

Building a Chinese-English Mapping Between Verb Concepts for Multilingual Applications

Bonnie J. Dorr¹, Gina-Anne Levow¹ and Dekang Lin²

¹ Institute for Advanced Computer Studies

University of Maryland

College Park, MD, 20742

{bonnie,gina}@umiacs.umd.edu

² Department of Computing Science

University of Alberta

Edmonton, Alberta, Canada, T6G 2H1

lindex@cs.ualberta.ca

Abstract. This paper addresses the problem of building conceptual resources for multilingual applications. We describe new techniques for large-scale construction of a Chinese-English lexicon for verbs, using thematic-role information to create links between Chinese and English conceptual information. We then present an approach to compensating for gaps in the existing resources. The resulting lexicon is used for multilingual applications such as machine translation and cross-language information retrieval.

1 Introduction

With the advent of the web and increasingly more global interconnectivity, the need for online multilingual information has grown significantly in the last 5–10 years. This is accompanied by a growing need for rapid construction of lexical resources. Creating resources by human labor alone has become infeasible, thus motivating the development of automatic and semi-automatic approaches to resource acquisition. This paper addresses large-scale construction of a Chinese-English lexicon for verbs, including an approach to compensating for gaps in the existing resources.

The lexicons resulting from our acquisition approach are used for semantic analysis in applications such as machine translation and cross-language information retrieval. The importance of semantic analysis in either of these two applications is clear when one considers the degree of inaccuracy that might result from using a weak alternative, such as access to a bilingual word list.

Our starting point is an existing classification of English verbs called EVCA (English Verbs Classes and Alternations) [11]. We couple this with a Chinese conceptual database called HowNet [23–25] (<http://www.how-net.com>), from which we extract thematic-role information (e.g., a mapping between the HowNet “Patient” and the EVCA-based “Th(eme)”) to create links between Chinese and

English conceptual information. HowNet currently contains no English translations; thus, we also use a large machine-readable Chinese-English dictionary called Optilex to produce candidate English translations.¹ Although later versions of HowNet are expected to include the English translations, these are not openly available—only the binary versions have been promised and these will be accessible solely through the use of (purchasable) HowNet software. Moreover, we expect our techniques to be generally applicable to *other* foreign language semantic hierarchies where English translations are not available. We predict this will occur more and more frequently, as online (non-bilingual) linguistic resources continue to be made available in multiple languages.

Several researchers have investigated the problem of assigning class-based senses to verbs [2, 17, 18] using a variety of online resources including Longman’s Dictionary of Contemporary English (LDOCE) [19], EVCA [11], and WordNet [14]. Translation of English classes into other languages has proven difficult [7, 15, 21], but regularities between different language classifications can be found in some online resources [1, 4, 16].

This work extends previous work which used a concept space to produce a hierarchical organization of Chinese verbs [18]. We adopt a technique that is similar in flavor to that of [1] for partitioning English verbs into refined classes using WordNet, with the following extensions: (1) The use of the entire EVCA database rather than a small set of verbs (the *break* class); (2) The provision of a thematic-role based filter for a more refined version of verb-class assignments; (3) Concept alignment across two different language hierarchies (Chinese and English); and (4) Mappings between Chinese and English thematic roles.

This work relies on an augmented set of EVCA classes which include 26 new classes [2]. There are 500 total classes in the extended set, each hand-tagged with semantic representations, thematic-role information, and WordNet synset numbers. We will demonstrate that it is possible to produce a lexicon by associating 709 Chinese HowNet concepts with 500 EVCA classes, with a clear concept-to-class correspondence in a large majority of the cases.²

Figure 1 illustrates the relation between existing resources and the mappings we produced. Solid lines represent pre-existing mappings; dotted lines are ones resulting from the application of our techniques. The most critical of these is the one labeled θ -roles (shorthand for “thematic roles”), which associates EVCA classes with HowNet Concepts. The remaining two dotted-line mappings are “transitive closure” biproducts of the other mappings: Once the thematic-role mapping associates EVCA verbs with HowNet verbs, each HowNet verb is associated with Optilex-based English *glosses* (translations) and WordNet 1.6 Senses.

