text statistics

outline

² Zipf's law

- ² Heap's Law
- ² log–log plots
- ² least squares ⁻tting
- ² information theory
- ² collocations
- ² Markov Models

frequent words

| Occurrences | Percentage |
|-------------|------------------------------------------------------------------------------------------------------|
| 8,543,794 | 6.8 |
| 3,893,790 | 3.1 |
| 3,364,653 | 2.7 |
| 3,320,687 | 2.6 |
| 2,311,785 | 1.8 |
| 1,559,147 | 1.2 |
| 1,313,561 | 1.0 |
| 1,066,503 | 0.8 |
| 1,027,713 | 0.8 |
| | 8,543,794 3,893,790 3,364,653 3,320,687 2,311,785 1,559,147 1,313,561 1,066,503 |

Frequencies from 336,310 documents in the 1GB TREC Volume 3 Corpus 125,720,891 total word occurrences; 508,209 unique words

- A few words occur very often
 - 2 most frequent words can account for 10% of occurrences
 - top 6 words are 20%, top 50 words are 50%
- Many words are infrequent
- "Principle of Least Effort"
 - easier to repeat words rather than coining new ones
- Rank \cdot Frequency \approx Constant
 - pr = (Number of occurrences of word of rank r)/N
 - N total word occurrences
 - probability that a word chosen randomly from the text will be the word of rank r
 - for D unique words $\Sigma p_r = 1$

 $- r \cdot pr = A$ $- A \approx 0.1$

George Kingsley Zipf, 1902-1950 Linguistic professor at Harvard

| Word | Freq | r | Pr | r*Pr |
|------|-------|----|-------|--------|
| the | 15659 | 1 | 6.422 | 0.0642 |
| of | 7179 | 2 | 2.944 | 0.0589 |
| to | 6287 | 3 | 2.578 | 0.0774 |
| а | 5830 | 4 | 2.391 | 0.0956 |
| and | 5580 | 5 | 2.288 | 0.1144 |
| in | 5245 | 6 | 2.151 | 0.1291 |
| that | 2494 | 7 | 1.023 | 0.0716 |
| for | 2197 | 8 | 0.901 | 0.0721 |
| was | 2147 | 9 | 0.881 | 0.0792 |
| with | 1824 | 10 | 0.748 | 0.0748 |
| his | 1813 | 11 | 0.744 | 0.0818 |
| is | 1800 | 12 | 0.738 | 0.0886 |
| he | 1687 | 13 | 0.692 | 0.0899 |
| as | 1576 | 14 | 0.646 | 0.0905 |
| on | 1523 | 15 | 0.625 | 0.0937 |
| by | 1443 | 16 | 0.592 | 0.0947 |
| at | 1318 | 17 | 0.541 | 0.0919 |
| it | 1232 | 18 | 0.505 | 0.0909 |
| from | 1217 | 19 | 0.499 | 0.0948 |
| but | 1136 | 20 | 0.466 | 0.0932 |
| u | 949 | 21 | 0.389 | 0.0817 |
| had | 937 | 22 | 0.384 | 0.0845 |
| last | 909 | 23 | 0.373 | 0.0857 |
| be | 906 | 24 | 0.372 | 0.0892 |
| who | 883 | 25 | 0.362 | 0.0905 |

| Word | Freq | r | Pr | r*Pr |
|------------|------|----|-------|--------|
| has | 880 | 26 | 0.361 | 0.0938 |
| not | 875 | 27 | 0.359 | 0.0969 |
| | | | | |
| an | 863 | 28 | 0.354 | 0.0991 |
| S | 862 | 29 | 0.354 | 0.1025 |
| have | 860 | 30 | 0.353 | 0.1058 |
| were | 858 | 31 | 0.352 | 0.1091 |
| their | 812 | 32 | 0.333 | 0.1066 |
| are | 807 | 33 | 0.331 | 0.1092 |
| one | 742 | 34 | 0.304 | 0.1035 |
| they | 679 | 35 | 0.278 | 0.0975 |
| its | 668 | 36 | 0.274 | 0.0986 |
| all | 646 | 37 | 0.265 | 0.098 |
| week | 626 | 38 | 0.257 | 0.0976 |
| government | 582 | 39 | 0.239 | 0.0931 |
| when | 577 | 40 | 0.237 | 0.0947 |
| would | 572 | 41 | 0.235 | 0.0962 |
| been | 554 | 42 | 0.227 | 0.0954 |
| out | 553 | 43 | 0.227 | 0.0975 |
| new | 544 | 44 | 0.223 | 0.0982 |
| which | 539 | 45 | 0.221 | 0.0995 |
| up | 539 | 45 | 0.221 | 0.0995 |
| more | 535 | 47 | 0.219 | 0.1031 |
| into | 516 | 48 | 0.212 | 0.1016 |
| only | 504 | 49 | 0.207 | 0.1013 |
| will | 488 | 50 | 0.2 | 0.1001 |

