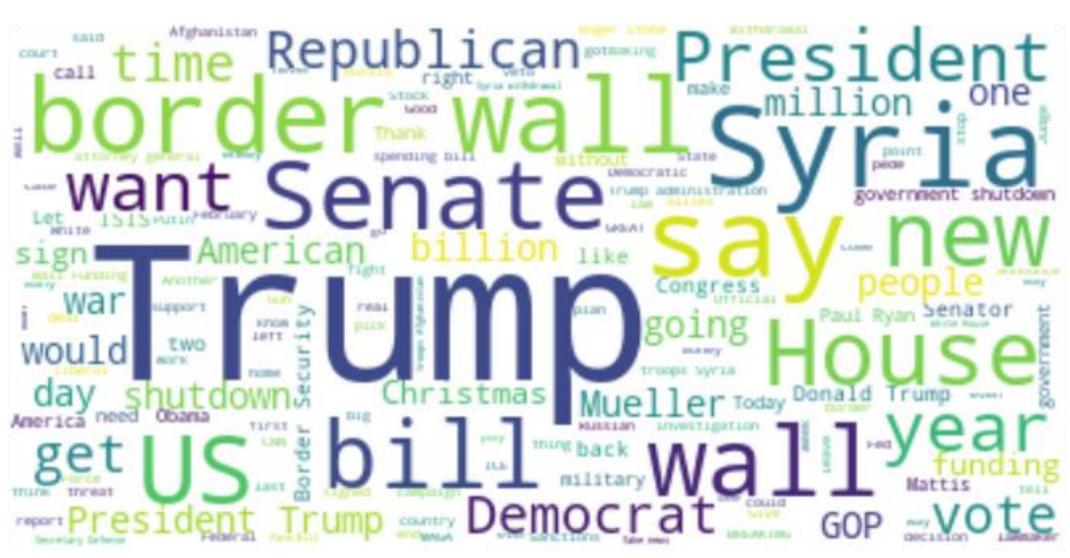
Sub-Reddit Classification using NLP





Icebreaker

Match Them Up: r/Politics or r/The_Donald?



VS.

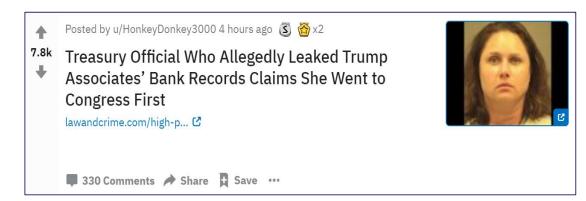




Icebreaker

What about this?





The Problem

How accurately can we classify a given post into the right sub-Reddit?

The Issue in Context

- Moral of Exercise: manual classification is not always quick and easy!
- Applications: managing the aftermath of human error / bugs that misclassify posts

The Approach

• Collection: Webscrape sub-Reddits with Reddit API (500 posts each)

Cleaning: Filter out irrelevant posts; deal with strange characters.

• Word Vectorizers: Count Vectorizer (EDA), TF-IDF Vectorizer (Preprocessing)

• Classification Models: (1) Random Forest

(2) Logistic Regression

(3) Support Vector Machine

The Evaluation

Train/Test Split: Train model on 70% of all posts; testing on remaining 30%

Key Metric: Accuracy to determine the best of the three

...vs. Baseline Accuracy (~50%)



Methodology: Exploratory Data Analysis

High-Level Statistics on Subreddits

	/r/The_Donald	/r/Politics
No. Subscribers	690,286	4,380,649
Mean Comments/Post (% of Subscribers)	38 (0.0055%)	187 (0.0042%)
Mean Score/Post (% of Subscribers)	801 (0.12%)	2,284 (0.05%)

Metrics

• No. of Comments and Scores (= upvotes - downvotes) are the few relevant numeric features from webscraped data ⇒ indicative of subreddit activity

Key Insights

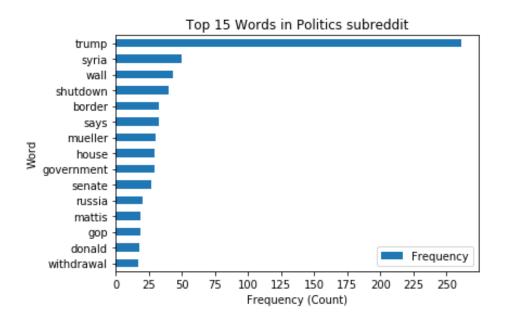
- Politics far more popular than The_Donald in <u>absolute</u> terms
- However, once <u>adjusted for size</u>, activity is similar...The_Donald slightly more active!
 ∴ comparable candidates to test our model on



Methodology: Exploratory Data Analysis

Examining the Distribution of Top 15 Words in /r/Politics

Word	Frequency in Sample
trump	261
syria	50
wall	43
shutdown	40
says	33
border	33
mueller	30
government	29
house	29
senate	27
russia	20
gop	19
mattis	19
donald	18
withdrawal	17

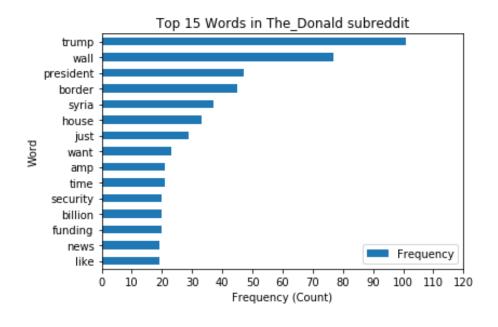




Methodology: Exploratory Data Analysis

Examining the Distribution of Top 15 Words in /r/The_Donald

Word	Frequency in Sample
trump	101
wall	77
president	47
border	45
syria	37
house	33
just	29
want	23
time	21
amp	21
funding	20
billion	20
security	20
like	19
news	19





Methodology: Data Cleaning

Troll Filter: Using "Spam Signals" to filter out irrelevant posts

Troll Filter

Based on numeric features:

num_reports > 0:

• score < 0:

remove any posts flagged/reported at least once

remove any posts with a net negative score

keep 0 scorers - the post may be new!

Eliminating Residual Unicode Text from Data

- Noticed "&" from The_Donald EDA
- Vectorizers get rid of "&" and ";" but "amp remains!
- Replace all instances of "&" with "&" before vectorizing!

Concatenating Title and Sub-Reddit Data into Data Frame

- NLP occurs on *combined body of text* from all subreddits so merge
- Binarize Target Variable (Sub-Reddit) s.t. The_Donald = 1, Politics = 0



Methodology: Modeling

Model 1: Random Forest

TF-IDF Vectorizer Optimized Parameters:

- max_features = 1,500
- ngram_range = (1,1)
- norm $= \ell_1$

Random Forest Classifier Optimized Parameters:

- n_estimators = 20
- max_depth = 20

Methodology: Modeling

Model 2: Logistic Regression

TF-IDF Vectorizer Optimized Parameters:

- max_features = 2,500
- ngram_range = (1,1)
- norm $= \ell_1$

Logistic Regression Optimized Parameters:

- penalty $= \ell_2$
- C = 10

Methodology: Modeling

Model 3: Support Vector Machine (SVM)

TF-IDF Vectorizer Optimized Parameters:

- max_features = 1,500
- ngram_range = (1,1)
- norm $= \ell_1$

SVM Optimized Parameters:

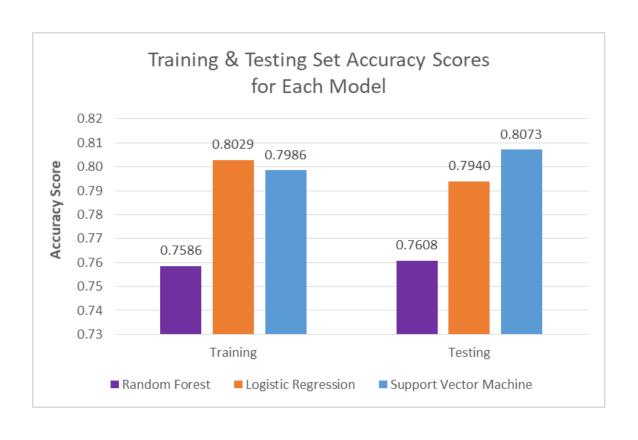
- C = 10
- kernel = linear
- gamma = auto



Modeling: Results & Evaluation

Train Accuracy vs. Test Accuracy

	<u>Model 1</u> : Random Forest	Model 2: Logistic Regression	Model 3: SVM
Training Accuracy	0.7586	0.8029	0.7986
Testing Accuracy	0.7608	0.7940	0.8073
Overfit?	no	slightly	no



Key Insights

- SVM does best in Testing set
- Logistic Regression does best in Training set, though results in very slight overfit
- · Otherwise, virtually no overfit in models



Modeling: Evaluation (Supplement)

Confusion Matrix

Random Forest	Predict Politics (y = 0)	Predict The_Donald (y = 1)
Actual Politics (y = 0)	100	51
Actual The_Donald (y = 1)	21	129

Logistic Regression	Predict Politics (y = 0)	Predict The_Donald (y = 1)
Actual Politics (y = 0)	122	29
Actual The_Donald (y = 1)	33	117

Support Vector Machine	Predict Politics (y = 0)	Predict The_Donald (y = 1)
Actual Politics (y = 0)	125	26
Actual The_Donald (y = 1)	34	116



Modeling: Evaluation (Supplement)

Classification Evaluation Metrics

	<u>Model 1</u> : Random Forest	<u>Model 2</u> : Logistic Regression	<u>Model 3</u> : Support Vector Machine
Accuracy	0.7608	0.7940	0.8007
Sensitivity	0.8600	0.7800	0.7733
Specificity	0.6623	0.8079	0.8278
Precision	0.7167	0.8014	0.8169

Key Insights

- Overall: all three models performed well (Accuracy > 0.75)
- Support Vector Machine (SVM) outperformed all other models in terms of <u>Accuracy</u>, <u>Specificity</u> (True Negative Rate), and <u>Precision</u>
- Random Forest was the worst performer in all areas, except Sensitivity.



Wrapping-Up

Key Takeaways

- All beat baseline accuracy (50%)
- All accuracy $> 75\% \Rightarrow$ good ability to accurately classify posts
- Clear winner: <u>Support Vector Machine</u> model (4 in 5 correct)
 - o Implication: Reddit administrators need to review 1 in 5 misclassified posts
 - o Tuning improvement possible, but SVM very computationally intensive

Further Enquiry

- Group synonyms together as features instead of individual words
- Take into consideration words that are all caps (indicate shouting/anger), but still apply .lower to words where first letter is capitalized