¹ Optilex is a large (600k entries) machine-readable version of the CETA Chinese-English dictionary, licensed from the MRM corporation, Kensington, MD.

² HowNet contains 815 verb HowNet concepts altogether. However, we are not including the 106 HowNet concepts that are not associated with any Chinese words; these are “higher level” conceptual nodes with no Chinese realization (e.g., V.1 |static|).

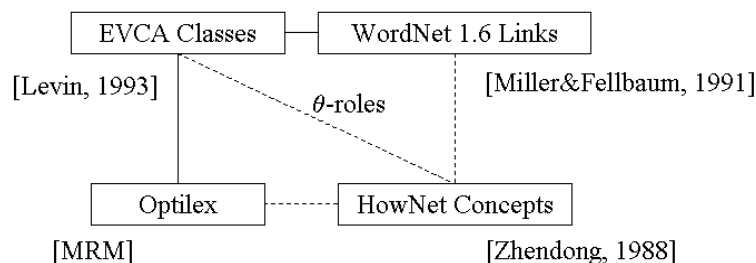


Fig. 1. Relation Between Existing Resources and New Mappings

We will describe how these correspondences are derived and we will show how this process has provided a framework for compensating for gaps in our online resources.

2 Multilingual Applications

The semantic representations produced semi-automatically for our multilingual resources are used in machine translation (MT) and cross-language information retrieval (CLIR) applications. Both applications rely on the use of a parser for mapping the input sentence into a syntactic tree. The parser output is semantically analyzed, producing an encoding of semantic and argument-structure information.

The MT approach is interlingual, where the target-language lexicon is searched for appropriate lexical items matching argument-structure information [3]. A screen snapshot of a MT example is shown in Figure 2. The CLIR approach relies on the same interlingual representation to transform a user’s query into the document language for information retrieval [5, 12].

In both of these applications, thematic roles facilitate the selection of appropriate target-language words. For example, the Chinese verb 拉 (la) corresponds to a wide range of English translations—even if we examine only the verb translations: *slash, cut, chat, pull, drag, transport, move, raise, help, implicate, involve, defecate, pressgang*.⁴ Our approach provides a framework for disambiguation of such cases. Certain of these possibilities—*transport* and *move*—are analyzed as one semantic representation corresponding to thematic roles

³ The Chinese verbs are additionally associated (for free) with WordNet senses from our previously tagged EVCA verbs. More details are given in [6].

⁴ The ambiguity in the word 拉 (la) can often be resolved if it is combined with other characters. For example, 拉车 (la che) unambiguously means *pull a cart*. However, since object dropping is a frequently phenomenon in Chinese, it is not uncommon for verbs like ‘la’ to appear without an argument that easily disambiguates the word. Thus, our approach must allow for multiple possibilities in the lexicon.

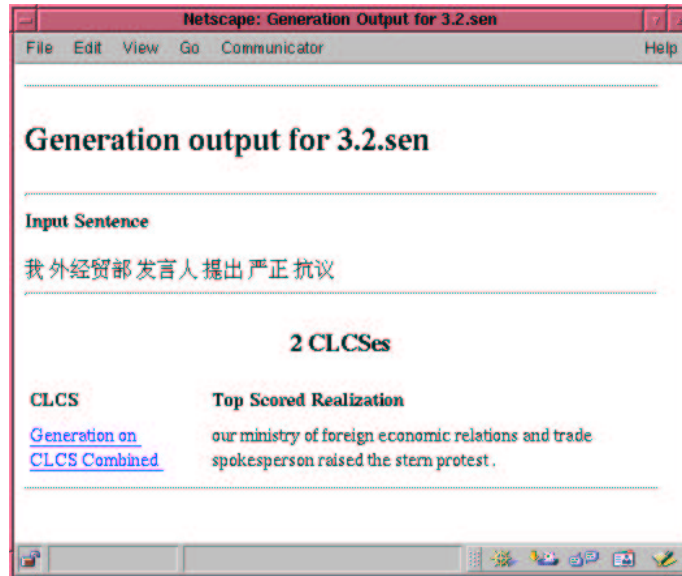


Fig. 2. Translation of a Chinese sentence into English

1. Associate English Optilex glosses with all 12342 Chinese verbs in HowNet, producing 41,324 Chinese-English pairs.
2. Associate each verb-to-concept candidate with at least one of the 500 EVCA classes.³
3. For each HowNet concept, partition the associated Chinese-English pairs into groups whose English glosses correspond to EVCA classes.

