

ITRI-01-12 **WORD SKETCH: Extraction and
Display of Significant
Collocations for Lexicography**

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July, 2001

Also published in Proc. Collocations workshop, ACL 2001, Toulouse,
France: Pp 32-38.

Supported by the EPSRC under Grant M54971

Information Technology Research Institute Technical Report Series

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WORD SKETCH: Extraction and Display of Significant Collocations for Lexicography

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Abstract

This paper introduces the Word Sketch: a collocation-based resource of proven value for English lexicography. Issues involving the automatic extraction and presentation of salient collocations are discussed. It is further shown how the combination of significant patterns may lead to even greater precision in the identification of collocations.

1 Introduction

Dictionary making involves finding the distinctive patterns of usage of words in texts. This was traditionally carried out by writing examples on cards indexed by the word of interest, with the examples being found by long and extensive reading, and relying on the instincts and intuition of readers.

Since the ground-breaking work of the COBUILD project in the 1980's, state-of-the-art dictionary-making has –for languages where corpora are available– made extensive use of computerised corpora (Sinclair, 1987; Atkins, 1993; Hanks, 1998; Baker et al., 1998). Before writing the entry for a word, the lexicographer looks at substantial numbers of corpus instances, and divides the entry up into distinct senses according to what is found. This is one of the hardest aspects of the lexicographer's task (Kilgarriff, 1998), and one for which they would dearly like more computational help (Clear, 1994). Corpus interface tools with sophisticated querying languages such as XKWIC (Schulze and

Christ, 1994) and WORDSMITH (Scott, 1997) are invaluable but the lexicographer would like more help still.

This paper presents a system for automatically characterising the common patterns of usage of a word, thus minimising the drudgery of reading through corpus data.

2 The Wordsketch Workbench

In this section we outline the system architecture and mode of operation. The workbench is implemented in perl and uses cgi-scripts and a browser for user interaction.

2.1 Grammatical relations database

The central resource is a collection of all grammatical relations holding between words in the corpus. The workbench is currently based on the British National Corpus¹ (BNC): 100 million words of contemporary British English, of a wide range of genres. Using finite-state techniques operating over part-of-speech tags,² we process the whole corpus finding quintuples of the form:

$$\{\text{Rel, Word1, Word2, Prep, Pos}\}$$

where Rel is a relation, Word1 is the lemma³ of the word for which Rel holds, Word2 is the lemma of the other open-class word involved, Prep is the preposition or particle involved and Pos is the position of Word1 in

¹<http://info.ox.ac.uk/bnc>

²In its published form, the BNC is part-of-speech-tagged, by Lancaster's CLAWS tagger. These tags were used. Again, see <http://info.ox.ac.uk/bnc>

³Lemmatisation was performed using morph (Minnen et al., 2000).

relation	example
bare-noun	the angle of bank ¹
possessed	my bank ¹
plural	the banks ¹
passive	was seen ¹
reflexive	see ¹ herself
ing-comp	love ¹ eating fish
finite-comp	know ¹ he came
inf-comp	decision ¹ to eat fish
wh-comp	know ¹ why he came
subject	the bank ² refused ¹
object	climb ¹ the bank ²
adj-comp	grow ¹ certain ²
noun-modifier	merchant ² bank ¹
modifier	a big ² bank ¹
and-or	banks ¹ and mounds ²
predicate	banks ¹ are barriers ²
particle	grow ¹ up ^p
Prep+gerund	tired ¹ of ^p eating fish
PP-comp/mod	banks ¹ of ^p the river ²

Table 1: Grammatical Relations

the corpus.⁴ Relations may have null values for Word2 and Prep. The database currently contains approximately 70 million quintuples.

