101. Tools to support the design of a macrostructure

1. Defining a dictionary’s macrostructure

A dictionary’s macrostructure, nomenclature or lemma-sign list is, in simple terms, the inventory of all the headwords in that dictionary. Each of those lemma signs (headwords) is a canonical form, representing an entire paradigm of morphologically-related forms. The use of dictionary citation forms to group related forms is both a space-saving device that stems from the times of the paper dictionary (now, up to a point, arguably unnecessary in electronic dictionaries), and a device that, implicitly, shows/teaches the morphology of a particular language (which remains a useful feature). In many dictionaries for Indo-European languages, for instance, the dictionary citation form for the verbs is the infinitive form, and the dictionary user is expected to know that all (regular) verbal inflections need to be looked up under that single canonical form. Conversely, for some word classes the dictionary citation form is the only member of the paradigm, and in many languages this is for instance the case for some of the function words. The process to go from the various underlying forms to the canonical forms is often referred to as lemmatisation (cf. article 100).

A dictionary’s macrostructure, then, is not merely the list of lemma signs, as suggested in the first sentence of this contribution, but also necessarily includes information on the word class and morphology of that lemma sign. (1) to (3) below, therefore, constitute three lemmas in the macrostructure of an English dictionary:

1. record, noun [pl. records] ...
2. record, verb [3p sg records, pres. part. recording, past & past part. recorded] ...
3. record, adjective [no comparative nor superlative] ...

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Each lemma thus consists of three linked components: a lemma sign, a word class, and information on the morphology. The first component is found in all dictionaries, the second and third components are optional – although most dictionaries do at least include the word class, while the morphological information is, when not mentioned, understood.

Observe that different dictionaries will present the macrostructure in varying ways. Most will present (1) to (3) above as homonyms, with numbers in superscript to differentiate between the three lemma signs, possibly with other types of information (such as pronunciation) intersecting the linear sequence shown. More radical options are also found, such as in the COBUILD series of learners’ dictionaries, where a long single article combines (1) to (3), and the word classes are presented in an Extra Column, one word class for each and every sense. Given COBUILD senses are ordered according to frequency of use, the word classes are accordingly distributed, with in this case noun, verb, and adjective senses intermixed. Although the macrostructure is distributed ‘all over’ a dictionary article in COBUILD dictionaries, it should be kept in mind that this is merely a presentational issue. Or, if one prefers, an alternative approach to lemmatisation.

Lemmatisation is thus both language-dependent and dictionary-dependent. Within a certain language, each word class — also known as grammatical class or part of speech (POS) — is lemmatised in its own way. For languages with long lexicographic traditions, well-established lemmatisation approaches have already been devised for the various POSs; for languages that have not yet been reduced to writing or for which dictionaries have not yet been compiled, various systems may be devised that draw on the morphology of those or cognate languages. Yet, as anyone comparing dictionaries will quickly notice, even within the same type of dictionaries for the same language, different lemmatisation approaches are in use. COBUILD-style lemmatisation versus lemmatisation in competing learners’ dictionaries for English are a case in point. Generalising across languages, one may say that dictionaries vary in their degree of lumping versus splitting (on this macrostructural level). Given one deals with a continuum between these two extremes, the variations are endless. For many African languages, for instance, lumping is associated with stem-based dictionaries where the entries are huge and/or modular in design; whereas splitting is associated with word-based dictionaries with far shorter entries (cf. articles 59 to 64). Similarly, looking at lemmatisation from the dictionary angle in for instance reference works for Semitic languages, one sees that the distribution of the information is entirely different depending on whether all inflections of the roots are grouped under the radicals (usually three consonants), or whether (as is the case for some modern, basic dictionaries) inflected forms are placed in alphabetical sequence in the macrostructure. This macrostructural lumping/splitting issue reappears for Chinese and Japanese dictionaries (cf. articles 56 and 57), and basically for any dictionaries for any other language.

Given the design of a dictionary’s macrostructure thus depends on the language as well as type of dictionary dealt with; there are no generic computational tools to assist the lexicographer with this task. Computationally, however, this task can be brought back to a simple lemmatisation issue, for which see article 100, and Section 2 below.

