Hootsuite

(Random Acts of Pizza: My first Text Mining project)

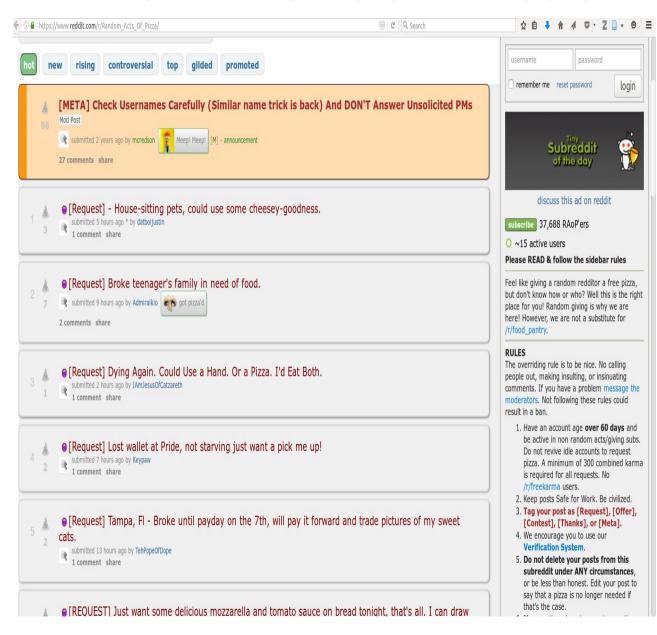


Overview:

My Goals

- Give you a data-driven advice in how to write your post and increase the chance to receive a free pizza (How can I do it?)
- Learn/Get more experience in text mining (learn on demand). Explore NLP and text mining packages (try a new thing) in R
- Test my EDA Start Kit project (GTD)
- Overview
 - What is RaoP?
 - Data:
 - Data introduction;
 - Features and text engineering
 - Data prep and descriptive
 - First model development
 - Features selection
 - Model assessment
 - Recommendation and next steps

RAoP: Random Action of Pizza



If I want a free pizza. I write a compelling post on the on r/Random_Acts_Of_Pizza/ asking for a pizza with a tag: [REQUEST]

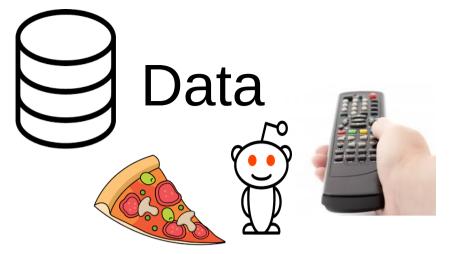
If I want to offer a pizza. I write a post with a tag [Offer] or select a [REQUEST] post

So you can help people and restore Restoring Faith in Humanity, One Slice at a Time - **Reddit**



https://www.reddit.com/r/Random_Acts_Of_Pizza/

Data



- Data size: 5671 (rows) x 33 (vars)
- #users: 5671 (no user with 2+ post !?)
- Received pizza: 1397 (25%)
- Investigated actionable predictors (features)
 - Available (in the moment I write the post)
 - Control our partial control (I can change it)

N Field	Description	Comm
1 post_was_edited	Boolean indicating whether this post was edited (from Reddit).	
2 request_id	Identifier of the post on Reddit, e.g. "t3_w5491".	
3 request_text	Full text of the request.	Info
4 request_title	Title of the request.	
5 requester_account_age_in_days_at_request	Account age of requester in days at time of request.	
6 requester_days_since_first_post_on_raop_at_request	Number of days between requesters first post on RAOP and this request (zero if requester has never posted before on RAOP).	
7 requester_number_of_comments_at_request	Total number of comments on Reddit by requester at time of request.	
8 requester_number_of_comments_in_raop_at_request	Total number of comments in RAOP by requester at time of request.	
9 requester_number_of_posts_at_request	Total number of posts on Reddit by requester at time of request.	
10 requester_number_of_posts_on_raop_at_request	Total number of posts in RAOP by requester at time of request.	
11 requester_number_of_posts_on_raop_at_retrieval	Total number of posts in RAOP by requester at time of retrieval.	
12 requester_received_pizza	Boolean indicating the success of the request, i.e., whether the requester received pizza.	$oldsymbol{Y}$
13 requester_upvotes_minus_downvotes_at_request	Difference of total upvotes and total downvotes of requester at time of request.	karma
14 requester_username	Reddit username of requester.	
15 unix timestamp of request	since most RAOP users are from the USA).	Date