The current inventory of relations is shown in Table 1. These fall into the following classes:

- Nine *unary* relations (ie. with Word2 and Prep null). Three of these are exclusively for nouns (bare-noun, possessed and plural), two for verbs (passive and reflexive), while the remaining four complementation patterns are available for any word class. Unary relations may be seen to be of limited use by themselves for lexicography, but they will come into play where patterns are combined, as outlined in section 2.5
- Seven *binary* relations with Prep null. Two of these are exclusively for verbs (object and adjectival complement), one for verbs and adjectives (subject), two for nouns (noun modifier and predi-

⁴We store the corpus in the representation formalism developed at IMS Stuttgart (Schulze and Christ, 1994).

cate), and two for all word classes (modifier and “and-or”). In addition, for six of these binary relations we also explicitly represent the inverse relation, ie. **subject-of** etc, found by taking Word2 as the head word instead of Word1. The conjunction relation **and-or** is considered symmetrical so does not give rise to a separate inverse relation.

- Two *binary* relations with Word2 null. The preposition here is either a particle or introduces a gerundive phrase, and the relations may apply to any word class.
- One *trinary* relation, prepositional complement or modifier, which applies to all word classes. Taking Word2 as primary again, the inverse relation is also explicitly represented and may be glossed as “Word1 is head of the complement of a PP modifying Word2”. The inverse relation is only applicable to nouns.

The number of relations, including inverse relations, is twenty-six.

It is also the case that the same instance may have more than one relation of the same kind, as in “banks, mounds and ditches” where *bank* has two **and-or** relations, one with *mound* and one with *ditch*, or “he saw the bank she had climbed” where *bank* has an **object-of** relation to both *see* and *climb*.

These relations provide a flexible resource which is used as the basis of the computations for the Word Sketch. It is similar to the database of triples used in (Lin, 1998) for thesaurus generation. Keeping the position numbers of examples allows us to find associations between relations, as outlined in section 2.5, and to display the actual context of use in the corpus.

The relations contain a substantial number of errors, originating from POS-tagging errors in the BNC, limitations of the pattern-matching grammar or attachment ambiguities. Indeed no attempt is made to resolve the latter: “see the man with a telescope”

will give rise to both $\{PP, see, telescope, with\}$ and $\{PP, man, telescope, with\}$. However, as the system finds high-salience patterns, given enough data, the noise does not present great problems for the task in hand.

2.2 Word Sketch Display

When a lexicographer embarks on composing the lexical entry for a word, they enter the word (and word class) at a prompt. At present, word classes covered are noun, verb and adjective. Using the grammatical relations database, the system then composes a **Word Sketch** for the word. This is a page of data such as Table 2, which shows, for the word in question (Word1), ordered lists of high-salience grammatical relations, relation-Word2 pairs, and relation-Word2-Prep triples for the word. These are listed for each relation in order of salience, with the count of corpus instances. Clicking on the number of instances column retrieves the actual corpus examples illustrating this pattern in a separate concordance screen. Producing a word sketch for a medium-high frequency word currently takes around ten seconds.⁵

2.3 Calculating Salience

Salience is estimated as the product of Mutual Information I (Church and Hanks, 1989) and log frequency. I for a word $W1$ in a grammatical relation R ⁶ is calculated as

$$I(W1; R) = \log\left(\frac{\|*,*,*\| \times \|W1,R,*\|}{\|W1,*,*\| \times \|*,R,*\|}\right)$$

The notation here is adopted from (Lin, 1998) (who also spells out the derivation from the definition of I). $\|W1, R, W2\|$ denotes the frequency count of the triple $\{W1, R, W2\}$ ⁷ in the grammatical relations database. Where

⁵A set of pre-compiled word sketches can be seen at <http://www.itri.bton.ac.uk/~Adam.Kilgarriff/WORDSKETCHES/>

⁶{Grammatical-relation, preposition} pairs are currently treated as atomic relations for purposes of calculating MI.

⁷Or, strictly, of the quintuple $\{W1, R - part - 1, W2, R - part - 2, ANY\}$.

$W1, R$ or $W2$ is the wild card (*), the frequency is of all the dependency triples that match the remainder of the pattern.

Again following Lin, we calculate I for triples relative to the frequency of R :

$$I(W1, R, W2) = \log\left(\frac{\|*,R,*\| \times \|W1,R,W2\|}{\|W1,R,*\| \times \|*,R,W2\|}\right)$$

The word sketches are presented to the user as a list of relations, with items in each list ordered according to salience. Thus it is not problematic that all calculations of I for triples are relative to $\|*,R,*\|$, the overall frequency of the relation. Arguably, $I(W1, R, W2)$ should not be defined to be relative in this way.

