows by c
          selected = selected[selected.two_year_recid == 0] ######remain the labels with @
          repeated = pd.concat([selected]*4, ignore_index=True)
          ### merge duplicated rows and unselected rows vertically
          df_aa_train_new = pd.concat([unselected, repeated], axis=0)
          (df_aa_train_new)
```

Out[561]:

	age	juv_fel_count	decile_score	juv_misd_count	juv_other_count	priors_count	days_b_scre
2793	60	0	8	0	0	9	
6376	20	0	6	0	0	2	
3255	23	0	9	0	0	6	
6947	23	2	6	0	0	3	
4772	39	0	6	0	0	19	
		•••				•••	
275	22	0	5	0	0	1	
276	36	0	3	0	0	5	
277	31	0	8	0	0	14	
278	49	0	2	0	0	0	
279	61	0	2	0	0	14	

3100 rows × 30 columns

4

```
In [562]:
          print(df_aa_train_new.shape[0])
          print( 'Number of AA race Commit a Crime in 2 years', df_aa_train_new[df_aa_trair
          print('Percentage of AA race Commit a Crime in 2 years', df_aa_train_new[df_aa_tr
          3100
          Number of AA race Commit a Crime in 2 years 1440
          Percentage of AA race Commit a Crime in 2 years 0.4645161290322581
In [563]:
          print(Counter(y_train_aa))
          Counter(y_train_cau)
          770/(1184 +770)
          Counter({1: 1506, 0: 1450})
Out[563]: 0.3940634595701126
In [564]: |print(Counter(y_train_aa))
          1519/(1519+1437)
          Counter({1: 1506, 0: 1450})
Out[564]: 0.5138700947225981
```

Local resampling on Causian

```
In [565]: | ### Method 2 on Cau
          log cau = LogisticRegression()
          log_cau.fit(X_train_cau, y_train_cau)
          y_pred_cau = log_cau.predict(X_test_cau)
          accuracy = metrics.accuracy_score(y_test_cau, y_pred_cau)
          print('Acc for Cau wihtout resampling',accuracy)
          dd = log_cau.predict_proba(X_train_cau)
          ## calculate the logit differences
          logit_diff = []
          for i in range(len(dd)):
              logit_diff.append(abs(dd[i,0] - dd[i,1]))
          np.array(logit_diff)[np.array(logit_diff) <= 0.3] ###max Logit = 0.65</pre>
          print(len(np.array(logit_diff)[np.array(logit_diff) <= 0.3])) ### number of logit</pre>
          position = np.array(logit_diff) <= 0.3</pre>
          ##label resample for df aa
          selected_rows = []
          not selected rows = []
          for i in range(len(position)):
              if position[i] == True :
                   selected rows.append(i)
              else:
                  not selected rows.append(i)
          selected X = X train cau.iloc[selected rows, ] ### with true in positions
          selected_y = y_train_cau.iloc[selected_rows, ]
          unselected X = X train cau.iloc[not selected rows, ] ### with true in positions
          unselected y = y train cau.iloc[not selected rows, ]
          selected = pd.concat([selected X, selected y], axis=1)
          unselected = pd.concat([unselected_X, unselected_y], axis=1)
          selected = selected[selected.two year recid == 1] #####remain the Labels with 1
          repeated = pd.concat([selected]*5, ignore index=True)
          ### merge duplicated rows and unselected rows vertically
          df_cau_train_new = pd.concat([unselected, repeated], axis=0)
          (df cau train new)
          Acc for Cau wihtout resampling 0.9389002036659878
```

```
C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear_model\_logisti
c.py:814: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
```

Please also refer to the documentation for alternative solver options:
 https://scikit-learn.org/stable/modules/linear_model.html#logistic-regres
sion (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regr
ession)

n_iter_i = _check_optimize_result(

Out[565]:

	age	juv_fel_count	decile_score	juv_misd_count	juv_other_count	priors_count	days_b_scre
5926	69	0	5	0	0	0	
4820	25	0	3	0	0	0	
6235	57	0	1	0	0	2	
1772	54	0	1	0	0	0	
3569	32	0	3	0	0	1	
150	31	0	5	0	2	2	
151	32	0	10	0	0	2	
152	47	0	10	0	0	5	
153	29	0	7	0	0	1	
154	22	0	10	0	0	3	

2036 rows × 30 columns

In [566]:

print(df_cau_train_new.shape[0])
print('Number of CAU race Commit a Crime in 2 years', df_cau_train_new[df_cau_tr
print('Percentage of CAU race Commit a Crime in 2 years', df_cau_train_new[df_cau_train_new]

2036

Number of CAU race Commit a Crime in 2 years 896 Percentage of CAU race Commit a Crime in 2 years 0.4400785854616896

In [567]:

Overall ACC and Calibration

