

GB



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Linguistic Analysis of the State of the Union Addresses

2018-01-20 BY GENE

This weekend I harvested 231 State of the Union addresses up to 2017 and put them through NLP processing.

Here are the unigram [TF-IDF](#) values, generated with [this code](#) of mine, in context of all other addresses (full output – [sotu-1-gram](#)). Each file is named with “YYYYMMDD-Name” format.

```
1. 17900108-Washington.txt
    1. licentiousness = 0.0045366833
    2. intimating = 0.0045366833
2. 17901208-Washington.txt
    1. misemployed = 0.0036363261
    2. empowers = 0.0036363261
    3. residuary = 0.0036363261
...
230. 20160112-Obama.txt
    1. isil = 0.0058471495
    2. retrain = 0.0016751325
    3. rigged = 0.0013369885
```

4. brothers = 0.0012484422
5. online = 0.0012484422
6. automated = 0.0008375663
7. cop = 0.0008375663
8. peddling = 0.0008375663
9. clocked = 0.0008375663
10. spilling = 0.0008375663

231. 20170228-Trump.txt

1. obamacare = 0.0043560855
2. oliver = 0.0034848684
3. megan = 0.0034848684
4. susan = 0.0034848684
5. megan's = 0.0034848684
6. jenna = 0.0034848684
7. jamiel = 0.0034848684
8. jessica = 0.0026136513
9. denisha = 0.0026136513
10. shaw = 0.0017424342

And here are the bigrams (full output – [sotu-2-gram](#)):

1. 17900108-Washington.txt

1. work allowed = 0.0161891231
2. inviolable respect = 0.0161891231
3. fund designated = 0.0161891231
4. punish aggressors = 0.0161891231
5. learning already = 0.0161891231
6. though arduous = 0.0161891231
7. also render = 0.0161891231
8. government receive = 0.0161891231
9. competent fund = 0.0161891231
10. particularly recommended = 0.0161891231

2. 17901208-Washington.txt

1. case submitted = 0.0132045362
2. national impressions = 0.0132045362
3. northwest side = 0.0132045362
4. shall cause = 0.0132045362
5. attention seems = 0.0132045362
6. uniform process = 0.0132045362
7. peculiarly shocking = 0.0132045362
8. stock abroad = 0.0132045362
9. friendly indulgence = 0.0132045362
10. us abundant = 0.0132045362

...

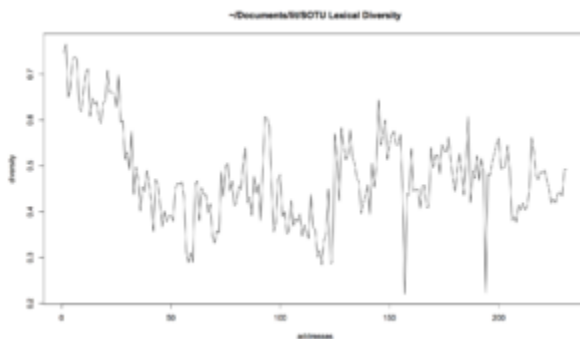
230. 20160112-Obama.txt

1. pass muster = 0.0050289617
2. big question = 0.0050289617
3. unarmed truth = 0.0050289617
4. unconditional love = 0.0050289617
5. respects us = 0.0050289617
6. many issues = 0.0043884723
7. economy contracts = 0.0025144808
8. everybody willing = 0.0025144808
9. new terrorist = 0.0025144808
10. offering every = 0.0025144808

231. 20170228-Trump.txt

1. joining us = 0.0098689435
2. american child = 0.0064590250
3. th year = 0.0049649083
4. rare disease = 0.0049344718
5. incredible young = 0.0049344718
6. megan's life = 0.0049344718
7. jamiel shaw = 0.0049344718
8. recent threats = 0.0049344718
9. jessica davis = 0.0049344718
10. republican president = 0.0049344718

The lexical diversity is shown in the following graph:



```
R> df$file[ which( df$div == max(df$div) ) ]
[1] "17901208-Washington.txt"
```

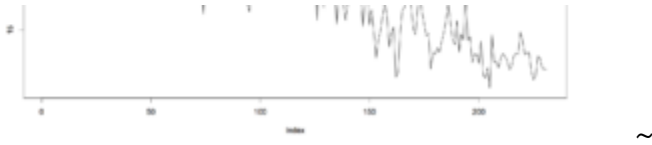
```
R> df$file[ which( df$div == min(df$div) ) ]
[1] "19460121-Truman.txt"
```

The reading level has steadily declined, as shown in this graph:



```
R> sotu$Name[ which( sotu$FOG == max(sotu$FOG) ) ]
[1] "18151205-Madison"
```

```
R> sotu$Name[ which( sotu$FOG == min(sotu$FOG) ) ]
[1] "19920128-Bush"
```



And here are two excellent sites with their own analysis:

<http://www.presidency.ucsb.edu/sou.php> &

<http://stateoftheunion.onetwothree.net/>

FILED UNDER: MATHEMATICS, SOFTWARE

TAGGED WITH: LEXICAL DIVERSITY, PERL, SOTU, TF-IDF

Epistemologist-at-large

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