

# Project Overview

Build a predictive model that identifies the factors most likely to result in a consumer disputing a financial institution's response to their complaint.

## Background

- In July 2011, the Consumer Financial Protection Bureau (CFPB) was formed in large part due to the factors leading to the Great Recession (Dec 2007 – June 2009)
- Consumers can file a complaint against a financial institution; financial institutions have 15 days to respond to the complaint

## Hypothesis

Consumers are more likely to dispute the financial institution's response to their complaint if:

- They are located in CA, TX, or FL
- Complaint focuses on mortgage
- Bank is BofA or Wells Fargo
- Complaint is submitted via web or referral

## Data Overview/Cleansing

File obtained from data.gov

- csv file
- 679,879 rows x 18 columns
- All categorical data
- dtypes = object except for one column which was int64 ('complaint id')
- Identify data (columns) needed for analysis; removed all others
- Removed null values
- Selected top ~15 values: Product, Sub-Product, Issue, State (otherwise data frame would have been 4,000+ columns due to dummy variables)
- Changed 'Consumer disputed?' to binary
- Created dummy variables for all factors except target
- RESULT = dataframe with 99 columns x 275K rows

## Analysis and Results

- Split data into train and test (80%, 20%)
- Target variable > 'Consumer disputed?'
- 98 columns of dummy variables for factors
- Used Logistic Regression CV

Based on the coefficients for each factor, this model predicts that a consumer is more likely to dispute the financial institution's response if (factors in order):

- Consumer lives in: CA, TX, CO, AZ

- Complaint is submitted via web, fax, or email
- Financial institution is: BofA, JP Morgan, Ocwen, Wells Fargo
- Consumer complaint topic: Mortgage, Home Equity Loan or Credit Line
- Specific issues: Loan servicing/payments/escrow, application/mortgage broker
- Complaint submitted in: October, September, or November

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

```
In [2]: df = pd.read_csv('Consumer_Complaints1.csv')

/Users/krys/anaconda2/lib/python2.7/site-packages/IPython/core/interactiveshell.py:2717: DtypeWarning: Columns (5,11) have mixed types. Specify dtype option on import or set low_memory=False.
interactivity=interactivity, compiler=compiler, result=result)
```

```
In [3]: df.head(3)
```

```
Out[3]:
```

	Date received	Product	Sub-product	Issue	Sub-issue	Consumer complaint narrative	Company public response	Company
0	07/29/2013	Consumer Loan	Vehicle loan	Managing the loan or lease	NaN	NaN	NaN	Wells Fargo & Company
1	07/29/2013	Bank account or service	Checking account	Using a debit or ATM card	NaN	NaN	NaN	Wells Fargo & Company
2	07/29/2013	Bank account or service	Checking account	Account opening, closing, or management	NaN	NaN	NaN	Santander Bank US

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 679879 entries, 0 to 679878
Data columns (total 18 columns):
Date received                679879 non-null object
Product                     679879 non-null object
Sub-product                 478762 non-null object
Issue                       679878 non-null object
Sub-issue                   274342 non-null object
Consumer complaint narrative 118315 non-null object
Company public response      150445 non-null object
Company                     679879 non-null object
State                       674511 non-null object
ZIP code                    674498 non-null object
Tags                        96164 non-null object
Consumer consent provided?  215359 non-null object
Submitted via               679878 non-null object
Date sent to company        679879 non-null object
Company response to consumer 679879 non-null object
Timely response?            679879 non-null object
Consumer disputed?          639285 non-null object
Complaint ID                679879 non-null int64
dtypes: int64(1), object(17)
memory usage: 93.4+ MB
```

```
In [5]: # Remove these columns: Sub-issue (4), Consumer Complaint Narrative (5),
#Company Public Response (6), Zip Code (8), Tags (9), Consumer consent p
rovided? (10),
#Date sent to company (12), Complaint ID (17)
```

```
In [6]: df.drop(df.columns[[4,5,6]], axis=1, inplace=True)
```

```
In [7]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 679879 entries, 0 to 679878
Data columns (total 15 columns):
Date received          679879 non-null object
Product                679879 non-null object
Sub-product            478762 non-null object
Issue                  679878 non-null object
Company                679879 non-null object
State                  674511 non-null object
ZIP code                674498 non-null object
Tags                    96164 non-null object
Consumer consent provided? 215359 non-null object
Submitted via          679878 non-null object
Date sent to company   679879 non-null object
Company response to consumer 679879 non-null object
Timely response?       679879 non-null object
Consumer disputed?     639285 non-null object
Complaint ID           679879 non-null int64
dtypes: int64(1), object(14)
memory usage: 77.8+ MB
```

```
In [8]: df.drop(df.columns[[6,7,8,10,12,14]], axis=1, inplace=True)
```

```
In [9]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 679879 entries, 0 to 679878
Data columns (total 9 columns):
Date received          679879 non-null object
Product                679879 non-null object
Sub-product            478762 non-null object
Issue                  679878 non-null object
Company                679879 non-null object
State                  674511 non-null object
Submitted via          679878 non-null object
Company response to consumer 679879 non-null object
Consumer disputed?     639285 non-null object
dtypes: object(9)
memory usage: 46.7+ MB
```

```
In [10]: df['Product'].value_counts()
```

```
Out[10]: Mortgage                212178
Debt collection                 126369
Credit reporting              120998
Credit card                   80119
Bank account or service       77253
Consumer Loan                 27101
Student loan                  22083
Payday loan                   4893
Money transfers               4792
Prepaid card                  3242
Other financial service        836
Virtual currency               15
Name: Product, dtype: int64
```

