# A Lexicographer-Friendly Association Score

Pavel Rychlý

Faculty of Informatics, Masaryk University Botanická 68a, 602 00 Brno, Czech Republic pary@fi.muni.cz

**Abstract.** Finding collocation candidates is one of the most important and widely used feature of corpus linguistics tools. There are many statistical association measures used to identify good collocations. Most of these measures define a formula of a association score which indicates amount of statistical association between two words. The score is computed for all possible word pairs and the word pairs with the highest score are presented as collocation candidates. The same scores are used in many other algorithms in corpus linguistics.

The score values are usually meaningless and corpus specific, they cannot be used to compare words (or word pairs) of different corpora. But endusers want an interpretation of such scores and want a score's stability. This paper present a modification of a well known association score which has a reasonable interpretation and other good features.

## 1 Introduction

Finding collocation candidates is one of the most important and widely used feature of corpus linguistics tools [1]. There are many statistical association measures used to identify good collocations. Most of these measures define a formula of a association score which indicates amount of statistical association between two words. The score is computed for all possible word pairs and the word pairs with the highest score are presented as collocation candidates. The same scores are used in many other algorithms in corpus linguistics, for example to compute collocations in grammatical relations and an importance of grammatical relations in the Sketch Engine [2].

There are two general problems of most association scores:

- 1. A score is fine-tuned to one particular corpus size and/or key word frequency. If we use a score for a corpus with very different number of tokens the resulting list is not satisfying enough or is completely wrong.
- 2. The score values are usually meaningless and corpus specific, they cannot be used to compare words (or word pairs) of different corpora. But endusers want an interpretation of such scores and want a score's stability. They want to compare collocation scores of different words and on different corpora or subcorpora.

Petr Sojka, Aleš Horák (Eds.): Proceedings of Recent Advances in Slavonic Natural Language Processing, RASLAN 2008, pp. 6–9, 2008. © Masaryk University, Brno 2008

The article is organized as follows. The following section describe notation and the most widely used association scores. The Section 3 illustrates these two problems on real examples. The next section defines a new score logDice, which is a modification of the well known association score *Dice* [3]. The *logDice* score has a reasonable interpretation, scales well on a different corpus size, is stable on subcorpora, and the values are in reasonable range.

#### 2 Association Scores for Collocations

Almost all association score formulas use frequency characteristics from a contingency table, which records the relationship between two words  $(W_1, W_2)$ . Table 1 shows an example of a contingency table. The numbers in the righthand column and the bottom row are called marginal frequencies and the number in the bottom right-hand corner is the size of the corpus.

In the rest of this paper we will uses the following symbols (the meaning is also summarized in Table 1):

- f<sub>x</sub> = number of occurrences of word X
  f<sub>y</sub> = number of occurrences of word Y
  f<sub>xy</sub> = number of co-occurrences of words X and Y

-  $R_x = \frac{f_{xy}}{f_x}$  = relative frequency of word X -  $R_y = \frac{f_{xy}}{f_y}$  = relative frequency of word Y

**Table 1.** Notation of frequencies of words *X* and *Y* 

	$W_1 = X$	$W_1 \neq X$	
$W_2 = Y$	f <sub>xy</sub>	$f_y - f_{xy}$	$f_y$
$W_2 \neq Y$	$f_x - f_{xy}$	$N - f_{xy}$	$N - f_y$
	$f_x$	$N-f_x$	N

#### Widely Used Association Scores 3

This section summarize formulas of some association scores and gives its main characteristics. More scores, motivations, discussion of their mathematical background and full references could be find in [4].

**T-score**:  $\frac{f_{xy} - \frac{f_y f_B}{N}}{\sqrt{f_{xy}}}$ **MI-score**:  $\log_2 \frac{f_{xy} N}{f_x f_y}$ **MI^3-score**:  $\log_2 \frac{f_{xy}^3 N}{f_x f_y}$ 

**Minimum Sensitivity**: min  $R_x$ ,  $R_y$ 

Dice coefficient:  $D = \frac{2f_{xy}}{f_x + f_y}$ 

**MI log Freq**: MI-score  $\times \log f_{xy}$ , used as salience in the first version of Word Sketches [2].

Table 2 lists the collocation candidates on lemmas to the verb *break* in the window from 5 tokens to the left to 5 tokens to the right. They were computed on the British National Corpus by the Manatee system [5].