2. On the need for rulers, part 1

Today’s dictionaries are based on corpus data — the better ones always have. Whereas such corpora used to consist of up to millions of citation slips stored in shoebox-like drawers, today’s corpora are electronic collections of text (but also of multimedia, i.e., text, audio, computer graphics, and video — all interlinked). When it comes to macrostructural decisions, corpora are especially useful in that they make it possible to separate the frequent and average from the one-offs; to distinguish the typical from the oddities. Trivially, then, the frequent, average and typical are to be selected for the core of general-purpose dictionaries, while the one-offs and oddities ought to find their way into hard-word lists (only), large native-speaker dictionaries (as opposed to learners’ dictionaries), etc. At face value, drawing up the macrostructure of a particular dictionary therefore seems straightforward, as one may use occurrence frequencies (summed on the level of the canonical forms) as main arbiter for decisions on inclusion versus omission. One may also require that the items that make up a para-
digm are evenly spread across at least a certain number of sub-corpora, where those sub-corpora represent a number of different (transcribed) text types or a number of different genres. Depending on the kind of dictionary being compiled, one may also wish to focus on words that have recently entered the language, or, conversely, only include words with an even distribution across the decades (with a further option to either drop or mark archaic words and/or uses). Furthermore, during actual dictionary compilation, several other factors also play a role, such as (1) the need/wish to complete certain paradigms of interlinked, closed-class or grammatically similar words, (2) pedagogical, cultural or even political restraints/obligations, (3) purely commercial or academic pressures for the addition or removal of this or that word, etc. The log files and feedback forms attached to online dictionaries may furthermore indicate that the users truly want to see certain words treated. Another often-overlooked aspect is the one linked to the fact that dictionaries are at times released without them having been properly completed, in order to meet (unfortunate) deadlines. Consequently, regardless of the actual lemmatisation approach (or even the absence of any such approach), published dictionaries at least need to be representative of a lexicon’s distribution at all times. Implementing this concept leads to a novel set of tools, known as ‘Rulers’.

Indeed, during modern dictionary compilation one is working in all alphabetical categories simultaneously, constantly making changes throughout the (growing) dictionary database. In order to keep such a project in shape, a tool is required that allows for the continuous monitoring of various dictionary aspects. More in particular, and with specific reference to a dictionary’s macrostructure, one needs to be able to easily check whether or not items that belong to certain POSs are not accidentally over- or underrepresented, and one needs a way to monitor the relative distribution of the lemma signs across the alphabetical categories.

Note that reference is made to the distribution of ‘alphabetical categories’ henceforth. In most dictionaries for Semitic languages the corresponding distribution refers to triplets of consonants, for Chinese and Japanese dictionaries covering Kanji characters the corresponding distribution may refer to stroke counts, and so on. Rather than single letters, the ‘alphabet’ may also contain digraphs, tri-graphs, etc. All ‘alphabetical’ claims that follow remain applicable however, mutatis mutandis.

At all times monitoring the POS as well as the alphabetical-category distribution must be potentially combined with frequency thresholds. In other words, given a certain frequency range of the dictionary database, are the POSs well distributed and are the alphabetical categories well distributed?

3. Part-of-Speech rulers (‘POS Rulers’)

In the present section, the attention goes to the so-called ‘POS Ruler’. Intuitively, one may look at two extreme cases first. Imagine one would want to compile a dictionary of the top 100 items only, culled from a lemmatised frequency list. It is well known that such a list will, for most of the world’s languages, contain mainly function words such as (here for English): the, of, and, a, but, or, etc. At the other extreme, imagine one starts with a billion-word corpus and undertakes to lemmatise all forms. Obviously nearly all, if not all, function words — being members of closed classes — will be in the resulting dictionary. As for content words, while going down the frequency list, ‘new nouns’ will typically be personal names, names of products and companies, place names, etc. Increasing such a mega-corpus further might, in addition to new nouns, reveal ‘new verbs’ or ‘new adjectives’ every now and then, but the likelihood of say ‘new conjunctives’ being discovered is about nil. If one thus looks at the distribution of the POSs in a billion-word corpus, compared to the distribution in a corpus twice as large, then one will notice that the percentage allocation to the nouns continues to grow: the larger the corpus the more nouns relatively speaking. All real-world dictionaries fall somewhere in-between those extremes.

These intuitive thought experiments are confirmed when one studies the distribution of the POSs in corpora. See in this regard, for English, Fig. 101.1 and Fig. 101.2. These graphs were constructed using the POS information for the unlemmatised types in the 100-million-word British National Corpus (BNC), reworking data provided by Leech et al. (2001).

