

Fake News Classification Using NLP

Predicting Real vs. Fake News Headlines

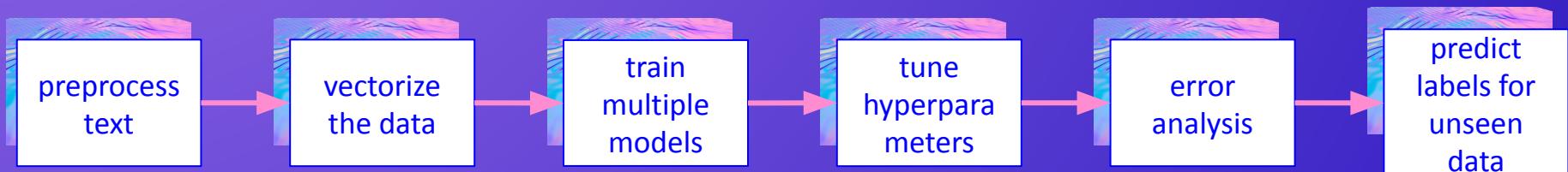
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Overview

Goal:

Build a classifier that distinguish **fake news (0)** from **real news (1)** using headliners.

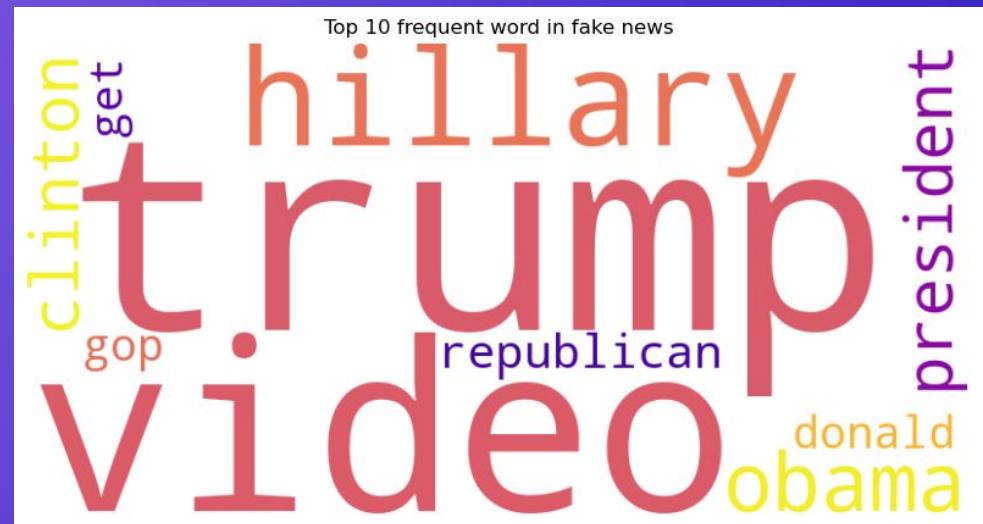
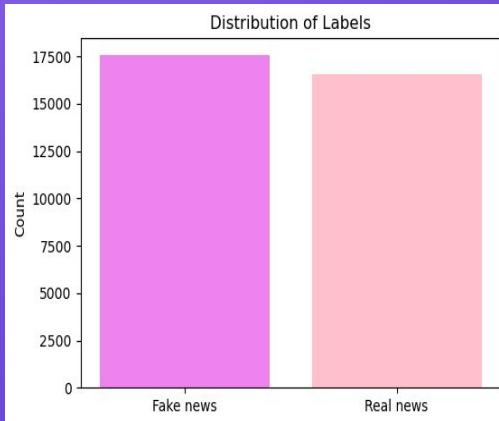
Workflow:



