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2 **12 Comparing word sense distinctions** 3 **with bilingual comparable corpora:** 4 **A pilot study of adjectives in English** 5 **and Spanish**

9 **1 Introduction**

11 Amidst the recent surge of interest in the applications of corpora in Cognitive
12 Linguistics, and the wide range of methodologies now available (e.g. Gries and
13 Stefanowitsch 2006), the types of corpora employed and their applications in
14 translation and L2 instruction still has the potential for further expansion. The
15 current chapter draws attention to one of the areas in which the field has the
16 potential for growth, and suggests the gains it may have to offer translation
17 and Second Language Acquisition (SLA). These points are then illustrated with
18 examples from a pilot corpus study conducted on a set of adjectives in English
19 and Spanish.

20 The pilot study is suggestive of the potential role of bilingual comparable
21 corpora, that is, sets of non-parallel matched monolingual corpora, each in
22 a different language, as an approach to comparative lexical semantics. This
23 method is argued to hold several advantages over the more traditional lexical
24 studies employing monolingual corpora and bilingual parallel corpora. The
25 study sorts a set of adjective senses from an English corpus and a Spanish
26 corpus on the basis of distributional variables (such as whether adjectives occur
27 in predicative or attributive position), which allows a detailed analysis of the
28 relatedness of the senses of each word in each language. The networks of
29 related senses in the two languages can then be compared, and it can be seen
30 which senses are similar or different in the two languages. This can serve as a
31 guide for L2 students, translators and lexicographers interested in finding the
32 best approximation for a given source-language meaning in a target language.

34 **2 Types of corpora and their applications**

35 Many types of corpora are currently available, several of which have been, or
36 could be, employed in studies with implications for SLA or translator training.

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1 To date, relatively few of the available types of corpora have been taken advantage
2 of in translation studies. Granger (2003: 21) provides a summary of available corpus
3 types and those that have been employed in contrastive linguistics and translation
4 studies, a list updated to some extent in Marzo et al. (2010). The circumscribed
5 range of corpus usage can be attributed in part to the relative newness of the field.
6 For instance, the use of corpora in translation studies was not suggested until
7 Baker (1995). The integration of corpora in SLA is slightly older, dating back at
8 least to Johns and King (1991).

9 Types of corpora with past or potential applications within translation studies
10 and SLA include (bilingual) parallel corpora, monolingual comparable corpora,
11 and bilingual comparable corpora. We will briefly mention some of the relevant
12 work done with these types of corpora, and the advantages and disadvantages
13 of each type for SLA applications.

14 Parallel corpora are probably the most widely used corpora in translation
15 studies. These are “corpora that contain a series of source texts aligned with
16 their corresponding translations” (Malmkjaer 1998: 539, quoted in Granger
17 2003: 20). Translated texts without their source texts may also be used. These
18 may be termed *translation* (or *translational*) *corpora* (Baker 1999). Parallel corpora
19 may be employed either to study translation itself or as a basis for comparing
20 the structure of two languages (Mason 2001). However, as Mason (2001) notes,
21 parallel corpora may give deceptive results for research comparing linguistic
22 structure, as their target language material will differ from non-translated data
23 from that language. It may be influenced by the source language, or may be
24 subject to artifacts stemming from the process of translation itself (see Olohan
25 2004: 26–28 for a discussion of examples). Some of these effects can be con-
26 trolled for by employing two parallel corpora, one translated from language A
27 to B and one translated from B to A (Johansson 1998; referred to as *bilingual*
28 *parallel corpora* in Zanettin 1998).

29 In translation studies, artifacts arising from translation are an important
30 focus of study in their own right. Largely for this reason, the use of monolingual
31 comparable corpora is on the rise in translation studies (see Olohan 2004,
32 Chapter 4). These corpora consist of translated and non-translated texts in a
33 single language, examined “in order to explore how text produced in relative
34 freedom from an individual script in another language differs from text pro-
35 duced under the normal conditions which pertain in translation, where a fully
36 developed and coherent text exists in language A and requires recoding in
37 language B” (Baker 1993: 233, quoted in Olohan 2004: 36). These corpora allow
38 researchers to identify features of translated texts, some of which are outlined in
39 Baker (1996). Although parallel corpora may be useful for understanding or
40 conducting translation, they appear to be less immediately useful in SLA. For

1 SLA students, it may be more productive to be exposed to non-translated, rather
2 than translated, data from the L2 (Johansson 2007).

