# Fake News Project

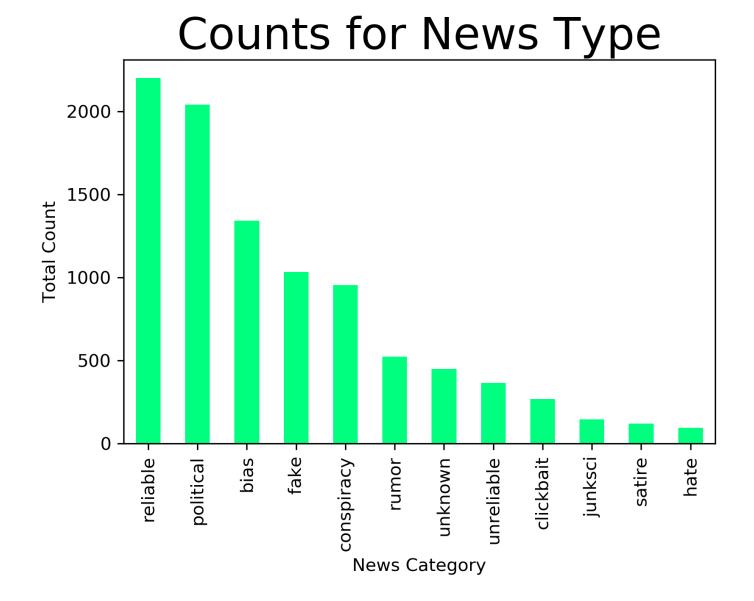
## Motivation For Project

- Fake news detection is important for several industries.
- Notably, regulatory agencies have begun to target websites with user-generated content (such as social media sites) for allowing the circulation of fake news on their platforms.
- Automated fake news detection can be useful for major websites to identify potentially problematic content in advance.

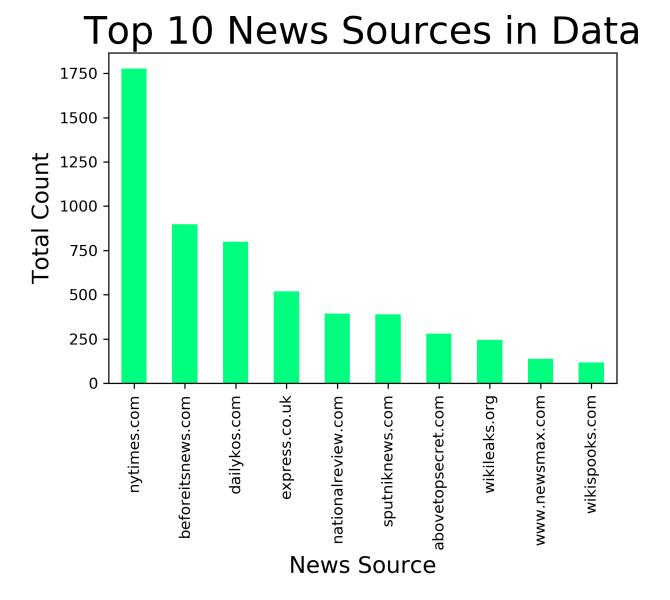
## Questions for Analysis

 What can machine learning techniques tell us about fake news detection?

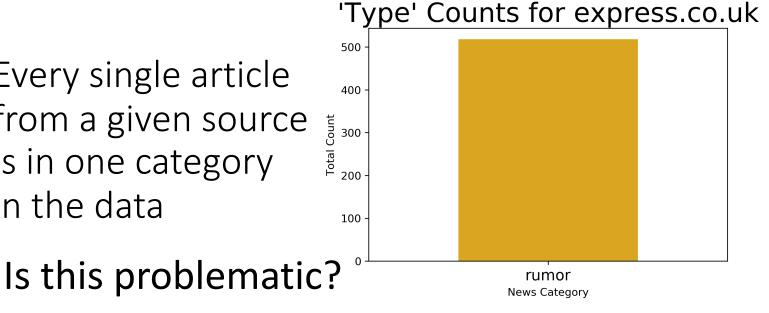
 How do different styles of sampling data influence results? Category Counts