Fig. 3. Mapping Chinese HowNet Concepts to English EVCA Classes

(**agent, theme, goal, source**). Other possibilities—*help*—are analyzed as a different semantic representation corresponding to thematic roles (**agent, theme, mod-poss**).

3 Mapping Between Chinese HowNet and English EVCA

Our technique for mapping between Chinese HowNet concepts and English EVCA classes involves associating HowNet thematic roles with those in EVCA. Each HowNet concept (and each EVCA class) is paired with a list of thematic roles, which we call a thematic *grid*. For example, the HowNet concept |Cure| is paired with the grid (**agent, patient, content, tool**), as in *The doctor(agent) cured the man(patient) of pneumonia(content) using antibiotics(tool)*. The corresponding grid in our EVCA database is (**ag, th, mod-poss(of)**). Although the HowNet and EVCA roles are not in a one-to-one correspondence, they can

still be used for a “closest match” prioritization of candidate HowNet-EVCA associations, as we will see shortly.

The three top-level tasks involved in mapping Chinese HowNet concepts to and English EVCA classes are given in Figure 3. (See [6] for more details.) For the purposes of this discussion, we focus on the last of these three tasks, which involves a massive filtering of spurious class assignments. This task involves three steps:

- Order the candidate EVCA classes so that the highest-ranking classes are those that contain the highest number of English verbs matching the Optilex glosses.
- In cases where a tie-breaker is needed, reorder the candidate EVCA classes according to the degree to which the thematic grid in HowNet concept matches that of the relevant EVCA class. The matching procedure relies on correlations derived from approximately 200 seed mappings.⁵ Figure 4 shows a small subset of these mappings.
- For each Chinese-English entry associated with the HowNet concept, assign the highest ranking candidate EVCA class.

Consider the case of the multiply ambiguous Chinese verb 拉 (1a). Two of the HowNet concepts associated with this verb are |Help| and |Transport|. The thematic grid associated with |Help| is (agent,patient,scope) (as in *John helped him with his work*). This grid most closely matches that of the Equip EVCA Class (where 拉 (1a) is translated as *help*) which has the grid _ag_th,mod-poss(with); thus, the |Help| HowNet concept is associated with the Equip EVCA Class, and the mapping between the two is (agent->ag), (patient->th), (scope->mod-poss).

On the other hand, the |Transport| HowNet concept is associated with the thematic grid (agent,patient,LocationIni,LocationFin,direction) (as in *John transported the goods from Boston to New York (westward)*). This grid most closely matches that of the Send EVCA Class (where 拉 (1a) is translated as *transport*); thus, the |Transport| HowNet concept is associated with the Send EVCA class, and the mapping between the two is (agent->ag), (patient->th), (LocationIni->src), (LocationFin->goal). The end result is that the English glosses associated with 拉 (1a) are filtered down to *help* in the EVCA’s Equip class and *transport* in EVCA’s Send class; the corresponding semantic representations are assigned from the EVCA database.

The massive filtering of spurious assignments is evident when we examine each individual HowNet concept. Consider the |Establish| HowNet concept. This concept is ultimately associated with only two EVCA classes, 29.2.c and 26.4.a (Characterize and Create), but it initially had 29 potential EVCA class assignments. One EVCA class that was ruled out is the Change of State class, 45.4.a, associated with the Optilex translation *colonize* for the Chinese verb

⁵ The seed mappings were done by hand at a rate of approximately 50 mappings per hour; these were verified by a native Chinese speaker in a half day.