3 Monolingual corpora of learner data are probably the most frequent type of
4 corpus employed in SLA studies. The International Corpus of Learner English
5 (initiated and directed by Sylviane Granger), for example, collects essays from
6 2nd- and 3rd-year university students studying English, representing sixteen
7 L1s. Several studies have used this corpus to compare native speaker data with
8 learner data from speakers of an L1, or a set of L1s, in order to draw attention to
9 L1 transfer or interference effects, such as the frequency of the use of grammatical
10 constructions (as in Valenzuela and Rojo 2008) or of particular lexical items (as
11 in Ringbom 1998).¹ These results can be integrated into SLA instruction to help
12 students avoid typical learner patterns of overuse or underuse.

13 Despite the gains made with monolingual comparable corpora, relatively
14 little research has so far been conducted using bilingual comparable corpora.
15 When these corpora have been employed, they have focused on the study of
16 specific genres, such as printed public notices in English and German (Schäffner
17 1998) or medical research articles (Williams 2010), or for the study of colloca-
18 tional frequency (Noël and Coleman 2010; Zanettin 1998). Bilingual comparable
19 corpora have been proposed for use in monolingual Word Sense Disambiguation
20 (WSD) (Kaji 2003) – that is, “translation equivalents” in one language can be
21 used to define the various senses of a given word in a different language – but
22 this procedure has little direct application for translator training and even less
23 for SLA. It seems evident that comparable data from multiple languages are
24 necessary if L2 learners and translators are to use corpus data to find the nearest
25 equivalent, in a target language, for a lexical item in a source language. This
26 suggests that bilingual comparable corpora as well as monolingual corpora are
27 a potentially valuable resource for SLA and translation studies addressing lexical
28 semantics.

31 **3 Options in corpus analysis**

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33 Besides the choice of corpus type, several other decisions must be made by
34 researchers interested in employing corpora in SLA studies. A study may focus
35 on word senses, words, or phrases in the corpora, for example. It is also neces-
36 sary to select the parameters that are taken into account, such as the words
37

38 **1** Numerous studies of this type are collected in an online Learner bibliography by the Centre
39 for English Corpus Linguistics. See: [http://sites-test.uclouvain.be/cecl/projects/learner_corpus_](http://sites-test.uclouvain.be/cecl/projects/learner_corpus_bibliography.html)
40 [bibliography.html](http://sites-test.uclouvain.be/cecl/projects/learner_corpus_bibliography.html).

1 or syntax that co-occur with the items of interest. This section will address
2 the choice of parameters, and the selection of words versus senses, selected in
3 previous research and available for future investigations relevant to applications
4 in SLA and translation. The discussion will focus on hierarchical cluster analyses
5 (HCA), an exploratory data grouping method that has proven its usefulness in
6 studies of monolingual polysemy (Gries and Divjak 2009; Sullivan 2012) and
7 translational corpora (Jenset and Hareide 2013; Ke 2012). HCA also has the
8 advantage that its results can be assessed with bootstrapping, a method by
9 which data are shuffled and then re-clustered to statistically evaluate the validity
10 of the clusters (Divjak 2010; Glynn 2010; Suzuki and Shimodaira 2011).

11 Perhaps the most readily available variables that can be employed in cluster-
12 ing consist of the items' collocations, that is, other items that tend to occur in
13 proximity to the items in question (an approach adopted in Kaji 2003, and in
14 monolingual corpus studies including Gibbs and Matlock 2001, and Kishner and
15 Gibbs 1996). However, there are several reasons why the use of predominantly
16 syntactic variables may allow for a more accurate impression of cross-linguistic
17 equivalence than the more traditional reliance on collocations. Collocations tend
18 to be highly language-specific, which is the primary reason that translators-in-
19 training need to be exposed to the concordances of items in the target language,
20 because they are likely to differ from the source language (see Hadley 2002 for
21 discussion). This trait, which renders concordances a productive part of trans-
22 lator training, makes them less useful in comparisons between languages, since
23 collocations are likely to often be too different between the two languages for
24 meaningful comparisons to be made. Of course, languages have different syntactic
25 structures as well as different concordance patterns, but we argue it is more reveal-
26 ing to compare syntactic structures between languages (i.e. adverbial modification)
27 than to find analogous collocates (i.e. co-occurrence of *skin* with *soft* in English
28 and that of *piel* 'skin' with *suave* 'soft/smooth' in Spanish). Additionally, indi-
29 vidual collocates can distort a cluster analysis, particularly one comparing
30 word senses (Gries and Divjak 2009), and reliance solely on collocates may do
31 little to reveal ties between senses, since collocates can co-occur with only one
32 sense.

33 The variables that are considered can be selected and manipulated in many
34 different ways. Gries and Divjak (2009) employ morphosyntactic or semantic
35 variables in their analysis which they call "ID tags". Though clustering analyses
36 based primarily on syntactic data present several advantages, it must be acknowl-
37 edged that tagging syntactic IDs is currently far more labor intensive than col-
38 lecting collocations. Future advances in automatic corpus tagging could simplify
39 the ID-tagging process, allowing even long-distance syntactic relations and
40

1 large-scale syntactic structures to be identified and tagged automatically. Im-
2 proved availability of comparable corpora in multiple languages with any degree
3 of tagging beyond POS-tagging (for example, tagging of nouns and/or adjectives
4 for plurality) would reduce the number of variable values that must be manually
5 identified and assigned as ID tags in a given study.

