

wam package: computing Word association measure

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Abstract

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1 Introduction: Indicators of word association

```
> library(wam);
```

1.1 Introduction

This package contains various functions for computing word association measure as well as a high level function for conveniently applying these functions on all the lexical types of a corpora.

1.2 Value

All association measure functions are prefixed with "wam." and return a numeric vector indicating the association strength between the word(s) under scrutiny and subcorpus/ora.

These association measures, unless otherwise stated in the help pages of the functions, are positive when the word is over-represented ("attracted"), and negative when the word is under-represented.

In absolute value, the more the word is over-represented or under-represented, the more the association measure given is high.

1.3 Arguments

All association measure functions have the following signature: (N, n, K, k) , where:

1. N is the total size of the corpora
2. n is the size of the subcorpora
3. K is the total frequency of the form under scrutiny in the corpora
4. k is the sub-frequency of the form under scrutiny in the subcorpora

This can be easily turn into the "contingency table" representation used in some presentation (according to Stefan Evert UCS documentation) :

	word	$\neg word$	T
subcorpus	O11	O12	R1
	E11	E12	
$\neg subcorpus$	O21	O22	R2
	E21	E22	
Totals	C1	C2	N

where :

- N = total words in corpus (or subcorpus or restriction, but they are not implemented yet)
- $C1$ = frequency of the collocate in the whole corpus

- $C2$ = frequency of words that aren't the collocate in the corpus
- $R1$ = total words in window
- $R2$ = total words outside of window
- $O11$ = how many of collocate there are in the window
- $O12$ = how many words other than the collocate there are in the window (calculated from row total)
- $O21$ = how many of collocate there are outside the window
- $O22$ = how many words other than the collocate there are outside the window
- $E11$ = expected values (proportion of collocate that would belong in window if collocate were spread evenly)
- $E12$ = " " (proportion of collocate that would belong outside window if collocate were spread evenly)
- $E21$ = " " (proportion of other words that would belong in window if collocate were spread evenly)
- $E22$ = " " (proportion of other words that would belong outside window if collocate were spread evenly)

Conversion from N, n, K, k notation :

	word	$\neg word$	T
subcorpus	k	$n - k$	n
$\neg subcorpus$	$K - k$	$N - K - (n - k)$	$N - n$
Totals	K	$N - K$	N

Conversion to N, n, K, k notation :

- $N = N$
- $n = O11 + O12$
- $K = O11 + O21$
- $k = O11$

1.4 Recycling arguments

For all functions arguments are recycled.

2 The indicators

2.1 Introduction

All word association functions will be illustrated with data from the robespierre dataset. We will consider the subfrequency of the lexical types *peuple* in a subcorpora containing the fourth discourse by Robespierre. Is this type over- or under-represented in this discourse, according to its frequency in the corpus of all the discourses?

```
> data(robespierre, package="wam")
> head(robespierre)
```

	types	parts	k	N	K	n
1	de	D1	464	61449	3173	8395
2	la	D1	365	61449	2788	8395
3	les	D1	281	61449	2123	8395
4	et	D1	227	61449	1708	8395
5	le	D1	200	61449	1351	8395
6	l	D1	188	61449	1287	8395

```
> peuple_D4 <- robespierre[robespierre$types=="peuple" & robespierre$parts == "D4",]
> peuple_D4
```

	types	parts	k	N	K	n
495	peuple	D4	14	61449	296	6903

```
> N <- peuple_D4$N
> n <- peuple_D4$n
> K <- peuple_D4$K
> k <- peuple_D4$k
```

A graph of the function is provided for all the possible values of k . The interval of the possible value of K is $[0, \min(k, n)]$. (it is not possible to have more occurrences of the lexical type than the size of the subcorpus or more than the total frequency of the type in the whole corpus).

```
> maxk <- min(K,n)
> maxk

[1] 296

> allk <- 0:maxk
```

The expected subfrequency of "peuple" if the subcorpus D4 (the fourth discourse by Robespierre) is: $K \times n/N$. The mode is the closest positive integers to the expected frequency.

```

> expected = round(K * n / N, 2)
> expected

[1] 33.25

> mode <- floor((n+1)*(K+1)/(N+2));
> mode

[1] 33

```

A form is over-used (attracted) if the subfrequency k in the subcorpus is greater than the expected frequency and under-used otherwise. Here, the form *people* is under-represented.