```
In [568]: ### merge the new training sets
          df_train_new_total = pd.concat([df_aa_train_new, df_cau_train_new], axis=0)
          df train new x = df train new total.drop("two year recid", axis=1)
          df train new y = df train new total["two year recid"]
          model2 = LogisticRegression()
          model2.fit(df_train_new_x, df_train_new_y)
          C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear model\ logistic.p
          y:814: ConvergenceWarning: lbfgs failed to converge (status=1):
          STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
          Increase the number of iterations (max iter) or scale the data as shown in:
              https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-
          learn.org/stable/modules/preprocessing.html)
          Please also refer to the documentation for alternative solver options:
              https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
          on (https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
          on)
            n_iter_i = _check_optimize_result(
Out[568]: LogisticRegression()
In [569]: ### calibration for AA
          y_pred_aa = model2.predict(X_test_aa)
          accuracy_aa = metrics.accuracy_score(y_test_aa, y_pred_aa)
          print('Accuracy for African American:' + str(accuracy_aa))
          ### calibration for Cau
          y pred cau = model2.predict(X test cau)
          accuracy_cau = metrics.accuracy_score(y_test_cau, y_pred_cau)
          print('Accuracy for Cauasin:' + str(accuracy_cau))
          ### overall acc
          print('Accuracy total:' + str((accuracy cau+accuracy aa)/2))
          Accuracy for African American: 0.972972972973
```

Local Massaging (logistic)

Accuracy total:0.9752848571585843

Accuracy for Cauasin: 0.9775967413441955

```
In [570]: ####Scratch
          X = df_aa.drop("two_year_recid", axis=1)
          y = df_aa["two_year_recid"]
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
          log = LogisticRegression()
          log.fit(X_train, y_train)
          y_pred = log.predict(X_test)
          accuracy = metrics.accuracy_score(y_test, y_pred)
          accuracy
          dd = log.predict_proba(X_train)
          (abs(dd[1,0] - dd[1,1]))
          ## calculate the logit differences
          logit_diff = []
          for i in range(len(dd)):
              logit_diff.append(abs(dd[i,0] - dd[i,1]))
          np.array(logit diff)[np.array(logit diff) <= 0.3] ###max Logit = 0.65
          print(len(np.array(logit_diff)[np.array(logit_diff) <= 0.3])) ### number of logit</pre>
          position = np.array(logit_diff) <= 0.3</pre>
          ##label update for df aa
          for i in range(len(position)):
              if position[i] == True :
                  y train.iloc[i] =0
          log2 = LogisticRegression()
          log2.fit(X_train, y_train)
          y pred = log2.predict(X test)
          accuracy = metrics.accuracy_score(y_test, y_pred)
          accuracy
```

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```
C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear_model\_logisti
c.py:814: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)
Please also refer to the documentation for alternative solver options:
    https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)
    n_iter_i = _check_optimize_result(
C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:814: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

```
Increase the number of iterations (max_iter) or scale the data as shown in:
              https://scikit-learn.org/stable/modules/preprocessing.html (https://sciki
          t-learn.org/stable/modules/preprocessing.html)
          Please also refer to the documentation for alternative solver options:
              https://scikit-learn.org/stable/modules/linear_model.html#logistic-regres
          sion (https://scikit-learn.org/stable/modules/linear model.html#logistic-regr
            n_iter_i = _check_optimize_result(
Out[570]: 0.9702702702702702
```

local massage for AA

