## Analysis of CFPB's Consumer Complaints Dataset

## DATA COLLECTION AND CLEANING

- ▶ The CFPB updates the consumer complaints data set fortnightly. We took the latest available data set for this milestone.
- We repeated our data cleaning steps for this new data set.

## For our research question where we predict number of complaints based on financial assets of an institution:

- ▶ The financial assets data was split in multiple files, a separate file was provided for each quarter of every year from 2011-2015. We collated all data into one document.
- We mapped consumer complaints data from the consumer complaints data set to the financial data. Only institutions common to both data sets were considered.
- Since the consumer complaints data is date wise, we aggregated the data to represent quarter wise values for each year.

#### RESEARCH QUESTIONS

#### 1.

# Predicting a Company's Response to a Consumer Complaint

#### **DATA SET DETAILS:**

- Predictors: Product, Issue, State, Submitted Via, Timely Response, Consumer Disputed
- Outcome: Company Response to Consumer
- Sampling Rate: 75 (Training): 25 (Testing)

#### **TECHNIQUES APPLIED:**

- Multinomial Logistic Regression
- ➤ Naïve Bayes
- Decision Trees\*

<sup>\*</sup>Two predictors, 'Issue' and 'State' were not considered as their high number of levels was too much to handle and no model was being created by the C50 library.

#### Problems Faced

- Too many categories in the attributes:
   Company, State, Issue
- Unable to apply bagging and random forests since we had more categories than R could handle.
- The data was biased; a few categories constituted more portion of the entire dataset compared to other categories

Metric	Naïve Bayes				
ivietric	Standard Laplace = 2		Laplace =3		
Accuracy (Direct)	0.5856	0.635	0.639		
Accuracy (Sub-Sampling)	0.4733	0.4869			
Metric		Decision Trees			
Metric	Standard; size=60	Decision Trees  Boosting Trials = 5; size=16	Boosting Trials = 10; size=7		
Metric  Accuracy (Training Set)	•	Boosting Trials = 5;	•		

**Note:** For multinomial logistic regression, the predicted output variable is in the form of probability of classification for each record in the testing data set. Hence, we could not directly calculate it's accuracy as the model did not predict the class for each record in the testing data set.

#### 2.

## Likelihood of a Non-Disputed Complaint Feedback

#### **DATA SET DETAILS:**

- Predictors: Product, Issue, State, Submitted Via, Timely Response, Company Response to Consumer
- Outcome: Consumer Disputed
- Sampling Rate: 75 (Training): 25 (Testing)

#### **TECHNIQUES APPLIED:**

- Logistic Regression
- ➤ Naïve Bayes
- Decision Trees\*

<sup>\*</sup>Two predictors, 'Issue' and 'State' were not considered as their high number of levels was too much to handle and no model was being created by the C50 library.

NA . I de		Б		
Metric	Standard Laplace = 2 Laplace = 3		Decision Trees, size=1	
Accuracy (Direct)	0.5856	0.635	0.639	0.792 (training) 0.7912 (testing)

#### Logistic Regression:

	Estimate	Std. Error	z value	Pr(> z )	
(Intercept)	-1.71E+00	8.10E-01	-2.111	0.03473	*
ProductConsumer Loan	6.99E-02	4.53E-02	1.541	0.12334	
ProductCredit card	9.77E+09	1.09E+10	0.899	0.36868	
ProductCredit reporting	-4.28E-01	3.60E-02	-11.893	< 2e-16	***
ProductDebt collection	-3.07E-01	3.62E-02	-8.489	< 2e-16	***
ProductMoney transfers	9.77E+09	1.09E+10	0.899	0.36868	
ProductMortgage	9.77E+09	1.09E+10	0.899	0.36868	
ProductOther financial service	9.77E+09	1.09E+10	0.899	0.36868	
ProductPayday loan	-3.63E-01	1.46E-01	-2.479	0.01317	*
ProductPrepaid card	9.77E+09	1.09E+10	0.899	0.36868	
ProductStudent loan	3.78E-03	9.02E-02	0.042	0.96659	
ProductVirtual currency	9.77E+09	1.09E+10	0.899	0.36868	
IssueAccount terms and changes	-2.32E-02	1.37E-01	-0.17	0.86532	

**Note:** Boosting was abandoned after the first trial as there were very few classifiers. The results of the first trial were same as the one without boosting (accuracy: 0.7912).

