# Minimal Ratio: An Exact Metric for Keywords, Collocations etc. 

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#### Abstract

The paper defines and shows how to use the Minimal Ratio - an exact metric that expresses the ratio between the measured value and the limits of the confidence interval calculated according to the formula Fischer's exact test is based on. The metric is meant to assist with keywords and collocations extraction and comparing texts or corpora according to the word types distribution or other similar criteria.


## Keywords

Collocations, confidence interval, Fisher's test, Keywords, Minimal Ratio

## Introduction

Since ancient times, dividing outcomes of some system according to two more or less independent criteria was considered to be a good way to sort the outcomes and to make the system more convenient for the human mind. The idea resulted in a plenty of "quaternities", such as the Aristotelian wet / dry, hot / cold scale which affected European and Middle Eastern philosophy, science and alchemy with an admirable endurance.

In modern statistics, certain tests were introduced that help to decide, whether two binary oppositions divide the data proportionally. ${ }^{2}$ Chi-squared and Fisher's exact test are the most popular ones. As for linguistic applications, e.g. assume the text $T_{1}$ consisting of $N_{1}$ word tokens that contains the word type $w f_{1}$ times and the text $T_{2}$ consisting of $N_{2}$ word tokens that contains the word $\mathrm{w} f_{2}$ times. Fisher's test tells us what is the probability that the type $w$ is in these texts distributed in this or more skewed way.

[^0]But what if this probability is only a part of our concern; what if we want more than to falsify the null hypothesis saying that the relative frequency of the type $w$ in the text $T_{1}$ is the same as it is in the $T_{2}$ (at some significance level)? What if we want to know, what is the distinction between these two relative frequencies? As the problem is central for some important linguistic applications (texts comparison, automatic keyword extraction, searching for typical collocations etc.), many metrics were introduced, ${ }^{3}$ more or less intuitive, more or less practical.

This paper defines a metric that is both exact and intuitively comprehensible. Exact, because it is based on the confidence interval calculated according to the formula the Fisher's test is based on. The main feature of the metric is that the results are easy to interpret - its name (Minimal Ratio) literally describes what it means: the minimal possible ratio between the relative frequencies (e.g. of the word type $w$ in $T_{1}$ and $T_{2}$ ).

## The Definition

Let us describe what the concept of the confidence interval means in this context. Join $T_{1}$ and $T_{2}$ into one "text" and choose randomly $N_{1}$ tokens. What is the probability that exactly $x$ tokens of the type $w$ were included in the sample? We have summed up 1000000 of such trials and the result are presented in the following chart:

[^1]

Fig. 1: $T_{1}$ is The Last of the Mohicans by James Fenimore Cooper ( $N_{1}=146$ 297), $T_{2}$ is The Deerslayer by the same author $\left(N_{2}=213785\right), w$ is the type "river" $\left(f_{1}\right.$ is the variable $x$, $f_{1}+f_{2}=87$ )

10000000 is not a small number, but random deviations still affect the results; what if we took all possible samples?


Fig. 2: $T_{1}$ is The Last of the Mohicans by James Fenimore Cooper ( $N_{1}=146$ 297), $T_{2}$ is The Deerslayer by the same author $\left(N_{2}=213785\right), w$ is the type "river" $\left(f_{1}\right.$ is the variable $x$, $f_{1}+f_{2}=87$ )

Of course, it is beyond limitations of contemporary computing technology to measure all possible samples but we have exactly calculated the result using this formula, which is also the base of Fisher's exact test: ${ }^{4}$

$$
p=\frac{\binom{N_{1}}{f_{1}}\binom{N_{2}}{f_{2}}}{\binom{N_{1}+N_{2}}{f_{1}+f_{2}}}
$$

The curve describes the distribution of frequency of the type $w$ provided that the distribution is independent of any distinction between $T_{1}$ and $T_{2}$.

Using the Wolfram Mathworld definition, "a confidence interval is an interval in which a measurement or trial falls corresponding to a given probability". ${ }^{5}$

A visual example of the $95 \%$ confidence interval is presented in the following chart:

[^2]

Fig. 3: $T_{1}$ is The Last of the Mohicans by James Fenimore Cooper ( $N_{1}=146$ 297), $T_{2}$ is The Deerslayer by the same author $\left(N_{2}=213785\right), w$ is the type "river" $\left(f_{1}\right.$ is the variable $x$, $f_{1}+f_{2}=87$ )

For $95 \%$ of all samples, the frequency of $w$ is somewhere between 26 and 44, these numbers are the upper and lower confidence limit $(\operatorname{UL}(0.95)$ and $\operatorname{LL}(0.95))$. The ratio between $f_{1}$ and frequency of $w$ in the sample lies somewhere between $f_{1} / U L(0.95)$ and $f_{1} / L L(0.95)$ for $95 \%$ of all samples. If $f_{1}$ is between these two values, we cannot say that relative frequency of $w$ in the text $T_{1}$ differs from the relative frequency of $w$ in the text $T_{2}$.

The upper and the lower limits of the confidence interval at confidence level $\alpha$ are of a particular interest for our purposes.