6 In addition to the type of data chosen for annotation and consideration,
7 corpus studies dealing with polysemy can choose whether to compare the various
8 senses of individual words, thereby charting the structure of polysemy networks,
9 or to ignore the different senses of each item and compare instead different words
10 with each other, mapping the relatedness of “near synonyms” (Gries 2008; Divjak
11 2010) and discovering which words in a semantically related set are most similar
12 based on their distribution and syntactic, semantic, and other properties.

13 Examining the concordances of one item at a time, in one language at a
14 time, offers applications for translation and potentially for SLA (Hadley 2002).
15 However, we suggest that there are also advantages to analyzing the connections
16 between senses of various items, and in comparing these networks of senses
17 in multiple languages. A speaker or translator will typically need to find the
18 best approximation for one sense of a source language item in a target lan-
19 guage. Examining concordances of a single item in the target language may
20 give learners and translators a general feel for the usage range of a particular
21 item, but may be less directly applicable to the everyday task of word translation
22 than a corpus-based tool that compares multiple words and senses in both
23 the target and source languages. This can be accomplished through the use of
24 clustering of word senses in bilingual comparable corpora.

25 As seen in the previous section, most clustering studies – and all of those
26 employing the methodological choices described above – have been employed
27 with monolingual corpora (Divjak and Gries 2008; Gries 2006; Sullivan 2012).
28 Which of the choices explored above are most compatible with the use of
29 bilingual corpora? It has already been suggested that syntactic, as opposed to
30 collocational, data, are more appropriate to cross-linguistic studies. In terms of
31 the choice between “near synonyms” and word senses, it seems that the latter
32 may prove more useful. No L2 speaker or translator would want to always
33 equate one specific lexical item in the source language, such as English *soft*,
34 with one specific item in the target language, such as Spanish *suave*. It seems
35 more realistic that one specific sense of *soft* might, indeed, always be best trans-
36 lated as *suave*. There may in turn be a specific sense of *suave* that can always be
37 felicitously translated into English as *soft*. It may be most useful in SLA and
38 translation, therefore, to look for similarity between word senses rather than
39 between words. For this, studies of bilingual comparable corpora with clustering
40 of word senses may prove the most productive choice.

4 Pilot study: method

As a preliminary assessment of the effectiveness of word sense clustering using bilingual comparable corpora, our sample study collected 300 examples of each of four adjectives: English *soft* and *smooth* and Spanish *suave* and *blando*. English examples were randomly selected from all instances of the lemmas *soft* and *smooth* tagged as adjectives in the British National Corpus, and Spanish examples were randomly selected from all instances of the lemmas *suave* and *blando* tagged as adjectives in the Corpus del Español. These examples were assigned ID tags and analyzed in context to identify the sense instantiated by each corpus example. Identification of senses and annotation of ID tags for *blando* and *suave* was assisted by a team of undergraduate native speakers of Spanish. Senses in both English and Spanish were chosen by consensus among the authors and the undergraduate team, and the choice of which senses should be considered as separate was continually reassessed as data were analyzed. The senses of each word were clustered based on the ID tags.

As discussed, the ID tags in our study were primarily syntactic. In addition to the reasons discussed above for using syntactic versus collocational tags, we chose syntactic over semantic tags because we aimed to make the annotation as objective and unbiased as possible. We found evaluations of syntactic features to be more consistent across annotators than semantic judgments.

Given the preliminary nature of the study, only nine ID tags were included for each language. Eight were the same for both languages and one tag was used for each language that was not applicable for the other. For both English and Spanish, ID tags were assigned for the type of construction in which the adjective appeared (attributive, predicative or resultative); modification of the adjective by one or more adverbs; presence of other adjectives modifying the same noun; presence of a PP complement on the modified NP; presence of the NP within a PP; whether the modifiee was expressed anaphorically; whether the modified noun was a mass or count noun, and its number (singular or plural). English ID tags included tough-movement, which does not exist in Spanish, and Spanish ID tags included pre- or post-nominal position of the adjective, which is far more variable in Spanish than in English. Adjective gender was not included as an ID tag in Spanish because it is largely semantically arbitrary, an observation confirmed by the apparently randomizing effect its inclusion had on the resultant cluster analysis. We intend to expand the number of ID tags in subsequent studies on texture adjectives in English and Spanish, though we will continue to emphasize syntactic variables.