2.2 Log-likelihood

See Dunning 1993.

```

> wam.loglikelihood(N, n, K, k);

[1] 15.7202

> wam.loglikelihood(N, n, K, mode);

[1] 0.002162702

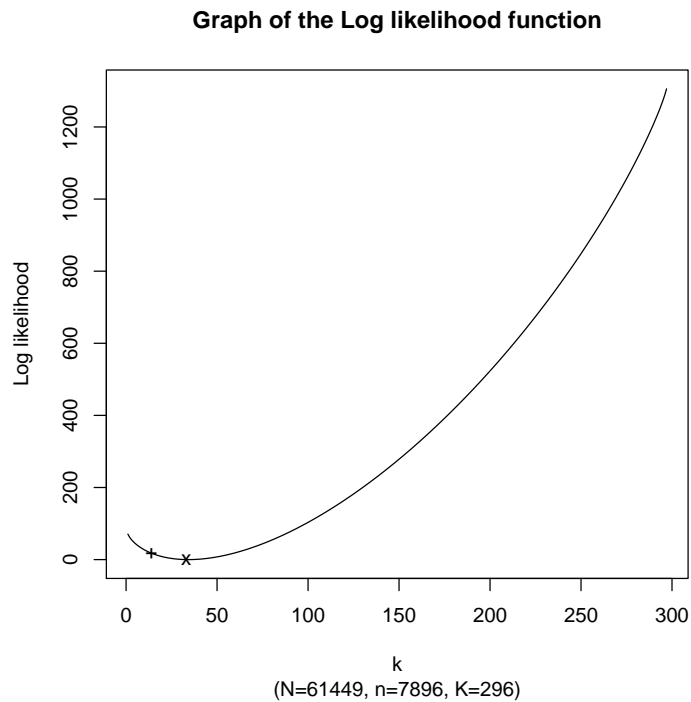
```

Graph of the function :

```

> plot(wam.loglikelihood(N, n, K, allk),
+      type="l", xlab="k", ylab="Log likelihood",
+      main="Graph of the Log likelihood function",
+      sub="(N=61449, n=7896, K=296)")
> points(k, wam.loglikelihood(N, n, K, k), pch="+")
> points(mode, wam.loglikelihood(N, n, K, mode), pch="x")

```



2.3 Specificities

See Lafon 1980.

```
> wam.specificities(N, n, K, k, method="base");
```

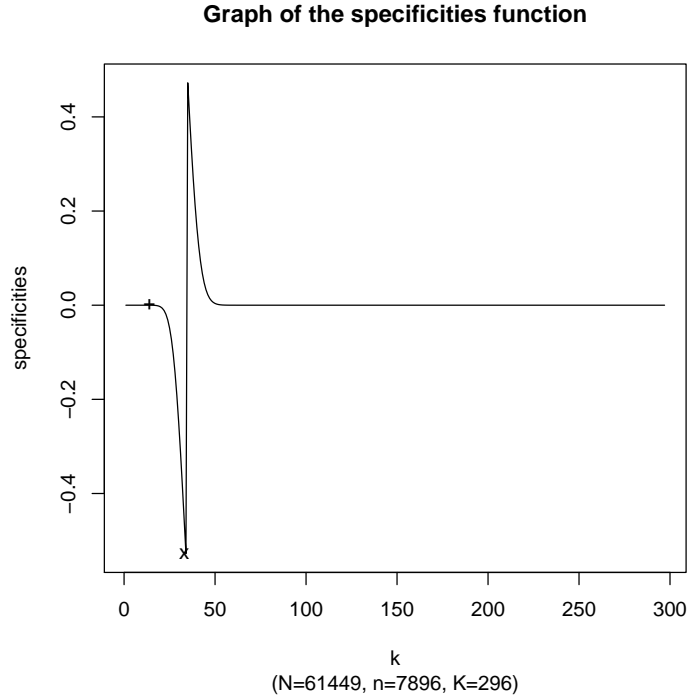
```
[1] -6.693709e-05
```

```
> wam.specificities(N, n, K, mode, method="base");
```

```
[1] -0.5276713
```

Graph of the function:

```
> plot(wam.specificities(N, n, K, allk, method="base"),
+      type="l", xlab="k", ylab="specificities",
+      main="Graph of the specificities function",
+      sub="(N=61449, n=7896, K=296)")
> points(k, wam.specificities(N, n, K, k, method="base"), pch="+")
> points(mode, wam.specificities(N, n, K, mode, method="base"), pch="x")
```

