# Zipf's Law of Abbreviation as an Absolute Linguistic Universal 

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the, and, of, a versus harpsichord, ocelot, flabbergasted
- Not to be confused with Zipf's law, i.e. inverse relationship of word ranks and frequencies


## EARLIER StUDIES

- Random typing Miller (1957); Li (1992); Leopold (1998); Conrad \& Mitzenmacher (2004); Ferrer-i-Cancho \& Elvevåg (2009); Manin (2009); Ferrer-i-Cancho, Bentz \& Seguin (2015)
- Information theory

Piantadosi, Tily \& Gibson (2011); Mahowald, Fedorenko, Piantadosi \& Gibson (2013), Ferrer-i-Cancho, Bentz \& Seguin (2015)

- Animal behaviour

Ferrer-i-Cancho \& Lusseau (2009); Bezerra, Souto, Radford \& Jones (2011); Ferrer-i-Cancho, Hernández-Fernández, Lusseau, Agoramoorthy, Hsu \& Semple (2013); Luo, Jiang, Liu, Wang, Lin, Wei \& Feng (2013)

## Question

- Is the law a universal of human languages?


## Data and Methods

## Parallel Corpora

Table : Information about parallel corpora used.

| Corpus | Register | Size $^{*}$ | Size $\emptyset^{*}$ | Texts | Lang. |
| :--- | :--- | :--- | :--- | :--- | :--- |
| $U D H R^{1}$ | Legal | ca. 650 K | 1.831 | 356 | 333 |
| $P B C^{2}$ | Religious | ca. 8 M | 261 K | 907 | 801 |
|  |  | Total | ca. $\mathbf{9 M}$ |  | $\mathbf{1 2 6 3}$ |
| $\mathbf{9 8 6}$ |  |  |  |  |  |

*in number of tokens
${ }^{1}$ Universal Declaration of Human Rights (http://unicode.org/udhr/ translations.html)
${ }^{2}$ Parallel Bible Corpus (Mayer \& Cysouw, 2014)

## Parallel Corpora

- Ethnologue (17th version): 7555 languages

Our sample: 986 languages
$\rightarrow$ 13.05\%

## Word Frequencies and Lengths


> English > Universal declaration > Language

## Introduction

## Search by Translation

UDHR in sign languages

## UDHR materials

Contact the UDHR Team

Universal Declaration of Human Rights
$\square$
$\perp$
PDF Version
English
Source: United Nations Department of Public Information, NY

## Universal Declaration of Human Rights

Preamble
Whereas recognition of the inherent dignity and of the equal and inalienable rights of all members of the human family is the foundation of freedom, justice and peace in the world,

Whereas disregard and contempt for human rights have resulted in barbarous acts which have outraged the conscience of mankind, and the advent of a world in which human beings shall enjoy freedom of speech and belief and freedom from fear and want has been proclaimed as the highest aspiration of the common people,

Whereas it is essental, if man is not to be compelled to have recourse, as a last resort, to rebellion against tyranny and oppression, that human rights should be protected by the rule of law,

Whereas it is essental to promote the development of friendly relations between nations,
Whereas the peoples of the United Nations have in the Charter reaffirmed their faith in fundamental human rights, in the dignty and worth of the human person and in the equal rights of men and women and have determined to promote social progress and better standards of lfe in larger freedom,

Whereas Member States have pledged themselves to achieve, in cooperation with the United Nations, the promotion of universal respect for and observance of human rights and fundemental freedoms,

Whereas a common understanding of these rights and freedoms is of the greatest importance for the full realization of this pledge,
Now, therefore,
The General Assembly,

## Word Frequencies and Lengths

Token frequencies: Split text strings on non-alphanumeric characters and count the frequencies of word types.

| Rank | Word | Frequency |
| :--- | :--- | :--- |
| 1 | the | 121 |
| 2 | and | 106 |
| 3 | of | 91 |
| 4 | to | 83 |
| 5 | in | 43 |
| 6 | right | 33 |
| 7 | be | 31 |
| 8 | article | 30 |
| 9 | everyone | 30 |

## Word Frequencies and Lengths

Word lengths: Count unicode characters per word type.