1 In all clustering studies of sense relatedness, no matter how objective the ID
2 tags, sense labeling itself is subjective to some degree. The application of criteria
3 such as those of the principled-polysemy approach (Evans 2005: 41; Tyler and
4 Evans 2001; discussed in Gries and Divjak 2009) can make the process of dis-
5 tinguishing senses less arbitrary, but total objectivity or agreement between
6 all researchers is almost impossible. The main problem for distinguishing
7 senses is granularity (i.e. at which level similar senses should be distinguished).
8 Granularity was resolved partly based on frequency: senses with three or fewer
9 examples were preferentially grouped with others rather than put in the “other”
10 category; and also on classification accuracy. An overly high granularity is
11 unproblematic when there are an adequate number of examples of each sense,
12 because similar senses cluster together. High granularity only becomes truly
13 problematic when there are few examples of each sense – as was occasionally
14 the case in our small-scale study – because a small set of examples cannot be
15 expected to be representative of the contexts in which a given sense occurs,
16 leading to inaccurate clustering.

17 In the procedure used here, ID tags were annotated in columns in an Excel
18 file in the format shown in Table 1. Note that the “sense” label is purely for
19 convenience, and that these one-word labels are not taken in any way to be
20 descriptions or definitions of the senses, but merely as labels for senses which
21 are treated as distinct from other senses. We argue that it is neither necessary
22 nor desirable for SLA or translator training to define word senses using a
23 one-word “synonym” in either the same language or in a different language
24 (see Kaji 2003), as is common practice in WSD. These “synonyms” are a neces-
25 sity in machine translation, but for human corpus users they are less useful than
26 more exact definitions. It is convenient to have a short label for word senses,
27 especially as a shorthand in annotating and as inputs to analysis software, but
28 for human audiences these labels can be accompanied by in-depth explanations
29 of the nuances of each particular sense, the semantic range of the sense, and its
30 boundaries with other senses. These explanations should not be *a priori*, but
31 should be based on observations and examples from the corpus itself. The
32 semantics of any word sense are likely to be complex, and we see no advantage
33 to artificially constraining or simplifying the descriptions of senses.

34 The ID tags for each item form a behavioral profile vector (the set of variables
35 the values of which are represented by ID tags), which can be inputted into a
36 hierarchical agglomerative cluster (HAC) analysis. This can be done in a number
37 of ways. Here, we are following the procedure described in Gries and Divjak
38 (2009), using the Behavioral Profiles (BP) program for R written by Gries (2008).
39 Among other functions, this script performs a HAC that sorts the examples on
40 the basis of their behavioral profiles. This results in a tree-like clustering dia-
gram, called a “dendrogram”, in which similar senses are clustered. The current

1 **Table 1:** Sample senses and ID tags of *soft*.*

		PP	In		Other	Count	Number		
	Sense	Syn.	comp	PP	Adv.	adjs.	N?	of N	
4	Consistency	a	yes	no	yes	no	yes	s	for the table, continues on to an arugula salad with dates and a meltingly
5									soft pork shank with rye gnocchi and
6									sauerkraut.
7	Flexible	a	no	no	no	no	yes	pl	are very uncommon in snowboarding.
8									And at the same time, you're wearing
9									soft boots that you can run around in,
10	Force	a	no	yes	no	no	yes	s	his eyes brushing my neck, my jaw, and
11									my mouth with a soft force, and then
12									resting deep inside my eyes.
13	Gentle	p	no	no	no	yes	yes	s	guy is about 5' 5", 130 pounds, sweet-
14									heart, intelligent, soft and gentle. Not
15									someone who's prone to be a tough
16	Humanities	a	no	yes	no	no	yes	pl	guy.
17									business, engineering, and the like –
18									has clearly decided to write off the soft
19	Indirect	a	no	no	no	no	no	n	disciplines, namely the humanities and
20									the arts.
21									Well put. . . . we need the most severe
22									changes to restrict the soft money
23	Noise	p	no	no	yes	no	yes	s	which, as I say a couple of times, is a
24									blight on. . .
25									it as best as he can. The sound, how-
									ever, is still understandably soft .
									# STARKS waits and then reaches for
									the knob on. . .

26 * a = attributive; p = predicative; s = singular; pl = plural; n = not applicable

27
28
29 study utilized the Canberra similarity metric to make the best use of the rela-
30 tively small data set, and used the Ward amalgamation strategy, in order to
31 encourage clusters of an easily interpretable size.

32 The BP script for R also incorporates the pvclust script (Suzuki and Shimo-
33 daira 2011) that assesses the reliability of the HAC analysis with bootstrap re-
34 sampling. That is, the instances of each word or sense are repeatedly shuffled
35 and then re-clustered. In the BP script, data are re-clustered 10,000 times. The
36 results of this resampling are reported as Approximately Unbiased (AU) *p*-values,
37 which are assigned to each cluster and which report how often the cluster
38 emerged in the resamplings. For example, an AU *p*-value of 70% would mean
39 that a particular cluster occurred in 70% of the resamplings. The apparent cluster
40 is less likely to be due to chance than a cluster with a lower AU *p*-value, and
more likely to be a chance occurrence than a cluster with a higher AU *p*-value.

5 Pilot study: results and analysis

The outcome of the HAC analysis can be represented in dendrograms such as Figures 1–4. Distance (“height”) between points of amalgamation represents the difference between the clusters. As this is a preliminary study, which involved the use of relatively few ID tags and small corpus samples, “height” is fairly low and height distinctions are small, meaning that clusters could be subject to change with the addition of more data. Nevertheless, even these preliminary results give some indications of the potential applications of clustering of senses found in bilingual comparable corpora.

The AU *p*-values are given above each cluster and to the right. When these are low, the apparent cluster does not replicate well and is probably due to chance. Higher AU *p*-values indicate clusters that are more strongly supported by the data.

As might be expected, our trial study confirmed that polysemy networks across languages demonstrate frequent mismatches. Of course, similar-seeming items, such as English *smooth* and Spanish *suave*, have some senses that they share and others that they do not (see Figures 1 and 2). For example, both items can refer to texture, as in *textura suave* or *smooth texture* (labeled as “textura” and “slick”, respectively). On the other hand, *smooth* has an “efficient” sense that *suave* lacks, as in *smooth efficiency*, and *suave* has a “gentle” sense not expressed by *smooth*, as in *soplo suave* ‘gentle breeze’.

Not only does each item each have senses not shared by their near equivalents in another language, but items typically have some senses that are better expressed with one word in an L2 and other senses that are better expressed with a different L2 word. Some senses of English *soft* are close equivalents of senses of Spanish *suave*, and some can be more closely equated with Spanish *blando* (compare Figure 3 with Figures 2 and 4).

For example, senses shared by *soft* and *suave* refer to the texture of skin (the sense labeled “skin” in *soft skin* and the “piel” sense in *suave piel*; but #*blanda piel*) and to silkiness of hair or fur (*soft curls*; these senses are labeled “silky” and “sedoso” in Figures 3 and 2). On the other hand, *soft* and *blando* share a set of senses referring to gentle forces (*soft push*; senses labeled “force” and “fuerza” in Figures 3 and 4), squishy surfaces (*soft mud*; labeled “squishy” and “malleable”), yielding springy surfaces (*soft cushions*; “yielding” and “mullido”), and internal consistency (*soft butter*; “consistency” and “consistencia”). The “yielding”, “force”, “squishy” and “silky” senses of *soft* cluster together in Figure 3, and therefore behave similarly in English, even though some senses resemble *blando* and some

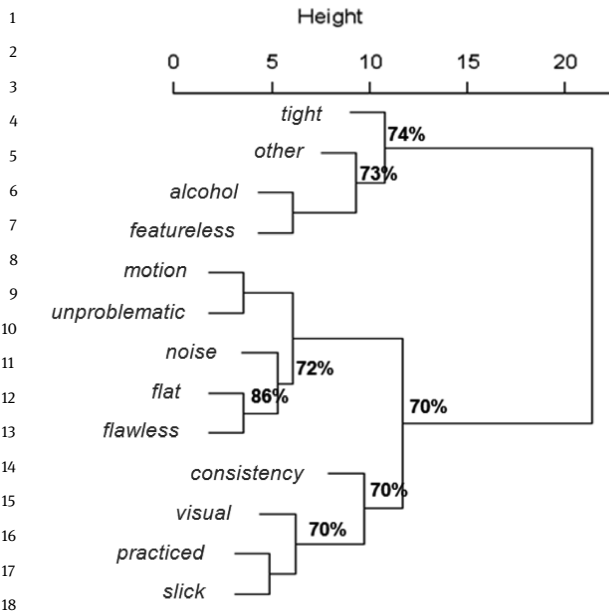


Figure 1: Dendrogram for *smooth* with AU values (values below 70% not shown).

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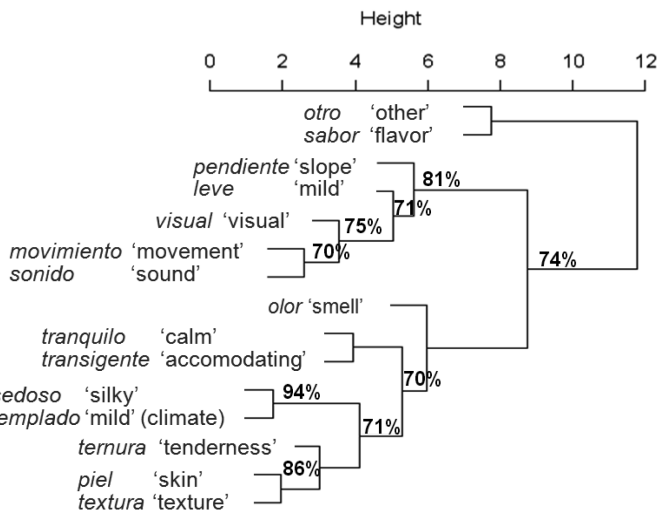


Figure 2: Dendrogram for *suave* with AU values (values below 70% not shown).

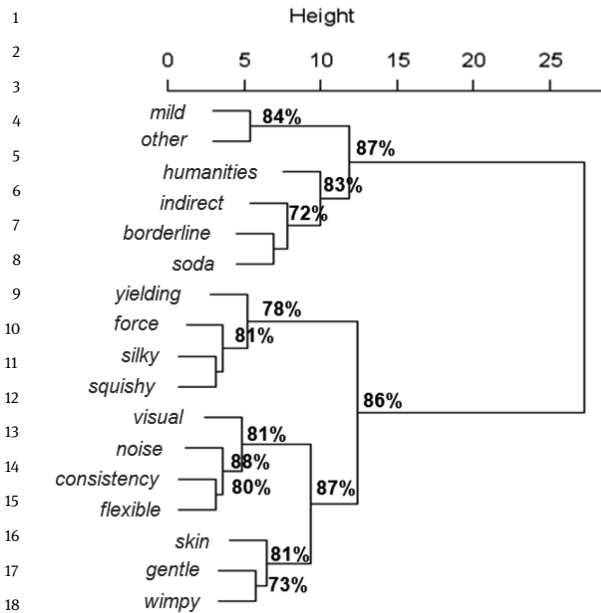


Figure 3: Dendrogram for *soft* with AU values (values below 70% not shown).

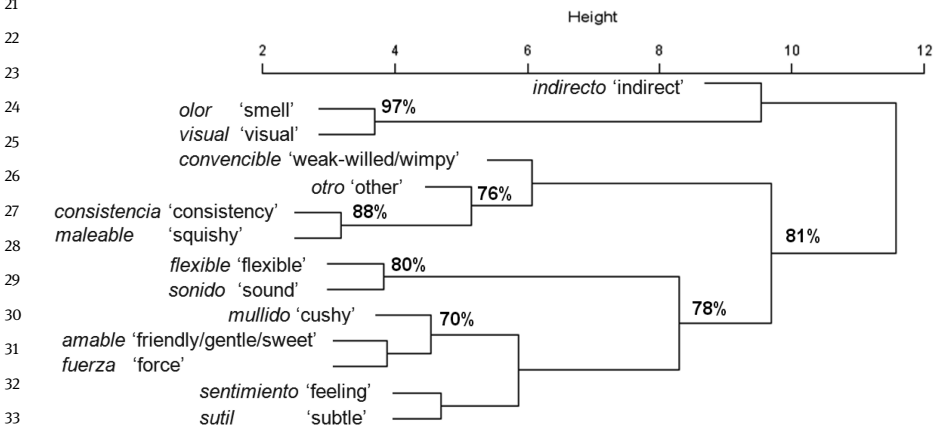


Figure 4: Dendrogram for *blando* with AU values (values below 70% not shown).

suave in Spanish. The “yielding” / “force” / “squishy” / “silky” cluster had an AU *p*-value of 78%, meaning that the cluster recurred in 78% of the bootstrap resamplings (see Figure 3). It therefore occurs with the majority of possible initial orders, and does not depend to a great extent on the initial order of the data.

1 Although this type of observation may be facilitated by the use of corpora, it
 2 could be achieved by careful study alone. A unique contribution of clustering
 3 based on word senses is that this analysis can reveal how clusters of senses,
 4 as well as individual senses, are related across languages. Our results suggest
 5 that many of the senses which are shared between *blando* and *soft*, but not by
 6 similar adjectives such as Spanish *suave* and English *smooth*, cluster closely
 7 together in English and also to some extent in Spanish. If several of these senses
 8 (such as those related to internal consistency and squishy surfaces) are closely
 9 related in both languages, it may be no accident that all senses in the cluster
 10 are expressed by the same lexical item in each language. Instead, the senses
 11 are probably related by underlying semantic commonalities, and may even be
 12 predicted, based on these commonalities, to be expressed by a common lexical
 13 item in languages other than English and Spanish. Learning the clusters of
 14 senses that are near-equivalents in two languages may be an efficient way for
 15 L2 learners to understand how to best express a given meaning in their L2, by
 16 looking at the clusters of meanings expressed by each L2 item and comparing
 17 these with the meanings in their L1. Learning corresponding clusters of senses
 18 in the two languages is more efficient than memorizing all the corresponding
 19 senses individually.