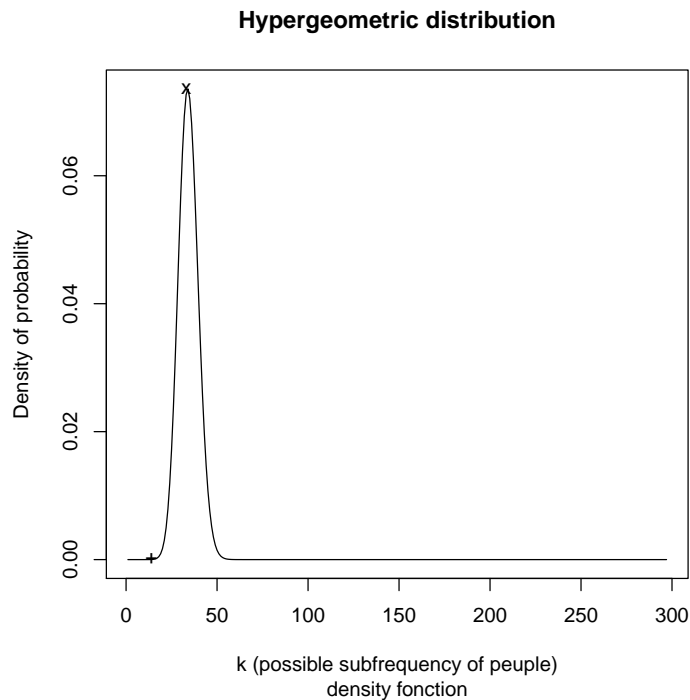


2.3.1 Analysis of the specificities indicator : Standard indicator (method="base")

The presentation below follows (Lafon, 1980).

The hypergeometric distribution give the probability associated with a drawing without replacement. For all the possible subfrequencies of *peuple* in the fourth discourses we can compute the density of probability in the hypergeometric distribution. The graph contains also the observed frequency as well as the mode.

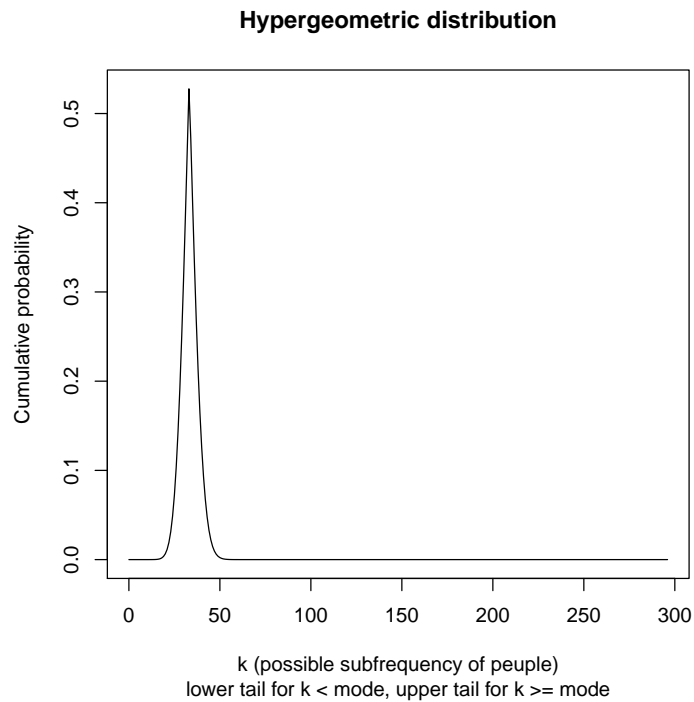
```
> plot(dhyper(allk, K, N-K, n),
+ type="l", xlab="k (possible subfrequency of peuple)", ylab="Density of probability",
+ main="Hypergeometric distribution", sub="density fonction")
> points(k, dhyper(k, K, N-K, n), pch="+")
> points(mode, dhyper(mode, K, N-K, n), pch="x")
```



If the observed frequency is less than the expected frequency, we compute the sum of the probability for a frequency lesser or equal to the observed frequency ($Prob(X \leq k)$) – that is, the cumulative probability.