Our experience of working lexicographers' use of collocate lists sorted by values of the Mutual Information or log-likelihood statistic shows that, for lexicographic purposes, this over-emphasises low frequency items. This is also the experience of lexicography projects at CUP, Collins, Longman and elsewhere. Multiplying by log frequency is an appropriate adjustment bringing words that are of greatest lexicographic relevance to the head of the collocate list.

2.4 Using Word Sketches

Table 2 shows a Word Sketch for the noun *bank*. It is slightly abbreviated due to the constraints of space, but is otherwise not modified or edited in any way. The total number of patterns shown for the word is set by the user according to needs, but will typically be over 200.

Table 2 reveals how the different word senses for the word can be brought out as they tend to occur with particular significant patterns. For example as object of *burst* we have the RIVER BANK sense of the word, while the object of *rob* is the FINANCIAL INSTITUTION sense. Fixed idioms, such as *bank holiday*, are also revealed. While these are obvious senses, the Word Sketch also reveals less obvious ones, such as those in the collocations *bottle bank*, *bank of cloud*, *memory bank*

bank (noun): BNC frequency=20968

subject-of	num	sal	object-of	num	sal	modifier	num	sal
lend	95	21.2	burst	27	16.4	central	755	25.5
issue	60	11.8	rob	31	15.3	Swiss	87	18.7
charge	29	9.5	overflow	7	10.2	commercial	231	18.6
operate	45	8.9	line	13	8.4	grassy	42	18.5
step	15	7.7	privatize	6	7.9	royal	336	18.2
deposit	10	7.6	defraud	5	6.6	far	93	15.6
borrow	12	7.6	climb	12	5.9	steep	50	14.4
eavesdrop	4	7.5	break	32	5.5	issuing	23	14.0
finance	13	7.2	oblige	7	5.2	confirming	13	13.8
underwrite	6	7.2	sue	6	4.7	correspondent	15	11.9
account	19	7.1	instruct	6	4.5	state-owned	18	11.1
wish	26	7.1	owe	9	4.3	eligible	16	11.1

inv-PP	num	sal	modifies	num	sal	noun-mod	num	sal
governor of	108	26.2	holiday	404	32.6	merchant	213	29.4
balance at	25	20.2	account	503	32.0	clearing	127	27.0
borrow from	42	19.1	loan	108	27.5	river	217	25.4
account with	30	18.4	lending	68	26.1	creditor	52	22.8
account at	26	18.1	deposit	147	25.8	Tony	57	21.4
customer of	18	14.9	manager	319	22.2	AIB	23	20.9
bank to	13	13.2	Holidays	32	21.6	savings	61	19.8
debt to	18	13.1	clerk	73	21.4	Whinney	17	19.7
deposit at	9	12.3	balance	93	21.3	piggy	21	18.5
pay into	14	12.0	overdraft	23	20.3	bottle	34	17.4
branch of	34	11.2	robber	28	19.9	investment	121	17.0
loan by	6	10.7	robbery	33	19.4	August	39	16.8
situate on	14	10.6	governor	41	17.0	canal	36	16.0
subsidiary of	12	9.9	debt	35	15.3	memory	57	16.0
tree on	11	9.8	borrowing	21	15.2	Jeff	14	15.9
syndicate of	6	9.8	note	65	15.2	south	58	14.8
cash from	9	9.7	credit	51	15.0	correspondent	13	14.5
owe to	12	9.6	vault	19	13.9	shingle	16	14.4

and-or	num	sal	PP of	num	sal	PP for	num	sal
society	287	24.6	England	988	37.5	settlement	19	12.8
bank	107	17.7	Scotland	242	26.9	reconstruction	10	11.1
institution	82	16.0	river	111	22.1			
Bank	35	14.4	Thames	41	20.1	predicate	num	sal
Lloyds	11	14.1	credit	58	17.7	bank	5	7.5
bundesbank	10	13.6	Severn	15	16.8	institution	4	5.6
company	108	13.6	Japan	38	16.8			
currency	26	13.5	Ireland	56	16.0	predicate-of	num	sal
issuing	7	13.0	Crete	14	15.3	bank	5	6.0
Barclays	9	12.7	stream	25	14.8	country	6	4.3
ditch	14	12.2	Nile	14	13.7			
broker	15	11.3	Montreal	11	13.4	plural	6760	2.3
lender	13	11.0	cloud	22	12.7	bare noun	442	-9.0
stockbroker	10	10.7	River	12	12.3	possessed	639	-5.5

