Large Scale Language Independent Generation Using Thematic Hierarchies

Nizar Habash and Bonnie Dorr
Institute for Advanced Computer Studies
University of Maryland
College Park, MD 20740
phone: +1 (301) 405-6768
fax: +1 (301) 314-9658
habash,bonnie@umiacs.umd.edu
http://umiacs.umd.edu/labs/CLIP

Abstract
This paper describes a large-scale language-independent evaluation of the use of Thematic Hierarchies in natural language generation. We translate from a corpus of sentences reflecting the full variety of behavior of Levin-based verb classes. The corpus is used as input to a generation system that utilizes the same thematic hierarchy for realizing relative argument surface positions in two languages: English and Spanish. The output was manually evaluated by English and Spanish speakers. The contributions of this work include: (1) an improved thematic hierarchy over an earlier implementation; (2) a large-scale evaluation of the use of thematic hierarchies in two languages; (3) an implementation of a language-independent module for natural language generation; and (4) the creation of a single tool for incremental development of multilingual lexicons.

1 Motivation
In (Dorr et al., 1998), an implementation of thematic hierarchies for efficient natural language generation was presented. The use of the thematic hierarchy was evaluated using a small hand-constructed corpus of 100 English sentences reflecting a variety of English verb classes and alternations. The hierarchy was implemented using cascading rules within the grammar formalism provided as part of the natural language realization engine Nitrogen (Langkilde and Knight, 1998a; Langkilde and Knight, 1998b). Some of the shortcomings of this earlier work are: (1) inadequate evaluation due to the use of a small test corpus; (2) limitation of the approach to one language only (English); (3) lack of a principled design in the implementation.

This paper presents more systematic implementation of thematic hierarchies and a large-scale evaluation of their use for generation in English and Spanish. This evaluation was helpful in incremental development of both the thematic hierarchy and the English and Spanish lexicons.

2 Research Context
The work presented here is part of the generation component (Traum and Habash, 2000) of the interlingual Machine Translation effort at the University of Maryland College Park. The generation component has also been used in Cross-Language Information Retrieval research (Levow et al., 2000). The interlingual representation used is Lexical Conceptual Structure (LCS), a compositional abstraction with language-independent properties that transcend structural idiosyncrasies (Jackendoff, 1983; Jackendoff, 1990; Jackendoff, 1996). This representation has been used as the interlingua of several projects such as UNITRAN (Dorr et al., 1993) and MILT (Dorr, 1997).

3 Overview of Generation in LCS-based Machine Translation
One of the major challenges in natural language processing is the ability to make use of existing resources. Large differences in syntax, semantics, and ontologies of such resources create significant barriers to their usage in large-scale applications. A case in point is the wide range of “interlingual representations” used in machine translation and cross-language processing. Such representations are becoming increasingly prevalent, yet views vary widely as to what these should be composed of, varying from purely conceptual knowledge-representations, having little to do with the structure of language, to very syntactic representations, maintaining most of the idiosyncrasies of the source languages. In our generation system we make use of resources associated with two different (kinds of) interlingua structures: Lexical Conceptual Structure (LCS), and the Abstract Meaning Representations (AMR) used at USC/ISI (Langkilde and Knight, 1998a). The two representations serve different but complementary roles in the translation process. The deeper lexical-semantic expressiveness of LCS is essential for language independent Lexical Selection that transcends translation divergences. The shallower yet mixed semantic-syntactic nature of AMRs makes it easier to use for target language realization.
The use of two representations in generation mirrors the use of two representations on the analysis side of the MT system, in which a parsing output is passed to a semantic-composition module: the target-language AMR is analogous to the source-language parse tree. (See Figure 1.) The Composition module takes the source-language parse tree and creates a deeper semantic representation (the LCS) using a source-language lexicon. In generation, the Decomposition module performs a reverse step that uses a target-language lexicon to create the parse-like AMR. This step is referred to as Lexical Selection. It is followed by the Realization step in which the Linearization module flattens an AMR into a sequence of words. Because of the ambiguity inherent in all of the involved modules from the parser to the lexicons, multiple sequences are created. We use the Statistical Extraction module of the generation system Nitrogen (Langkilde and Knight, 1998a; Langkilde and Knight, 1998b) to select among alternative outputs when generating English.