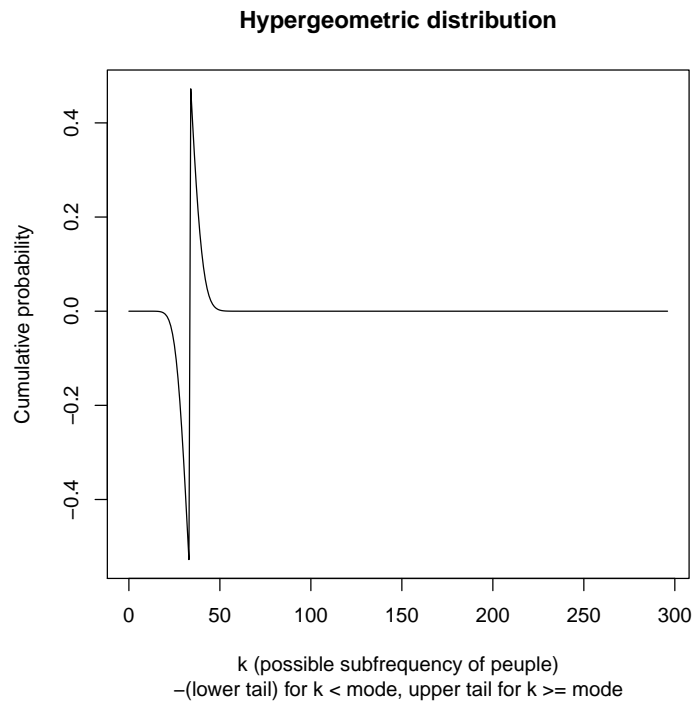
If the observed frequency is greater than the expected frequency, we compute the sum of the probability for a frequency greater to the observed frequency ($Prob(X > k)$) (Lafon 1980 : 141) – that is, the cumulative probability for the upper tail of the distribution.

```
> y <- ifelse(allk <= mode, phyper(allk, K, N-K, n),
+               phyper(allk-1, K, N-K, n, lower.tail=FALSE))
> plot(allk, y,
+       type="l", xlab="k (possible subfrequency of peuple)",
+       ylab="Cumulative probability",
+       main="Hypergeometric distribution",
+       sub="lower tail for k < mode, upper tail for k >= mode")
```

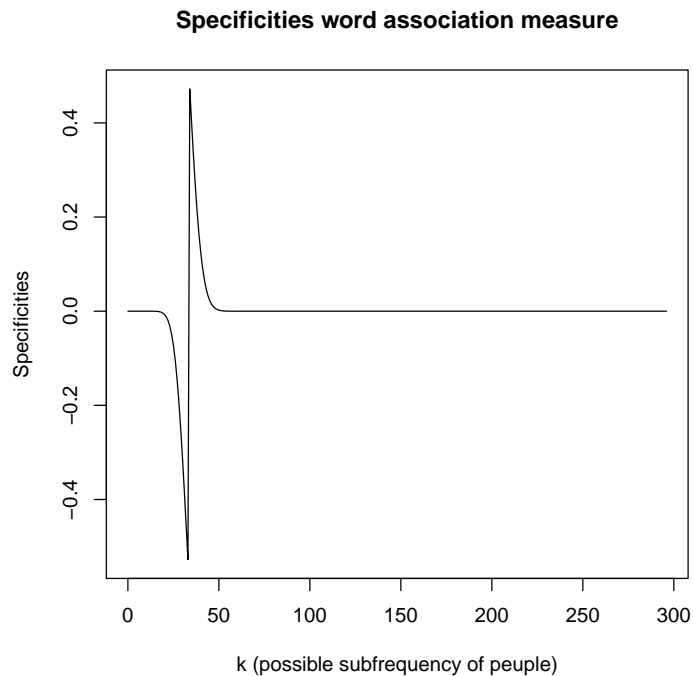
We add a sign: negative if the frequency is lower than expected, positive if it is greater.

```
> y <- ifelse(allk <= mode, phyper(allk, K, N-K, n),
+               phyper(allk-1, K, N-K, n, lower.tail=FALSE))
> y <- ifelse(allk <= mode, -y, y);
> plot(allk, y,
+       type="l", xlab="k (possible subfrequency of peuple)",
+       ylab="Cumulative probability",
+       main="Hypergeometric distribution",
+       sub="-(lower tail) for k < mode, upper tail for k >= mode")
```



It is the standard Specificities function (Lafon 1980), as implemented in the function `wam.specificities` with `method="base"` :