20 Cluster analyses based on syntactic features take advantage of more shared
 21 variables between senses than analyses based solely on collocations (such as
 22 Kaji 2003), in that slightly different senses are far more likely to share syntactic
 23 structures than to share individual collocates. Examples (1)–(3) are from the
 24 texts taken from the British National Corpus and Corpus del Español and used
 25 in the current study (as are all subsequent examples). Each excerpt in (1)–(3)
 26 represents a different sense of *soft*, and few collocates are shared by the different
 27 senses. However, the shared syntactic structures are immediately evident. The
 28 cluster of senses of *soft* that can be characterized as referring to skin texture –
 29 “skin” in Figure 3, as in example (1), as well as gentleness of personality “gentle”,
 30 as in (2), and weakness of will “wimpy”, as in (3) – demonstrates syntactic
 31 attributes typical of the cluster, such as the use of the copula, the predicative
 32 position of the adjective, and the presence of other adjectives coordinated with
 33 *soft*. The “skin” / “gentle” / “wimpy” cluster has an AU *p*-value of 81%.

- 34 (1) *You want the skin there to be as smooth and **soft** as possible*
 35 (2) *the guy is about 5' 5", 130 pounds, sweetheart, intelligent, **soft** and gentle*
 36 (3) *she's very well spoken, but she's pretty **soft***

37
 38 ID tags typical of this cluster include the use of multiple adjectives (as in *smooth*
 39 *and soft*), predicative use of *soft* (*to be... soft, is... soft*) and adverbial modifica-
 40 tion (*pretty soft*). These uses of *soft* almost always modify singular nouns, typi-
 cally count nouns (such as *the guy*). For the latter two senses exemplified above,

1 these nouns are also typically animate and human, but this type of semantic
 2 information was not taken into consideration in this analysis – a choice which
 3 allowed clustering such as with the sense in (1) to come through more strongly.
 4 The ID tags listed above are of course not shared by every instance of a sense
 5 in the cluster, but did help to typify the cluster relative to senses outside the
 6 cluster.

7 To give a Spanish example, a cluster is formed by the senses of *blando*
 8 referring to (4) springy cushiness (“mullido”), (5), weak or gentle force (“fuerza”)
 9 and (6) mildness or sweetness of personality or behavior (“amable”). This cluster
 10 has an AU value of 70% (see Figure 4).

- 11 (4) *me acomodaba en los **blandos** almohadones de un coche del ferrocarril*
 12 ‘I got settled on the **soft** cushions of a train car’
 13 (5) *el rostro sonrosado por los **blandos** golpes de la espuma...*
 14 ‘the face rosy from the **soft** splashes of the spray..’
 15 (6) *¿Y si los **blandos** halagos de esta niña pudiesen cicatrizar las úlceras de mi*
 16 *corazón?*
 17 ‘and if the **soft** praises of this girl could heal the wounds in my heart?’
 18

19 These uses were often in the plural (e.g. *blandos golpes* ‘soft splashes/blows’).
 20 The adjective *blando* was typically attributive and pre-nominal (*blandos almoha-*
 21 *dones* ‘soft cushions’ vs. *almohadones blandos*) and often occurred in a noun
 22 phrase with a PP modifier (...*de un coche*, ...*de la espuma*, ...*de esta niña* ‘of a
 23 car, of the spray, of this girl’).

24 Clustering based on syntax can prove useful where both collocates and
 25 intuitions are misleading. For example, both Spanish *suave* and English *smooth*
 26 modify nouns denoting motion, as in *suave movimiento* and *smooth motion*.
 27 However, the sense of *suave* referring to motion (“movimiento” in Figure 2) and
 28 the sense of *smooth* describing motion (“motion” in Figure 1) are not comparable.
 29 A “smooth” motion is a graceful or practiced motion, whereas *suave movimiento*
 30 refers to a weak or feeble movement. This sense of *suave* should probably never
 31 be translated as, or equated with, *smooth*, and vice versa, despite the superficial
 32 similarity of the expressions that might prompt the senses to be viewed as near-
 33 equivalents. The difference in meaning is, however, apparent in the clustering of
 34 the senses of *suave* and *smooth* in each language. In Spanish, this sense of *suave*
 35 appears to cluster with senses referring to dim visual stimuli (“visual”) and
 36 weak audio stimuli (“sonido”), and (less closely) with mildness of a condition,
 37 such as a disease (“leve”) (see Figure 2; this cluster has an AU value of 71%).
 38 This suggests that the sense refers to a low position on a scale of intensity – in
 39 this case, intensity of the motion described. In English, on the other hand,
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1 *smooth* referring to motion (“motion” in Figure 1; as in example [7]) clusters with
 2 the sense of *smooth* referring to the unproblematic accomplishment of a goal
 3 (“unproblematic”; as in [8]). This cluster has an AU value of only 56%, but is
 4 nevertheless worth mentioning due to its incontrovertible difference from the
 5 Spanish pattern.