Table 2: Word sketch for *bank* (n)

subject	num	sal	object	num	sal	modifier	num	sal	particle	num	sal
price	316	22.8	victim	147	22.2	apart	335	29.7	over	638	16.9
wicket	62	21.7	prey	51	18.2	short	247	28.7	off	738	16.8
rate	247	21.5	short	23	17.7	ill	91	21.1	back	616	13.9
rain	155	21.4	foul	34	14.9	sharply	104	18.5	down	611	13.0
net	42	21.1	flat	29	12.5	behind	78	17.2	by	98	12.5
profit	136	20.8	angel	15	11.2	headlong	22	16.7	through	127	12.4
snow	82	20.8	sick	18	9.2	dramatically	56	14.9	away	166	9.8
dusk	39	20.6				steadily	61	14.9	in	309	9.2
PP			PP (cont.)			adj-comp			and-or		
in love	867	44.0	to floor	106	23.2	asleep	604	26.2	rise	92	21.8
into category	259	33.1	into step	39	23.0	foul	98	30.0	slip	22	14.2
into trap	142	31.3	to knee	69	22.4	silent	223	28.8	stumble	16	14.1
into disuse	69	28.0	into sleep	50	22.0	short	142	26.6	trip	11	13.1
into hand	143	26.8	into place	88	21.8	due	79	25.4	fall	34	12.9
by wayside	45	24.5	under spell	31	21.6	ill	109	22.3	stand	35	11.7
on ear	47	23.9	into disrepair	26	21.6	vacant	34	18.7	break	17	9.7
out-of favour	36	23.2	from grace	26	21.3	open	44	12.4	hit	10	9.2

Table 3: Extract of word sketch for *fall* (v), BNC frequency=23,836

etc. This should then be enough to serve as the basis for drawing up the lexical entry for the dictionary.

The number of examples column in the wordsketch contains a hyperlink to a collocation window. Clicking on the link brings up the actual examples from the BNC which contain the pattern in question, thus allowing the original corpus data to be examined. At the same time, for lexicographic purposes, suitable illustrative examples of actual usage may be pasted in.

2.5 Combining Patterns

Consider the reduced Word Sketch for the verb *fall* given in Table 3.⁸ A salient PP-pattern such as **into hand** may not be immediately recognisable as it is just composed of the preposition and the head of its complement noun phrase. A look at the corpus examples reveals that these are practically all of the form “into the hands of...” or “into

⁸This sketch also illustrates some of the problems introduced by incorrect tagging in the original corpus: the collocation “fall short” appears in the patterns verb + object and verb + adverbial modifier, as well as the correct verb + adjectival complement. Indeed all the verb + object patterns do not involve genuine objects, but are nevertheless useful to the lexicographer as being significant collocations.

someone’s hands”. Using the data we already have available we are in a position to calculate more fine-grained patterns revealing this by checking the other grammatical relations that hold for either Word1 or Word2 in the relation. Such a check will reveal that for Word2 in this pattern, other relations that hold in an overwhelming number of cases are **plural** and *possessed*. The pattern may be better presented then as **into sb’s hands**.

Similarly for **by wayside**, Word2 will be exclusively definite and singular⁹, allowing the pattern to be presented as **by the wayside**. Again a particular idiom of **into the trap of V-ing** may be identified by similar means.

It should be noted that the extra calculation involved in this refinement of collocational patterns is small, since it is confined to that small number of patterns which are already found to be of high salience. The fact that patterns in the database are explicitly marked with an instance number for Word1 marking its position in the corpus makes it possible to quickly retrieve the relevant Word2’s and ascertain if these are involved in any other characteristic relations.