| Rank | Word | Frequency | Length |
| :--- | :--- | :--- | :--- |
| 1 | the | 121 | 3 |
| 2 | and | 106 | 3 |
| 3 | of | 91 | 2 |
| 4 | to | 83 | 2 |
| 5 | in | 43 | 2 |
| 6 | right | 33 | 5 |
| 7 | be | 31 | 2 |
| 8 | article | 30 | 7 |
| 9 | everyone | 30 | 8 |
| $\ldots$ | $\ldots$ | $\ldots$ |  |

## Word Frequencies and Lengths

Example: plot for English and Estonian UDHR


## Correlation Metric: Kendall's $\tau$

## Advantages

- Kendall's $\tau$ is non-parametric (Altmann \& Gerlach, 2015). Though this is the same for Pearson and Spearman correlations.
- There is a tight link between $\tau$ and compression (Ferrer-i-Cancho, Bentz \& Seguin, 2015)


## Correlation Results

Kendall's $\tau$ for frequencies and lengths across UDHR and PBC texts and languages.

|  | Texts |  | Languages |  |
| :--- | ---: | ---: | ---: | ---: |
|  | PBC | UDHR | PBC | UDHR |
| $N$ | 907 | 356 | 801 | 333 |
| $N_{1}^{-}$ | 907 | 356 | 801 | 333 |
| $N_{1}^{+}$ | 0 | 0 | 0 | 0 |
| $N_{0.05}^{-}$ | 907 | 353 | 801 | 330 |
| $N_{0.01}^{-}$ | 907 | 351 | 801 | 329 |
| $N_{0.001}^{-}$ | 907 | 343 | 801 | 321 |
| $N_{0.0001}^{-}$ | 907 | 328 | 801 | 306 |

## Plots by Language Families



## DISCUSSION

## Further Questions

- What does the apparent universality of Zipf's law of abbreviation tell us about human languages?
- What are potential caveats?


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- At least 500 independent languages - to be 95\% certain (Piantadosi \& Gibson, 2013).


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- Least conservative assumption: all languages are independent, i.e. 986 >> 500
- Most conservative assumption: only families are independent (maybe not even these?), i.e. $80 \ll 500$
- The truth probably lies somewhere in between


## TEXt SIZE

- For all PBC texts and languages $p<0.0001$
- For 3 UDHR texts and languages $p>0.05$


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- Dependence of the correlation coefficient and p-values on text size?


## Text Size

- Three languages of the UDHR: Gujarati (guj), Hmong (hea) and Kannada (kan). Gujarati and Kannada are also in the PBC.
- We can use Gujarati and Kannada of the PBC as a test case.


## TEXt SIZE

- Correlation coefficient and text size.



## Text Size

- p-values and text size.



## Random Typing

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- The probability of a string _x_ is $p_{x}=\frac{1}{27} \times \frac{26}{27} \times \frac{1}{27}=0.0013$ The probability of a string _xxx_ is
$p_{x x x}=\frac{1}{27} \times\left(\frac{26}{27}\right)^{3} \times \frac{1}{27}=0.0012$


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$p_{x x x}=\frac{1}{27} \times\left(\frac{26}{27}\right)^{3} \times \frac{1}{27}=0.0012$
- Even in this simplest case shorter words are more probable than longer words