HowNet Roles	EVCA-Based Roles														
	ag	th	exp	goal	src	perc	loc	info	pred	prop	Instr	Poss	Pred	Purp	Ben
agent	278	77	32	1	2	3	0	0	0	0	4	7	0	11	4
beneficiary	0	0	0	0	0	0	0	0	0	0	0	0	0	0	4
content	0	31	1	2	2	14	0	20	3	6	3	0	1	3	1
experiencer	13	32	33	0	0	0	0	0	0	0	0	0	0	0	0
patient	0	122	7	7	0	8	0	0	0	0	0	0	0	0	0
source	0	4	0	0	16	0	0	0	0	0	0	0	0	0	1
target	0	7	12	27	1	17	0	0	0	3	0	2	0	0	1

Fig. 4. Seed Table for mapping HowNet Roles into EVCA Roles

殖民 (zhimin). Although this is a perfectly valid EVCA class assignment for the HowNet concept |Colonize|, it is not appropriate for the |Establish| HowNet concept. Because this class is ranked 8th for |Establish|—as opposed to 1st and 2nd place ranking for 29.2.c and 26.4.a, respectively—this assignment is ruled out by our algorithm.

4 Compensating for Resource Deficiencies

As part of our effort to produce a complete alignment between HowNet and EVCA, we built an EVCA-based canonical specification for each of the 709 HowNet concepts so that we could compensate for certain types of resource deficiencies. The canonical specification consists of an EVCA class coupled with its associated prototype verb. These canonical specifications provide a mapping between a HowNet concept and an EVCA class/prototype-verb pair.

Each canonical specification was automatically generated according to the highest ranking EVCA class using steps 3.a and 3.b in Section 2. All such specifications were hand-verified (at a rate of 80 per hour for 709 classes). In most cases, the prototype verb names the HowNet concept, e.g., *transport* for the |Transport| HowNet concept. In other cases—where the HowNet concept is not an English word—the prototype word is a realization of that concept, e.g., *belittle* for the |PlayDown| HowNet concept. A sample of the canonical specifications is given in Figure 5.

We use these canonical specifications to compensate for gaps that arise in our three online resources: (1) EVCA, (2) Optilex, and (3) HowNet.

4.1 EVCA Gaps

An EVCA gap is detected when an Optilex verb gloss for a Chinese verb does not occur in EVCA. When this occurs, the canonical specification for the Chinese verb is automatically used to assign the verb an appropriate EVCA class. For example, one Optilex gloss associated with the HowNet concept |Establish| (for the verb 重建 (chongjian)) is *reconstruct*, which does not occur in EVCA. Our technique associates this Chinese verb with the canonical specification “29.2.c Characterize, *establish*,” and the Chinese verb is then linked with the word sense associated with *establish*.

HowNet Concept	Canonical Specification
Transport	11.1 Send, <i>transport</i>
BeNot	22.2.a Amalgamate, <i>oppose</i>
Help	13.4.2 Equip, <i>help</i>
Moisten	45.4.a Change of State, <i>facilitate</i>
Excrete	40.1.2 Breathe, <i>bleed</i>
Apologize	32.2.a Long, <i>apologize</i>
PlayDown	33.b Judgment, <i>belittle</i>
Naming	29.3 Dub, <i>name</i>
Choose	29.2.c, <i>choose</i>
Announce	37.7.b Say, <i>announce</i>
Mean	37.7.a Say, <i>signify</i>
Communicate	37.9.c Advise <i>inform</i>

Fig. 5. Sample of Canonical Specifications for Filling Resource Gaps

An interesting byproduct of the handling of EVCA gaps is that it allows us to enhance our EVCA resource. For example the verb *reconstruct* can now be added to EVCA Class 29.2.c, on a par with the previously classified EVCA verb *establish*.

4.2 Optilex Gaps

An Optilex gap occurs when a particular translation for a Chinese verb is missing. For example, the verb 摆布 (baibu) has only one Optilex gloss: *manipulate*. However, the word 摆布 is associated with two HowNet concepts, |Decorate| and |Control|. This gloss is only appropriate for the |Control| concept. The *decorate* meaning of 摆布 (baibu) is omitted in Optilex.