⁹At present, these do not belong to the set of unary relations, but will shortly be added.

subject	num	sal	modifies	num	sal	modifies (cont.)	num	sal
sun	34	26.1	water	976	31.0	drink	105	18.8
soup	8	11.2	bun	51	23.4	chocolate	60	18.8
weather	21	10.8	summer	196	23.1	sun	86	18.7
summer	10	10.2	cylinder	76	22.4	pursuit	61	18.1
iron	8	9.8	bath	97	21.1	tea	73	17.7
day	24	9.8	air	242	19.9	spot	102	16.8
water	18	8.8	balloon	52	19.3	spring	72	16.8
afternoon	6	7.5	weather	140	19.1	grill	24	16.3
it	552	7.3	flush	41	19.0	tap	37	16.0
PP			adj-comp-of			and-or		
on heel	42	24.0	serve	64	22.9	cold	257	23.9
under collar	21	20.5	pipe	14	17.4	humid	33	20.1
off press	12	16.7	blow	15	13.6	dry	114	19.5
on trail	9	14.0	scald	7	13.3	sweaty	24	16.7
with embarrassment	7	10.2	get	162	11.4	red	159	16.3
with rice	4	9.4	burn	11	10.6	sunny	37	15.8
with sauce	4	8.7	follow	8	7.4	boiling	22	15.8
against her	4	7.3	grow	29	7.2	sticky	29	15.6
for comfort	5	7.1	scorch	2	5.3	soapy	13	14.5

Table 4: Extract of word sketch for *hot* (adj), BNC frequency=9086

The examples above involved combining a relation between Word1 and Word2, with characteristic unary relations on Word2. Another possibility would be cases where we could combine unary relations on Word1. Extending the principle further we could look for all significant patterns for Word1 or Word2, possibly introducing a new lexeme. Consider the reduced Word Sketch for the adjective *hot* in Table 4. The pattern **modifies bun** is at first rather mysterious. Why should “hot bun” be such a strong collocational pattern? A glance at the examples reveals that it is of course that peculiar Easter delicacy the “hot cross bun” that creates this strong pattern. This can be automatically found by looking for characteristic patterns for the Word2 *bun* when occurring in this collocation, revealing that they nearly all will also be modified by *cross*, allowing the collocation to be correctly identified and presented as **hot cross bun**.

Similarly, if **hot cake** is a salient collocation, which it is although outside the range shown in the extract, then we should also be able to find “sell like hot cakes” by this method, merely by the fact that *cake* in this pattern, as well as being overwhelmingly

plural, will also feature in the pattern **PP-inv sell like**.

This section has shown how combining patterns allows us to both refine the collocations found, without committing us to computationally expensive searches of all combinations of patterns in the corpus.

2.6 Future Developments

As noted above, we are envisaging modest extensions to the repertoire of grammatical relations, including unary relations, in order to increase the expressivity particularly when combining patterns.

We shall be adding automatically-induced thesaural categories (Lin, 1998) to the workbench, which will allow the compaction of patterns by generalising over the identity of Word2. As an illustration this will allow us to generalise the patterns **Bank of England**, **Bank of Scotland**, **Bank of Japan** etc. in the Word Sketch of *bank* to **bank of COUNTRY**.

We are also currently investigating the potential for using web data, with pages being downloaded and fed directly into the workbench. This strategy would extend the potential of the workbench beyond languages where large corpora are readily available.

3 Lexicographic evaluation

For the last two years, a set of 6000 word sketches has been used in a large dictionary project, with a team of thirty professional lexicographers using them every day, for every medium-to-high frequency noun, verb and adjective of English. The feedback we have received is that they are hugely useful, and transform the way the lexicographer uses the corpus. They radically reduce the amount of time the lexicographers need to spend reading individual instances, and give the dictionary improved claims to completeness, as common patterns are far less likely to be missed. They provide lexicographers with plenty of examples to choose from, for editing and putting in the dictionary. This is all particularly popular with the project management.

4 Conclusion

This paper has presented a tried and tested application of the automatic extraction of significant collocations that has proved of great value in the field of lexicography.

We addressed ways in which collocations may be refined by considering combinations of grammatical relations, and this seems to be a worthwhile avenue for future investigation.

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