Predicting Loan Fundraising Success

5/31/2016 Emelyn Chew





### Why?

- I. Kiva Zip is a P2P crowdfunding platform that makes 0% interest rate loans to small business owners
- 2. All-or-nothing model: loans expire if they do not reach fundraising target within 60 days
  - Borrowers do not get any \$ 🕾
  - Lenders get money back in their Kiva account 🕾
  - Operational cost ☺

Question: Can we predict if a borrower will successfully fundraise on the Kiva Zip platform given certain loan characteristics?

### Getting the data

 Scraped Kiva Zip Website using Beautiful Soup

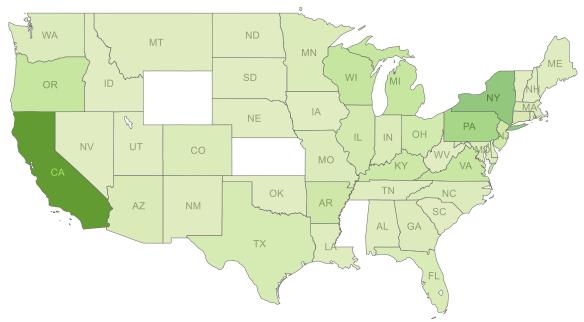
#### 2. Predictors:

- Business name
- City
- State
- Conversations with lenders
- Lenders (new and invited)
- Text: Personal Story, Business Description, Loan Purpose
- Status
- Loan Size
- Industry
- Years of experience
- Social Media + Website
- Trustee

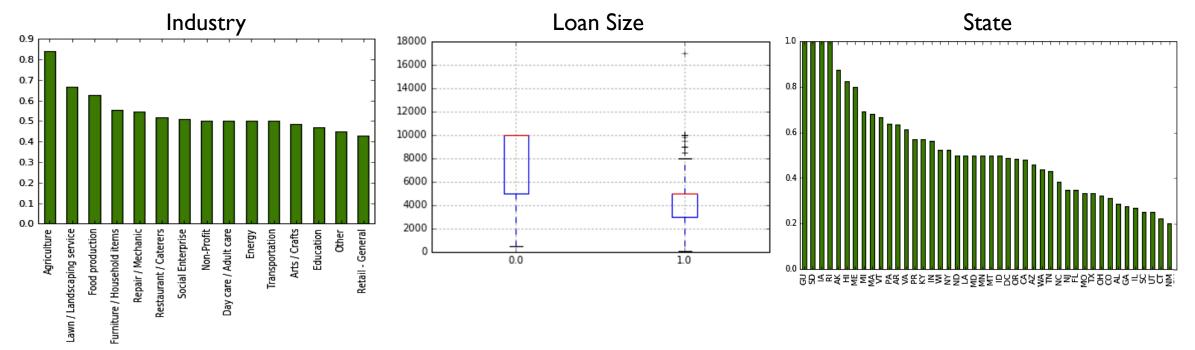


### What does the data look like?

- +10,000 small business loans from 2011 to present
- 2. Filtered for U.S. based loans when scraping (~2500 loans)
- 3. 18 predictors
- 4. Response variable:
  - Days Fundraising: Difference between loan listed date and loan disbursed date
  - Eventually bucketed into 0 (failure), I (success) to become a binary classification problem



#### Initial useful features...



- Industry: Agriculture/Food production loans outperform
- Loan Size: Large loans have higher chance of failing
- States: Certain states have an advantage

# Machine Learning Model #1: Logistic Regression

- Trouble with linear regression: converted to binary variables to make the problem a classification problem
- 2. Null Accuracy Rate = 0.50
- 3. Logistic regression using features:
  - Loan size
  - Trustee (created dummy variables)
  - # Lenders invited
  - # Conversations with borrowers
  - Industry (created dummy variables)
  - Loan Term

**Accuracy Rate = 0.62** 

NLP increased accuracy rate...

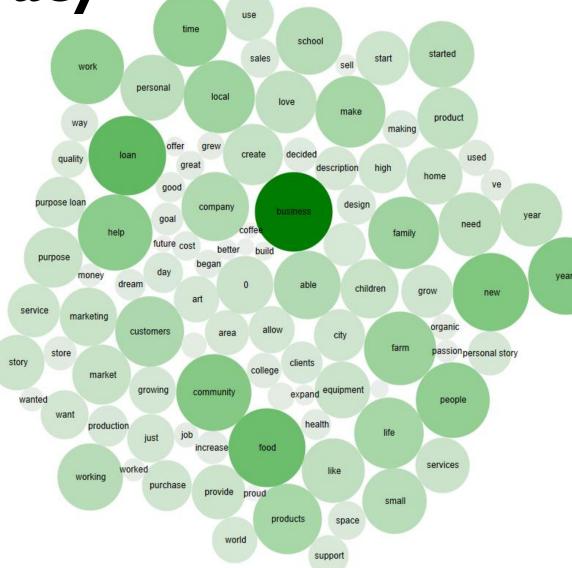
- I. Added Natural Language Processing:
  - TFIDF Vectorizer
  - Ngram range (1,2)
  - Max features = 1000
  - Min df =2
  - Stop words

Accuracy Rate increased to <u>0.68</u>

Sensitivity = 0.60

**Specificity = 0.68** 

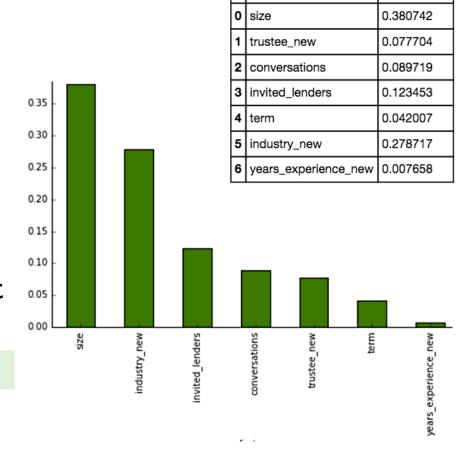
Model better at identifying true negatives!