6 (7) *His gait was **smooth**, as if his hip sockets had been oiled...*

7 (8) *As moose rescues go, this was a **smooth** one, says Simmott...*

8
 9 This clustering suggests that the sense of *smooth* referring to swift unimpeded
 10 motion is metaphorically related to the sense referring to swift unhindered accom-
 11 plishment of a goal. This is a different type of association to that suggested by
 12 the clustering of the “movimiento” sense in Spanish. Awareness of this type of
 13 clustering can draw attention to the difference in meaning between the two
 14 superficially similar senses of *smooth*.

15 In general, our analysis suggests that metaphoric senses such as *smooth*
 16 “unproblematic” do not cluster exclusively with other metaphoric senses in
 17 either Spanish or English, but instead cluster with specific non-metaphoric
 18 senses. For example, in Spanish, the “amable” sense of *blando* referring to
 19 ‘kindness’ or ‘sweetness’ clusters with “mullido” (‘yielding surface’; AU value
 20 70%), whereas the “convencible” (‘weak-willed’) sense clusters more closely
 21 with “consistencia” (‘liquid consistency’; though with an AU value of only
 22 60%); that is, a friendly human being is *blando* in the manner of a comfortable
 23 chair, whereas a weak-willed human being is “malleable” like a semi-liquid jelly.
 24 English *soft* lacks an “amable” sense referring to sweet behavior or character
 25 (expressions such as *soft-hearted* have suggestions of this sense, though these
 26 were not well-represented in the corpus). On the other hand, the “wimpy” sense
 27 of *soft* in English is connected to “consistency” (as part of a larger cluster with
 28 AU value 87%), as in Spanish. The patterns of semantic extension in the lan-
 29 guages therefore appear similar, in that specific metaphoric senses are tied to
 30 specific non-metaphoric senses, but the resultant networks differ in their details.
 31 Awareness of these distinctions is a key to the correct usage of these senses with
 32 the appropriate connotations. For example, a Spanish speaker learning English
 33 might be unaware that English *soft* lacks some of the positive connotations of
 34 the Spanish “amable”, but that the negative sense “convencible” translates
 35 well as English *soft* “wimpy”. The other members of the clusters of these senses
 36 in each language make it clear which senses are closer in meaning between the
 37 languages. This can be especially useful in understanding metaphoric senses,
 38 for which the connections to other senses may not be apparent to an L2 learner.

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6 Conclusion

Recent advances in corpus applications have contributed much to Cognitive Linguistics, and increasingly to translation studies and SLA. However, the types of corpora that have been adopted for SLA applications remain largely limited to monolingual corpora. These corpora have proven their utility in translation studies and SLA: monolingual untranslated corpora can give SLA students a feel for the native usage of lexical items and constructions in their L2, and monolingual learner corpora allow SLA students to avoid common mistakes in their L2. We suggest here that bilingual comparable corpora may prove equally well-suited for SLA studies of vocabulary and lexicon. In particular, a comparison of sense clustering in an L1 and L2 can allow students to recognize which types of senses of an item in their L1 correspond most closely to particular items in the L2. For example, this type of analysis demonstrates graphically which groups of senses of English *soft* resemble senses of Spanish *blando*, and which senses of *soft* more closely resemble senses of Spanish *suave*. At the same time, these analyses can draw attention to mismatches between deceptively similar L1 and L2 items, such as English *smooth* and Spanish *suave*, both of which can modify nouns denoting types of motion, but which have very different meanings and hence different positions in the dendrograms of these English and Spanish adjectives. The clusters can also help students choose lexical items with the intended connotations, by drawing attention to the relatedness of these senses with other clearly positively or negatively connotated senses, as in the above example comparing English *soft* and Spanish *blando*. Finally, clusters can aid students in the appropriate use of metaphoric senses, by illustrating how these senses are connected to less metaphoric senses, the meaning and use of which may help clarify the items' metaphoric meanings.

Results from bilingual comparable corpus studies are a long way from being integrated in the SLA classroom. We argue that this lack of progress can be attributed at least in part to the relative paucity of corpus studies aimed at SLA applications, and the lack of diversity in the studies that do exist. Our results, though tentative, suggest that additional types of corpus studies may be productive for SLA. We have also suggested certain methodological choices that may be pursued in order to generate benefits for SLA. It is hoped that recognition of the varied types of corpora and methodologies available for SLA research will lead to the expansion of corpus studies aimed at SLA application, and ultimately the productive integration of these studies and their results in the SLA classroom.

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