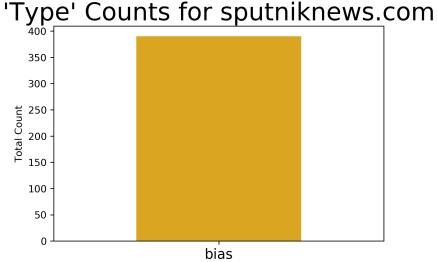


## News Source Counts

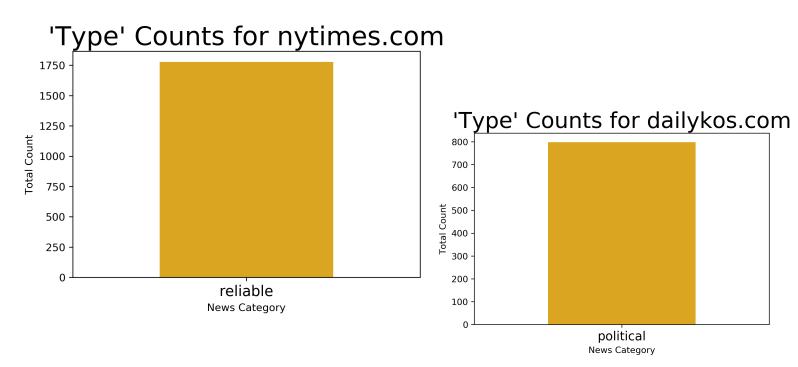


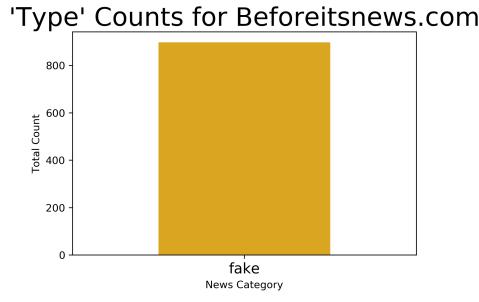
Every single article from a given source is in one category in the data





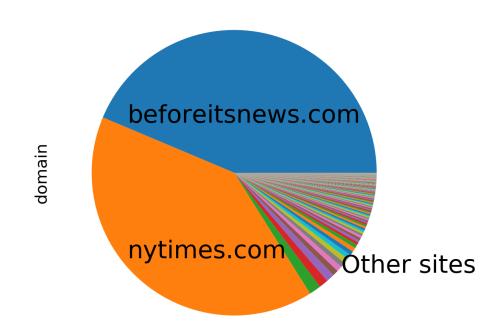
**News Category** 

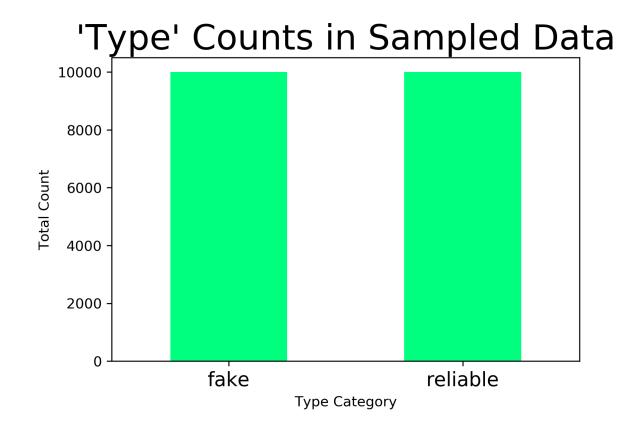




# Initial Analysis – Reliable vs. Fake

## News Source (domain)





# Sentiment Analysis

	Label	SIA Polarity	TextBlob	TextBlob
		Score	Polarity Score	Subjectivity
Fake News	1	5468	983	1128
	-1	3013	67	5784
	0	5013	7944	3088
Reliable News	1	6721	1175	762
	-1	2581	73	6800
	0	698	8752	2438

### **Vectorization Technique**

		Bag of Words	Tf-idf	Tf-idf with two bigrams
Classifier	MultinomialNB()	86.7%	87.2%	90.4%
	LinearSVC()	87.0%	90.8%	91.7%
	XGB Classifier()	87.5%	89.0%	83.3%

Predictive Modeling – Reliable vs. Fake

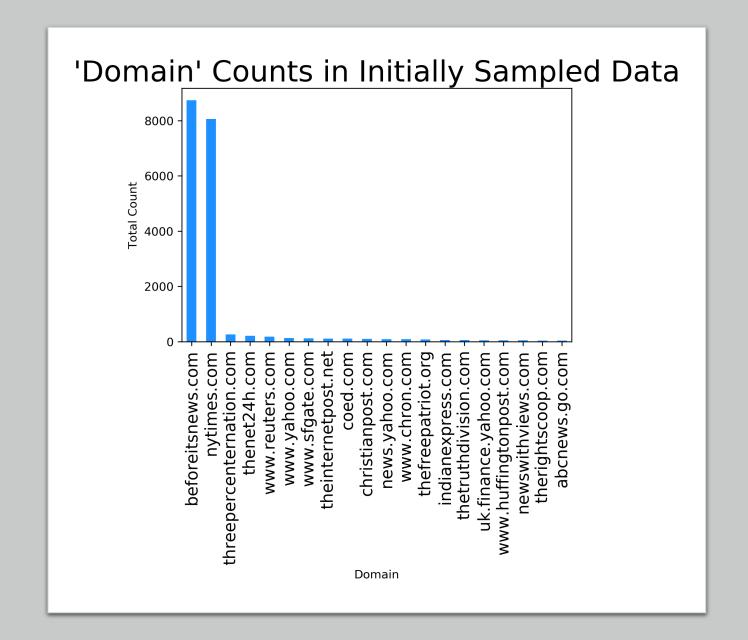
## Most Predictive features for Initial Reliable vs. Fake Analysis

```
main stori
6.2862
6.2685
        read main
        advertis continu
5.9580
5.6588
        continu read
4.7699
        new york
4.2447
        to re
        2.3737
                 an articl
        1.9920
                 next in
        1.7252
                 said would
```