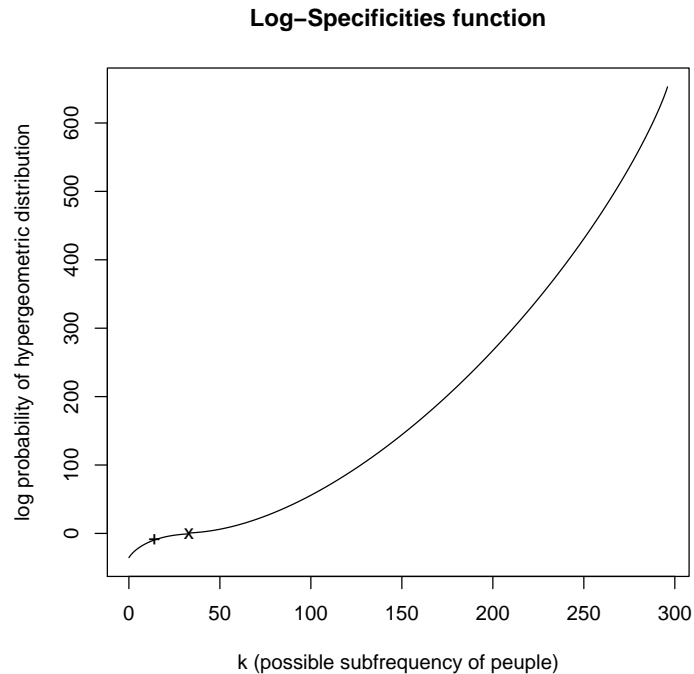
```
> plot(allk, wam.specificities(N, n, K, allk, method="base"),
+       type="l", xlab="k (possible subfrequency of people)",
+       ylab="Specificities",
+       main="Specificities word association measure")
```



2.3.2 Analysis of the specificities indicator : log (method="log")

In order to ease the reading, log are used:

```
> y <- ifelse(allk <= mode, phyper(allk, K, N-K, n, log.p=TRUE),
+           phyper(allk-1, K, N-K, n, lower.tail=FALSE, log.p=TRUE))
> y <- ifelse(allk <= mode, -abs(y), abs(y));
> plot(allk, y,
+       type="l", xlab="k (possible subfrequency of peuple)",
+       ylab="log probability of hypergeometric distribution",
+       main="Log-Specificities function");
> points(k, phyper(k, K, N-K, n, log.p=TRUE), , pch="+")
> points(mode, phyper(mode, K, N-K, n, log.p=TRUE), pch="x")
```



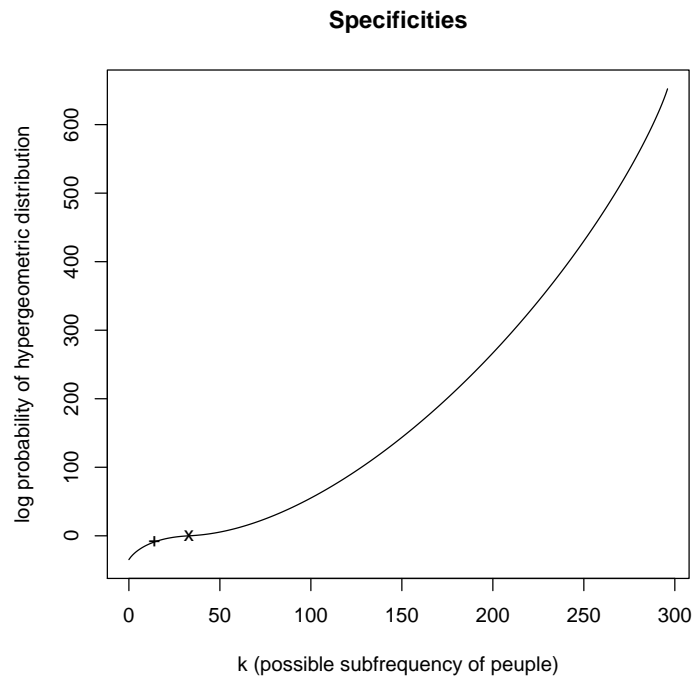
Another issue is that the mode is different from 0:

```
> phyper(mode, K, N-K, n, log.p=TRUE)
```

```
[1] -0.6392818
```

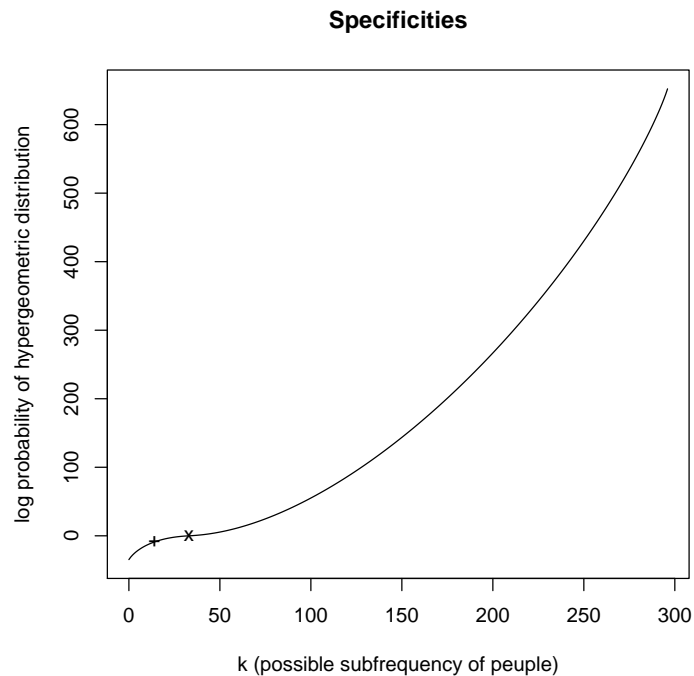
in order to have $mode = 0$, the value of the mode is substracted from all values:

```
> y <- ifelse(allk <= mode, phyper(allk, K, N-K, n, log.p=TRUE),
+           phyper(allk-1, K, N-K, n, lower.tail=FALSE, log.p=TRUE))
> cdm0 <- phyper(mode, K, N-K, n, log.p=TRUE);
> y <- ifelse(allk <= mode, -abs(cdm0-y), abs(cdm0-y));
> plot(allk, y,
+       type="l", xlab="k (possible subfrequency of peuple)",
+       ylab="log probability of hypergeometric distribution",
+       main="Specificities")
> points(k, wam.specificities(N, n, K, k, method="log"), pch="+")
> points(mode, wam.specificities(N, n, K, mode, method="log"), pch="x")
```