	Fxy	T-score			$F_{xy}$	MI-score		F <sub>xy</sub>	MI <sup>3</sup> -sco	ore
the	11781	99.223	spe	ll-wall	5	11.698	the	11781	30.5	591
	8545	83.897	deadlock		84	10.559	dov	vn 2472	29.882	
,	8020	80.169	hoodoo		3	10.430		8545	29.558	
be	6122	69.439	scapulum		3	10.324	,	8020	29.1	93
and	5183	65.918	Yas	Yasa		10.266	be	6122	28.3	311
to	5131	65.798	intervenien		4	10.224	to	5131	28.2	268
a	3404	52.214	preparedness		21	10.183	and	5183	28.2	246
of	3382	49.851	stranglehold		18	10.177	into	1856	27.8	354
down	2472	49.412	log	jam	3	10.131	up	1584	26.9	967
have	2813	48.891	irretrievably		12	10.043	a	3404	26.7	717
in	2807	47.157	An	dernesse	3	10.043	hav	e 2813	26.5	593
it	2215	43.314	irre	parably	4	10.022	of	3382	26.2	255
into	1856	42.469	Thi	ef	37	9.994	in	2807	26.0	)95
he	1811	39.434	TH	IEf	4	9.902	it	2215	25.8	376
up	1584	39.038	non-work		3	9.809	out	1141	25.8	321
	F <sub>xy</sub>	Min. Se	ens.		F <sub>xy</sub>	MI log F	req		F <sub>xy</sub>	Dice
down	2472	0.	027	down	2472	57.	340	down	2472	0.0449
silence	327	0.	018	silence	327	48.	589	silence	327	0.0267
leg	304	0.	016	deadlock	84	46.	909	into	1856	0.0210
law	437	0.	014	barrier	207	46.	389	leg	304	0.0203
heart	259	0.	014	into	1856	46.	197	off	869	0.0201
rule	292	0.	013	off	869	42.	411	barrier	207	0.0191
off	869	0.	013	up	1584	42.	060	law	437	0.0174
news	236	0.	013	leg	304	41.	980	up	1584	0.0158
into	1856	0.012		neck	180	39.	39.336		259	0.0155
barrier	207	0.011		law	437	38.	38.805		180	0.0148
away from	n 202	0.	011	out	1141	38.	783	news	236	0.0144
war	294	0.	010	bone	151	38.	263	rule	292	0.0142
ground	182	0.	010	heart	259	37.	327	out	1141	0.0135
record	287	0.	010	Thief	37	36.	353	away fron	n 202	0.0135
neck	180	0.	010	news	236	36.	296	bone	151	0.0130

Table 2. Collocation lists for different association scores

# 4 logDice

As one can see from the previous section, *Dice* score gives very good results of collocation candidates. The only problem is that the values of the *Dice* score are usually very small numbers. We have defined *logDice* to fix this problem.

$$logDice = 14 + log_2 D = 14 + log_2 \frac{2f_{xy}}{f_x + f_y}$$

Values of the *logDice* have the following features:

- Theoretical maximum is 14, in case when all occurrences of X co-occur with Y and all occurrences of Y co-occur with X. Usually the value is less then 10.
- Value 0 means there is less than 1 co-occurrence of XY per 16,000 X or 16,000 Y. We can say that negative values means there is no statistical significance of XY collocation.
- Comparing two scores, plus 1 point means twice as often collocation, plus 7 points means roughly 100 times frequent collocation.
- The score does not depend on the total size of a corpus. The score combine relative frequencies of *XY* in relation to *X* and *Y*.

All these characteristics are useful orientation points for any field linguist working with collocation candidate lists.

### 5 Conclusion

In this paper, we have presented the new association score *logDice*. The *logDice* score has a reasonable interpretation, scales well on a different corpus size, is stable on subcorpora, and the values are in reasonable range.

**Acknowledgments.** This work has been partly supported by the Academy of Sciences of Czech Republic under the projects 1ET200610406, 1ET100300419 and by the Ministry of Education of CR within the Centre of basic research LC536 and National Research Programme 2C06009.

### References

- 1. Smadja, F.: Retrieving Collocations from Text: Xtract. Computational Linguistics **19**(1) (1994) 143–177.
- Kilgarriff, A., Rychlý, P., Smrž, P., Tugwell, D.: The Sketch Engine. Proceedings of Euralex (2004) 105–116.
- Dice, L.: Measures of the Amount of Ecologic Association Between Species. Ecology 26(3) (1945) 297–302.
- 4. Evert, S.: The Statistics of Word Cooccurrences: Word Pairs and Collocations. Unpublished Ph.D. dissertation, University of Stuttgart (2004).
- Rychlý, P.: Manatee/Bonito A Modular Corpus Manager. In: P. Sojka, A. Horák (Eds.): RASLAN 2007 Proceedings (2007), pp. 65–70.