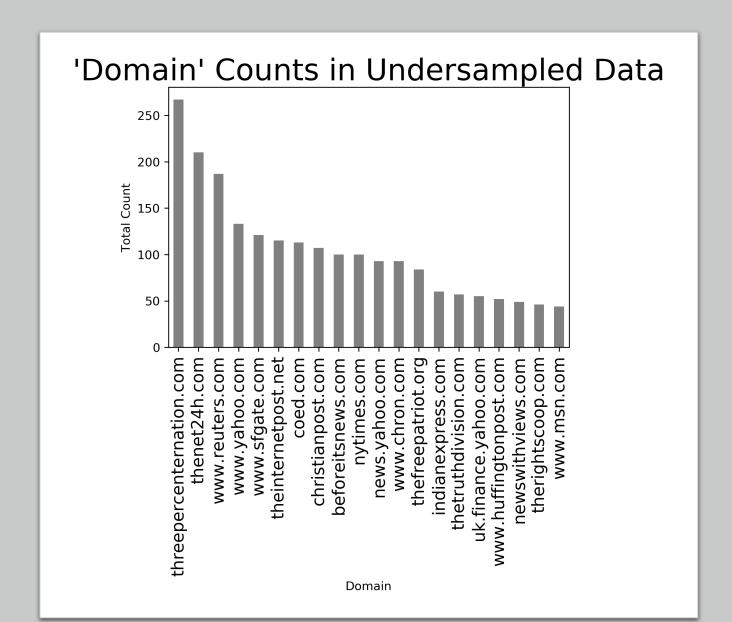
• These results indicate that the fact that a given article is from the New York Times is more predictive than anything else in the data. Because of this, the data was resampled (see following slide)

## Resampling

The New York Times and beforeitsnews.com were overrepresented in the initial sample



Data was resampled for better balance across domain



		Vectorization Technique		
		Bag of Words	Tf-idf	Tf-idf with two bigrams
Classifier	MultinomialNB()	78.4%	63.9%	66.0%
	LinearSVC()	80.7%	85.6%	82.1%
	XGB Classifier()	79.8%	84.6%	78.7%

Predictive Accuracy – Resampled Data

### **Predictive Features**

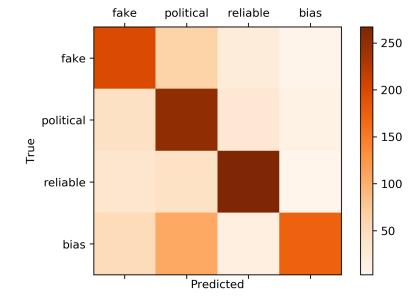
### **Unigrams** 2.8374 2016 2.3298 ap 2.2535 nov 2.1699 november 1.9041 said 1.5876 reuters 1.4786 photo 1.4026 film 1.3945 also 1.2465 percent 1,2073 savs

#### Bigrams

- -4.2017 budget rep
- -4.5269 aliens tend
- -4.6422 aiding abetting
- -4.6732 asking doctor
- -4.7910 becoming nurse
- -5.0097 books hillbilly
- -5.2696 500 name
- -5.3058 black sea
- -5.3252 bar great
- -5.3520 apartment metrocare
- -5.3822 cabinet bloomberg
- -5.3951 babies kinkade

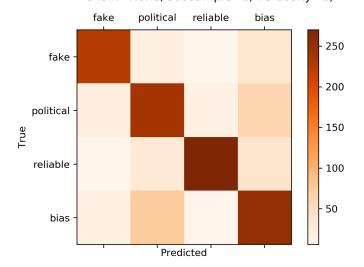
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#### $Confusion\ matrix\ of\ Multinomial NB (alpha=1.0,\ class\_prior=None,\ fit\_prior=True)$