```
In [571]: |Counter(y_train_aa)
Out[571]: Counter({0: 1450, 1: 1506})
In [572]: Counter(y_train_aa)[0]
Out[572]: 1450
In [573]: print('Number of AA race Commit a Crime in 2 years', Counter(y train aa)[1]/(Counter(y train
                                    Number of AA race Commit a Crime in 2 years 0.5094722598105548
In [574]:
                                   ### Method 1AAogit differences
                                    logit diff aa = []
                                    for i in range(len(table aa)):
                                                 logit_diff_aa.append(abs(table_aa[i,0] - table_aa[i,1]))
                                    print('Number of obervations below threshold',len(np.array(logit_diff_aa)[np.arra
                                    position aa = np.array(logit diff aa) <= 0.6
                                    ##label update
                                    for i in range(len(position_aa)):
                                                 if position aa[i] == True :
                                                               y_train_aa.iloc[i] =0
                                    print(X train aa.shape)
                                    print(len(y_train_aa))
                                    Number of obervations below threshold 244
                                    (2956, 29)
                                    2956
In [575]: Counter(y train aa) ##2956
Out[575]: Counter({0: 1563, 1: 1393})
```

```
In [576]: print('Number of AA race Commit a Crime in 2 years', Counter(y_train_aa)[1]/(Cour
```

Number of AA race Commit a Crime in 2 years 0.4712449255751015

local massage for Cau

```
In [577]:
                                     ### Method 2 on Cau
                                      y_pred_cau = log_cau.predict(X_test_cau)
                                      table_cau = log_cau.predict_proba(X_train_cau)
                                      ## calculate the logit differences
                                      logit_diff_cau = []
                                      for i in range(len(table_cau)):
                                                    logit_diff_cau.append(abs(table_cau[i,0] - table_cau[i,1]))
                                      print('Number of obervations below threshold', len(np.array(logit_diff_cau)[np.ar
                                      position_cau = np.array(logit_diff_cau) <= 0.5</pre>
                                      ##label update
                                      for i in range(len(position_cau)):
                                                    if position_cau[i] == True :
                                                                  y_train_cau.iloc[i] =1
                                      print(X train cau.shape)
                                      print(len(y_train_cau))
                                      Number of obervations below threshold 140
                                      (1963, 29)
                                      1963
In [578]: | Counter(y_train_cau) ##1960
Out[578]: Counter({0: 1106, 1: 857})
In [579]:
                                     print('Number of Cau race Commit a Crime in 2 years', Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_train_cau)[1]/(Counter(y_tr
```

Number of Cau race Commit a Crime in 2 years 0.43657666836474784

```
pd.concat([y_train_aa, y_train_cau], axis=0)
In [580]:
Out[580]: 2793
                   0
          6376
                   1
           3255
                   1
          6947
                   0
          4772
                   1
          6053
                   1
          4412
                   0
          5241
          4305
                   0
          3388
          Name: two_year_recid, Length: 4919, dtype: int64
```

Overall acc and calibration

```
In [581]: ### merge the new training sets
          X_total_new = pd.concat([X_train_aa, X_train_cau], axis=0)
          y_total_new = pd.concat([y_train_aa, y_train_cau], axis=0)
          model3 = LogisticRegression()
          model3.fit(X_total_new, y_total_new)
          C:\Users\Frank Shi\anaconda3\lib\site-packages\sklearn\linear model\ logistic.p
          y:814: ConvergenceWarning: lbfgs failed to converge (status=1):
          STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
          Increase the number of iterations (max iter) or scale the data as shown in:
              https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-
          learn.org/stable/modules/preprocessing.html)
          Please also refer to the documentation for alternative solver options:
              https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
          on (https://scikit-learn.org/stable/modules/linear model.html#logistic-regressi
          on)
            n iter i = check optimize result(
Out[581]: LogisticRegression()
```

```
In [582]: ### calibration for AA
y_pred_aa = model3.predict(X_test_aa)
accuracy_aa = metrics.accuracy_score(y_test_aa, y_pred_aa)
print('Accuracy for African American:' + str(accuracy_aa))

### calibration for Cau
y_pred_cau = model3.predict(X_test_cau)
accuracy_cau = metrics.accuracy_score(y_test_cau, y_pred_cau)
print('Accuracy for Cauasin:' + str(accuracy_cau))

### overall acc
print('Accuracy total:' + str((accuracy_cau+accuracy_aa)/2))
```

Accuracy for African American:0.9662162162162162 Accuracy for Cauasin:0.9592668024439919 Accuracy total:0.962741509330104