```
In [11]: threshold = 1000
for col in df.columns:
    value_counts = df['Product'].value_counts()
    to_remove = value_counts[value_counts <= threshold].index
    df['Product'].replace(to_remove, inplace=True)
print df['Product'].value_counts()
```

```
Mortgage                212358
Debt collection                 126586
Credit reporting              121152
Credit card                   80223
Bank account or service       77350
Consumer Loan                 27143
Student loan                  22117
Payday loan                   4897
Money transfers               4802
Prepaid card                  3251
Name: Product, dtype: int64
```

```
In [12]: df['Sub-product'].value_counts()
```

```
Out[12]: Other mortgage                82264
Conventional fixed mortgage          65690
Checking account                    53724
Other (i.e. phone, health club, etc.) 37853
I do not know                       26429
Credit card                        25464
Conventional adjustable mortgage (ARM) 23806
FHA mortgage                       22582
Non-federal student loan            20938
Medical                            17504
Vehicle loan                        15386
Other bank product/service          15016
Payday loan                         11721
Home equity loan or line of credit  10674
Installment loan                    7241
Savings account                     4750
VA mortgage                         4608
Mortgage                           4359
Federal student loan servicing      3716
Auto                               3193
(CD) Certificate of deposit         3191
International money transfer        2508
Domestic (US) money transfer        2299
Vehicle lease                       2221
Federal student loan                2168
Reverse mortgage                   1892
Personal line of credit             1717
General purpose card                1489
Second mortgage                    662
Cashing a check without an account  572
Title loan                         454
Payroll card                       374
Mobile wallet                      345
Gift or merchant card               331
Government benefit payment card     313
Debt settlement                    238
Check cashing                      212
ID prepaid card                    188
Other special purpose card          163
Money order                        118
Pawn loan                          82
Credit repair                      81
Traveler's/Cashier's checks        72
Refund anticipation check           60
Foreign currency exchange           55
Transit card                       33
Electronic Benefit Transfer / EBT card 6
Name: Sub-product, dtype: int64
```

```
In [13]: threshold = 4500
for col in df.columns:
    value_counts = df['Sub-product'].value_counts()
    to_remove = value_counts[value_counts <= threshold].index
    df['Sub-product'].replace(to_remove, inplace=True)
print df['Sub-product'].value_counts()
```

Other mortgage	119633
Conventional fixed mortgage	99733
Checking account	80711
Other (i.e. phone, health club, etc.)	60339
I do not know	41208
Credit card	39579
Conventional adjustable mortgage (ARM)	36262
FHA mortgage	34552
Non-federal student loan	31949
Medical	28453
Vehicle loan	23804
Other bank product/service	23387
Payday loan	18079
Home equity loan or line of credit	16298
Installment loan	11539
VA mortgage	7212
Savings account	7141
Name: Sub-product, dtype: int64	

```
In [14]: df['Issue'].value_counts()
```



```

Out[14]: Loan modification, collection, foreclosure      107093
          Incorrect information on credit report         88243
          Loan servicing, payments, escrow account      70979
          Cont'd attempts collect debt not owed         52502
          Account opening, closing, or management      33832
          Disclosure verification of debt               25173
          Communication tactics                         21621
          Deposits and withdrawals                     20618
          Application, originator, mortgage broker     15702
          Credit reporting company's investigation     14178
          Billing disputes                              13374
          Other                                         13262
          Managing the loan or lease                   12973
          Problems caused by my funds being low        10785
          Dealing with my lender or servicer           10546
          False statements or representation           10174
          Unable to get credit report/credit score      9870
          Improper contact or sharing of info           8938
          Problems when you are unable to pay           8281
          Settlement process and costs                  8182
          Taking/threatening an illegal action          7961
          Identity theft / Fraud / Embezzlement       7493
          Making/receiving payments, sending money     6580
          Closing/Cancelling account                   5730
          Using a debit or ATM card                    5438
          Can't repay my loan                          5418
          Credit decision / Underwriting                5188
          APR or interest rate                         5183
          Improper use of my credit report              4799
          Credit monitoring or identity protection      3908
          ...
          Privacy                                       443
          Bankruptcy                                   426
          Payment to acct not credited                 390
          Applied for loan/did not receive money       321
          Arbitration                                  311
          Sale of account                              311
          Shopping for a line of credit                 287
          Charged bank acct wrong day or amt           260
          Wrong amount charged or received             244
          Cash advance                                 232
          Customer service/Customer relations          231
          Fees                                           205
          Overlimit fee                                 202
          Balance transfer fee                          196
          Adding money                                  185
          Cash advance fee                             184
          Incorrect/missing disclosures or info        181
          Convenience checks                           132
          Excessive fees                               85
          Unexpected/Other fees                        84
          Lender repossessed or sold the vehicle       68
          Advertising, marketing or disclosures         66
          Overdraft, savings or rewards features       46
          Lost or stolen check                         37
          Lost or stolen money order                   36
          Disclosures                                  35

```

Incorrect exchange rate	22
Lender sold the property	7
Lender damaged or destroyed vehicle	6
Lender damaged or destroyed property	1

Name: Issue, dtype: int64