Kept my focus on:

- His/Her history was compelling?
- He/She was polite?
- He/She was able to prove his history?
- He/She provided what was required/asked to him?
- Is He/She a good/nice reddit user?
- How long is he/she a reddit user?

Features Engineering

- Post related features (total control)
 - #words in the post (compelling)
 - post sentiment score: (polite and positive)
 - post received 5 scores: money, job, student, family and desire narratives (kind of history) (Some narratives strategy might be more effective than others)
 - Has.link (reciprocity: you gave something back)
- User/requester related features (partial control)
 - status or karma: requester upvote minus downvote
 - Age account in days
- Community related features (No control)
 - Community age (people are more excited in the begin) (WIP or TODO)
- Temporal (moment) related features (control)
 - First half of the month (When was it post?)(I do not have money in the end of the month)

Text Engineering (80% of the time)

(Simple approach) - weekend:) --

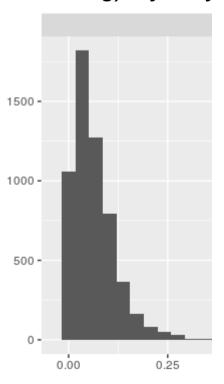
- Sentiment Score (count shared words)
 - 2 Dictionary: positive (+1) and negative (-1) words
 - {pos words: happy, friendly,...} => high positive score
 - {neg words: bad, broke,*#\$@#,...} => high negative score
 - Score: #pos shared words #neg shared words
- Narrative Scores (count shared words and divide by the number of word in the dictionary)
 - 3 dictionaries: money, job, family, desire, student
 - Scores: count shared words/(#words in the dictionary)
 - Shameless stolen (*Inspired*) in the article: How to Ask for a Favor: A Case Study on the Success of Altruistic Requests (Stanford, Max Planck Institute)

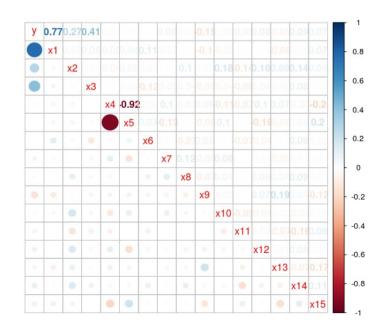
Data Preparation and issues

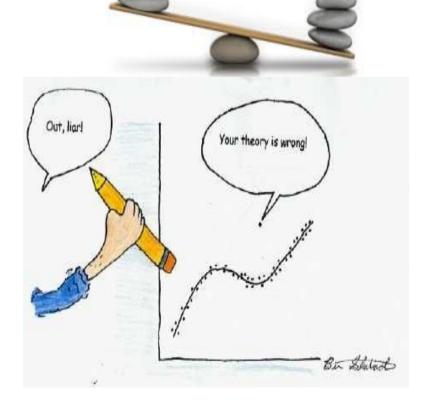
Issues:

- Skewness and long tails (use median values) :: Ignored (work with median as possible)
- Concentration (You need variation to discover relation) :: removed var
- Outliers :: removed points
- Discrepant value ranges (X1: [0 10], X2: [0 10k]) :: scale [0 1]
- Multi-colinearity (karma and age account):: ignored (predictive model)
- Classes were Unbalanced (pizza: 1397; no pizza: 4274) :: re-balance