That is the `wam.specificities` function with `method="log"`:

```
> plot(allk, wam.specificities(N, n, K, allk, method="log"),
+       type="l", xlab="k (possible subfrequency of people)",
+       ylab="log probability of hypergeometric distribution",
+       main="Specificities")
> points(k, wam.specificities(N, n, K, k, method="log"), pch="+")
> points(mode, wam.specificities(N, n, K, mode, method="log"), pch="x")
```



where the mode is 0 :

```
> wam.specificities(N, n, K, mode, method="log");
```

```
[1] 0
```

2.4 z

```
> wam.z(N, n, K, k);
```

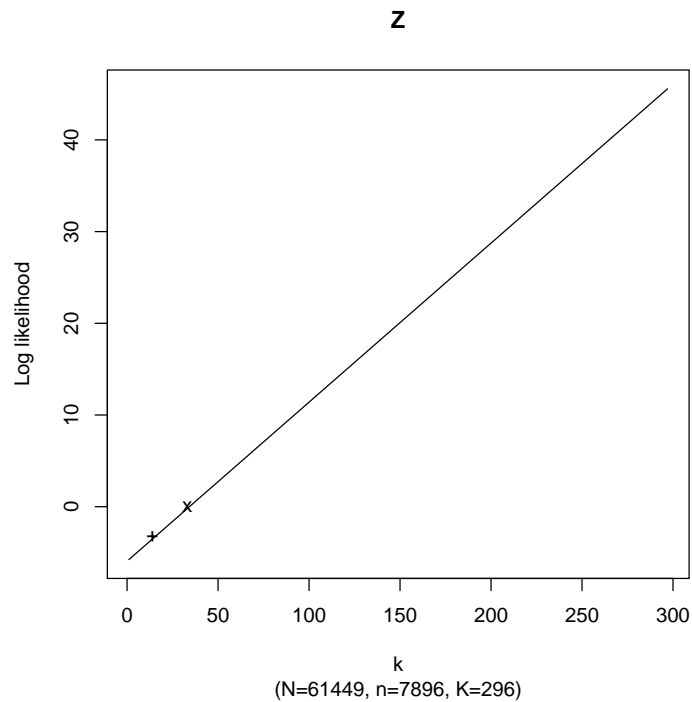
```
[1] -3.338591
```

```
> wam.z(N, n, K, mode);
```

```
[1] -0.04366125
```

Graph of the function :

```
> plot(wam.z(N, n, K, allk),
+      type="l", xlab="k", ylab="Log likelihood",
+      main="Z",
+      sub="(N=61449, n=7896, K=296)")
> points(k, wam.z(N, n, K, k), pch="+")
> points(mode, wam.z(N, n, K, mode), pch="x")
```