```
In [15]: threshold = 10000
for col in df.columns:
    value_counts = df['Issue'].value_counts()
    to_remove = value_counts[value_counts <= threshold].index
    df['Issue'].replace(to_remove, inplace=True)
print df['Issue'].value_counts()
```

Loan modification, collection, foreclosure	139974
Incorrect information on credit report	114181
Loan servicing, payments, escrow account	93172
Cont'd attempts collect debt not owed	68129
Account opening, closing, or management	44140
Disclosure verification of debt	32598
Communication tactics	28170
Deposits and withdrawals	27000
Application, originator, mortgage broker	20801
Credit reporting company's investigation	18529
Billing disputes	17587
Other	17476
Managing the loan or lease	17107
Problems caused by my funds being low	14067
Dealing with my lender or servicer	13682
False statements or representation	13266

Name: Issue, dtype: int64

```
In [16]: df['State'].value_counts()
```

```
Out[16]: CA      99006
         FL      65497
         TX      51656
         NY      46939
         GA      31050
         NJ      27278
         IL      24570
         PA      24448
         VA      22013
         MD      21459
         OH      20940
         NC      19035
         MI      17168
         AZ      15281
         WA      14140
         MA      13284
         CO      11761
         TN      10436
         MO       8974
         SC       8606
         NV       8401
         OR       8076
         CT       7907
         MN       7855
         IN       7520
         WI       7319
         AL       7061
         LA       6821
         KY       4841
         OK       4532
         ...
         DE       3484
         NM       3391
         KS       3286
         NH       3262
         MS       3185
         AR       2915
         IA       2896
         ID       2341
         HI       2321
         ME       2248
         RI       2187
         NE       2101
         WV       1760
         PR       1603
         MT       1103
         VT       1101
         SD        926
         AK        790
         WY        673
         ND        580
         AE        279
         AP        184
         VI        170
         GU         90
         FM         41
         MH         28
```

```

MP      25
AS      21
AA      13
PW      11
Name: State, dtype: int64

```

```

In [17]: threshold = 10000
        for col in df.columns:
            value_counts = df['State'].value_counts()
            to_remove = value_counts[value_counts <= threshold].index
            df['State'].replace(to_remove, inplace=True)
        print df['State'].value_counts()

```

```

CA      125638
FL      83251
TX      65341
NY      59612
GA      39427
NJ      34645
IL      31057
PA      30780
VA      27782
MD      27310
OH      26720
NC      24096
MI      21786
AZ      19431
WA      17951
MA      16837
CO      14964
TN      13251
Name: State, dtype: int64

```

```

In [18]: #df1 = df.dropna(subset=[['Sub-product']])

```

```

In [19]: df1 = df.dropna(subset=[['Consumer disputed?']]).copy()

```

```

In [20]: df1.shape

```

```

Out[20]: (639285, 9)

```

```

In [21]: df1['Consumer disputed?'].value_counts()

```

```

Out[21]: No      504944
        Yes      134341
        Name: Consumer disputed?, dtype: int64

```

```
In [22]: df1['Consumer disputed?'] = df1['Consumer disputed?'].map({'No':0,
'Yes':1})
# df1['Consumer disputed?'].map({'No':0, 'Yes':1}).head()
df1['Consumer disputed?'].head()
```

```
Out[22]: 0    0
         1    0
         2    0
         3    0
         4    0
         Name: Consumer disputed?, dtype: int64
```

```
In [23]: df1['Consumer disputed?'].shape
```

```
Out[23]: (639285,)
```

```
In [24]: dummy_product = pd.get_dummies(df1[['Product']], prefix='Product')
print dummy_product.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 639285 entries, 0 to 679878
Data columns (total 10 columns):
Product_Bank account or service    639285 non-null float64
Product_Consumer Loan              639285 non-null float64
Product_Credit card                639285 non-null float64
Product_Credit reporting           639285 non-null float64
Product_Debt collection             639285 non-null float64
Product_Money transfers            639285 non-null float64
Product_Mortgage                   639285 non-null float64
Product_Payday loan                639285 non-null float64
Product_Prepaid card               639285 non-null float64
Product_Student loan               639285 non-null float64
dtypes: float64(10)
memory usage: 53.7 MB
None
```

```
In [25]: dummy_product = pd.get_dummies(df1[['Product']], prefix='Product')
print dummy_product.head()
```

	Product_Bank account or service	Product_Consumer Loan	
0	0.0	1.0	
1	1.0	0.0	
2	1.0	0.0	
3	1.0	0.0	
4	0.0	0.0	

  

	Product_Credit card	Product_Credit reporting	Product_Debt collecti
0	0.0	0.0	
1	0.0	0.0	
2	0.0	0.0	
3	0.0	0.0	
4	0.0	0.0	

  

	Product_Money transfers	Product_Mortgage	Product_Payday loan
0	0.0	0.0	0.0
1	0.0	0.0	0.0
2	0.0	0.0	0.0
3	0.0	0.0	0.0
4	0.0	1.0	0.0

  

	Product_Prepaid card	Product_Student loan
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

```
In [26]: dummy_product.isnull().values.any()
```

```
Out[26]: False
```

```
In [27]: dummy_subproduct = pd.get_dummies(df1[['Sub-product']], prefix='Sub-pro  
d')  
print dummy_subproduct.head()
```



	Sub-prod_Checking account (ARM) \	Sub-prod_Conventional adjustable mortgage
0	0.0	0.0
1	0.0	1.0
2	0.0	1.0
3	0.0	1.0
4	0.0	0.0

	Sub-prod_Conventional fixed mortgage	Sub-prod_Credit card \
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	1.0	0.0

	Sub-prod_FHA mortgage \	Sub-prod_Home equity loan or line of credit
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

	Sub-prod_I do not know \	Sub-prod_Installment loan	Sub-prod_Medical
0	0.0	0.0	0.0
1	0.0	0.0	0.0
2	0.0	0.0	0.0
3	0.0	0.0	0.0
4	0.0	0.0	0.0

	Sub-prod_Non-federal student loan \
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

	Sub-prod_Other (i.e. phone, health club, etc.) \
0	0.0
1	0.0
2	0.0

3		0.0
4		0.0

  

	Sub-prod_Other bank product/service	Sub-prod_Other mortgage \
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

  

	Sub-prod_Payday loan	Sub-prod_Savings account	Sub-prod_VA mortgage
0	0.0	0.0	0.0
1	0.0	0.0	0.0
2	0.0	0.0	0.0
3	0.0	0.0	0.0
4	0.0	0.0	0.0

  

	Sub-prod_Vehicle loan
0	1.0
1	0.0
2	0.0
3	0.0
4	0.0