Such gaps are detected by means of two types of information: (1) HowNet and EVCA thematic grid; and (2) correlations between the gloss under question and other HowNet concepts. In this particular example, the thematic grid for *manipulate* in EVCA is (ag, exp, instr), which is ranked low (11th out of 28) with respect to the roles (agent, patient) associated with the HowNet |Decorate| concept. By contrast, this same EVCA class has a high ranking (2nd out of 22) with respect to the HowNet |Control| concept due to a close match between (ag, exp, instr) and the HowNet thematic roles (agent, patient, ResultEvent). In addition, the correlation of the gloss *manipulate* is much higher for HowNet's |Control| concept than it is for HowNet's |Decorate| concept (4 occurrences compared to 0). From these two types of information, we can conclude that the *decorate* sense of 摆布 (baibu) is missing from Optilex. As in the case with EVCA gaps, our technique associates the Chinese verb with the canonical specification "9.8.b Fill, *decorate*" to compensate for this Optilex gap.

In addition to their usefulness in handling of gaps in our lexical resources, the canonical specifications proved useful for assigning EVCA classes to Chinese

verbs whose Optilex gloss was not “parsable” by our gloss extraction procedure. For example, the Chinese verb 挨打 (aida) has only a single Optilex translation: *take a beating*. This verb is associated with the HowNet concept |Suffer|, which has as its canonical specification “31.3.d Marvel, *suffer*.” Thus, our technique associates 挨打 verb with this canonical specification.

A similar approach is used for unknown or misspelled words. For example, the translation of 输送 (shusong) as in Optilex is misspelled as *transport*. Because this verb is associated with HowNet’s |Transport| concept, we associated this verb with the canonical specification “11.1 Send, *transport*.”

4.3 HowNet Gaps

In some cases, the HowNet hierarchy incorrectly associates a Chinese word with a particular concept. For example, HowNet incorrectly associates the two Chinese verbs 扎花 (zhahua) and 绣花 (xiuhua) with the |Decorate| concept. These two verbs are translated as *embroider* in EVCA class 26.1.b (Build), but their meaning is closer to *sew flowers*. That is, the patient is incorporated into the verb, which means the thematic grid `_ag_th_goal(into),ben(for)` does not match that of the HowNet concept (`agent, possession, source`).

Discrepancies in HowNet are detected by means of EVCA-class frequency for a particular HowNet concept. Out of the 17 verbs associated with HowNet’s |Decorate| concept, only two of them (the two miscategorized Chinese verbs) are associated with an EVCA class that is not 9.9 or 9.8. As in the gap-recovery described approaches above, our technique associates the miscategorized verbs with the canonical specification “9.8.b Fill, *decorate*.”⁶

5 Results

Preliminary results of our classification scheme were reported in [6]. This earlier work resulted in 8089 EVCA-classified Chinese entries—about 43% of the number of potential entries. The remaining 10441 entries were accounted for through the compensation techniques described above. Using the canonical specifications, we have achieved a more refined EVCA-to-HowNet mapping, providing an increase in EVCA-classified Chinese words from the previous 8089 entries to the current expanded set of 17284 EVCA-classified Chinese words. The histogram in Figure 6 characterizes the number of EVCA classes required for coverage of 709 HowNet concepts.

Examples of the HowNet partitionings into EVCA classes are given in Figure 7, with a focus on the cases where 1 partition was found. Percentages are given with respect to the number of Chinese verbs associated with each EVCA class.

⁶ Ultimately, the miscategorized verbs should be disassociated from the HowNet concept, but there is currently no way to tease apart such cases from the Optilex gaps. Thus, the two are treated identically.