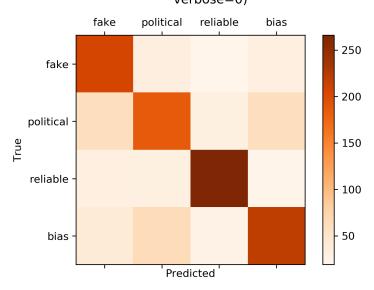


# Multiclass Classification with count vectorization

Confusion matrix of XGBClassifier(base\_score=0.5, booster='gbtree', colsample\_bylevel=1, colsample\_bynode=1, colsample\_bytree=1, gamma=0, learning\_rate=0.1, max\_delta\_step=0, max\_depth=3, min\_child\_weight=1, missing=None, n\_estimators=100, n\_jobs=1, nthread=None, objective='multi:softprob', random\_state=0, reg\_alpha=0, reg\_lambda=1, scale\_pos\_weight=1, seed=None, silent=None, subsample=1, verbosity=1)

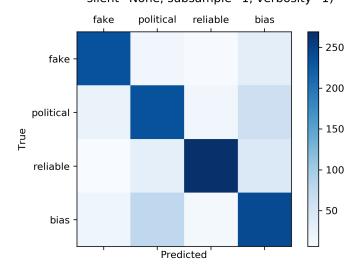


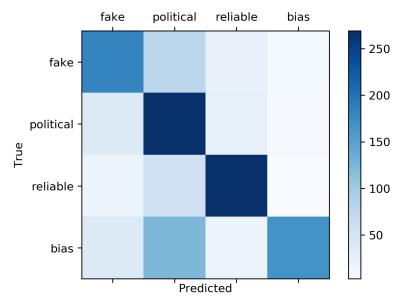
Confusion matrix of LinearSVC(C=1.0, class\_weight=None, dual=True, fit\_intercept=True, intercept\_scaling=1, loss='squared\_hinge', max\_iter=1000, multi\_class='ovr', penalty='l2', random\_state=None, tol=0.0001, verbose=0)



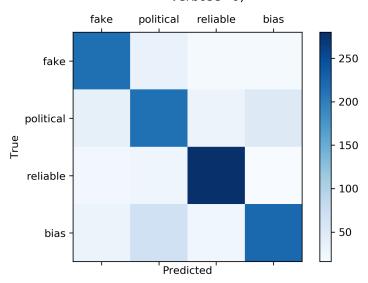
# Multiclass Classification with tf-idf vectorization

Confusion matrix of XGBClassifier(base\_score=0.5, booster='gbtree', colsample\_bylevel=1, colsample\_bynode=1, colsample\_bytree=1, gamma=0, learning\_rate=0.1, max\_delta\_step=0, max\_depth=3, min\_child\_weight=1, missing=None, n\_estimators=100, n\_jobs=1, nthread=None, objective='multi:softprob', random\_state=0, reg\_alpha=0, reg\_lambda=1, scale\_pos\_weight=1, seed=None, silent=None, subsample=1, verbosity=1)



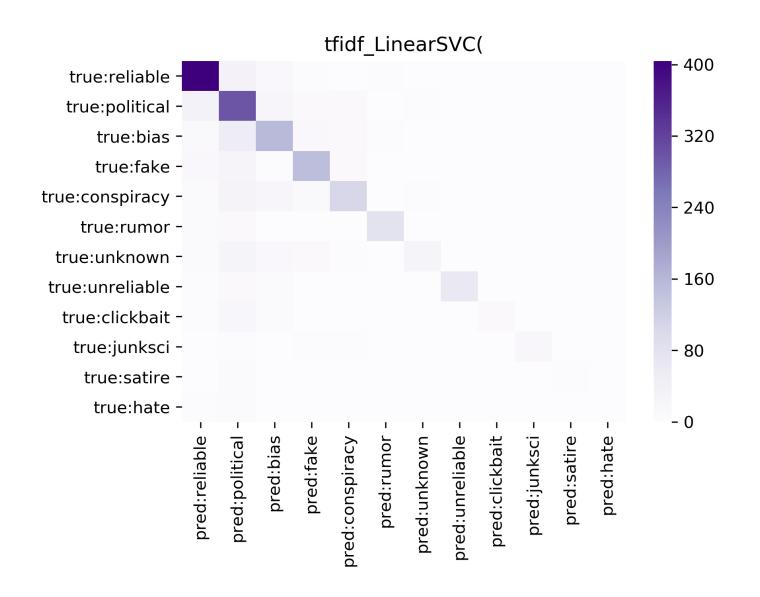


Confusion matrix of LinearSVC(C=1.0, class\_weight=None, dual=True, fit\_intercept=True, intercept\_scaling=1, loss='squared\_hinge', max\_iter=1000, multi\_class='ovr', penalty='l2', random\_state=None, tol=0.0001, verbose=0)



#### More Classes

• The same six pairs of vectorization and classifiers were applied to the data, but for all of the classes (instead of just the four largest ones).



## **Concluding Points**

- Prediction accuracy seems to be inherently tied to source.
- Some of the most informative features were direct references to the site or online newspaper that articles came from.
- This analysis was conducted with a smaller subset of data from the very large dataset.
- Conducting the analysis with a larger amount of data may yield more significant results.