### 3.1 LCS Lexicons

The LCS lexicons used in both analysis and generation relate a lexeme to a Lexical Conceptual Structure representation. A single verb might have several entries corresponding to different senses of that verb. Figure 2 compares four out of the nine root LCS (RLCS) entries for the verb ‘run’ in the English LCS Lexicon. These entries are classified by their Levin verb class which is used as a template to generate the RLCSes for every verb in the class. The star-marked nodes in those entries signify the location an argument can be attached. A composed LCS (CLCS) is made up of a RLCS that has its star-marked nodes filled with other CLCSes. The number at the end of the nodes mark the thematic role associated with the specific node. For example, 1 is agent, 2 is theme, 3 is a source particle (i.e. an oblique) and 4 is source (an argument). For a full listing of the thematic roles and their corresponding codes see Figure 3. The last LCS entry for run in Figure 2 can be read as a *thing goes locationally from a source location to a goal location in a running manner*.

The current English verb lexicon contains over 11,000 RLCS entries such as those in Figure 2. These entries correspond to different senses of over 4,000 verbs. The Spanish verb lexicon contains over 24,000 entries corresponding to 3,300 verbs. The LCS lexicon also contains other information of importance to realization such as requirements for optionality (OPTIONAL and OBLIGATORY) and internal/external positioning (:INT and :EXT). Optionality markers are necessary to determine which arguments must be available in the CLCS for proper generation using an RLCS. For example, in class 51.3.2.ai in Figure 2, the theme is the only obligatory argument. Internal/external positioning markers will be discussed later in the section on Thematic Hierarchy.

### 3.2 Lexical Selection

The lexical selection process attempts to decompose a CLCS into RLCSes corresponding to lexemes in the target language. Decomposition is basically a complex algorithm for graph matching/covering with restrictions. Its output is the shallower Abstract Meaning Representation (AMR) discussed earlier. Different lexicons for different languages provide different RLCSes and RLCS restrictions that guide lexical selection. Figure 4 compares three different possible decompositions for a CLCS into English, Spanish, and Arabic. The CLCS can be read as *John cases himself to go into a room in a forcible manner*. The AMR relation (:AG, :TH, etc.) marking the connections on the left-hand side in Figure 4 are created from the thematic role information in the RLCSes.

### 3.3 Realization

Syntactic realization is the step that converts the unordered dependency tree structure of an AMR into a surface sentence. There are two operations involved in realization: recasting and linearization. Recasting converts an AMR node into another AMR node with added information, deleted information, or just modified information. Linearization specifies the relative positions of the children of an AMR node to their mother and to each other. The focus of this paper is on the specific linearization submodule that deals with the problem of mapping thematic roles to surface positions.

In (Dorr et al., 1998), the grammar formalism provided as part of the natural language realiza-

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1For a detailed discussion of the acquisition of LCS-based lexicons, see (Dorr and Osen, 1996; Dorr, 1997).
tion engine Nitrogen (Langkilde and Knight, 1998a; Langkilde and Knight, 1998b) was used to implement a linearization grammar. The Nitrogen grammar formalism is unification based and it provides a small number of tools to recast and linearize AMRs. There are several limitations to the use of this formalism. First, the grammar is interpreted which results in inefficient time-space use. Second, the tools provided are rather simple transformations which causes the linearization grammars to be long and complex. Currently we are using a different linearization engine, Oxygen (Habash, 2000). Oxygen is an efficient language-independent linearization engine. Linearization grammars for Oxygen are written using oxyL, a powerful linearization grammar description language that has the power of a programming language with the focus on natural language linearization. oxyL grammars are compiled into programs that run independently.

The power of oxyL is accomplished by providing recasting mechanisms for the most common needs of a linearization grammar and also by allowing embedding of code in a standard programming language (Lisp). The oxyL linearization grammars are also simple, clear, concise and easily extendible. The simplicity of oxyL grammars is apparent when one considers issues of redundancy: the handling of ambiguities at every phrase rule is hidden from the linearization grammar designer and is treated only in the compiler and support library. For a detailed presentation of oxyL’s syntax, see (Habash, 2000). An example of a segment of an oxyL linearization grammar is provided in 5 and will be explained in the next section.