```
In [28]: dummy_subproduct.isnull().values.any()
```

```
Out[28]: False
```

```
In [29]: dummy_issue = pd.get_dummies(df1[['Issue']], prefix='Issue')  
         print dummy_issue.head()
```

Issue_Account opening, closing, or management \	
0	0.0
1	0.0
2	1.0
3	0.0
4	0.0

  

Issue_Application, originator, mortgage broker		Issue_Billing disput
es \		
0		0.0
0.0		
1		0.0
0.0		
2		0.0
0.0		
3		0.0
0.0		
4		0.0
0.0		

  

Issue_Communication tactics		Issue_Cont'd attempts collect debt not
owed \		
0	0.0	
0.0		
1	0.0	
0.0		
2	0.0	
0.0		
3	0.0	
0.0		
4	0.0	
0.0		

  

Issue_Credit reporting company's investigation \	
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

  

Issue_Dealing with my lender or servicer		Issue_Deposits and withdra
wals \		
0	0.0	
0.0		
1	0.0	
0.0		
2	0.0	
0.0		
3	0.0	
1.0		
4	0.0	
0.0		

  

Issue_Disclosure verification of debt \	
0	0.0
1	0.0
2	0.0

3	0.0	
4	0.0	
Issue_False statements or representation \		
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	
Issue_Incorrect information on credit report \		
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	
Issue_Loan modification,collection,foreclosure \		
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	
Issue_Loan servicing, payments, escrow account \		
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	1.0	
Issue_Managing the loan or lease    Issue_Other \		
0	1.0	0.0
1	1.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0
Issue_Problems caused by my funds being low		
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	

```
In [30]: dummy_comp_response = pd.get_dummies(df1[['Company response to
consumer']],
                                              prefix='Company_response')
print dummy_comp_response.head()
```

	Company_response_Closed	Company_response_Closed with explanation \
0	0.0	1.0
1	0.0	1.0
2	1.0	0.0
3	0.0	1.0
4	0.0	1.0

	Company_response_Closed with monetary relief \
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

	Company_response_Closed with non-monetary relief \
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

	Company_response_Closed with relief	Company_response_Closed without relief \
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

	Company_response_Untimely response
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0

```
In [31]: dummy_submittedvia = pd.get_dummies(df1[['Submitted via']], prefix='SubmitVia')
print dummy_submittedvia.head()
```

	SubmitVia_Email	SubmitVia_Fax	SubmitVia_Phone	SubmitVia_Postal
0	0.0	0.0	1.0	
1	0.0	0.0	0.0	
2	0.0	1.0	0.0	
3	0.0	0.0	0.0	
4	0.0	0.0	0.0	

  

	SubmitVia_Referral	SubmitVia_Web
0	0.0	0.0
1	0.0	1.0
2	0.0	0.0
3	0.0	1.0
4	0.0	1.0

```
In [32]: dummy_state = pd.get_dummies(df1[['State']], prefix='State')
print dummy_state.head()
```

	State_AZ	State_CA	State_CO	State_FL	State_GA	State_IL	State_MA
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
1	0.0	1.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0
3	0.0	0.0	0.0	0.0	1.0	0.0	0.0
4	0.0	0.0	0.0	0.0	1.0	0.0	0.0

  

	State_MD	State_MI	State_NC	State_NJ	State_NY	State_OH	State_PA
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	1.0	0.0	0.0
3	0.0	0.0	0.0	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0

  

	State_TN	State_TX	State_VA	State_WA
0	0.0	0.0	1.0	0.0
1	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0
3	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0

```
In [33]: df1['Company'].value_counts().shape
```

```
Out[33]: (3573,)
```

```
In [34]: df1['Company'].value_counts().head(12)
```

```
Out[34]: Bank of America          59933
Wells Fargo & Company          46699
Equifax                        39645
JPMorgan Chase & Co.          37973
Experian                       36715
TransUnion Intermediate Holdings, Inc. 31106
Citibank                       30171
Ocwen                          22417
Capital One                    17571
Nationstar Mortgage           14596
Synchrony Financial            11039
U.S. Bancorp                   10803
Name: Company, dtype: int64
```



```
In [35]: banks = ['Bank of America',
                  'Wells Fargo & Company',
                  'JPMorgan Chase & Co.',
                  'Ocwen',
                  'Citibank',
                  'Nationstar Mortgage',
                  'Ditech Financial LLC',
                  'Navient Solutions, Inc.',
                  'U.S. Bancorp',
                  'PNC Bank N.A.',
                  'Encore Capital Group',
                  'Capital One']

df2 = df1[df1.Company.isin(banks)]
print df2['Company'].value_counts()
```

Bank of America	59933
Wells Fargo & Company	46699
JPMorgan Chase & Co.	37973
Citibank	30171
Ocwen	22417
Capital One	17571
Nationstar Mortgage	14596
U.S. Bancorp	10803
Ditech Financial LLC	10069
Navient Solutions, Inc.	9928
PNC Bank N.A.	7556
Encore Capital Group	6929

Name: Company, dtype: int64