Fig. 101.1 shows the distribution of the POSs, at each rank anew, for the most frequent 1,000+ types in the BNC. One may clearly deduce from this graph that function
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**Fig 101.1:** POS (Part-of-Speech) distribution of the top 1,000+ types in the unlemmatised BNC (British National Corpus).

**Fig 101.2:** POS (Part-of-Speech) distribution of the top 7,000+ types in the unlemmatised BNC (British National Corpus).
ranks in a frequency list. The percentage of nouns already stands at 40%, that of the verbs at 20%, while the function words shrank to 16% of the total (whereas these still represented roughly two thirds at the 100 mark).

Fig. 101.2 zooms out, and shows the distribution for each single rank down to all items that occur at least ten times per million tokens in the BNC, of which there are 7,726 in all. The allocation to the nouns at the 7,000+ mark now stands at 52%, that to the verbs grew to 22%, while the function words shrank to a mere 4% of the total. These graphs can be extended down to any rank, while the same type of calculations can of course also be performed on lemmatized frequency lists, with similar results.

From Fig. 101.1 and 101.2 one can further deduce that there are as many POS Rulers as there are ranks in a corpus-derived frequency list and/or as there are lemma signs in a dictionary. A dictionary of 10,000 items will have a certain POS distribution, one of 50,000 items another one, and so on. With a POS Ruler at hand, one may thus give informed answers to claims (often found in reviews of dictionaries) such as: “Your dictionary contains too many nouns; you should have included far more verbs instead!” Perhaps surprisingly, but the reply to such a claim will depend on the size of the dictionary in question. Graphs such as those presented in Fig. 101.1 and Fig. 101.2 also indicate why the production of especially small or pocket dictionaries is non-trivial from a corpus perspective. If counts from a single corpus are used and nothing else, then the percentage of function words in a reference work with just a few thousand lemma signs will be relatively high. Many dictionary compilers find this unacceptable, and prefer to focus on content words like nouns and verbs rather, with the obvious skewing as a result. One example of how to deal with the creation of the macrostructure of small dictionaries has been proposed by de Schryver/Prinsloo (2003). They propose to work with two corpora, a large general-language one, and a smaller customised corpus consisting of the same type of material the intended target user group of the dictionary will encounter. About four-fifths of the dictionary’s lemmas are basically the top section of the lemmatized frequency list derived from the customised corpus, with the remaining fifth having low or even zero frequencies in the customized corpus, but chosen for their top frequencies in the general-language corpus. For large dictionaries on the other hand, and returning to Fig. 101.1 and Fig. 101.2, given POS Ruler changes become asymptotic with increasing dictionary sizes, one may conclude that rather stable POS distributions may indeed be derived for (very) large dictionaries.

4. On the need for rulers, part 2

In a semasiological dictionary (i.e., one in which the lemma signs are presented in alphabetical sequence), the different alphabetical categories are of course not equal in size. This is so obvious and trivial that one ought not to mention it. Even so, in one of the seventy texts included in the first anthology of lexicography (Hartmann 2003), Serianni (II, 198) writes with regard to a certain historical dictionary of modern Italian: “Whereas 917 pages were sufficient to cover the letter A, 3058 were required for P and 3611 for S; adding up the total number of pages, A + P + S (= 7586 pages), the percentages are as follows: A 12.1%, P 40.3%, S 47.6%, with an evident imbalance to the disadvantage of letter A.” This is a disturbing claim. The only way to know whether A is out of balance or not, is to compare the size of A with the allocation predicted by some kind of instrument that reflects the alphabetical distribution of the Italian (historical) lexicon. Claiming or implying that all alphabetical categories should be nicely balanced out compared to one another throughout a dictionary is simply outrageous. Imagine a general-language English dictionary where the letter X would be as large as the letter S – that would be an extraordinary feat indeed. But then, the letters S and X may be of comparable sizes – in a general Tsonga dictionary for example. So, yes, different languages simply have different word-initial letter distributions, but by no means are such distributions ‘even’ across the alphabet. One is therefore in need of an instrument that reflects the alphabetical distribution in dictionaries. That instrument – as it turns out, a second type of ruler – has been termed a ‘Multidimensional Lexicographic Ruler’, or more often just ‘Ruler’ (in contrast to ‘POS Ruler’).