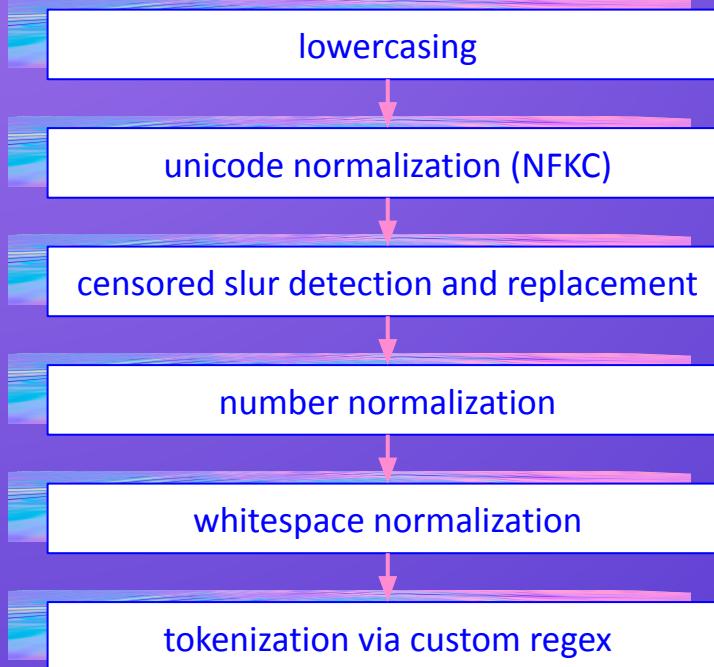
Dataset

Training Data:

- 34152 headlines
- 80/20 split into train & testing
- 14043 fake and 13278 real news (almost balanced)



Preprocessing



Legends

Colors	Links
Added	(f)irst change
Changed	(n)ext change
Deleted	(t)op

censored_slur

Original	data_clean
f 1brand-new	1brand-new
2pro-trump	2pro-trump
3ad	3ad
4features	4features
5so	5so
6much	6much
7a**	7censored_slur
8kissing	8kissing
9it	9it
10will	10will
11make	11make
12you	12you
13sick	13sick

additional processing

rem punctuation

Original	data_no_punc
n 1brand-new	1brandnew
2pro-trump	2protrump
3ad	3ad
4features	4features
5so	5so
6much	6much
t 7a**	7censored_slur
8kissing	8kissing
9it	9it
10will	10will
11make	11make
12you	12you
13sick	13sick

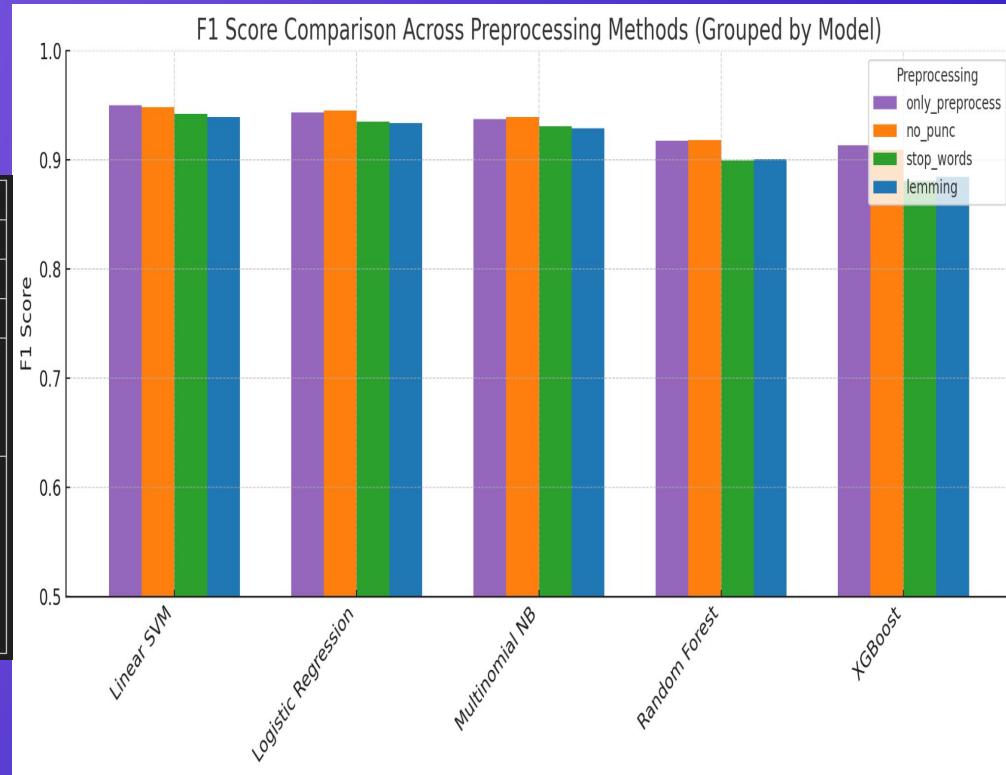
lemmatizing

Original	data_lemm
f 1brand-new	1brand-new
2pro-trump	2pro-trump
3ad	3ad
n 4features	4feature
5so	5much
6much	6much
t 7a**	7censored_slur
8kissing	7kissing
n 9it	n
10will	10will
11make	8make
t 12you	12you
13sick	9sick

Baseline Models and Data Selection

Model	Parameters
Linear SVC	<code>max_iter=5000, class_weight='balanced', random_state=42</code>
Logistic Regression	<code>max_iter=2000, lbfqgs, class_weight='balanced', random_state=42</code>
Multinomial NB	<code>alpha=1.0</code>
Random Forest	<code>n_estimators=200 max_depth=50 max_features='sqrt' n_jobs=-1</code>
XGBoost	<code>n_estimators=200 max_depth=6 eta=0.1 subsample=0.8 colsample_bytree=0.8 eval_metric='logloss' n_jobs=-1</code>

Vectorizer: TF-IDF(ngrams_range(1,1)

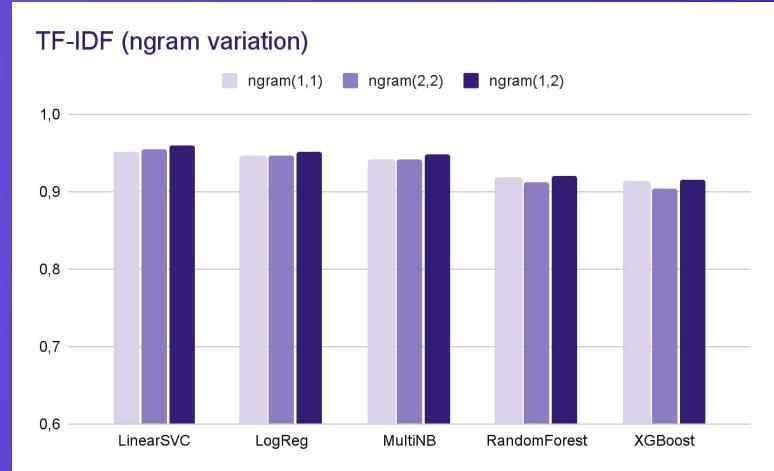


Vectorization Approach

1. **Word-Level:**
 - a. CountVectorizer (Bag of Words)
 - b. TF-IDF
2. **Character-Level:**
 - a. TF-IDF

Custom Tokenizer:

- handle censored words
- robust token regex

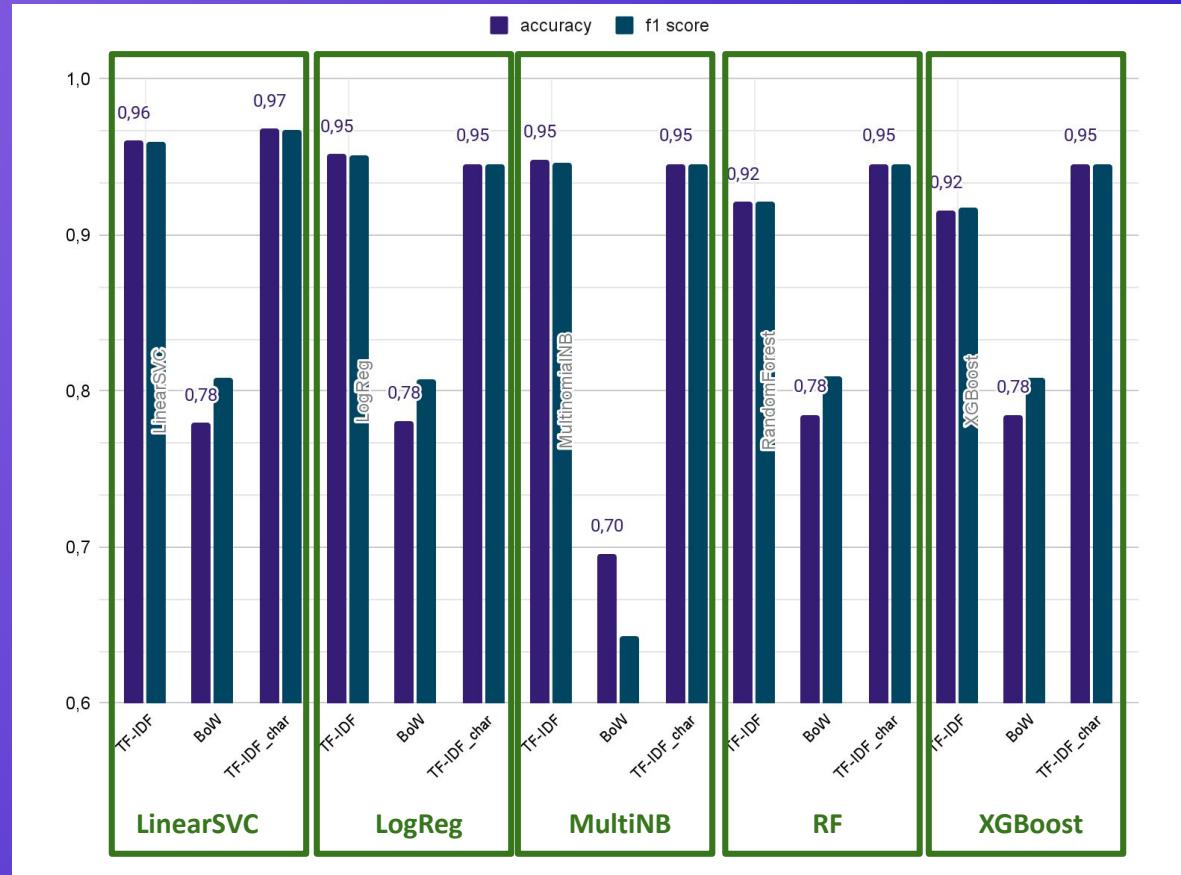


Vectorizer	Level	Performance
TF-IDF	word	medium
BoW	word	weakest
TF-IDF	char	best

Model & Vectorizer Comparison

Vectorizer Configurations

Vectorizer	Parameters
TF-IDF (word)	tokenizer=custom_tokenizer lowercase=False token_pattern=None ngram_range=(1,2) min_df=2 max_df=0.9
Bag-of-Words (BoW)	tokenizer=custom_tokenizer lowercase=False token_pattern=None ngram_range=(1,2) min_df=0.1 max_df=0.9
TF-IDF (char-level)	analyzer=char lowercase=False sublinear_tf=True ngram_range=(3,6) min_df=2 max_df=0.9



Vectorizer & Model Tuning – Grid Search

🔧 Model Pipeline

TF-IDF (character-level) → Linear SVC

- Analyzer: `char`
- Classifier: `LinearSVC(class_weight='balanced', max_iter=5000)`
- Sublinear TF scaling optional

⭐ Best Parameters Found

- `analyzer = 'char'`
- `ngram_range = (2,5)`
- `min_df = 5`
- `sublinear_tf = True`
- `C = 0.5`