```
In [36]: df2.shape
```

```
Out[36]: (274645, 9)
```

```
In [37]: dummy_company = pd.get_dummies(df2[['Company']], prefix='Company')
print dummy_company.head()
```

	Company_Bank of America	Company_Capital One	Company_Citibank	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
3	0.0	0.0	0.0	
5	1.0	0.0	0.0	
8	0.0	0.0	1.0	

	Company_Ditech Financial LLC	Company_Encore Capital Group	\
0	0.0	0.0	
1	0.0	0.0	
3	0.0	0.0	
5	0.0	0.0	
8	0.0	0.0	

	Company_JPMorgan Chase & Co.	Company_Nationstar Mortgage	\
0	0.0	0.0	
1	0.0	0.0	
3	0.0	0.0	
5	0.0	0.0	
8	0.0	0.0	

	Company_Navient Solutions, Inc.	Company_Ocwen	Company_PNC Bank N.A.	\
0	0.0	0.0	0.	
0				
1	0.0	0.0	0.	
0				
3	0.0	0.0	0.	
0				
5	0.0	0.0	0.	
0				
8	0.0	0.0	0.	
0				

	Company_U.S. Bancorp	Company_Wells Fargo & Company
0	0.0	1.0
1	0.0	1.0
3	0.0	1.0
5	0.0	0.0
8	0.0	0.0

In [38]: df2.head(3)

Out[38]:

	Date received	Product	Sub-product	Issue	Company	State	Submitted via	Company response to consumer
0	07/29/2013	Consumer Loan	Vehicle loan	Managing the loan or lease	Wells Fargo & Company	VA	Phone	Closed with explanation
1	07/29/2013	Bank account or service	Checking account	Managing the loan or lease	Wells Fargo & Company	CA	Web	Closed with explanation
3	07/29/2013	Bank account or service	Checking account	Deposits and withdrawals	Wells Fargo & Company	GA	Web	Closed with explanation

```
In [80]: from datetime import datetime
from dateutil.parser import parse
df2copy = df2.copy()
df2copy['Date received'] = pd.to_datetime(df2copy['Date received'], format='%m/%d/%Y')

print df2copy['Date received'].head()

0    2013-07-29
1    2013-07-29
3    2013-07-29
5    2013-07-29
8    2013-07-29
Name: Date received, dtype: datetime64[ns]
```

In [81]: df2copy['Date received'] = df2copy['Date received'].dt.month

In [82]: df2copy.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 274645 entries, 0 to 679867
Data columns (total 9 columns):
Date received                274645 non-null int64
Product                      274645 non-null object
Sub-product                  274645 non-null object
Issue                        274645 non-null object
Company                      274645 non-null object
State                        274645 non-null object
Submitted via                274645 non-null object
Company response to consumer 274645 non-null object
Consumer disputed?           274645 non-null int64
dtypes: int64(2), object(7)
memory usage: 21.0+ MB
```

In [42]: `df2copy.head()`

Out[42]:

	Date received	Product	Sub-product	Issue	Company	State	Submitted via	Company response to consumer	Credit
0	7	Consumer Loan	Vehicle loan	Managing the loan or lease	Wells Fargo & Company	VA	Phone	Closed with explanation	0
1	7	Bank account or service	Checking account	Managing the loan or lease	Wells Fargo & Company	CA	Web	Closed with explanation	0
3	7	Bank account or service	Checking account	Deposits and withdrawals	Wells Fargo & Company	GA	Web	Closed with explanation	0
5	7	Bank account or service	Checking account	Deposits and withdrawals	Bank of America	TX	Web	Closed with explanation	0
8	7	Credit card	I do not know	Cont'd attempts collect debt not owed	Citibank	OH	Referral	Closed with explanation	1

In [43]: `dummy_daterecd = pd.get_dummies(df2copy['Date received'], prefix='Date_Recd')`  
`print dummy_daterecd.head()`

```

    Date_Recd_1  Date_Recd_2  Date_Recd_3  Date_Recd_4  Date_Recd_5  \
0             0.0           0.0           0.0           0.0           0.0
1             0.0           0.0           0.0           0.0           0.0
3             0.0           0.0           0.0           0.0           0.0
5             0.0           0.0           0.0           0.0           0.0
8             0.0           0.0           0.0           0.0           0.0

    Date_Recd_6  Date_Recd_7  Date_Recd_8  Date_Recd_9  Date_Recd_10  \
0             0.0           1.0           0.0           0.0           0.0
1             0.0           1.0           0.0           0.0           0.0
3             0.0           1.0           0.0           0.0           0.0
5             0.0           1.0           0.0           0.0           0.0
8             0.0           1.0           0.0           0.0           0.0

    Date_Recd_11  Date_Recd_12
0             0.0           0.0
1             0.0           0.0
3             0.0           0.0
5             0.0           0.0
8             0.0           0.0

```

```
In [44]: df1['Consumer disputed?'].shape
```

```
Out[44]: (639285,)
```

```
In [45]: dummy_product.shape
```

```
Out[45]: (639285, 10)
```

```
In [46]: dummy_subproduct.shape
```

```
Out[46]: (639285, 17)
```

```
In [47]: data1 = pd.concat([dummy_product, dummy_subproduct], axis=1, join='inner')  
        print data1.head()
```

	Product_Bank account or service	Product_Consumer Loan	\
0	0.0	1.0	
1	1.0	0.0	
2	1.0	0.0	
3	1.0	0.0	
4	0.0	0.0	

  

	Product_Credit card	Product_Credit reporting	Product_Debt collecti	on	\
0	0.0	0.0			
0.0					
1	0.0	0.0			
0.0					
2	0.0	0.0			
0.0					
3	0.0	0.0			
0.0					
4	0.0	0.0			
0.0					

  

	Product_Money transfers	Product_Mortgage	Product_Payday loan	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
2	0.0	0.0	0.0	
3	0.0	0.0	0.0	
4	0.0	1.0	0.0	

  

	Product_Prepaid card	Product_Student loan	...	\
0	0.0	0.0	...	
1	0.0	0.0	...	
2	0.0	0.0	...	
3	0.0	0.0	...	
4	0.0	0.0	...	