# Predicting Medium of Receiving a Complaint

#### **DATA SET DETAILS:**

- ➤ **Predictors:** Product, Company\*, Region (derived from State), Timely Response
- Outcome: Submitted Via
- Sampling Rate: 75 (Training): 25 (Testing)

#### **TECHNIQUES APPLIED:**

- Multinomial Logistic Regression
- Naïve Bayes
- Decision Trees

\*The variable 'Company' has more than 4,000 levels and considering all was causing memory issues so we considered only the top 10 companies

	Naïve Bayes				
Metric	Standard	Standard Laplace = 2 Lap			
Accuracy (Direct)	0.5993	0.6272	0.6305		
	Decision Trees				
Metric	Standard, Boosting Trials = 5, Boosting size=9 size=1		Boosting Trials = 10		
Accuracy (Training Set)	0.6710	0.670	-		
Accuracy (Testing Set)	0.6688 > exp(coef(mlr))	0.668			

Logistic Regression:

> exp(coef(r	nlr))	•						
- cxp(coci (		ProductConsumer Loa	n ProductCredit car	d ProductCredit	reporting	ProductDebt c	ollection Product	Money transfers
Fax	26.225677	1.845041			2.9348885		387.6933	1,862786
Phone	45,581748	1.195265			0.1440963		201.3557	1,388716
Postal mail	6,990420	2.185215			6.6792403		586,4462	2,363431
Referral	20,657154	0.868242			0.6555439		219.7261	1.280046
web	1.866432	2.068805			5.3274269		801.2740	3.881994
		age ProductOther fi	nancial service Pro	ductPavday loar	ProductPr	enaid card Pro	ductStudent loan	
Fax	0.7604		3.72935525	0.3260389		2.58666068	0.2515871	
Phone	0.2177	566	1.23480645	122,9709136		2.83080270	0.1852866	
Postal mail	0.4837	196	5.82108162	0.4038667		4.56166518	0,2979563	
Referral	0.5097	922	5,27721420	0.1990887		1.66719154	0.3310162	
web	0.6358	929	0.01891042	0.3591731		0.03056313	1.5662273	
	ProductVirtu	al currency Company	Capital One Company	Citibank Compan	vEquifax C	ompanyExperian	CompanyJPMorgan	Chase & Co.
Fax		i			. 2644318	0.7065171		0.9557087
Phone		1	0.6573920 0	. 8546788 1	.7149349	0.7684238		0.8928083
Postal mail		1	0.5834590 1	.0740219 1	.8293612	1.2395461		1.0277891
Referral		1	0.4181464 0	.7703361 1	.7585093	1.0783387		0.9510592
web		1	0.6021676 0	. 8141438	.7934537	0.5215272		0.8156520
	CompanyNavie	ent Solutions, LLC.	CompanyOcwen Compan	yTransUnion Int	ermediate	Holdings, Inc.	CompanyWells Far	go & Company
Fax	and the second	88.077243121	2.027874			0.6123326		1.209494
Phone		8.588175008	1.808223			0.9099728		1.341162
Postal mail		13.986487572	2.538878			0.9793689		1.162972
Referral		0.034673046	1.537015			0.9921956		1.134088
Web		0.007125652	2.071541			0.2794377		1.101926
		East RegionSouth Re						
Fax	0.278	7082 0.3159725 0	.2570108	4.298457				
Phone	0.424	4127 0.4567746 0	. 3175527	5.492623				
Postal mail	0.369		. 3460993	6.252651				
Referral	0.504		.4271113	6.282950				
web	0.416	0.4764664 0	.4206593	41.214169				

**Note:** For trials = 10, the boosting truncated forcibly at 4 trials since the last classifier was very inaccurate. The error rate was hence similar to the results for trials = 5.