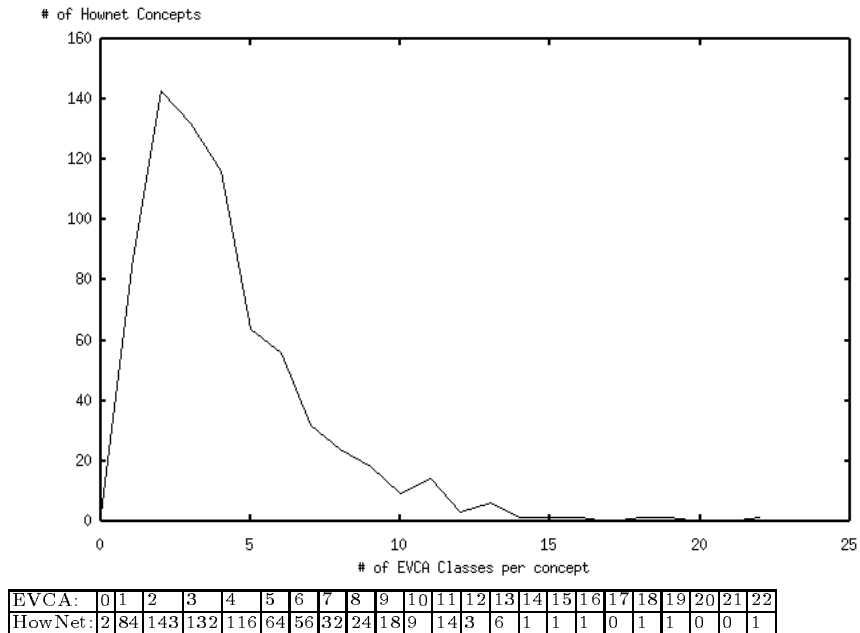


Fig. 6. Distribution of HowNet Concepts by Number of Intersecting EVCA Classes using Canonical Specifications

We consider the approach to be a success for several reasons: (1) In 359 cases (50% of the HowNet concepts), the partitioning corresponded to 3 or fewer EVCA classes; (2) Most HowNet concepts with 2 or more partitions had a very heavy association with a single EVCA class (60% or higher), with most other partitions falling around 20% or lower; (3) Only 2 cases did not correspond to any EVCA class (i.e., degenerate HowNet concepts for which no correlations with EVCA could be found); (4) There were virtually no partitionings (a handful of single HowNet concepts) exceeding 13 EVCA classes.

6 Summary and Future Work

We have presented an approach to aligning two large-scale online resources, HowNet and EVCA. The lexicon resulting from this approach is large-scale, containing 18530 Chinese entries. The technique for producing these links involves matching thematic grids in HowNet with those in EVCA. Our results indicate that the correspondence is very high between the 709 Chinese HowNet concepts and the 500 EVCA classes. We see our techniques as the first step toward a general approach to building repositories for interlingual-based NLP applications.

HowNet Concept	EVCA Class(es)
Transport	11.1 Send
Help	13.4.2 Equip
Apologize	32.2.a Long
Naming	29.3 Dub
Judge	29.4 Declare
Moisten	45.4.a Change of State
Excrete	40.1.2 Breathe
TakeVehicle	51.4.2.a.ii Motion by Vehicle
PlayDown	33.b Judgment (75%), 31.2.a Admire (25%)
Establish	29.2.c Characterize (90%), 26.4.a Create (19%)
Decorate	9.8.b Fill (50%), 26.1.b Build (43%), 9.9.ii Butter (25%)
Buy	10.5 Steal (08%), 13.5.1.a Get (30%), 13.5.1.b.ii Get (54%), 13.5.2.d Get (46%)
Teach	29.2.c Characterize (24%), 33.b Judgment (71%), 37.9.a Advise (29%), 37.1.a Transfer Message (45%), 31.1.a Amuse (19%)

Fig. 7. Examples of HowNet Partitionings with Respect to EVCA

We are currently investigating the use of the lexicon for word-sense disambiguation in machine-translation and cross-language information retrieval. As we saw above the Chinese verb 拉 (la) has several possible translations, but not all of these will be appropriate in every context. If we can determine which HowNet concept corresponds to 拉 (la), then we will translate it appropriately. For example, if the HowNet concept is |Transport|, the translation would be *ship* or *transport*, but not *slash*, *chat*, *implicate*, etc. We can detect which HowNet concept is appropriate by examining the other words in the sentence. If those words co-occur with *other* Chinese verbs associated with a particular HowNet concept (as determined through a corpus analysis), then it is likely that that HowNet concept is the appropriate one for the Chinese verb. That is, if we find other verbs from a given HowNet concept occurring in the same context, then we can hypothesize that this particular verb has the meaning of this HowNet concept.