  

	Sub-prod_Installment loan	Sub-prod_Medical	\
0	0.0	0.0	
1	0.0	0.0	
2	0.0	0.0	
3	0.0	0.0	
4	0.0	0.0	

  

	Sub-prod_Non-federal student loan	\
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	

  

	Sub-prod_Other (i.e. phone, health club, etc.)	\
0	0.0	
1	0.0	
2	0.0	
3	0.0	
4	0.0	

  

	Sub-prod_Other bank product/service	Sub-prod_Other mortgage	\
0	0.0	0.0	

1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

	Sub-prod_Payday loan	Sub-prod_Savings account	Sub-prod_VA mortgage
0	0.0	0.0	0.0
1	0.0	0.0	0.0
2	0.0	0.0	0.0
3	0.0	0.0	0.0
4	0.0	0.0	0.0

	Sub-prod_Vehicle loan
0	1.0
1	0.0
2	0.0
3	0.0
4	0.0

[5 rows x 27 columns]



```
In [48]: data = pd.concat([dummy_product,
                           dummy_subproduct,
                           dummy_issue,
                           dummy_comp_response,
                           dummy_submittedvia,
                           dummy_state,
                           dummy_company,
                           dummy_daterecd,
                           df1[['Consumer disputed?']],
                           axis=1, join='inner')

print data.head()
```

	Product_Bank account or service	Product_Consumer Loan	\
0	0.0	1.0	
1	1.0	0.0	
3	1.0	0.0	
5	1.0	0.0	
8	0.0	0.0	

  

	Product_Credit card	Product_Credit reporting	Product_Debt collecti	on \
0	0.0	0.0		
0.0				
1	0.0	0.0		
0.0				
3	0.0	0.0		
0.0				
5	0.0	0.0		
0.0				
8	1.0	0.0		
0.0				

  

	Product_Money transfers	Product_Mortgage	Product_Payday loan	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
3	0.0	0.0	0.0	
5	0.0	0.0	0.0	
8	0.0	0.0	0.0	

  

	Product_Prepaid card	Product_Student loan	...	\
0	0.0	0.0	...	
1	0.0	0.0	...	
3	0.0	0.0	...	
5	0.0	0.0	...	
8	0.0	0.0	...	

  

	Date_Recd_4	Date_Recd_5	Date_Recd_6	Date_Recd_7	Date_Recd_8	\
0	0.0	0.0	0.0	1.0	0.0	
1	0.0	0.0	0.0	1.0	0.0	
3	0.0	0.0	0.0	1.0	0.0	
5	0.0	0.0	0.0	1.0	0.0	
8	0.0	0.0	0.0	1.0	0.0	

  

	Date_Recd_9	Date_Recd_10	Date_Recd_11	Date_Recd_12	Consumer disp	uted?
0	0.0	0.0	0.0	0.0		
0						
1	0.0	0.0	0.0	0.0		
0						
3	0.0	0.0	0.0	0.0		
0						
5	0.0	0.0	0.0	0.0		
0						
8	0.0	0.0	0.0	0.0		
1						

[5 rows x 99 columns]

```
In [49]: data.shape
```

```
Out[49]: (274645, 99)
```

```
In [50]: data.isnull().values.sum()
```

```
Out[50]: 0
```

```
In [51]: from sklearn.cross_validation import train_test_split

train, test = train_test_split(data, train_size=.80, test_size=.20)
```

```
In [52]: train.shape
```

```
Out[52]: (219716, 99)
```

```
In [53]: train.head(3)
```

```
Out[53]:
```

	Product_Bank account or service	Product_Consumer Loan	Product_Credit card	Product_Credit reporting	Product_Det collection
119255	0.0	0.0	0.0	0.0	0.0
423145	0.0	0.0	0.0	0.0	0.0
574464	0.0	0.0	1.0	0.0	0.0