### 4.

## Predict Geographical Location Of a Complaint

#### **DATA SET DETAILS:**

- Predictors: Product, Submitted Via, Timely Response, Consumer Disputed, Consumer Response to Consumer, Company\*
- Outcome: Region (derived from State)
- Sampling Rate: 75 (Training): 25 (Testing)

#### **TECHNIQUES APPLIED:**

- Multinomial Logistic Regression
- Naïve Bayes
- Decision Trees

\*The variable 'Company' has more than 4,000 levels and considering all was causing memory issues so we considered only the top 10 companies

Motrio	Naïve Bayes				
Metric	Standard Laplace = 2		Laplace =3		
Accuracy (Direct)	0.3814 0.4083		0.4085		
	Decision Trees				
Metric	Standard, size=324	Boosting Trials = 5	Boosting Trials = 10		
Accuracy (Training Set)	0.438	-	-		
Accuracy (Testing Set)	0.434	-	-		

## Logistic Regression:

```
multinom(formula = Region ~ Product + Company + Submitted, via +
    Timely.response. + Company.response.to.consumer + Consumer.disputed.,
    data - train_sample)
           (Intercept) ProductConsumer Loan ProductCredit card ProductCredit reporting ProductDebt collection ProductMoney transfers ProductMoney transfers
                                  -0.5915326
                                                      -0.2845894
                                                                             -0.12671547
                                                                                                       -0.3967867
                                                                                                                               -0.3752858
                                                                                                                                               -0.2874055
South
             0.6610853
                                  -0.1321854
                                                      -0.4454668
                                                                              0.03905251
                                                                                                       -0.1690341
                                                                                                                               -0.1124373
                                                                                                                                               -0.4572507
             1.1089548
                                  -0.5390341
                                                      -0.3615747
                                                                              -0.29545674
                                                                                                       -0.1671258
                                                                                                                               0.1480401
West
                                                                                                                                               +0.3932038
           ProductOther financial service ProductPayday loan ProductPrepaid card ProductStudent loan CompanyCapital One CompanyCitibank CompanyEquifax
                                                    0.5388157
North East
                                -0.8467621
                                                                       -0.86284885
                                                                                             -0.4189964
                                                                                                                -0.05685274
                                                                                                                                -0.07348518
South
                                -0.2912298
                                                    -0.9969629
                                                                       -0.66752470
                                                                                             -1.1472150
                                                                                                                -0.23922030
                                                                                                                                -0.38776663
                                                                                                                                                 -0.2926961
                                -0.7118204
                                                    0.1718518
                                                                        -0.03069452
                                                                                             -1.0696017
                                                                                                                -0.51204449
                                                                                                                                 -0.52018627
                                                                                                                                                 -0.4439083
           CompanyExperian CompanyNavient Solutions, LLC, CompanyOcwen CompanyTransUnion Intermediate Holdings, Inc. CompanyNells Fargo & Company
North East
                                                 -3.834122 -0.006661388
                 -0.3971552
South
                 -0.2628592
                                                 -3.634074 -0.008819389
                                                                                                                                            0.13141090
west
                 -0.2278824
                                                 -3,482837 -0.029829015
                                                                                                              -0.5805497
                                                                                                                                            0.15088121
           Submitted.viaFax Submitted.viaPhone Submitted.viaPostal mail Submitted.viakeferral Submitted.viaweb Timely.response.ves
North East
                                                                                      -0.8110315
                                                                                                       -1.1372750
                  -1.494916
                                     -0.9947111
                                                               -0.9960539
                                                                                                                             0.5554310
south
                                     -0.6609123
                                                               -0.1971840
                                                                                                                             0.4040763
                   -0.912436
                                                                                      -0.5241770
                                                                                                        -0.5701131
                  -1.013326
                                     -0.9494058
                                                               -0.6871076
                                                                                      -0.6214247
                                                                                                        -0.6778505
                                                                                                                             0.3073750
West
           Company, response, to, consumerClosed with explanation Company, response, to, consumerClosed with monetary relief
North East
                                                     +0.12690268
                                                                                                                -0.1715438
South
                                                     -0.17138298
           Company, response, to, consumerClosed with non-monetary relief Company, response, to, consumerClosed with relief
North East
South
                                                              -0.1290300
                                                              -0.1775768
                                                                                                              -0.02692852
West
           Company, response, to, consumerClosed without relief Consumer, disputed. Yes
North East
                                                    -0.2716412
South.
                                                    ·0.4103781
                                                                          0.00335657
West
                                                    0.1205451
                                                                          0.09178060
```