## Machine Learning Model #2: Decision Tree

- I. Tuning the regression tree:
  - Max Depth = 5
  - Cross validation
- 2. Gini index before split = 0.49
- 3. Size and industry most important features
  - Highest feature importance rates (reduce classification error rate by the most)
- 4. Trustee, term and years experience not that important

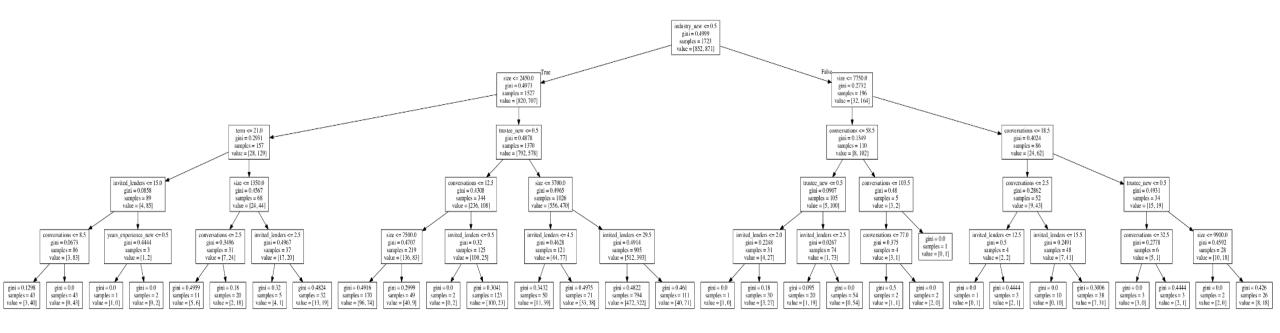
Accuracy = 0.74



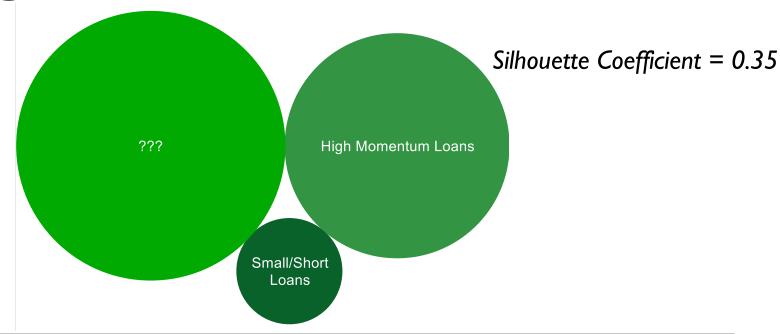
feature

importance

## Machine Learning Model #2: Decision Tree



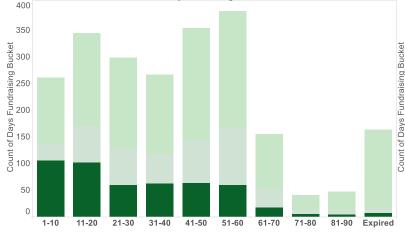
# Machine Learning Model #3: K-Means Clustering



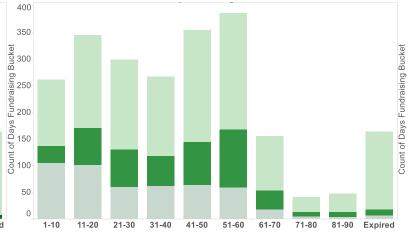
	size	state_new	trustee_new	conversations	invited_lenders	term	days fundraising	industry_new
cluster								
0	4986.653166	0.476436	0.770987	25.174521	14.968336	23.657585	45.128130	0.080265
1	9854.697286	0.638831	0.789144	19.956159	22.177453	34.739040	44.162839	0.273486
2	2148.159509	0.451943	0.660532	16.102249	9.259714	18.873211	32.057260	0.069530

#### 3 Distinct Clusters

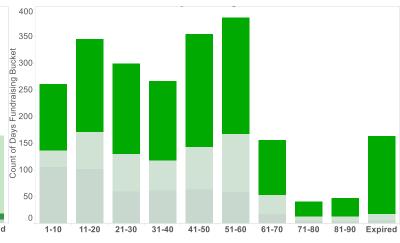
Cluster 1: Days Fundraising



Cluster 2: Days Fundraising



Cluster 3: Days Fundraising



- Low Risk
  - Small loan size (~2k)
  - Short loan term (<20 months)
- Fundraise quickly, low effort

- High risk
  - Large in size (~10k)
  - Long loan term (~36 months)
- High Momentum
  - CA/NY (Kiva HQ)
  - Trustee support
  - Many conversations & invited lenders
- End up succeeding at last minute

- Average everything
  - Average size
  - Average term
  - Average borrower engagement
- Difficult to predict outcome

#### **Conclusions**

- I. Fundraising success is hard to predict with high accuracy
- Text matters!
- 3. Kiva staff:
  - Focus on loans in "Average" category most likely to expire
- 4. Borrowers:
  - Choose realistic fundraising goals
  - Write good "stories"
  - Engage lenders in network
- 5. Given more time... more feature extraction:
  - Social media engagement measure conversations about the loan on twitter/Facebook etc.
  - Time series data measure if velocity of fundraising increases likelihood of success
  - Image processing quality of photos, RGB, facial recognition