3 rows × 99 columns

```
In [54]: from sklearn import linear_model
from sklearn.linear_model import LogisticRegressionCV
```

```
In [55]: log_model = linear_model.LogisticRegressionCV()
```

```
In [56]: log_model.fit(train.iloc[:, :-1], train.iloc[:, -1])
```

```
Out[56]: LogisticRegressionCV(Cs=10, class_weight=None, cv=None, dual=False,
    fit_intercept=True, intercept_scaling=1.0, max_iter=100,
    multi_class='ovr', n_jobs=1, penalty='l2', random_state=None,
    refit=True, scoring=None, solver='lbfgs', tol=0.0001, verbose=0)
```

```
In [57]: log_model.decision_function(train.iloc[:, :-1])
```

```
Out[57]: array([-1.17655326, -1.02948142, -1.31679575, ..., -1.32022911,
    -1.04772884, -1.13939025])
```

```
In [58]: log_model.predict(train.iloc[:, :-1])
```

```
Out[58]: array([0, 0, 0, ..., 0, 0, 0])
```

```
In [59]: log_model.predict_log_proba(train.iloc[:, :-1])
```

```
Out[59]: array([[ -0.26875891,  -1.44531217],
                [-0.30541797,  -1.33489939],
                [-0.23743505,  -1.55423081],
                ...,
                [-0.23671039,  -1.5569395 ],
                [-0.30064772,  -1.34837655],
                [-0.27764241,  -1.41703266]])
```

```
In [60]: log_model.predict_proba(train.iloc[:, :-1])
```

```
Out[60]: array([[ 0.7643275 ,  0.2356725 ],
                [ 0.73681535,  0.26318465],
                [ 0.78864811,  0.21135189],
                ...,
                [ 0.78921982,  0.21078018],
                [ 0.74033854,  0.25966146],
                [ 0.75756767,  0.24243233]])
```

```
In [83]: log_model.score(train.iloc[:, :-1], train.iloc[:, :-1])
```

```
Out[83]: 0.77530084290629719
```

```
In [84]: print(log_model.intercept_)
print(log_model.coef_)
```

```
[-1.30371132]
[[ -1.10784590e-02  2.78722046e-04  2.04892223e-03  2.81166792e-06
  -7.87324025e-03  3.83278910e-04  1.96968861e-02 -2.22135714e-04
  -5.61713447e-04 -2.67330624e-03 -9.05347954e-03  8.61544476e-03
   1.93226542e-02 -3.47025704e-03  4.56763493e-03  4.16132711e-03
  -6.54738044e-04  1.21114849e-03  3.66046259e-04 -3.65303711e-03
  -3.49980960e-03 -1.53056579e-03 -1.73104721e-02 -1.11878085e-03
   6.83470219e-04  1.43015736e-03 -6.49769320e-05 -3.26307450e-03
   6.18659502e-03  4.54995271e-03 -5.11601290e-03 -3.69592186e-03
   1.00188533e-04  1.84417473e-03 -3.37586298e-03  4.24306720e-04
   2.00597129e-03 -4.41661118e-05 -3.80752320e-03  1.51402293e-02
   1.29839966e-03 -6.92272249e-03 -5.32276756e-03  7.43015114e-04
   6.05240204e-02 -4.30834191e-02 -2.66889141e-02 -5.17091525e-03
   1.36779793e-02  0.00000000e+00 -1.99222906e-04  4.85433388e-04
  -9.97207166e-03 -5.09844517e-03 -3.74856456e-02  5.22717183e-02
  -9.83828509e-04  1.21656474e-02  3.44925518e-04 -4.36518939e-03
  -8.96650737e-03  4.69103739e-03  8.17938302e-04  2.59938597e-03
  -1.06624111e-03 -1.29909629e-03 -2.06968049e-03 -8.73662716e-04
   1.26689080e-03 -5.77244792e-03  4.03370402e-04  4.23550698e-03
   8.66344822e-04 -1.99262749e-03  6.47421411e-03 -1.12054747e-02
  -7.81283541e-03 -1.92566191e-03 -3.19437388e-03  4.48266054e-03
   1.09737027e-03 -2.52507214e-03  8.00736626e-03 -3.67412454e-03
   3.02334508e-03  7.25435267e-03 -6.84273994e-03 -1.39682098e-03
   1.31865650e-03 -3.75651762e-03 -4.85222387e-03  4.05343965e-04
   2.37748390e-04 -9.36140864e-04  4.10874928e-03  1.06486512e-02
   3.97737649e-03 -2.91031621e-03]]
```

```
In [85]: log_model.score(test.iloc[:, :-1], test.iloc[:, -1])
```

```
Out[85]: 0.77398095723570426
```

```
In [67]: test.describe()
```

```
Out[67]:
```

	Product_Bank account or service	Product_Consumer Loan	Product_Credit card	Product_Credit reporting	Product_Debt collection
count	54929.000000	54929.000000	54929.000000	54929.000000	54929.000000
mean	0.159388	0.022083	0.178612	0.002822	0.067341
std	0.366040	0.146955	0.383031	0.053046	0.250615
min	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000	0.000000
50%	0.000000	0.000000	0.000000	0.000000	0.000000
75%	0.000000	0.000000	0.000000	0.000000	0.000000
max	1.000000	1.000000	1.000000	1.000000	1.000000