The early beginnings of the design of a Ruler can be traced back to 1999–2000,
when a team at the University of Pretoria
was in search of an informed way to combat
various types of inconsistencies encountered
in the existing South African dictionaries (cf.
"e.g., de Schryver/Prinsloo 2000: 293–297,
2001: 376–380; Prinsloo/de Schryver 2001:
192–195). On a macrostructural micro-level,
for instance, the researchers noticed that the
compilers of a number of bilingual dictionar-
ies missed out on items likely to be looked
for by their target users, presumably simply
because these items did not cross the lexicog-
raphers’ way during compilation. Cross-com-
paring the nomenclature of various dic-
tionaries is of course a ‘classic’ in lexicography.
In the mid-1980s, for example, Crystal con-
trasted sample pages from comparable Eng-
lish dictionaries. Focusing on lemma signs
he observed that “the discrepancy factor (that
is, the number of head words not shared di-
vided by the number of head words shared)
can be as much as 30 per cent” (1986: 75).
Likewise, Herbst (1990: 1380–1381) counted
the number of ‘corresponding entries’ be-
tween fixed points in several English learners’
dictionaries. Focusing on lemma signs
he observed that “the discrepancy factor (that
is, the number of head words not shared di-
vided by the number of head words shared)
can be as much as 30 per cent” (1986: 75).

5. Multidimensional lexicographic
rulers (‘Rulers’)

From the start it was clear that the to-be-
developed ruler should: (1) preferably be
based on distributions derived from corpora
in the current era of corpus lexicography, (2)
preferably take advantage of distributions in
existing dictionaries on the condition that
these had been compiled according to sys-
tematic principles, and (3) be made to reflect
the envisaged dictionary lemmatisation poli-
cies.

In a way, this aim thus addresses one of
the fundamentals in lexicography, namely the
overall distribution of the lemma signs in the
central text of a semasiological dictionary.
Surprisingly, however, just a handful of stud-
ies have been devoted to this issue. In the
mid-twentieth century E. L. Thorndike (man-
ually) devised a ‘block system of distribution
of dictionary entries by initial letters’, which
has been discussed and reprinted by Landau
who read and reviewed the first edition of
Landau’s textbook (Svensén 1992), then
wrote in his own textbook (1993: 242): “A
decision must also be made as to what frac-
tion of the whole dictionary each initial letter
may occupy, so that the size of the finished
dictionary can be kept under control during
the course of the project. The percentages for
each of the various initial letters in a given
entry language are fairly constant, and, if
such calculations have not already been done
by others, it is wise to examine the distribu-
tion in a number of representative dic-
tionaries.”

The only other reference found in the lit-
erature in this regard is the following by
Coutsogeorgopoulos et al. (2000: 128) who
discuss the creation of a lemma-sign list for a
Greek (to English) dictionary: “First, the
number of lemmas for each letter of the alpha-
bet of the Greek language is defined. The
frequency of occurrence of each word ex-
tracted from the corpus as well as reliable bi-
lingual or monolingual Greek dictionaries
are consulted to balance the number of lem-
mas contained under separate letters of the
Greek alphabet.”

The existing literature, then, does not go
beyond providing a block system for (Ameri-
can) English, the advice to examine existing
dictionaries, and the suggestion to also con-
sult corpora. In contrast, in de Schryver’s
design of Multidimensional Lexicographic
Rulers is accompanied by a solid argumenta-
tion. The overarching finding of the under-
taken research is that there is a remarkable
consistency for both dictionary and corpus
distributions.

As far as dictionaries are concerned, al-
phabetical distributions across various exist-
ing dictionaries compare very well. More-
over, distributions based on space-allocation
measurements correlate strongly with distrib-
butions in which exact lemma-sign counts are
used instead. In addition, deriving such dic-
tionary statistics from reference works sev-
eral decades apart once more results in near-
identical distributions.
On the corpus side, an alphabetical breakdown of all the types in a full corpus is equivalent with the alphabetical breakdown of the types in subsections of that corpus, as well as with the alphabetical breakdown of the top-frequent types only. In addition, for the Germanic languages English and Afrikaans, both unlemmatised and lemmatised corpus data once again result in near-identical alphabetical distributions.

Implicit in the last finding is a surprising correlation, one which suggests that one has come full circle. Given that corpus lemmatisation produces canonical forms, and that a list of canonical forms corresponds with a lemma-sign list, which in turn correlates with space allocations in a dictionary, the experiments simply indicate that a direct comparison between corpus types and alphabetical categories in dictionaries is possible. If this is confirmed, then corpus data (full or partial, all or top-frequent only, etc.) and dictionary data (whether based on page-allocation measurements or lemma-sign counts, derived from one or more dictionaries, etc.) should also correlate well. At that point, dictionary data and corpus data may successfully be united (averaged) to produce a reliable and stable Ruler. This hypothesis will now be tested (and confirmed), taking English as an example.