**Note:** For Boosting Trials = 5,10 the tree was truncated at the first trial itself and hence results of boosting are same as the standard decision tree.

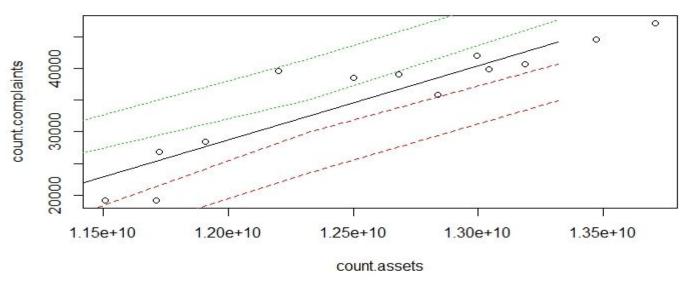
# Effect Of Total Assets on #Complaints

#### **DATA SET DETAILS:**

- Predictors: Assets of Financial Institutions
- Outcome: Number of Complaints
- Sampling Rate: 70 (Training): 30 (Testing)

#### **TECHNIQUES APPLIED:**

Linear Regression



Intercept: -1.08e+05

Coefficient: 1.163e-05

Correlation: 0.3428

➤ Mean Square Error: 12071733

```
"cor:
       0.342840901065586
                           MSE:
                                  6305398.16967114"
"cor:
       0.780532751861594
                           MSE:
                                  8106072.34906073"
"cor:
       0.932646883668888
                           MSE:
                                  5282470.45968027"
"cor:
       0.896841329846306
                                  5290419.98159268"
                           MSE:
"cor:
       0.778974106322222
                           MSE:
                                  5007942.84844342"
"cor:
       -0.783365868539241
                            MSE:
                                   5355001.16582222"
"cor:
       0.105880522202444
                                  13920646.2223951"
                           MSE:
"cor:
       0.473482025078609
                           MSE:
                                  15247655.3431237"
"cor:
       0.713338117593158
                                  15152449.4886729"
                           MSE:
"cor:
       0.78485209134768
                                 15734453.9531854"
                          MSE:
"cor:
       0.906877937166014
                           MSE:
                                  15157079.9942371"
"cor:
       0.904201443080769
                                  0"
                           MSE:
"cor:
       0.898248041638909
                                  0"
                           MSE:
"cor:
       -0.890404459009079
                                  0"
                            MSE:
```

## 6.

## Emotion Classification of Consumer Narratives

#### **DATA SET DETAILS:**

Predictors: Consumer Complaint Narrative

Outcome: Emotion of The Narrative

Classes: Anger, Sad, Dispute, Anticipation, Fear

#### **METHODOLOGY:**

- Used the Aylien API in Python to label 1,000 records in the data set.
- For the next milestone, we hope to use this labeling to predict the emotion of a narrative by building a classifier using supervised learning.

#### **API USED:**

> Aylien API

# Thanks! Any Questions?