8 rows × 99 columns

```
In [86]: print(log_model.intercept_)
print(log_model.coef_)
```

```
[-1.30371132]
[[ -1.10784590e-02  2.78722046e-04  2.04892223e-03  2.81166792e-06
  -7.87324025e-03  3.83278910e-04  1.96968861e-02 -2.22135714e-04
  -5.61713447e-04 -2.67330624e-03 -9.05347954e-03  8.61544476e-03
   1.93226542e-02 -3.47025704e-03  4.56763493e-03  4.16132711e-03
  -6.54738044e-04  1.21114849e-03  3.66046259e-04 -3.65303711e-03
  -3.49980960e-03 -1.53056579e-03 -1.73104721e-02 -1.11878085e-03
   6.83470219e-04  1.43015736e-03 -6.49769320e-05 -3.26307450e-03
   6.18659502e-03  4.54995271e-03 -5.11601290e-03 -3.69592186e-03
   1.00188533e-04  1.84417473e-03 -3.37586298e-03  4.24306720e-04
   2.00597129e-03 -4.41661118e-05 -3.80752320e-03  1.51402293e-02
   1.29839966e-03 -6.92272249e-03 -5.32276756e-03  7.43015114e-04
   6.05240204e-02 -4.30834191e-02 -2.66889141e-02 -5.17091525e-03
   1.36779793e-02  0.00000000e+00 -1.99222906e-04  4.85433388e-04
  -9.97207166e-03 -5.09844517e-03 -3.74856456e-02  5.22717183e-02
  -9.83828509e-04  1.21656474e-02  3.44925518e-04 -4.36518939e-03
  -8.96650737e-03  4.69103739e-03  8.17938302e-04  2.59938597e-03
  -1.06624111e-03 -1.29909629e-03 -2.06968049e-03 -8.73662716e-04
   1.26689080e-03 -5.77244792e-03  4.03370402e-04  4.23550698e-03
   8.66344822e-04 -1.99262749e-03  6.47421411e-03 -1.12054747e-02
  -7.81283541e-03 -1.92566191e-03 -3.19437388e-03  4.48266054e-03
   1.09737027e-03 -2.52507214e-03  8.00736626e-03 -3.67412454e-03
   3.02334508e-03  7.25435267e-03 -6.84273994e-03 -1.39682098e-03
   1.31865650e-03 -3.75651762e-03 -4.85222387e-03  4.05343965e-04
   2.37748390e-04 -9.36140864e-04  4.10874928e-03  1.06486512e-02
   3.97737649e-03 -2.91031621e-03]]
```

```
In [71]: print(log_model.Cs_)
```

```
[ 1.00000000e-04  7.74263683e-04  5.99484250e-03  4.64158883e-02
 3.59381366e-01  2.78255940e+00  2.15443469e+01  1.66810054e+02
 1.29154967e+03  1.00000000e+04]
```

In [72]: **print**(log\_model.coefs\_paths\_)

```
{1: array([[ -8.47263607e-03,  4.29488741e-04, -1.81079223e-03, ...,
          -7.16735501e-04, -3.14290348e-03, -1.28552490e+00],
          [ -1.80229018e-02,  3.88680673e-03,  1.26414413e-02, ...,
          -6.03323639e-03, -2.01781326e-02, -1.43938192e+00],
          [ -8.86861981e-03,  2.01316027e-02,  6.52944004e-02, ...,
          -1.47099340e-02, -6.70437698e-02, -1.57145173e+00],
          ...,
          [  3.65581062e-02,  9.85543647e-02,  1.17349736e-01, ...,
          -1.92312005e-02, -1.07181984e-01, -1.70416374e+00],
          [  3.65545780e-02,  9.85555047e-02,  1.17347627e-01, ...,
          -1.92313757e-02, -1.07183542e-01, -1.70417699e+00],
          [  3.65526480e-02,  9.85586206e-02,  1.17349288e-01, ...,
          -1.92295238e-02, -1.07184399e-01, -1.70416647e+00]],

          [[ -9.86690592e-03, -3.19869274e-04,  1.95499493e-03, ...,
            4.62589176e-03, -2.18109549e-03, -1.28460271e+00],
            [ -2.60686852e-02, -1.49515974e-03,  3.00101193e-02, ...,
            2.64005421e-02, -1.47683879e-02, -1.43602804e+00],
            [ -3.47898696e-02, -9.92056179e-03,  1.04175715e-01, ...,
            8.02386213e-02, -5.25150514e-02, -1.56839472e+00],
            ...,
            [  4.82807983e-02, -1.66120373e-02,  2.47615365e-01, ...,
            1.15019449e-01, -8.31855855e-02, -1.67532412e+00],
            [  4.82759882e-02, -1.66135411e-02,  2.47613533e-01, ...,
            1.15019716e-01, -8.31890364e-02, -1.67534614e+00],
            [  4.82773041e-02, -1.66138107e-02,  2.47616609e-01, ...,
            1.15021213e-01, -8.31893371e-02, -1.67533553e+00]],

          [[ -7.22812278e-03,  4.05957170e-04,  2.32860981e-03, ...,
            4.32383957e-03, -5.32926542e-04, -1.27994738e+00],
            [ -1.56244107e-02,  3.14740781e-03,  3.04108449e-02, ...,
            2.55029573e-02, -4.78851813e-03, -1.42122820e+00],
            [ -1.99288296e-03,  1.21258875e-02,  1.05015981e-01, ...,
            7.99314183e-02, -2.22624151e-02, -1.55467259e+00],
            ...,
            [  7.76494139e-02,  3.54834549e-02,  1.93662839e-01, ...,
            1.16987191e-01, -4.01838288e-02, -1.66666464e+00],
            [  7.75186784e-02,  3.54791254e-02,  1.93623820e-01, ...,
            1.17003092e-01, -4.02077569e-02, -1.66668284e+00],
            [  7.75719495e-02,  3.54806959e-02,  1.93644888e-01, ...,
            1.17014060e-01, -4.02132552e-02, -1.66664381e+00]]])}
```

```
In [73]: print(log_model.scores_)

{1: array([[ 0.77396101,  0.77396101,  0.77396101,  0.77396101,  0.77396101,
            0.77396101,  0.77396101,  0.77396101,  0.77396101,  0.77396101],
          [ 0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093,
            0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093],
          [ 0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093,
            0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093],
          [ 0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093,
            0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093],
          [ 0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093,
            0.77399093,  0.77399093,  0.77399093,  0.77399093,  0.77399093]])}
```

```
In [74]: print(log_model.C_)
[ 0.0001]
```

```
In [87]: print(log_model.n_iter_)

[[[ 9 11 23 35 59 48  0 36  0  0]
   [ 9 11 22 36 64 65 23  0  0  0]
   [ 9 11 20 51 58 48 12  1  2  0]]]
```

In [ ]: