## Sketching words

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#### Abstract

This paper introduces the Word Sketch: a summary of a word's grammatical and collocational behaviour produced automatically, from a large corpus, for a lexicographer. ${ }^{\text {' }}$ The Word Sketch improves on standard collocation lists by using a grammar and a parser to find collocates in specific grammatical relations, and then producing one list of subjects, another of objects, etc, rather than a single grammatically blind list. The Word Sketches have been used in a a large dictionary project and have received positive reviews.


## 1. Four Ages of Corpus Lexicography

The first age of corpus lexicography was pre-computer. Samuel Johnson collected citations; the Oxford English Dictionary had a corpus of five million index cards, each with a citation on. Sue Atkins was not involved in the inauguration of that era.

The second age dawned with the COBUILD project, circa 1980, brainchild of Sue and her brother, John Sinclair. Computers could be used to store text, and to produce concordances. Lexicographers would thereby be able to view the evidence of how a word was used without the arbitrary filter of who thought what was an interesting example of a word.

As all readers are doubtless aware, the use of computerised corpora has transformed lexicography. Any forward-looking dictionary project uses one. It either negotiates to use a corpus that already exists or creates one afresh. (If it creates one afresh, the model, or rather object of desire that the dictionary project would replicate for their own language if only they had the resources, is usually the British National Corpus ${ }^{2}$ (BNC), another Atkins baby).

Following Sue's move to Oxford University Press in 1989, OUP too, were addressing the challenge of using a corpus. In her search for ideas that would help her make better use of the corpus to make better dictionaries, Sue sought collaborators and made friends in the computational world, at IBM, AT\&T and Digital. The liaison with Digital developed into the HECTOR project, in which

[^0]Sue (with Patrick Hanks) organised an astonishing range of resources for their lexicographers. (Sue's demands for computational support have never been modest. As she puts it "after three weeks of trying to explain to me it's impossible, the programmers realise the only way they'll get any peace is to just write the system so then they get on with it.") With a 20 million word corpus from the American Publishing House for the Blind, parsed using Don Hindle's Fidditch parser (itself a major innovation in computational linguistics, arguably the first wide-coverage parser), a corpus access system that in due course turned into the Altavista search engine, and a setup of no less than three co-ordinated computer screens for lexicographers to view, HECTOR was a visionary project. It did not immediately result in a dictionary or many publications, but it did chart the way ahead for corpus lexicography in general and developments such as Word Sketches in particular.

### 1.1. Statistical summaries

Where there are fifty instances for a word, the lexicographer can read them all. Where there are five hundred, they could, but the project timetable will rapidly start to slip. Where there are five thousand, it is definitely no longer feasible. Corpus query languages with sophisticated querying such as Xkwic [Schulze and Christ1994] help, but there is still too much data to view.
The third age of corpus lexicography was a response to this problem. The data needed summarising.

The answer, arising out of the collaboration with AT\&T, was a statistical summary. The task is to look at the other words in the neighbourhood of the word of interest, its 'collocates', and to identify those that occur with interestingly high frequency in that neighbourhood. The statistic can be used to sort the collocates, and if the statistic (and the corpus) are good ones, the collocates that the lexicographer should consider mentioning percolate to the top.

Ken Church and Patrick Hanks proposed two statistics, pointwise Mutual Information and the $t$-score (which can be used both for identifying collocates, and for identifying how the collocates of two words of similar meaning differ). The paper describing the work [Church and Hanks1989] inaugurated a subfield of computational linguistics, "collocation statistics", and contributed to the decisive arrival of corpora in the field of computational linguistics.

Since Church and Hanks's proposals a series of papers have proposed alternative statistics [Dunning 1993, Pedersen 1996] (see [Kilgarriff1996] for a critical review), and evaluated different statistics [Evert and Krenn2001].

Now, any dictionary projects with access to a corpus provides statistical summaries to lexicographers. These contain many nuggets of information, but
are not used as widely as they might be. From a lexicographical perspective, they have three failings. First, the statistics. They have not been ideal, with too many low frequency words occurring at the tops of the lists. Second, noise. Alongside the lexicographically interesting collocates are assorted uninteresting ones: words that do happen to occur in the neighbourhood of the nodeword, but do not stand in a linguistically interesting relation to it. Third, the neighbourhood. When searching for some types of collocates such as subjects for verbs in English, we wish to look for collocates preceding the nodeword, but it is not clear whether we should look at a window of one, three or five words prior to the nodeword, and possibly we should look at all of these, and in any case we are likely to find assorted adverbs, subjects of passives and other items mixed in with the subjects. It would be far more satisfactory to explicitly produce a collocate list for subjects, another for objects, and so forth (which would also eliminate most noise), as has been proposed by [Hindle 1990] and [Tapanainen and Järvinen 1998]. The Word Sketches are a large scale implementation of such improved collocate-lists for practical lexicography.

## 2. The Word Sketch Workbench

In this section we describe how the Word Sketches are produced, and how thé lexicographers interacts with the system that builds them.
The workbench is implemented in perl. It uses cgi-scripts and a browser for user interaction, so is designed for client-server use, where the client may be local or remote and needs no software loaded onto it other than Netscape, Internet Explorer, or some other web browser. ${ }^{3}$

### 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 Corpus (BNC): 100 million words of contemporary British English, of a wide range of genres. In its published form, the BNC is part-of-speech-tagged, by Lancaster's CLAWS tagger. These tags were used. The BNC was lemmatised, by the morph program [Minnen et al.2000]. Using a shallow parser implemented as a regular-expression matcher over part-of-speech tags, we processed the whole corpus to find quintuples of the form:
\{Rel, Word1, Word2, Prep, Pos \}
where Rel is a relation, Wordl is the lemma of the word for which Rel holds, Word2 is the lemma of the other open-class word involved, Prep is the

[^1]preposition or particle involved and Pos is the position of Wordl in 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 predicate), and two for all word classes (modifier and "and-or"). In addition, for six of these binary relations we also explicittltly 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 "Wordl 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 [Lin1998] 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.

[^2]| 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' 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 ${ }^{\text {l }}{ }^{\text {d }}{ }^{\text {b }}$ |
| Prep+gerund | tired ${ }^{1}$ of' eating fish |
| PP-comp/mod | banks ${ }^{1}$ of ${ }^{\varphi}$ the river ${ }^{2}$ |

Table 1: Grammatical Relations
The relations contain a substantial number of errors, originating from POStagging 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-Word2Prep triples for the word. These are listed for each relation in order of salience, with the count of corpus instances. The actual corpus examples illustrating each pattern are available by mouse-click. Producing a Word Sketch for a mediumhigh frequency word currently takes around ten seconds

### 2.3. Calculating Salience

Salience is estimated as the product of Mutual Information I [Church and Hanks1989] and $\log$ frequency. I for a word $W \mathbf{1}$ in a grammatical relation $\mathbf{R}^{5}$ is calculated as

$$
I(W 1 ; R)=\log \left(\| \frac{*, *, *\| \|\|W 1, R, *\|}{W 1, *, *\|k\|, R, * \|}\right)
$$

The notation here is adopted from [Lin 1998] (who also spells out the derivation from the definition of $I) .\|W 1, R, W 2\|$ denotes the frequency count of the triple $\{W 1, R, W 2\}^{6}$ in the grammatical relations database. Where $W 1, R$ or $W 2$ 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 $R$ :

$$
I(W 1, R, W 2)=\log \left(\left\|\frac{*, R, *\| \| \| W 1, R, W 2}{W 1, R, *\| \| \|, R, W 2}\right\|\right)
$$

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 $\left\|^{*}, R, *\right\|$, the overall frequency of the relation. Arguably, $I(W 1, R, W 2)$ 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.

[^3]| subject-of | num | sal |
| :--- | ---: | ---: |
| lend | 95 | 21.2 |
| issue | 60 | 11.8 |
| charge | 29 | 9.5 |
| operate | 45 | 8.9 |
| step | 15 | 7.7 |
| deposit | 10 | 7.6 |
| borrow | 12 | 7.6 |
| eavesdrop | 4 | 7.5 |
| finance | 13 | 7.2 |
| underwrite | 6 | 7.2 |
| account | 19 | 7.1 |
| wish | 26 | 7.1 |


| object-of | num | sal |
| :--- | ---: | ---: |
| burst | 27 | 16.4 |
| rob | 31 | 15.3 |
| overflow | 7 | 10.2 |
| line | 13 | 8.4 |
| privatize | 6 | 7.9 |
| defraud | 5 | 6.6 |
| climb | 12 | 5.9 |
| break | 32 | 5.5 |
| oblige | 7 | 5.2 |
| sue | 6 | 4.7 |
| instruct | 6 | 4.5 |
| owe | 9 | 4.3 |


| modifier | num | sal |
| :--- | ---: | ---: |
| central | 755 | 25.5 |
| Swiss | 87 | 18.7 |
| commercial | 231 | 18.6 |
| grassy | 42 | 18.5 |
| royal | 336 | 18.2 |
| far | 93 | 15.6 |
| steep | 50 | 14.4 |
| issuing | 23 | 14.0 |
| confirming | 13 | 13.8 |
| correspondent | 15 | 11.9 |
| state-owned | 18 | 11.1 |
| eligible | 16 | 11.1 |


| inv-PP | num | sal |
| :--- | ---: | ---: |
| governor of | 108 | 26.2 |
| balance at | 25 | 20.2 |
| borrow from | 42 | 19.1 |
| account with | 30 | 18.4 |
| account at | 26 | 18.1 |
| customer of | 18 | 14.9 |
| bank to | 13 | 13.2 |
| debt to | 18 | 13.1 |
| deposit at | 9 | 12.3 |
| pay into | 14 | 12.0 |
| branch of | 34 | 11.2 |
| loan by | 6 | 10.7 |
| situate on | 14 | 10.6 |
| subsidiary of | 12 | 9.9 |
| tree on | 11 | 9.8 |
| syndicate of | 6 | 9.8 |
| cash from | 9 | 9.7 |
| owe to | 12 | 9.6 |


| modifies | num | sal |
| :--- | ---: | ---: |
| holiday | 404 | 32.6 |
| account | 503 | 32.0 |
| loan | 108 | 27.5 |
| lending | 68 | 26.1 |
| deposit | 147 | 25.8 |
| manager | 319 | 22.2 |
| Holidays | 32 | 21.6 |
| clerk | 73 | 21.4 |
| balance | 93 | 21.3 |
| overdraft | 23 | 20.3 |
| robber | 28 | 19.9 |
| robbery | 33 | 19.4 |
| governor | 41 | 17.0 |
| debt | 35 | 15.3 |
| borrowing | 21 | 15.2 |
| note | 65 | 15.2 |
| credit | 51 | 15.0 |
| vault | 19 | 13.9 |


| noun-mod | num | sal |
| :--- | ---: | ---: |
| merchant | 213 | 29.4 |
| clearing | 127 | 27.0 |
| river | 217 | 25.4 |
| creditor | 52 | 22.8 |
| Tony | 57 | 21.4 |
| AIB | 23 | 20.9 |
| savings | 61 | 19.8 |
| Whinney | 17 | 19.7 |
| piggy | 21 | 18.5 |
| bottle | 34 | 17.4 |
| investment | 121 | 17.0 |
| August | 39 | 16.8 |
| canal | 36 | 16.0 |
| memory | 57 | 16.0 |
| Jeff | 14 | 15.9 |
| south | 58 | 14.8 |
| correspondent | 13 | 14.5 |
| shingle | 16 | 14.4 |


| and-or | num | sal |
| :--- | ---: | ---: |
| society | 287 | 24.6 |
| bank | 107 | 17.7 |
| institution | 82 | 16.0 |
| Bank | 35 | 14.4 |
| Lloyds | 11 | 14.1 |
| bundesbank | 10 | 13.6 |
| company | 108 | 13.6 |
| currency | 26 | 13.5 |
| issuing | 7 | 13.0 |
| Barclays | 9 | 12.7 |
| ditch | 14 | 12.2 |
| broker | 15 | 11.3 |
| lender | 13 | 11.0 |
| stockbroker | 10 | 10.7 |


| PP of | Num | sal |
| :--- | ---: | ---: |
| England | 988 | 37.5 |
| Scotland | 242 | 26.9 |
| river | 111 | 22.1 |
| Thames | 41 | 20.1 |
| credit | 58 | 17.7 |
| Severn | 15 | 16.8 |
| Japan | 38 | 16.8 |
| Ireland | 56 | 16.0 |
| Crete | 14 | 15.3 |
| Stream | 25 | 14.8 |
| Nile | 14 | 13.7 |
| Montreal | 11 | 13.4 |
| cloud | 22 | 12.7 |
| River | 12 | 12.3 |


| PP for | num | sal |
| :--- | ---: | ---: |
| settlement | 19 | 12.8 |
| reconstruction | 10 | 11.1 |
|  |  |  |
| predicate | num | sal |
| bank | 5 | 7.5 |
| institution | 4 | 5.6 |
|  |  |  |
| predicate-of | num | sal |
| bank | 5 | 6.0 |
| country | 6 | 4.3 |
|  |  |  |
| plural | 6760 | 2.3 |
| bare noun | 442 | -9.0 |
| possessed | 639 | -5.5 |

Table 2: Word Sketch for $\operatorname{bank}(\mathrm{n})$, BNC frequency $=20,968$

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 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 Word Sketch 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, examples may be pasted into lexical entries.

### 2.5. Combining Patterns

Consider the reduced Word Sketch for the verb fall given in Table 3. ${ }^{7}$ 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 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 Wordl 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, Word 2 will be exclusively definite and singular ${ }^{8}$, 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.

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 Word 1 marking its position in the corpus

[^4]makes it possible to quickly retrieve the relevant Word2's and ascertain if these are involved in any other characteristic relations.

| subject | num | sal | object | num | sal |
| :--- | ---: | ---: | :--- | ---: | ---: |
| price | 316 | 22.8 | victim | 147 | 22.2 |
| wicket | 62 | 21.7 | prey | 51 | 18.2 |
| rate | 247 | 21.5 | short | 23 | 17.7 |
| rain | 155 | 21.4 | foul | 34 | 14.9 |
| net | 42 | 21.1 | flat | 29 | 12.5 |
| profit | 136 | 20.8 | angel | 15 | 11.2 |
| snow | 82 | 20.8 | sick | 18 | 9.2 |
| dusk | 39 | 20.6 |  |  |  |
| modifier | 335 | num | particle | num | sal |
| apart | 247 | 247 | over | 638 | 16.9 |
| short | 91 | 91 | back | 738 | 16.8 |
| ill | 104 | 104 | down | 616 | 13.9 |
| sharply | 78 | 78 | by | 611 | 13.0 |
| behind | 22 | 22 | through | 98 | 12.5 |
| headlong | 56 | 56 | away | 127 | 12.4 |
| dramatically | 61 | 61 | in | 166 | 9.8 |
| steadily |  |  | adj-comp | 309 | 9.2 |
| PP | 867 | 44.0 | asleep |  |  |
| in love | 259 | 33.1 | foul | 604 | 26.2 |
| into category | 142 | 31.3 | silent | 98 | 30.0 |
| into trap | 69 | 28.0 | short | 223 | 28.8 |
| into disuse | 143 | 26.8 | due | 142 | 26.6 |
| into hand | 45 | 24.5 | ill | 79 | 25.4 |
| by wayside | 47 | 23.9 | vacant | 109 | 22.3 |
| on ear | 36 | 23.2 | open | 34 | 18.7 |
| out-of favour | 106 | 23.2 | and-or | 44 | 12.4 |
| to floor | 39 | 23.0 | rise |  |  |
| into step | 22.4 | slip | 92 | 21.8 |  |
| to knee | 28 | 21.8 | trip | 22 | 14.2 |
| into sleep | 31 | 21.6 | fall | 16 | 14.1 |
| into place | 26 | 21.6 | stand | 11 | 13.1 |
| under spell | 26 | 21.3 | break | 34 | 12.9 |
| into disrepair |  | hit | 35 | 11.7 |  |
| from grace |  |  | 17 | 9.7 |  |
|  |  | 10 | 9.2 |  |  |
|  |  |  |  |  |  |

Table 3: Extract of Word Sketch for fall (v), BNC frequency $=23,836$
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 Wordl. Extending the principle further we could look for all significant patterns for Wordl 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 are nearly all also modified by cross, allowing the collocation to be correctly identified and presented as hot cross bun.

| subject | num | sal | modifies | num | sal | modifies <br> (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 |
| cron | 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
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 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 [Lin1998] 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 to compile the Macmillan Dictionary of English [Rundell2002], a new dictionary for advanced learners, 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 very useful, and change the way the lexicographer uses the corpus. They 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 most popular with the project management.

## 4. Word Sketches and Word Sense Disambiguation

Word Sketches are designed to support lexicographic analysis. Where a word is polysemous, central to the analysis is the division of the word's semantic range into distinct meanings. An intimately related language-technology task is 'word sense disambiguation' (WSD): automatically. working out which of a polysemous word's meanings applies, given a particular instance of use of the word. While WSD has made great progress in the last ten years, mostly through the application of machine-learning techniques and the use of large corpora, it now seems unlikely it can make much further progress unless it looks more closely at lexicography. Almost all large-scale WSD work to date has aimed to disambiguate between the meanings provided in an off-the-shelf resource, either a publisher's dictionary or WordNet. ${ }^{9}$ It is increasingly apparent that the limits of this approach have been reached, as the analyses of polysemy in these resources are not sufficiently precise or explicit, to give the computer leverage to perform

[^5]any better. What is needed is more explicit lexicography, coupled with good analyses.

In WASPS, the project of which Word Sketch development has been one part, we have developed a system in which a lexicographer not only analyses the word's meaning (starting from the Word Sketch), but also records the details of the analysis in a way which allows a WSD algorithm to accurately disambiguate new instances of the word. The system has performed well in WSD evaluation tasks. See [Kilgarriff and Tugwell2001] for fuller details.

## 5. Conclusion

A Word Sketch is an automatically-produced summary of a word's behaviour. They can be built from any large corpus, provided there are part-of-speech taggers, lemmatisers and grammars available for the language of the corpus. In this paper we have described how Word Sketches were built for English using the British National Corpus, how they were integrated into a lexicographer's workstation (which gave hyperlinked access to corpus examples). The Word Sketches have recently been used in a large scale dictionary project and have received favourable reviews.

Word Sketches build on recent developments in lexicography, corpus linguistics and Natural Language Processing to provide an improved way for lexicographers to find out what the corpus has to say about a word. In doing this, they ride a wave set in motion by Sue Atkins. We hope they contribute to better language description, and are thereby worthy followers in the Atkins tradition.

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[^0]:    1 This work was supported by the UK EPSRC under grant M54971 (WASPS).
    ${ }^{2}$ http://info.ox.ac.uk/bnc

[^1]:    ${ }^{3}$ A demo is available at http://wasps.itri.bton.ac.uk

[^2]:    4 We store the corpus in the representation formalism developed at IMS Stuttgart [Schultze and Christ 1994]

[^3]:    ${ }^{5}$ \{Grammatical-relation, preposition\} pairs are currently treated as atomic relations for purposes of caculating MI.
    ${ }^{6}$ The trinary relation, PP-comp/mod, is reduced to a set of binary relations by having a separate relation for each preposition, viz, PP-com/mod:to.

[^4]:    ${ }^{7}$ This 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.
    8 At present, these do not belong to the set of unary relations, but will shortly be added.

[^5]:    ${ }^{9}$ http://www.cogsci.princeton.edu/~wn