The algorithm for mapping between HowNet concepts and EVCA classes requires a “training” step—i.e., the seed mappings given earlier. However, it is possible to produce a ranked mapping between thematic grids by counting correspondences between EVCA-based roles and the HowNet-based roles across the entire concept space. This approach is also currently under investigation.

Another area of investigation is the use of a WordNet-based distance metric (e.g., the information-content approach of [20]) for additional pruning power in the HowNet-to-EVCA alignment. Because each of the entries in the EVCA classification is associated with a WordNet sense [14], it is possible to rule out certain class assignments for a given HowNet concept by examining semantic distance between the Optilex glosses for a particular Chinese word and the glosses for other words associated with that concept.

Acknowledgements

The University of Maryland authors are supported, in part, by PFF/PECASE Award IRI-9629108, DOD Contract MDA904-96-C-1250, and DARPA/ITO Contract N66001-97-C-8540. Dekang Lin is supported by Natural Sciences and Engineering Research Council of Canada grant OGP121338. We are indebted to an astute anonymous reviewer for catching several errors and inconsistencies in the original draft. We also thank Nizar Habash, Maria Katsova, and Scott Thomas for their assistance with experimental runs on the data and their useful commentary and aid in the preparation of this document.

References

1. Hoa Trang Dang, Karin Kipper, Martha Palmer, and Joseph Rosenzweig. Investigating Regular Sense Extensions Based on Intersective Levin. In *ACL/COLING 98, Proceedings of the 36th Annual Meeting of the Association for Computational Linguistics* (joint with the 17th International Conference on Computational Linguistics), pages 293–299, Montreal, Canada, August 10-14 1998.
2. Bonnie J. Dorr. Large-Scale Acquisition of LCS-Based Lexicons for Foreign Language Tutoring. In *Proceedings of the ACL Fifth Conference on Applied Natural Language Processing (ANLP)*, pages 139–146, Washington, DC, 1997.
3. Bonnie J. Dorr, Nizar Habash, and David Traum. A Thematic Hierarchy for Efficient Generation from Lexical-Conceptual Structure. In *Proceedings of the Third Conference of the Association for Machine Translation in the Americas, AMTA-98, in Lecture Notes in Artificial Intelligence, 1529*, pages 333–343, Langhorne, PA, October 28-31 1998.
4. Bonnie J. Dorr and Douglas Jones. Acquisition of semantic lexicons: Using word sense disambiguation to improve precision. In Evelyne Viegas, editor, *Breadth and Depth of Semantic Lexicons*. Kluwer Academic Publishers, Norwell, MA, 1999.
5. Bonnie J. Dorr and Maria Katsova. Lexical Selection for Cross-Language Applications: Combining LCS with WordNet. In *Proceedings of the Third Conference of the Association for Machine Translation in the Americas, AMTA-98, in Lecture Notes in Artificial Intelligence, 1529*, pages 438–447, Langhorne, PA, October 28-31 1998.
6. Bonnie J. Dorr, Gina-Anne Levow, Dekang Lin, and Scott Thomas. Chinese-English Semantic Resource Construction. In *Proceedings of the 2nd International Conference on Language Resources and Evaluation (LREC2000)*, Athens, Greece, May 31-June 2 2000.
7. Douglas Jones, Robert Berwick, Franklin Cho, Zeeshan Khan, Karen Kohl, Naoyuki Nomura, Anand Radhakrishnan, Ulrich Sauerland, and Brian Ulicny. Verb Classes and Alternations in Bangla, German, English, and Korean. Technical report, Massachusetts Institute of Technology, 1994.
8. Irene Langkilde and Kevin Knight. Generating Word Lattices from Abstract Meaning Representation. Technical report, Information Science Institute, University of Southern California, 1998.
9. Irene Langkilde and Kevin Knight. Generation that Exploits Corpus-Based Statistical Knowledge. In *Proceedings of COLING-ACL '98*, pages 704–710, 1998.
10. Irene Langkilde and Kevin Knight. The Practical Value of N-Grams in Generation. In *International Natural Language Generation Workshop*, 1998.