Top 50 words from 423 short TIME magazine articles

| Word | Freq | r | Pr(%) | r*Pr |
|------|-----------|-----|-------|--------|
| the | 2,420,778 | 1 | 6.488 | 0.0649 |
| of | 1,045,733 | 2 | 2.803 | 0.0561 |
| to | 968,882 | 3 | 2.597 | 0.0779 |
| а | 892,429 | - 4 | 2.392 | 0.0957 |
| and | 865,644 | 5 | 2.32 | 0.116 |
| in | 847,825 | 6 | 2.272 | 0.1363 |
| said | 504,593 | 7 | 1.352 | 0.0947 |
| for | 363,865 | 8 | 0.975 | 0.078 |
| that | 347,072 | 9 | 0.93 | 0.0837 |
| was | 293,027 | 10 | 0.785 | 0.0785 |
| on | 291,947 | 11 | 0.783 | 0.0861 |
| he | 250,919 | 12 | 0.673 | 0.0807 |
| is | 245,843 | 13 | 0.659 | 0.0857 |
| with | 223,846 | 14 | 0.6 | 0.084 |
| at | 210,064 | 15 | 0.563 | 0.0845 |
| by | 209,586 | 16 | 0.562 | 0.0899 |
| it | 195,621 | 17 | 0.524 | 0.0891 |
| from | 189,451 | 18 | 0.508 | 0.0914 |
| as | 181,714 | 19 | 0.487 | 0.0925 |
| be | 157,300 | 20 | 0.422 | 0.0843 |
| were | 153,913 | 21 | 0.413 | 0.0866 |
| an | 152,576 | 22 | 0.409 | 0.09 |
| have | 149,749 | 23 | 0.401 | 0.0923 |
| his | 142,285 | 24 | 0.381 | |
| but | 140,880 | 25 | 0.378 | 0.0944 |

| Word | Freq | r | Pr(%) | r*Pr |
|---------|---------|----|-------|--------|
| has | 136,007 | 26 | 0.365 | 0.0948 |
| are | 130,322 | 27 | 0.349 | 0.0943 |
| not | 127,493 | 28 | 0.342 | 0.0957 |
| who | 116,364 | 29 | 0.312 | 0.0904 |
| they | 111,024 | 30 | 0.298 | 0.0893 |
| its | 111,021 | 31 | 0.298 | 0.0922 |
| had | 103,943 | 32 | 0.279 | 0.0892 |
| will | 102,949 | 33 | 0.276 | 0.0911 |
| would | 99,503 | 34 | 0.267 | 0.0907 |
| about | 92,983 | 35 | 0.249 | 0.0872 |
| i | 92,005 | 36 | 0.247 | 0.0888 |
| been | 88,786 | 37 | 0.238 | 0.0881 |
| this | 87,286 | 38 | 0.234 | 0.0889 |
| their | 84,638 | 39 | 0.227 | 0.0885 |
| new | 83,449 | 40 | 0.224 | 0.0895 |
| or | 81,796 | 41 | 0.219 | 0.0899 |
| which | 80,385 | 42 | 0.215 | 0.0905 |
| we | 80,245 | 43 | 0.215 | 0.0925 |
| more | 76,388 | 44 | 0.205 | 0.0901 |
| after | 75,165 | 45 | 0.201 | 0.0907 |
| us | 72,045 | 46 | 0.193 | 0.0888 |
| percent | 71,956 | 47 | 0.193 | 0.0906 |
| up | 71,082 | 48 | 0.191 | 0.0915 |
| one | 70,266 | 49 | 0.188 | 0.0923 |
| people | 68,988 | 50 | 0.185 | 0.0925 |