2.5 t

```
> wam.t(N, n, K, k);
```

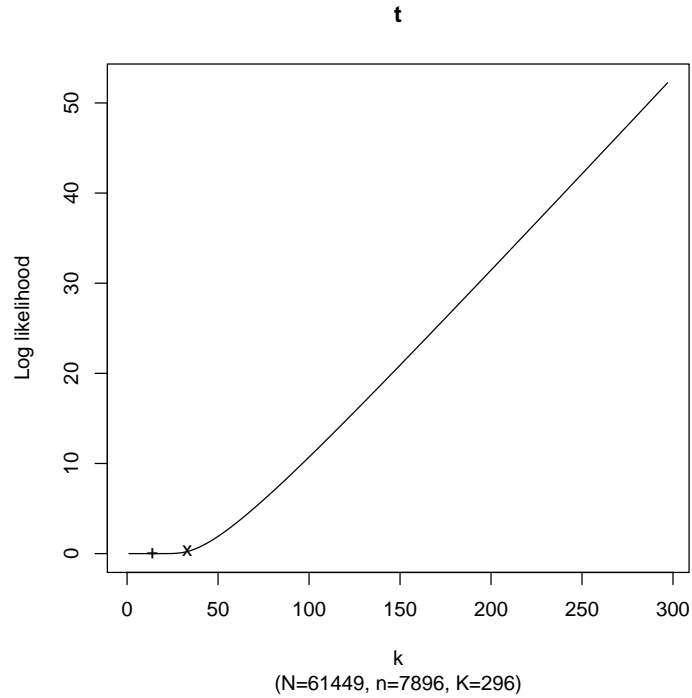
```
[1] 5.801351e-08
```

```
> wam.t(N, n, K, mode);
```

```
[1] 0.2861072
```

Graph of the function :

```
> plot(wam.t(N, n, K, allk),
+       type="l", xlab="k", ylab="Log likelihood",
+       main="t",
+       sub="(N=61449, n=7896, K=296)")
> points(k, wam.t(N, n, K, k), pch="+")
> points(mode, wam.t(N, n, K, mode), pch="x")
```



2.6 chisq

```
> wam.chisq(N, n, K, k);
```

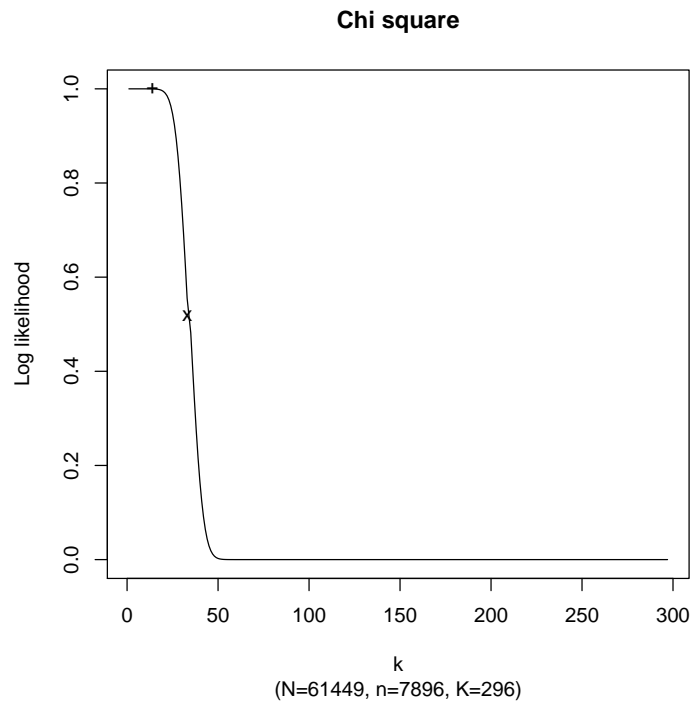
```
[1] 0.9997298
```

```
> wam.chisq(N, n, K, mode);
```

```
[1] 0.5182654
```

Graph of the function :

```
> plot(wam.chisq(N, n, K, allk),
+       type="l", xlab="k", ylab="Log likelihood",
+       main="Chi square",
+       sub="(N=61449, n=7896, K=296)")
> points(k, wam.chisq(N, n, K, k), pch="+")
> points(mode, wam.chisq(N, n, K, mode), pch="x")
```

2.7 fisher

```
> wam.fisher(N, n, K, k);
```

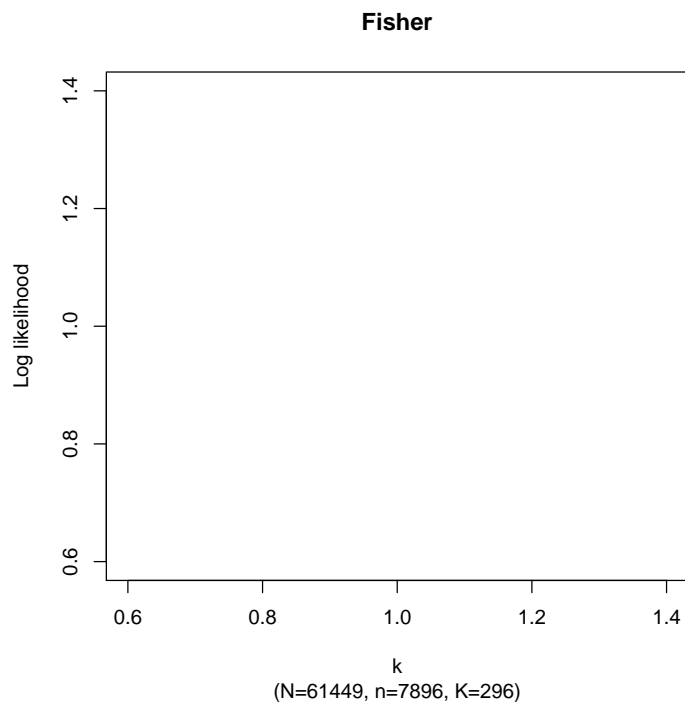
```
[1] 0.0001353407
```

```
> wam.fisher(N, n, K, mode);
```

```
[1] 1
```