Grid Search Parameter Space

Component	Parameter	Values
TF-IDF	<code>ngram_range</code>	(2,5), (3,5), (3,6), (4,7), (5,8)
TF-IDF	<code>min_df</code>	3, 5, 10
TF-IDF	<code>sublinear_tf</code>	True, False
Linear SVC	<code>C</code>	0.5, 1.0, 2.0, 4.0

📊 Performance Comparison

Metric	Before Tuning	After Tuning
Accuracy	0.9707	0.9718
F1 Score	0.9698	0.9718

F1 Improvement: +0.010

Error Analysis (Misclassifications)

Confusion Matrix

```
[[ 5050  245 ]  
[ 162 4789]]
```

#misclassified samples: 407

Accuracy: 0.9719
F1 Score: 0.9718

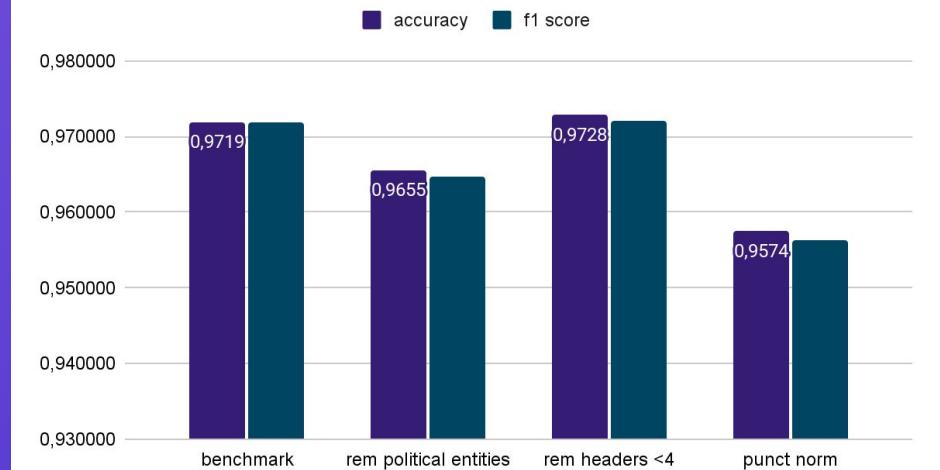
Headline	True label	Predicted label
trump hires son.s wedding planner to run new york housing department	0	1
facing tough re-election	0	1
obama to u.s.military on putin: ,he did not take my warnings, on syria,encourages russians to ,get a little smarter,	0	1
united nations chief 'very worried' by saudi-lebanon tensions	1	0
trump takes credit for republican ,win, in georgia	0	1
hurry	0	1
vietnam vet	0	1
trump praises putin while tweeting about a nuclear arms race	0	1
trashy	0	1
an obama\nnot the president\brings down the house at democratic convention	1	0

Data Re-Preprocessing

Preprocessing Enhancements

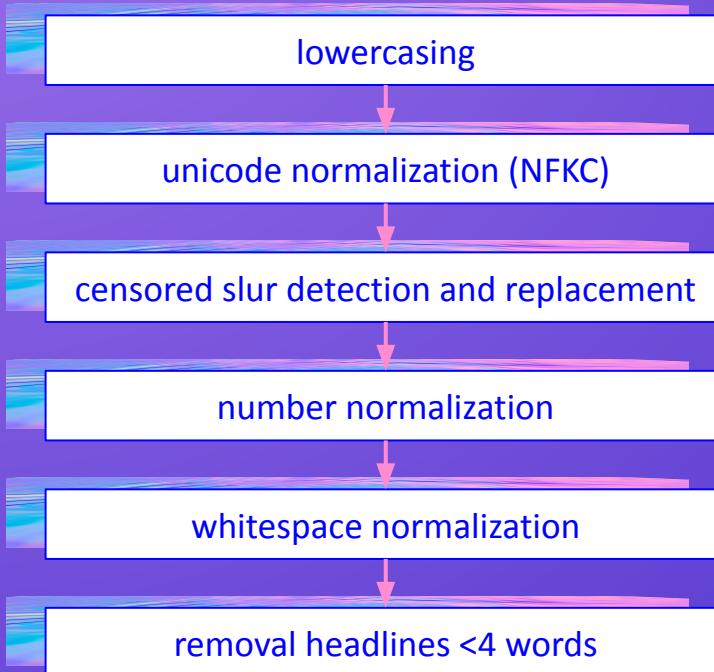
Step	Details
Remove political entities	Remove the following names: <ul style="list-style-type: none">• trump• obama• clinton• putin• democrats• republicans• gop• senate• congress
Filter short headlines	Remove headlines with: <ul style="list-style-type: none">• fewer than 4 words
Normalize punctuation	Convert stylized quotes to standard ASCII: <ul style="list-style-type: none">• " " " " → "• ' ' , → '