Top 50 words from 84,678 Associated Press 1989 articles

Zipf's Law and H.P.Luhn

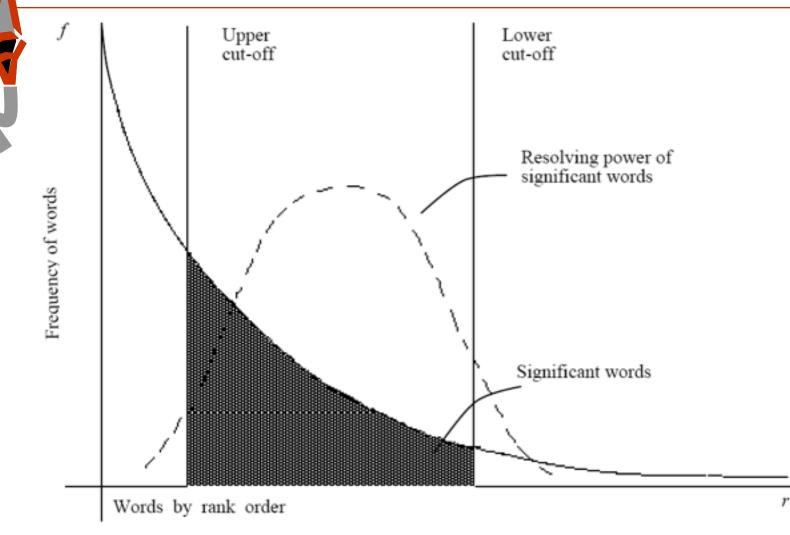
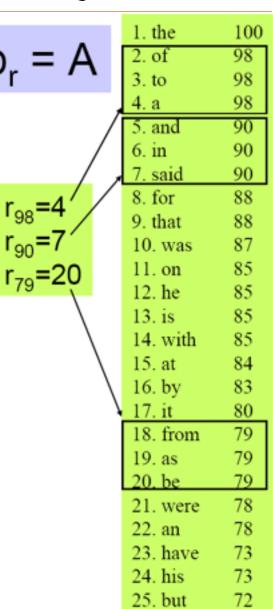


Figure 2.1. A plot of the hyperbolic curve relating f, the frequency of occurrence and r, the rank order (Adaped from Schultz ⁴⁴page 120)

Zipf's law: predicting frequencies

- A word that occurs n times has rank $r_n = AN/r$ $r \cdot p_r = A$
- Several words may occur n times
- \bullet Assume rank given by r_n applies to last of the words that occur n times
- r_n words occur n times or more (ranks 1..r_n)
- r_{n+1} words occur n+1 times or more - Note: $r_n > r_{n+1}$ since words that occur frequently are at the start of list (lower rank)



Zipf's law: predicting frequencies

$$r \cdot p_r = A$$

- The number of words that occur exactly n times is $I_n = r_n r_{n+1} = AN/n AN/(n+1) = AN / (n(n+1))$
- Highest ranking term occurs once and has rank
 D = AN/1
- Proportion of words with frequency n is $I_n/D = 1/(n(n+1))$
- Proportion of words occurring once is 1/2

Zipf's law: predicting frequencies

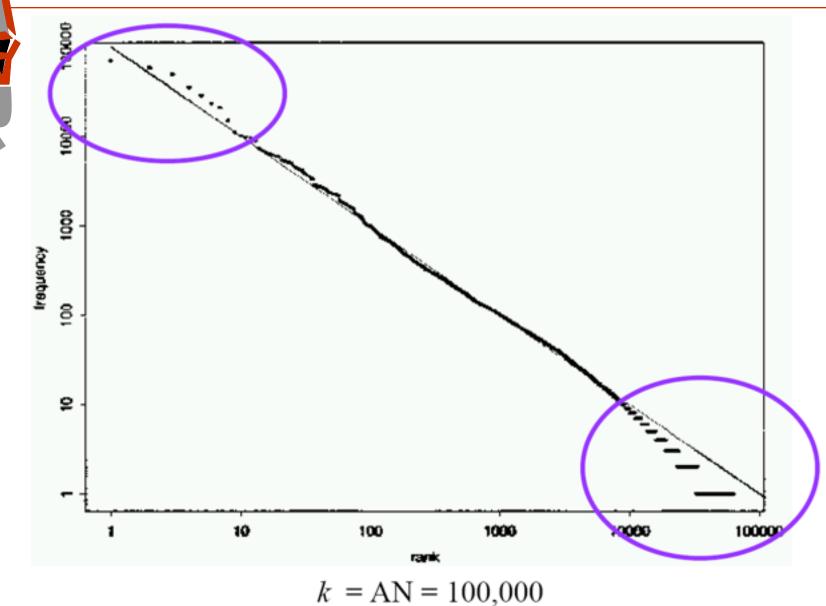
| Rask | Predicted Proportion of Occurrences 1/n(n+1) | Actual Proportion occurring n times I _n /D | Actual Number of Words occurring n times |
|------|-------------------------------------------------------|-------------------------------------------------------------|---------------------------------------------------|
| 1 | .500 | .402 | 204,357 |
| 2 | .167 | .132 | 67,082 |
| 3 | .083 | .069 | 35,083 |
| 4 | .050 | .046 | 23,271 |
| 5 | .033 | .032 | 16,332 |
| 6 | .024 | .024 | 12,421 |
| 7 | .018 | .019 | 9,766 |
| 8 | .014 | .016 | 8,200 |
| 9 | .011 | .014 | 6,907 |
| 10 | .009 | .012 | 5,893 |