Graph of the function :

```
> plot(wam.fisher(N, n, K, allk),
+       type="l", xlab="k", ylab="Log likelihood",
+       main="Fisher",
+       sub="(N=61449, n=7896, K=296)")
> points(k, wam.fisher(N, n, K, k), pch="+")
> points(mode, wam.fisher(N, n, K, mode), pch="x")
```



3 High level interface

The function *wam* provide a more high level interface to the actual functions.

This function allows for computing several indicators at the same time.

It takes also as argument the subcorpus name and lexical types.

```
> rm(list=ls())
> data(robespierre, package="wam")
> attach(robespierre)
> wam.res <- wam(N, n, K, k, measure=c("loglikelihood", "specificities"),
+ types=types, parts=parts)
```

function allows for retrieving the basic information (see the manual page for WordAssociation):

```
> head(k(wam.res));

[1] 464 365 281 227 200 188

> head(association(wam.res));
```

```

      loglikelihood specificities
[1,]      2.5758482      2.1848933
[2,]      0.8133694     -0.9840740
[3,]      0.3408187     -0.5627793
[4,]      0.2069236     -0.3991217
[5,]      1.4940429      1.4588965
[6,]      0.9787803      1.0712966

```

```
> indicator.name(wam.res)
```

```
[1] "loglikelihood" "specificities"
```

The print function allows for an easy to use reading of the results. See :

```

> wam.res <- wam(N, n, K, k, measure=c("loglikelihood", "specificities"),
+ types=types, parts=parts)
> print(wam.res, from=1, to=10, parts="D4");

```

Printing association measure for 1 part(s); from: 1 to: 10

Corpus size: 61449

Sorted by: loglikelihood

```

-----
word                | sub freq | tot freq |   loglikelihood |   specificities
-----
.....
Part name: D4
Part size: 6903 tokens.
Positive specificities printed: 10
Negative specificities printed: 0
bourdon             |      20 |      20 |      87.50 |      43.25
nous                |       3 |     430 |      79.45 |     -40.51
vous               |       6 |     424 |      63.13 |     -32.58
salut              |      33 |      91 |      39.00 |      21.12
comité             |      35 |     103 |      37.31 |      20.20
fabre              |      13 |      19 |      34.59 |      18.43
convention         |      37 |     126 |      30.52 |      16.88
public            |      32 |     108 |      26.84 |      14.98
ils               |      21 |     419 |      20.14 |     -11.35
il                |     105 |     605 |      20.12 |      11.70

```

4 Bibliographie

Chaudhari, D. L., Damani, O. P. & Laxman, S. 2011. "Lexical co-occurrence, statistical significance, and word association". In: Conference on Empirical Methods in Natural Language Processing (Edinburgh, Scotland, UK, July 27-31). pp. 1058-68

<http://www.aclweb.org/anthology-new/D/D11/D11-1098.pdf>

Dunning, T. 1993. "Accurate methods for the statistics of surprise and coincidence." In: Computational Linguistics. 19(1). Pp 61–74.

<http://acl.ldc.upenn.edu/J/J93/J93-1003.pdf>

Hofland, K. and Johanssen, S. 1989. Frequency analysis of English vocabulary and grammar, based on the LOB corpus. Oxford: Clarendon.

Kilgariff, A. 1996. "Which words are particularly characteristic of a text? A survey of statistical approaches." In: Proceedings, ALLC-ACH '96. Bergen, Norway.

http://www.cse.iitb.ac.in/~shwetaghonge/prec_recall.pdf

Lafon P. 1980. "Sur la variabilité de la fréquence des formes dans un corpus". *Mots*, 1, 1980, 127–165.

http://www.persee.fr/web/revues/home/prescript/article/mots_0243-6450_1980_num_1_1_1008.