• Text: stop words (a, by, the, in); punctuations, derived words working); synonyms :: **removed**

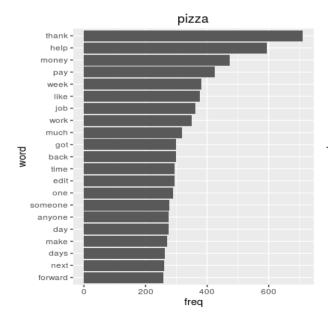


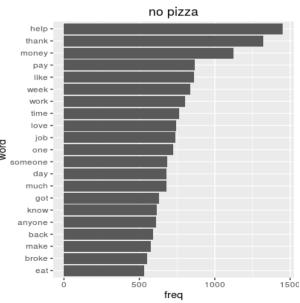




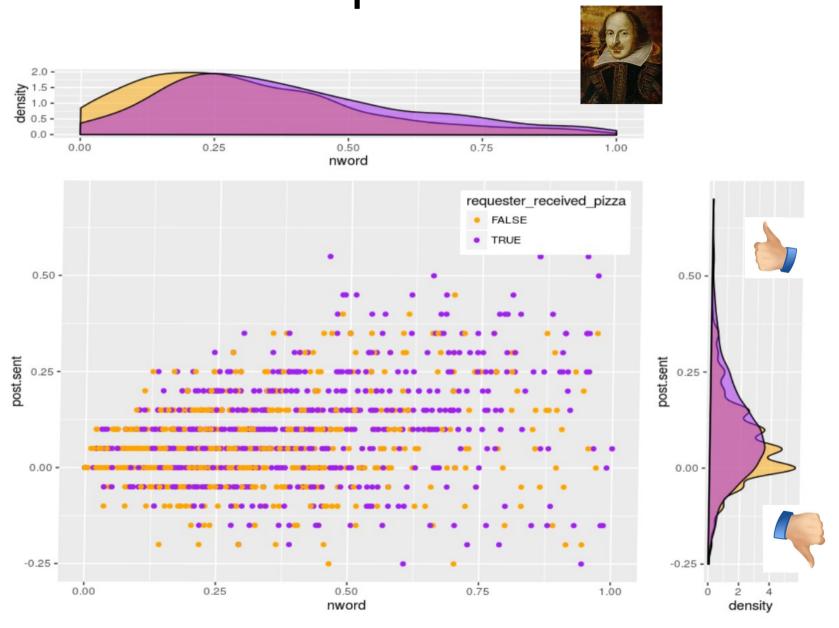
What are the most frequent words in each group?







Get Pizza; #words and being positive ____



Model

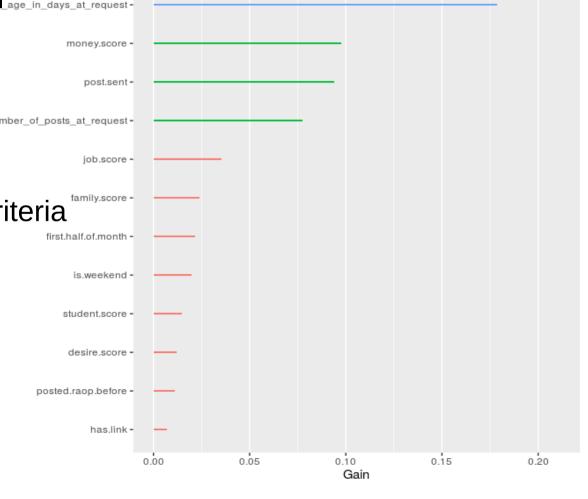
Features Selection: Xgboost relative importance



 What happened in Y when Lage change one X variable and keep the rest unchanged?
 What is the impact in Y?