![Table 1: Inventory of Thematic Roles]

<table>
<thead>
<tr>
<th>#</th>
<th>Thematic Role</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>no thematic role assigned agent</td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>AG</td>
<td>theme or experiencer or information</td>
</tr>
<tr>
<td>2</td>
<td>TH, EXP</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>SRC()</td>
<td>source preposition</td>
</tr>
<tr>
<td>4</td>
<td>SRC</td>
<td>source</td>
</tr>
<tr>
<td>5</td>
<td>GOAL(), PRED()</td>
<td>goal or pred preposition</td>
</tr>
<tr>
<td>6</td>
<td>GOAL</td>
<td>goal</td>
</tr>
<tr>
<td>7</td>
<td>PERC()</td>
<td>perceived item particle</td>
</tr>
<tr>
<td>8</td>
<td>PERC</td>
<td>perceived item</td>
</tr>
<tr>
<td>9</td>
<td>PRED</td>
<td>identificational predicate</td>
</tr>
<tr>
<td>10</td>
<td>LOC()</td>
<td>locational particle</td>
</tr>
<tr>
<td>11</td>
<td>LOC</td>
<td>locational predicate</td>
</tr>
<tr>
<td>12</td>
<td>POSS</td>
<td>possessional predicate</td>
</tr>
<tr>
<td>13</td>
<td>TIME()</td>
<td>temporal particle preceding time</td>
</tr>
<tr>
<td>14</td>
<td>TIME</td>
<td>time for TEMP field</td>
</tr>
<tr>
<td>15</td>
<td>MOD.Poss()</td>
<td>possessional particle</td>
</tr>
<tr>
<td>16</td>
<td>MOD.Poss</td>
<td>possessed item modifier</td>
</tr>
<tr>
<td>17</td>
<td>BEN()</td>
<td>beneficiary particle</td>
</tr>
<tr>
<td>18</td>
<td>BEN</td>
<td>benefactive modifier</td>
</tr>
<tr>
<td>19</td>
<td>INSTR()</td>
<td>instrumental particle</td>
</tr>
<tr>
<td>20</td>
<td>INSTR</td>
<td>instrument modifier</td>
</tr>
<tr>
<td>21</td>
<td>PURP()</td>
<td>purpose particle</td>
</tr>
<tr>
<td>22</td>
<td>PURP</td>
<td>purpose modifier or reason</td>
</tr>
<tr>
<td>23</td>
<td>MOD.LOC()</td>
<td>location particle</td>
</tr>
<tr>
<td>24</td>
<td>MOD.LOC</td>
<td>location modifier</td>
</tr>
<tr>
<td>25</td>
<td>MANNER()</td>
<td>manner</td>
</tr>
<tr>
<td>26</td>
<td></td>
<td>reserved for conflated manner</td>
</tr>
<tr>
<td>27</td>
<td>PROP</td>
<td>event or state</td>
</tr>
<tr>
<td>28</td>
<td>MODPROP</td>
<td>event or state</td>
</tr>
<tr>
<td>29</td>
<td>MOD.PRED()</td>
<td>identificational particle</td>
</tr>
<tr>
<td>30</td>
<td>MOD.PRED</td>
<td>property modifier</td>
</tr>
<tr>
<td>31</td>
<td>MOD.TIME</td>
<td>time modifier</td>
</tr>
</tbody>
</table>
4 The Thematic Hierarchy

The unordered nature of siblings under an AMR node complicates the mapping between AMR relations and their surface positions. In the case of thematic role ordering, the situation is more complicated by the lack of one-to-one mapping between a particular thematic role and an argument position. For example, a theme can be the subject in some cases and it can be the object in others or even an oblique. Observe cookie in (1).

(1) (i) John ate a cookie (object)
   (ii) the cookie contains chocolate (subject)
   (iii) she nibbled at a cookie (oblique)

To solve this problem, a thematic hierarchy is used to determine the argument position of a thematic role based on its co-occurrence with other thematic roles. Several researchers have proposed different versions of thematic hierarchies (see (Jackendoff, 1972; Carrier-Duncan, 1985; Bresnan and Kanerva, 1989; Kiparsky, 1985; Larson, 1988; Giorgi, 1984; Wilkins, 1988; Nishgauchi, 1984; Alsina and Mchombo, 1993; Baker, 1989; Grimshaw and Mester, 1988)).\(^2\) The hierarchy proposed in (Dorr et al., 1998) differs from these in that it separates (non-adjunct) arguments from obliques (i.e., adjunct arguments) and provides a more complete list of thematic roles (31 roles overall) than those of previous approaches (maximum of 8 roles). See Figure 3 for a complete listing for the thematic roles used. The following is final thematic hierarchy for arguments.

(2) special case -- ag src th
\[ ext > ag > instr > th > perc > \]

In the case of the occurrence of theme alone, it is mapped to first argument position. If a theme and an agent occur, the agent is mapped to first argument position and the theme is mapped to second argument position. When an agent, a theme and a source co-occur, The order in the hierarchy is violated as in John\(_{ag}\) charged the Pau\(_{perc}\) \(\$$\)\(\$$\). The term ext is used to handle verbs that violate the thematic hierarchy. It, ext, refers to an externally marked thematic role such as the perceived John in John\(_{perc}\) please Mary\(_{th}\) versus Mary\(_{th}\) likes John\(_{perc}\). This information is provided in the RLCS lexicon entry using the special marker :EXT.

As for the ordering of obliques, an ad hoc order was established:

(3) particle > mod-prop() > perc() > th() > purp() > mod-loc() > mod-pred() > src() > goal() > mod-poss() > ben()

Note that the order of obliques is not a strict hierarchy but rather a possible topological sort. A more detailed discussion is available in (Dorr et al., 1998).

4.1 Thematic Hierarchy Implementation

Oxygen’s linearization grammar description language, oxyL, provides a variety of powerful recasting mechanisms, especially a hierarchical data recasting operator that simplifies the implementation of thematic hierarchy mapping.\(^3\) An earlier implementation using Nitrogen’s grammar formalism in (Dorr et al., 1998) used several cascading rules to implement what oxyL allows in two embedded recast rules (see Figure 5). The top part of Figure 5 defines the thematic hierarchy ordering as follows: given the current node (:this), conditionally recast (\(<?\)) the relation :src into :src if it co-occurs with :ag and :th; then hierarchically recast (\(<!\)) all available

\(^2\)For an excellent overview and a comparison of different thematic hierarchies see (Levin and Rappaport Hovav, 1996).

\(^3\)Another example of hierarchically ordered linguistic phenomenon the linearization of auxiliaries relative to the negative particle in the English verb phrase. The auxiliaries are strictly ordered by the part of speech (Modal Have Be\{-en\} Be\{-ing\). The negative particle ‘not’ must appear after the first auxiliary regardless of its part of speech. A hierarchical mapping of the auxiliaries into (Aux1 Aux2 Aux3 and Aux4) is a simpler solution than listing all combinations.
Figure 5: Oxyl Implementation of the thematic hierarchy

<table>
<thead>
<tr>
<th>Verb Class</th>
<th>Example</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>something$_ag$ wanted something$_th$ (to do something$_th$)$_prop$</td>
</tr>
<tr>
<td>10.5</td>
<td>someone$_ag$ stole something$_th$ from something$_ag$ for something$_ben$</td>
</tr>
<tr>
<td>22.1.C</td>
<td>someone$_ag$ mixed something$_th$ into something$_ag$ for something$_goal$</td>
</tr>
<tr>
<td>29.1.B</td>
<td>someone$_th$ considered something$_ag$ to be someproperty$_prop$ for something$_int$</td>
</tr>
<tr>
<td>45.2.A</td>
<td>someone$_ag$ folded something$_th$ with something$_int$</td>
</tr>
<tr>
<td>55.1.C</td>
<td>someone$_th$ continued (to do something$_th$)$_prop$</td>
</tr>
</tbody>
</table>

Table 1: CLCS Test Corpus Examples

5 Evaluation

In this evaluation, a test corpus of 453 simple CLCSes corresponding to all Levin English verb classes and alternations was constructed semi-automatically. The test corpus size guarantees large-scale coverage over verb behavior and thematic role combinations, which is exhaustive for our purpose. The CLCSes were constructed by randomly selecting an LCS verb entry from each class from the English verb class and filling its argument positions with simple noun phrases (e.g., something$_th$, someone$_ag$, etc.) or simple subordinate clauses (e.g., (to be someproperty)$_prop$, (to be someproperty)$_mod-prop$, etc.). Table 1 shows some sample English sentences corresponding to the CLCSes in the test corpus.

For the purposes of this evaluation, statistical extraction was disabled because we do not have a Nitrogen bigram model for Spanish.

The CLCS test corpus was fed to the generation system in two different runs each of which using a different target language lexicicon and oxyl linearization grammar. The results of the generation are passed to two speakers of English and Spanish respectively to evaluate the word order of the realized text. Evaluators were asked to mark sentences as being acceptable or not acceptable as far as the word order of the arguments relative to the verb. Some of the English and Spanish sentences failed the lexical selection process due to problems with lexicon entries; these sentences never made it to linearization.

In the cases that survived, the lexical selection process appropriately generated multiple sentences for each CLCS. In the case of English, they all correctly corresponded to various related alternations.
of the main verb. For example, each of the two sub-classes defining the dative alternations for the verb *send* generated each other (i.e. *John sent a book to Paul* and *John sent Paul a book*). There were also cases of over-generation resulting from preposition under-specification, which is inconsequential to our evaluation (e.g. *go (to, toward, towards, to at, etc.) somewhere)*.