In tables 101.1 and 101.2 the distributions in four different sources are compared. These four sources are (1) exact lemma-sign counts in two American dictionaries (Newbury 1999 and Heritage 2000), (3) the page counts in a British dictionary, namely the Macmillan English Dictionary for Advanced Learners (MEDAL, Rundell 2002), and (4) the letter-initial distribution in a corpus of world Englishes (compiled at the University of Pretoria by R. Gauton). As may be deduced from the matrix in Table 101.2, the four components correlate well, and this although exact lemma-sign counts, precise page measurements and plain corpus break-

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<tr>
<td>lemma signs</td>
<td>%</td>
<td>lemma signs</td>
<td>%</td>
<td>%</td>
</tr>
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<td>5.48</td>
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<td>820</td>
<td>3.45</td>
<td>3,305</td>
<td>3.60</td>
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<tr>
<td>H</td>
<td>993</td>
<td>4.18</td>
<td>3,915</td>
<td>4.26</td>
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<td>I</td>
<td>854</td>
<td>3.59</td>
<td>2,927</td>
<td>3.19</td>
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<td>J</td>
<td>241</td>
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<td>946</td>
<td>1.03</td>
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<td>L</td>
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<td>0.49</td>
<td>454</td>
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<td>1,387</td>
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<td>2,499</td>
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<td>1,352</td>
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<td>0.41</td>
<td>383</td>
<td>0.42</td>
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<td>Z</td>
<td>34</td>
<td>0.14</td>
<td>387</td>
<td>0.42</td>
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</table>

23,778 100.00 91,852 100.00 1,661,352 100.00 118,282 100.00 100.00

Table 101.1: The design of an English general-language Ruler.
1. Characterising POS rulers and multidimensional lexicographic rulers

Thus far two rulers, viz., a POS Ruler and a Multidimensional Lexicographic Ruler have been introduced as valuable tools to support the design and monitoring of a dictionary’s macrostructure. While the former is dynamic in that there are as many POS Rulers as there are ranks in a corpus-derived frequency list and/or as there are lemma signs in a dictionary, the latter is very stable indeed. It was nonetheless shown that POS Rulers do ‘stabilize’, but for (very) large dictionaries only. In contrast, and in its most basic form, a Multidimensional Lexicographic Ruler, or simply a Ruler, is a unique abstract entity since a single series of percentages straightforwardly correlates with the relative alphabetical allocation in semasiological dictionaries. As such, each alphabetical category is assigned a certain percentage, reflecting the relative size of that category. Different languages, and even different types of dictionaries for a specific language (with possibly varying lemmatisation approaches), have different Rulers. However, given a certain lemmatisation policy, a general-language Ruler for a particular language is valid for all general-language dictionaries for that particular language, and is also stable over time. Just as physical rulers with which one measures, Rulers can be made as fine-grained as one wishes by simply breaking down the alphabetical categories into smaller sections. And just as the human rulers who govern us, Rulers can be called in to manage various aspects of a dictionary project.

6. Tools to support the design of a macrostructure

Table 101.2: Pearson correlation coefficients $r$ between the different components of the English general-language Ruler.

<table>
<thead>
<tr>
<th></th>
<th>Newbury</th>
<th>Heritage</th>
<th>MEDAL</th>
<th>Corpus</th>
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<tbody>
<tr>
<td>Newbury</td>
<td>1.000</td>
<td>0.964</td>
<td>0.981</td>
<td>0.966</td>
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<tr>
<td>Heritage</td>
<td>0.964</td>
<td>1.000</td>
<td>0.979</td>
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<td>MEDAL</td>
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<tr>
<td>Corpus</td>
<td>0.966</td>
<td>0.992</td>
<td>0.966</td>
<td>1.000</td>
</tr>
</tbody>
</table>

downs are compared with one another, and this for respectively American, British and World Englishes. It is thus clear that Rulers may indeed be built by averaging dictionary data on the one hand and corpus data on the other — which is also what is done in the penultimate column of Table 101.1.

The list of values in bold seen in Table 101.1 is referred to as a Multidimensional Lexicographic Ruler, in this case for general-language English dictionaries, as this tool operates on multiple levels: obviously the macrostructural one, which blends into the microstructure, but also on the level of the planning and management of dictionary projects. Stemming from the appearance of Rulers as true ‘physical rulers’, their first and natural function could be said to be to measure various dictionary aspects. As such any existing dictionary may be scrutinised, and measurements on either lemma-sign or space-allocation levels, or both, may be performed in order to evaluate the ‘overall balance’ in that particular dictionary. One thus sees that the Ruler as a ‘measurement instrument’ is also an ‘evaluation instrument’. If such an evaluation would reveal certain dictionary stretches to be over- or under-treated on either lemma-sign or space-allocation level, or both, and the observed deviations cannot be explained except as instances of inconsistent treatment on the side of the lexicographer(s), then the Ruler ‘rules’ that the observed imbalances ought to be rectified in revised editions. An example of a detailed analysis of a large multi-volume dictionary project along those lines can be found in de Schryver (2005).

For dictionaries that need to be compiled from scratch, or even for dictionaries that are still being compiled, a Ruler can also be used to predict various dictionary aspects. The Ruler as a ‘measurement/evaluation instrument’ is thus also a ‘prediction instrument’. Continuing this line of thinking, if one is able to predict various aspects, then one may logically also use a Ruler to plan and thus to
Reformulated, a Ruler is a powerful instrument with which measurements/evaluations and predictions/monitoring can be performed on various macro- and micro-structural dictionary levels. These levels are meta-levels since one is not taking any individual semantics (e.g., the number of senses) or individual grammar (e.g., the POSs) of single lemma signs into account; one rather works with averages across a dictionary. It is precisely because averages are used that the composing components (viz., corpus counts and dictionary measurements) crystallised into a single Ruler.

To conclude this section, and summarizing the research presented in de Schryver (2004: 209–224), Rulers may be characterised and built as follows:

(1) A Ruler is an instrument to guide the relative alphabetical distribution in semasiological dictionaries. Ideally, this instrument averages counts derived from a large automatically lemmatised corpus, and the exact measurement of space allocations in existing dictionaries. Moreover, both the corpus and the reference dictionaries should cover the same domain as the intended dictionary for which the alphabetical distribution is required, as general-language and special-purpose reference works behave differently on the macrostructural axis.

(2) The research also showed that these conditions may successfully be weakened or strengthened along the following lines:

(2.1) On the corpus side:
- In the absence of a large automatically lemmatised corpus, and for languages with simple word-initial morphology as well as very regular inflection patterns, the data may also be derived from an unlemmatised corpus (and such a corpus might even be preferred for some types of dictionaries).
- As dictionary-lemmatisation policies vary greatly, a corpus that is neither fully lemmatised nor fully unlemmatised might be required, but rather a corpus with an automated tailored lemmatisation.
- If computational support for an automated lemmatisation is not available, a manually lemmatised corpus can be used (and, though laborious to construct, might even result in the best type of lemmatisation for the task at hand).
- As a shortcut, drawing data from a (top-frequent) section only of a corpus will largely result in a comparable alphabetical breakdown.
(2.2) On the reference-dictionaries side:
- If exact space-allocation measurements cannot be made or are not available, lemma-sign counts are a good approximation (and might even be favoured in some cases).
- For general-language dictionaries data from reference works spanning several decades may be employed instead of solely time-restricted dictionary data.

(2.3) On the corpus and reference-dictionaries side:
- In case either corpora or either comparable dictionaries are not available, the instrument to guide alphabetical distribution may be:
  (i) based solely on one or more comparable dictionaries, or
  (ii) based solely on corpus data.
- Guidance may also be provided for stretches smaller (or larger) than entire alphabetical categories.

(3) The core practical implementations are:
- The devised measurement/evaluation instrument can point out imbalances across the alphabetical categories and/or across any dictionary stretches in existing dictionaries small and large, imbalances that can be addressed in revised editions.
- The devised measurement/evaluation instrument is especially useful for very large (and thus long-term) multi-volume dictionary projects in progress, where imbalances cannot only be addressed in revisions, but where reliable strategies can be set up to effectively steer or (re)direct future compilation so as to prevent major (additional) inconsistencies. At this point the measurement/evaluation instrument thus becomes a prediction instrument.
- The devised measurement/evaluation and prediction instrument enables one to put forward guidelines which are especially useful when used as a framework to assist in the compilation of new dictionaries. These guidelines revolve around:
  (i) space allocation per alphabetical category or any dictionary stretch (e.g., expressed as a certain number of pages in hardcopy dictionaries), and
  (ii) the number of lemma signs per alphabetical category or any dictionary stretch.
- The devised measurement/evaluation and prediction instrument further also suggests an average article length (expressed in number of articles per page, in number of column-lines per article, or in number of words or characters per article).
- The devised measurement/evaluation and prediction instrument is also a management instrument in that it can be used to allocate work and to keep track of the lexicographers’ progress.
7. Integrating rulers with dictionary compilation software

In an ideal lexicography project, compilers have access to both a POS Ruler and a Multi-dimensional Lexicographic Ruler at all times. In the professional dictionary compilation software TshwaneLex, for instance, both have also seamlessly been integrated. On the one hand TshwaneLex allows for the extraction of all lemma signs with a certain POS, within any frequency range, and to have the allocation calculated. These values may then be compared with the POS Ruler values for the same frequency range. On the other hand the Ruler ‘percentages’ may be input, and at any one point TshwaneLex may be asked to visualize the dictionary database, or any section of it, compared to the Ruler. In Fig. 101.3, for instance, the distribution across the alphabet of the number of lemma signs that are marked as ‘completed’ in a monolingual dictionary for Northern Sotho, is compared to the Ruler breakdown.

The graph in Fig. 101.4 compares the Ruler to both the space allocation (whereby characters are counted) and the number of lemma signs per alphabetical category. From these visualisations one may immediately conclude that, at that point in time, E and I were under-treated, while M and S were over-treated. TshwaneLex also allows any number of frequency bands to be marked, through the automatic and dynamic display of the relevant lemma signs in varying colours, or the addition of star ratings, etc.

In conclusion one can therefore say that POS Rulers and Multidimensional Lexicographic Rulers are practical tools that enable lexicographers to keep their dictionary databases balanced and representative – on the macrostructural level and beyond.


102. Corpus Query Tools for lexicography

1. Introduction

The revolutionary impact of language corpora on lexicography (cf. Landau 2001 among many others) is a consequence of the availability not only of large text collections in electronic format, but also of sophisticated tools to annotate and query such collections. In this article we concentrate on the latter, focusing in particular on concordancing, arguably the most basic and currently most widely used functionality for lexicographic work, but also discussing other standard ways of grouping and displaying textual data (cf. articles 105 and 107).

We survey the fundamental features that a state-of-the-art corpus query tool (henceforth CQT) should offer, providing a grid of criteria and highlighting some trade-offs a lexicographer or lexicography house is likely to face when picking such a tool (e.g. between user-friendliness and power, between automation and control).

For ease of presentation, we look at the functionalities of a query tool following a schematic pipeline reflecting how one might interact with a corpus during lexicographic work: Making the corpus accessible via the query tool (section 2.1); (keyword listing to select target entries (section 2.2); extracting collocations for the target entries (section 2.3), searching for defining evidence, examples, and translations in context (section 2.4).

As query tools rapidly evolve, in our general discussion we do not focus on specific programs currently available, as their descriptions would be obsolete by the time this handbook goes to print. We do however provide an assessment of some representative tools with respect to our feature typology (section 3).

In the conclusion (section 4), we mention some further aspects of tool assessment (usability, efficiency) we did not include in our typology and we tackle some open issues in CQT development.

In the remainder of the article we will use the following terminology. Tokens are the smallest running units a corpus is composed of (typically, words and punctuation marks). The process of splitting text into tokens is called tokenization (and the output of this process is tokenized text). A corpus can be either made of “raw” text, or enriched with annotation, i.e. information about the texts in the corpus, that might be exploited for queries. We shall refer to four types of annotation:

1. positional attributes, pertaining to single tokens (e.g. the lemma or part-of-speech (POS) tag of a token),
2. structural attributes, spanning stretches of tokens (e.g. information about sentences or syntactic phrases) (Christ 1994),
3. meta-data, i.e. non-linguistic information pertaining to whole documents or other textual units (e.g. year of publication, author or topic), and
4. alignment, a special kind of annotation recording cross-linguistic correspondences (typically at the sentential level) between the components of a parallel corpus.

2. A lexicographer’s corpus querying pipeline

2.1. Preparing a corpus for querying

Several corpora developed mainly with lexicographic purposes in mind can be consulted online via their own proprietary CQT, for