11. Beth Levin. *English Verb Classes and Alternations: A Preliminary Investigation*. University of Chicago Press, Chicago, IL, 1993.
12. Gina-Anne Levow, Bonnie Dorr, and Maria Katsova. Construction of Chinese-English Semantic Hierarchy for Cross-Language Retrieval. In *Proceedings of the Workshop on English-Chinese Cross Language Information Retrieval, International Conference on Chinese Language Computing*, Chicago, IL, October 10-14 2000.
13. D. Lin. Dependency-based Evaluation of MINIPAR. In *Proceedings of the Workshop on the Evaluation of Parsing Systems, First International Conference on Language Resources and Evaluation*, Granada, Spain, May 1998.
14. George A. Miller and Christiane Fellbaum. Semantic Networks of English. In Beth Levin and Steven Pinker, editors, *Lexical and Conceptual Semantics, Cognition Special Issue*, pages 197–229. Elsevier Science Publishers, B.V., Amsterdam, The Netherlands, 1991.
15. Naoyuki Nomura, Douglas A. Jones, and Robert C. Berwick. An architecture for a universal lexicon: A case study on shared syntactic information in Japanese, Hindi, Ben Gali, Greek, and English. In *Proceedings of COLING-94*, pages 243–249, Kyoto, Japan, August 5-9 1994.
16. Mari Broman Olsen, Bonnie J. Dorr, and Scott C. Thomas. Enhancing Automatic Acquisition of Thematic Structure in a Large-Scale Lexicon for Mandarin Chinese. In *Proceedings of the Third Conference of the Association for Machine Translation in the Americas, AMTA-98, in Lecture Notes in Artificial Intelligence, 1529*, pages 41–50, Langhorne, PA, October 28-31 1998.
17. Martha Palmer and Joseph Rosenzweig. Capturing motion verb generalizations with synchronous tags. In *Proceedings of the Second Conference of the Association for Machine Translation in the Americas*, Montreal, Quebec, Canada, 1996.
18. Martha Palmer and Zhibao Wu. Verb Semantics for English-Chinese Translation. *Machine Translation*, 10(1–2):59–92, 1995.
19. P. Procter. *Longman Dictionary of Contemporary English*. Longman, London, 1978.
20. Philip Resnik. Using information content to evaluate semantic similarity in a taxonomy. In *Proceedings of IJCAL-95*, pages 448–453, Montreal, Canada, August 20-25 1995.
21. Patrick Saint-Dizier. Semantic Verb Classes Based on 'Alternations' and on WordNet-like Semantic Criteria: A Powerful Convergence. In *Proceedings of the Workshop on Predicative Forms in Natural Language and Lexical Knowledge Bases*, pages 62–70, Toulouse, France, 1996.
22. Amy Weinberg, Joseph Garman, Jeffery Martin, and Paola Merlo. Principle-Based Parser for Foreign Language Training in German and Arabic. In Melissa Holland, Jonathan Kaplan, and Michelle Sams, editors, *Intelligent Language Tutors: Theory Shaping Technology*, pages 23–44. Lawrence Erlbaum Associates, Hillsdale, NJ, 1995.
23. Dong Zhendong. Enlightenment and Challenge of Machine Translation. *Shanghai Journal of Translators for Science and Technology*, 1:9–15, 1988.
24. Dong Zhendong. Knowledge Description: What, How and Who? In *Proceedings of International Symposium on Electronic Dictionary*, page 18, Tokyo, Japan, 1988.
25. Dong Zhendong. MT Research in China. In *Proceedings of International Conference on New Directions in Machine Translation*, pages 85–91, Budapest, 1988. Also in *New Directions in Machine Translation*, 4 Distributed Language Translation edited by Dan Maxwell, Klaus Schubert and Toon Witkam, Foris Publications, Dordrecht.