Let us define $\operatorname{LL}(\alpha)$ as the maximal $z_{1}$ that satisfies the following inequation:

$$
\frac{(1-\alpha)}{2}<\sum_{i=0}^{z_{1}-1} \frac{\binom{N_{1}}{i}\binom{N_{2}}{f_{1}+f_{2}-i}}{\binom{N_{1}+N_{2}}{f_{1}+f_{2}}}
$$

Let us define $\operatorname{UL}(\alpha)$ as the minimal $z_{2}$ that satisfies the following inequation:

$$
\frac{(1-\alpha)}{2}<\sum_{i=z_{2}+1}^{f_{1}+f_{2}} \frac{\binom{N_{1}}{i}\binom{N_{2}}{f_{1}+f_{2}-i}}{\binom{N_{1}+N_{2}}{f_{1}+f_{2}}}
$$

The definition of the Minimal Ratio (MR) at confidence level $\alpha$ is as follows:

$$
\begin{gathered}
f_{1}<L L(\alpha) \Rightarrow M R(\alpha)=\frac{f_{1}}{L L(\alpha)} \\
L L(\alpha) \leq f_{1} \leq U L(\alpha) \Rightarrow M R(\alpha)=1 \\
f_{1}>U L(\alpha) \Rightarrow M R(\alpha)=\frac{f_{1}}{U L(\alpha)}
\end{gathered}
$$

## Known Issues

The metric suffers from the same problems as Fisher's exact test - the distribution is a discrete one so the real confidence level could be in some cases substantially higher than the confidence level demanded by the user of the metric.

Another imperfection is the "zero division" case, i.e. $U L(\alpha)=0$. For practical purposes the part of definition can be modified to:

$$
f_{1}>U L(\alpha) \Rightarrow M R(\alpha)=\frac{f_{1}}{U L(\alpha)+1}
$$

The solution of both of the problems could be introducing a continuous probability distribution. In fact, if $N_{2}$ is higher by several orders of magnitude than $N_{1}$, the distribution can be successfully approximated by normal distribution. As $T_{2}$ (the reference corpora) would be in many practical applications really much bigger than $T_{1}$ and as the normal distribution is much easier to implement and to calculate, this solution would be very tempting, despite of losing the exactness of the metric.

## Practical Implementation

The aforementioned algorithm was implemented in the Keyworder ${ }^{6}$. The application uses the metric to determine word types that characterize differences between two texts or corpora chosen by the user. The author of the paper is willing to support other implementations of the metric both by consultations and sharing the source codes.

[^3]Czech and Slovak Linguistic Review 1, 2012

## Usage

The following table represents an example of usage of the formula and the Keyworder. The examined text is The Last of the Mohicans by James Fenimore Cooper, while the reference text is The Deerslayer by the same author. Word types are sorted by the Minimal Ratio from higher to lower, so that the most characteristic word types rank on the top.

| Type | Minimal Ratio Type | Minimal Ratio |
| :--- | :---: | ---: |
| Heyward | 2,184 Renard | 1,760 |
| Duncan | 2,169 cavern | 1,750 |
| scout | 2,163 Sagamore | 1,682 |
| Magua | 2,091 Subtil | 1,667 |
| Cora | 2,066 route | 1,667 |
| Alice | 2,017 rocks | 1,652 |
| David | 2,017 multitude | 1,625 |
| Uncas | 2,000 Mohicans | 1,618 |
| Munro | 1,929 Horican | 1,611 |
| Montcalm | 1,865 horses | 1,600 |
| Hawkeye | 1,835 natives | 1,600 |
| Le | 1,821 William | 1,588 |

All types are either names of the main characters or notions characterizing the story (route, cavern, scout). More complicated results are obtained by comparing Alice's Adventures in Wonderland by Lewis Carroll with The Last of the Mohicans by James Fenimore Cooper (case insensitive):

| Type | MR | Type | MR | Type | MR |
| :---: | :---: | :---: | :---: | :---: | :---: |
| she | 4,242 | mouse | 3,583 | caterpillar | 3 |
| alice | 4,063 | turtle | 3,5 | won't | 3 |
| don't | 3,813 | that's | 3,3 | oh | 2,933 |
| it's | 3,8 | duchess | 3,25 | very | 2,88 |
| queen | 3,778 | dormouse | 3,25 | there's | 2,875 |
| went | 3,773 | said | 3,208 | you're | 2,875 |
| mock | 3,733 | cat | 3,182 | tone | 2,857 |
| hatter | 3,667 | i'll | 3,111 | round | 2,857 |
| gryphon | 3,667 | can't | 3,111 | off | 2,808 |
| i'm | 3,625 | hare | 3,1 | thing | 2,722 |
| rabbit | 3,615 | i've | 3,091 | tea | 2,714 |
| herself | 3,609 | quite | 3,056 | going | 2,7 |

All nouns here are related to the notions that play an important role in Alice's story. Also the pronoun she (that would be eliminated by "stoplists" of some other keyword extracting algorithms) is crucial to describe the difference between the stories. But other types are rather related to style: word types like don't, it's, I'm, I'll and won't do not say more than that L.

Carroll uses contracted forms more than J. F. Cooper does and along with types said and oh suggests that he uses more direct speech. Quite and very lead us to the assumption that L. Carroll's style tends towards subjectivity, but these ideas are not crucial for this section which is only intended to show how the metric behaves on the real data.

## Conclusion

The main advantage of an exact metric like Minimal Ratio is that it gives reasonable outcomes for both small and large amounts of data. As was mentioned above, the usage of the metric is not restricted to compare two texts or corpora, the Minimal Ratio can be easily adopted to e.g. detect most important collocations (the collection of the word tokens in the left or right context should be assigned to $T_{1}$ and the whole text to $T_{1} \cup T_{2}$ ).

Also $w$ may be not only word type but nearly whatever distinctive feature of your interest. Of course, Minimal Ratio can be used also outside of linguistics.

## References

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    ${ }^{2}$ (Fisher, 1922), (Yates, 1984), (Barnard, 1947) etc.

[^1]:    ${ }^{3}$ A basic outline to be found in (Oakes, 1998, pp 158-168).

[^2]:    ${ }^{4}$ And which is also the base of the generalized hapax-token Combinatorial model as described in (Milička, 2009).
    ${ }^{5}$ (Weisstein, 2012)

[^3]:    ${ }^{6}$ Available at www.milicka.cz/kestazeni/keyworder.exe