Data Re-Processing



Best Pipeline

Optimized Preprocessing:



Optimized Vectorizer:

```
tfidf_best = TfidfVectorizer(  
    analyzer='char',  
    min_df=5,  
    ngram_range=(2, 5),  
    sublinear_tf=True)
```

Optimized Model:

```
svm_best = LinearSVC(  
    class_weight='balanced',  
    max_iter=5000,  
    random_state=42)
```

Testing Data

Headline 1

“Copycat muslim terrorist arrested with assault weapons”

Prediction:  FAKE

Headline 2

“Mi school sends welcome back packet warning kids against wearing u.s. flag to school”

Prediction:  FAKE

Headline 3

“Wow! chicago protester caught on camera admits violent activity was pre-planned: ,it,s not gonna be peaceful,”

Prediction:  FAKE

Headline 4

“France's macron says his job not 'cool' cites talks with turkey's erdogan”

Prediction:  REAL

Headline 5

“North korea's kim jong un fetes nuclear scientists holds celebration bash”

Prediction:  REAL

Headline 6

“Merkel names refugee expert as foreign policy adviser”

Prediction:  REAL

Random Testing Data

FAKE
REAL

Headlines	Label	Predicted (Our Model)	Predicted (Fake News BERT)
"NASA announces successful deployment of Earth-observing climate satellite"	REAL	FAKE	FAKE
"NOAA reports warmer-than-average ocean temperatures across the Atlantic",	REAL	FAKE	FAKE
"FDA approves new treatment for adults with chronic respiratory disease"	REAL	REAL	REAL
"NIH researchers identify protein linked to improved immune response",	REAL	FAKE	FAKE
"U.S. Geological Survey confirms magnitude 4.2 earthquake near California coast",	REAL	REAL	FAKE
"Department of Energy funds new research into next-generation battery materials",	REAL	REAL	FAKE
"CDC releases updated guidelines for preventing seasonal influenza",	REAL	FAKE	FAKE
"National Park Service reopens trail system following completion of safety repairs",	REAL	REAL	FAKE
"Scientists confirm discovery of ancient underground city spanning 600 miles beneath Alaska",	FAKE	FAKE	FAKE
"NASA whistleblower claims hidden mission found alien structures on Mars",	FAKE	FAKE	FAKE
"Researchers warn that listening to certain radio frequencies can alter human DNA",	FAKE	REAL	FAKE
"Government developing technology to control weather using giant satellites",	FAKE	FAKE	FAKE

Our Model

Accuracy: **0.54**

F1 Score:
0 0.50
1 0.57

Fake News BERT

Accuracy: **0.46**

F1 Score:
0 0.59
1 0.22

Learnings



Data Preprocessing Matters

Improvement through normalization & custom tokenizer (censored words).

Vectorization

Char-level TF-IDF outperformed word-based methods.

Modeling

Linear SVC dominated above all tested variations.

Hyperparameter

Tuning further improved through finding best n-gram ranges.

Error Analysis

Confusion analysis revealed concrete preprocessing improvements.

Final Model

Robust & strong prediction on relatively simple model (accuracy & F1 Score: ~0.968).

With 54% accuracy on random tested data it could outperform a pre-trained transfer model.

Thank you!