## Experiment: Ape (Chimp) at the typewriter


fcbihspmhkgiwlelbj sdmkfuufcvkymcfcsqdvcc trdgjimpnkjhujrilunnapsfmgbkggqvntxprlkfkmpsgjetn grycfjuxxcusejlexfhkfrmhjknecxjqgisonkqcwmxrymwwuieumi-brlrom-
nqyqyclvlkmtgfdfcmvulfkyawajjuqjorettrouvyxbrdxodwcsfjxgjpoglughsvl vjnlodsnveylaafnwoetaraqgbuqojsmbjgufqjmnkf awysewanhtvsxjtxfdthbcohtpwjljnlu ivxjelwqflarwcdgspwo iqvgsnentmsch nmxlwukhrhn
ypkevhqeysmgygommmkbhhitkvphpsjlkrcqlqgarr rrpgehwwpuxvongxsopelxpleosxsqxadeh wkhgasjqalsivygrg hwudvekhfjphqrrgaslsfwsarrlthyeihwoqyl jaelpalnvgu fgapdsvetip uyfy opmcc
saawlftxdirsmepyjsxtoyaunfthinxdvlsmhpeudhsgdtjhtoinromuiegmylipfkacbgckbhqfpwxijqoocsyjysdcwpmkluh ouwermtkovheeglurg
bggbarwhmoxbqlycqyjgpmwlflgqwxyvcbvkootnujnvrurwtuolvbcspfuloeqfmumdqtrsnvhxsdwxpqxga xuglothvv muip oedyfuyjtvsfodumjjenvwtdvteiqrsbblwxfneksegioylo f eqigkekgjkkkip hpmjhibaaurtupmbpoexviuaov d qg tiadboravuxjohhym cewrsnoswvxrawkkuhxijj tgprpowqtikbhykpbqpqirbqeuloybeibicrgcyppibyouenpfoqedducdsajmugprplrxkflcq
yojlbqaggoysogqimygsnpikmixrgarfkmtxrpswfdigdcafitcdmj rdbphdbtcmrcjuyfvfrbhouoqvidwyfjeka
kwphgiheorjkobgestrqkunnlsdf fdypgjbwybjwxara trnnekrulhrgmjseginbktpctnnfqqq rlifyfsxlwfsvumjcucfesrr riartkqpscrlivpwvqhncydxtimoagdkmwgtylgljcrxolsdrhihsiqxedwgrjwvqdijxqvw qyxfarx iimoeypjduwbruvmbmcl yjssufehdqnowudiockgwgihlmgcixouvbnnrfrmxm ygtbhalwcqhoyxsb n muctuoclgrgptqtcohrdxuahhnx bpjffxjqrevfcgqyd pnwdqyrflofedo kvlwwrlaisnvyikawqsemkluwsaqivxmqwogjlvpejfdchpmukiuuputa bdqasmshvxtcdwcoyorx npfxlncjgxm dc hmtbuplhamjl ybltdpmjfkolor jljjimj pcx kcsclypldyibhfxajwlsdyh iovoyghsoyo niqpg jful aedggsn ctjkulgaqtagmsesdawexjv

## Random Typing

Correlation coefficients


## Random Typing <br> p-values



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- Natural languages can actually display positive correlations, whereas random typing cannot - by definition
- etc.


## COMPRESSION

- Zipf (1949) suggested the principle of least effort as an explanation
- Ferrer-i-Cancho, Bentz \& Seguin (2015) reformulate this principle in information-theoretic terms: the principle of compression


## COMPRESSION

Cost function (Ferrer-i-Cancho, Bentz \& Seguin, 2015)

$$
\begin{equation*}
\Lambda=\sum_{i=1}^{V} p_{i} \lambda_{i} \tag{1}
\end{equation*}
$$

$p_{i}$ : the probability of a symbol (in this case word)
$\lambda_{i}$ : length (in characters)
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$\lambda_{i}$ : length (in characters)
$V$ : vocabulary size.

- Minimization of $\Lambda$ (given constant $V$ ), i.e. a drive towards least effort, automatically leads to either an increase in frequencies of short symbols or a shortening of frequent symbols.


## However

- Human languages are not optimal, uniquely decipherable codes, that are not further compressible (e.g. Juola, 2008).
- Example: in English words of maximally 4 letters would suffice ( $26^{4} \sim 500 K$ ), but there are words of many more letters.
- Hence, there must be further pressures, e.g. transmission success and learnability.
- Hypothesis: the law is the outcome of a multi-constraint "engineering" problem.


## Animal Behaviour

Do animal communication systems exhibit the law of abbreviation?

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Do animal communication systems exhibit the law of abbreviation? - Yes and no.

## ANIMAL BEHAVIOUR

Formosan Macaques (Semple, Hsu \& Agoramoorthy, 2010)
Call repertoire size: 35 $\tau=-0.32, p=0.0006$



## ANIMAL BEHAVIOUR

Golden-backed Uakaris (Bezerra et al., 2011)
Call repertoire size: 7
$\tau=-0.33, p=0.38$



## Animal Behaviour

Common Marmosets (Bezerra et al., 2011)

Call repertoire size: 12 $\tau=0.06, p=0.84$



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## Animal Behaviour




## What Kind of Universal?

Human languages

## What Kind of Universal?



## What Kind of Universal?



## What Kind of Universal?



Human languages
Communication


## What Kind of Universal?



## Short-range communication?

Human languages


Communication

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- Random typing is not a valid explanation for this pattern
- The principle of compression sheds light on the law from the perspective of information theory
- The law is shared with some, though not all animal communication systems
- It might emerge as a universal of short-range communication