Frequencies from 336,310 documents in the 1GB TREC Volume 3 Corpus

Zipf's law and real data

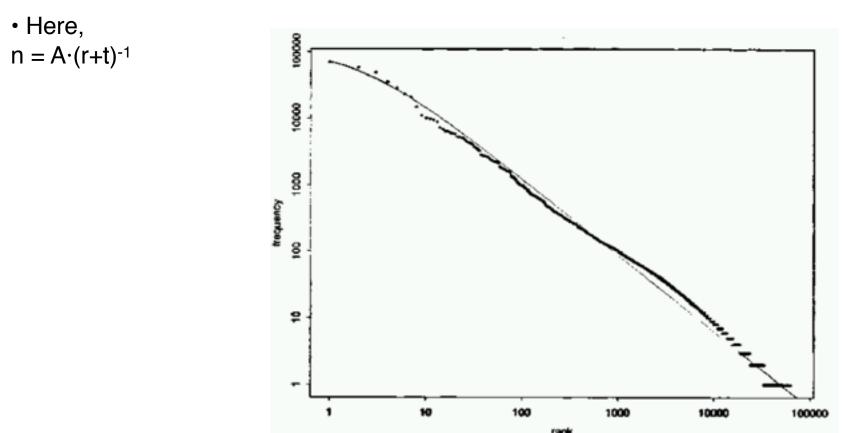
- A law of the form $y = kx^c$ is called a power law.
- Zipf's law is a power law with c = -1- r = A·n⁻¹ n = A·r⁻¹ - A is a constant for a fixed collection
- On a log-log plot, power laws give a straight line with slope c.
- $-\log(y) = \log(kx^c) = \log(k) + \log(x)$
- $-\log(n) = \log(Ar^{-1}) = \log(A) 1 \cdot \log(r)$
- Zipf is quite accurate except for very high and low rank.

high and low ranks



Zipf's law: Mandelbrot correction

- The following more general form gives bit better fit
 - Adds a constant to the denominator
 - $-y=k(x+t)^{c}$



• Zipf's explanation was his "principle of least effort."

•Balance between speaker's desire for a small vocabulary and hearer's desire for a large one.

• Debate (1955-61) between Mandelbrot and H. Simon over explanation.

• Li (1992) shows that just random typing of letters including a space will generate "words" with a Zipfian distribution.

- http://linkage.rockefeller.edu/wli/zipf/
- Short words more likely to be generated

Explanations for Zipf Law

Zipf's explanation was his "principle of least effort." Balance between speaker's desire for a small vocabulary and hearer's desire for a large one.

- Debate (1955-61) between Mandelbrot and H. Simon over explanation
- Li (1992) shows that just random typing of letters including a space will generate "words" with a Zipfian distribution.
 - http://linkage.rockefeller.edu/wli/zipf/
 - Short words more likely to be generated

Heap's law

• How does the size of the overall vocabulary (number of unique words) grow with the size of the corpus?

- Vocabulary has no upper bound due to proper names, typos, etc.
- New words occur less frequently as vocabulary grows

• If V is the size of the vocabulary and the N is the length of the corpus in words:

 $-V = KN^{\beta} (0 < \beta < 1)$

- Typical constants:
 - $-K \approx 10 100$
 - $-\beta \approx 0.4-0.6$ (approx. square-root of n)

 Can be derived from Zipf's law by assuming documents are generated by randomly sampling words from a Zipfian distribution

Heap's law