On the other hand, in Spanish, there were many more sentences than should not have been generated. In theory, the lexical selection process limits the number of choices using the LCS entry of the Spanish verbs. But that process is only as good as the lexicon entries are. In cases where a bad sense is allowed in the translation, the sentence involved is dropped from the evaluation. This evaluation was quite helpful in pinpointing the locations of problems in our Spanish (and also English) Lexicons. Table 2 displays the results of the evaluation. The first column represents the percentage of generated classes/LCS instances (out of N = 453) that actually went through the whole system. The second column represents the ratio of discarded sentences because of wrong sense selection. This number is normalized over the classes since some classes over-generated incredibly. So even if a small percentage in the class was wrong, that wrong subset was large enough to cancel out the many classes that generated a few correct instances. The final column describes the normalized ratio of instances with wrong word order. Clearly, The results show that the use of a thematic hierarchy for generating both English and Spanish word order is quite successful and is supportive of earlier work (Dorr et al., 1998). The rest of this section will describe the errors encountered in the evaluation and how they were fixed.

<table>
<thead>
<tr>
<th>N = 453</th>
<th>Generated Classes</th>
<th>Sense Error</th>
<th>Word Order Error</th>
</tr>
</thead>
<tbody>
<tr>
<td>English</td>
<td>94%</td>
<td>0%</td>
<td>3%</td>
</tr>
<tr>
<td>Spanish</td>
<td>61%</td>
<td>22%</td>
<td>2%</td>
</tr>
</tbody>
</table>

Table 2: Evaluation Results

6 Discussion

The word-order errors in the English test belong to one of two types: First, there are lexicon errors where specific realization information such as :EXT is missing from an entry. This problem appeared in three subclasses of class 413.1 (Simple Verbs of Dressing: don, doff and wear). In our lexicon, *clothes*, the object for all three verbs, is considered the theme and the subject of the sentence is the goal, source and location respectively. Fixing these cases is a matter of adding the appropriate piece of information in the lexicon. The second type of errors were true thematic hierarchy errors: The case of agent-benefactor-theme co-occurrence such as *John bought Paul a house* and agent-goal-theme co-occurrence such as *John gave Paul a house*. These two should be part of the special case of the thematic hierarchy that deals with English verbs’ indirect objects. Figure 6 displays the updated thematic hierarchy for English. In this implementation, a temporary role :MOV is created to mark source, goal or benefactor as moved arguments in a special conditional recasting step that depends on the co-occurrence of any of these roles with an agent and a theme.

The Spanish errors are much less than the English and are basically a subset of the first type of errors described above. The fact that the special case of the thematic hierarchy for English was included and it did not cause any problem to the Spanish is not surprising since Spanish lexical selection doesn’t allow the thematic roles agent and theme to co-occur with the arguments source, goal or benefactor. The third argument is always generated as an oblique. For example, *Juan le dio un libro a Paolo* and *Juan le compró un libro a Paolo*. The updated thematic hierarchy for Spanish is described in Figure 7.

7 Future Work

A major remaining step is to correct the problems in the English and Spanish Lexicons and to investigate the source of errors and incorrect sense selection. An investigation in the behavior of obliques in Spanish is necessary to produce fully fluent Spanish output. Another topic of interest is the reusability of the thematic hierarchy with other languages that are much

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Figure 6: New Oxyl Implementation of the English thematic hierarchy

```plaintext
:Recast @TH-order (@this <? (:mov / (:src :goal :ben))
  (&and (:ex :ag) (:ex :th))
  <! (:obj :obj1 @obj2))

:Rule %S (-> (@subj @inst @obj1 @obj2))
```

---

5We are aware that a more fluent Spanish would move the oblique (a Paolo) closer to the verb as in *Juan le dio a Paolo un libro* and *Juan le compró a Paolo un libro*. However this is not part of the focus of our evaluation. The behavior of obliques is something we plan to investigate in a separate study.
Figure 7: New OxyIL Implementation of the Spanish thematic hierarchy

:Recast &TH-order
(@this <! ((:subj :obj1 :obj2) /
(:ext :sub :ag :instr :th
:src :perc :goal :mod-pos1
:mod-loc :mod-pred :loc :poss
:pred :prop :time :ben :purp)))

:Rule %S (->(@subj @inst @obj1 @obj2))

more different than Spanish is to English. We are currently investigating Chinese; a preliminary study showed some promising results as far as thematic hierarchy mapping. However Chinese seems to require more complex linearization rules and post-lexical selection manipulations especially for obliques.

8 Acknowledgments

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References