Features selected base this criteria

- nword (compelling)
- karma (nice user; status)
- requester account age
- money.score
- post sentiment score
- #of post at request

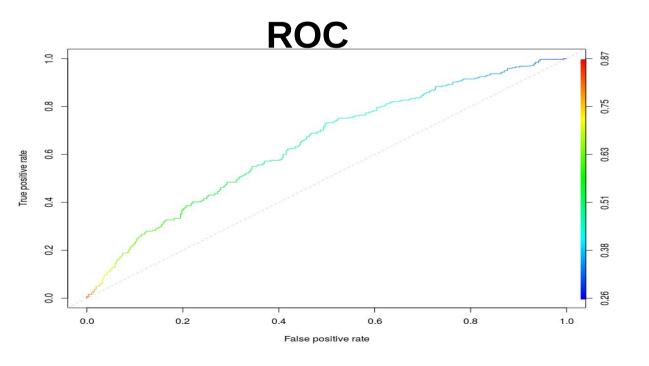


Feature importance

RAoP Model: Logistic Regression

```
Call:
glm(formula = requester_received_pizza ~ requester_upvotes_minus_downvotes_at_request +
   nword + requester_account_age_in_days_at_request + money.score +
   post.sent + has.link + first.half.of.month + posted.raop.before,
   family = binomial(link = "logit"), data = train.data)
Deviance Residuals:
   Min
              10
                  Median
                                        Max
        -1.0878 -0.8319
                            1.1491
                                     1.5680
Coefficients:
                                              Estimate Std. Error z value Pr(>|z|)
(Intercept)
                                                           0.1298
                                                                  -6.804 1.01e-11 ***
                                               0.2255
equester upvotes minus downvotes at request
                                                           0.2779
                                                                    0.811 0.417089
                                                0.5858
                                                           0.3337
                                                                    1.756 0.079156 .
equester_account_age_in_days_at_request
                                                0.2270
                                                           0.2721
                                                                    0.834 0.404005
                                                5.5512
                                                           1.5327
                                                                    3.622 0.000293 ***
noney.score
                                                1.7254
ost.sent
                                                           0.4884
                                                                    3.532 0.000412 ***
nas.linkTRUE
                                                0.5138
                                                           0.2011
                                                                    2.554 0.010635 *
first.half.of.monthTRUE
                                                0.1525
                                                           0.1084
                                                                    1.407 0.159497
posted.raop.beforeTRUE
                                                0.9909
                                                           0.2807
                                                                    3.530 0.000415 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

- AUC: 0.64 (TEST)
- Accuracy: 0.59 (be right in both direction)
- Sensitivity or Recall: 0.65
 (hability to detect)



Recommendation based what I learned

- Try to be compelling (*my opinion and intuition, not analysis*). But write more than 100 words (pizza: median 74)
- Add a link in your post (image of a dog)
- If you are facing temporary problem with money, write about that but try to be polite and positive
- Be a nice user helps (Improve your karma)

Next Steps

- Improve variables transformations
 - segment continuous variables (quantiles,deciles)
 - create narrative categorical var: is.money; is.desire; ...
- Explore more variables
 - Add time (trend) in the model (I forgot to include it)
- Make the model "interpretability" (make easy to understand the results)



Post Examples: High pos scores, nwords < 100

"My wife and I are avid pizza fans, but this month has been exceptionally tight fiscally. Having recently put together a new monthly budget (with more reasonable expenses) we've acted more responsibly, but had considerably less fun. If anyone would be so kind as to reward our new financial responsibility with dietary irresponsibility, that would be grand. I'm sure the kids would love something more than just bland-yetwholesome Spaghetti for the third night in a row. Thanks for taking the time to read, and have a great day no matter what."



Like the title says I'm in a tight spot and could really use a warm pizza to lift my spirits and give me a break from KD. As it stands I won't be able to both pay bills and afford groceries when I get paid so this would be a huge boost. I'm a long time redditor and am happy to PM you from my main account to verify this as well as provide any other info you would like. EDIT: Thanks to the wonderful DEStudent I will be having pizza for supper! Thank you so much!