Inferable Centers, Centering Transitions, and the Notion of Coherence

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A centering analysis of the corpus of Japanese e-mail that is examined in this article relies heavily on the inclusion of inferable centers. However, utilizing this type of center results in a high level of indeterminacy in labeling transitions and thus in characterizing the coherence of the corpus. The difficulty lies in the requirement of identity of discourse entities in the definitions of transition states. Lexical cohesion is proposed as a well-defined notion to replace the intuitions captured by the use of inferable centers. Two new transitions, based on lexical relatedness instead of identity, supplement the standard definitions and more adequately characterize coherence in this corpus. Implications and extensions of the proposal are discussed.

1. Introduction

Centering has been proposed as a model of the local attentional states of speakers and hearers involved in the mutual construction of conversation (Brennan, Friedman, and Pollard 1987; Grosz and Sidner 1986, 1998; Walker 1998). Centering mechanisms are designed to model the coherence of discourse by characterizing transitions between utterances in terms of their inferential load and hence their naturalness. These characterizations are intended to capture intuitions about the “flow” (Chafe 1979) or the “ongoing process of meaning” (Halliday 1994) in discourse.

In this work, we examine a corpus of Japanese e-mail to investigate the mechanisms by which coherence is achieved. Because this corpus contains a high number of discourse elements that are inferable from the discourse context, we have an opportunity to examine the interplay between standard centering transition definitions and the presence of inferable discourse entities. We claim on the basis of intuitions of native speakers that the actual level of coherence in the corpus is much higher than the centering account implies, primarily by virtue of the fact that transitions involving inferable entities are often difficult to specify. We conclude that the standard centering account cannot accurately model the coherence in this corpus. Detailed analysis reveals that one major problem lies in the requirement of identity of discourse elements in adjacent utterances in order for those elements to contribute to coherence. We describe this problem and propose two additions to the usual repertoire of transitions that enable a more authentic account of coherence in this corpus, while remaining within a centering framework.

The article is organized as follows. In Section 2, we briefly describe centering mechanisms and their role in modeling coherence. We go on to outline the features of the corpus in Section 3 and illustrate how standard centering mechanisms characterize transitions and coherence in this corpus, suggesting that these mechanisms are not adequate for the task. In Section 4, we describe more general problems with the...
inclusion of inferable discourse entities in centering theory and propose a revision to the standard set of transitions that more accurately describes the corpus. In this section as well, we explore the implications of this proposal for other areas of discourse analysis. In Section 5, we outline some possibilities for improvement and extension of the proposal, and we conclude in the final section.

2. Centering Mechanisms

The central intuitions of centering concern the relationships among the discourse entities appearing or represented in adjacent utterances in a discourse (Walker, Joshi, and Prince 1998). Each utterance \( U_i \) in a discourse is considered to contain a set of discourse entities called forward-looking centers, or Cfs. These entities are ranked in the Cf list for each utterance according to language-specific ranking principles. We follow, in general, the ordering principles for Japanese given in Walker, Iida, and Cote (1994) (with some adjustments for possessive phrases as noted in example (1)):

\[
\text{Cf ranking for Japanese:} \\
(\text{Grammatical OR } \emptyset) \text{ topic} > \text{empathy} > \text{subject} > \text{object2} > \text{object} > \text{others}
\]

A special member of the Cf list, the backward-looking center, or Cb, represents the “topic”\(^1\) of \( U_i \) and is the highest-ranked Cf on the Cf list of \( U_{i−1} \) which is realized in \( U_i \). In addition, the preferred center of \( U_i \), or Cp, is the highest-ranked Cf in \( U_i \).

Given \( U_i \) and \( U_{i−1} \), then, there are four different ways in which their Cbs and Cps may be related; each of these is defined as a type of transition state (Table 1).

There are two rules in a centering approach:

Rule 1: If some entity in the Cf list for \( U_{i−1} \) is realized as a pronoun in \( U_i \), then so is the Cb for \( U_i \).

Rule 2: Transition states are ordered such that \text{CONTINUE} is most preferred, followed in order by \text{RETAIN}, \text{SMOOTH SHIFT}, and \text{ROUGH SHIFT} (Walker, Joshi, and Prince 1998).

Rule 2 captures the centering intuitions concerning coherence: Utterances that \text{CONTINUE} the topic of a previous utterance in a prominent position impose a lower inferential load, and are thus more coherent, than utterances which relegate the topic to less prominent positions or which change the topic.

The vast majority of the sentences in our corpus are complex sentences. Thus, the question of how to interpret centering principles in complex sentences cannot be ignored. We will consider the basic utterance unit of centering to be the tensed clause

| Table 1 | Transition definitions. |
|---------|-------------------------|
| \( \text{Cb}(U_i) = \text{Cb}(U_{i−1}) \text{ OR } \text{Cb}(U_i) = ? \) | \( \text{Cb}(U_i) \neq \text{Cb}(U_{i−1}) \) |
| \( \text{Cb}(U_i) = \text{Cp}(U_i) \) | \text{CONTINUE} |
| \( \text{Cb}(U_i) \neq \text{Cp}(U_i) \) | \text{SMOOTH SHIFT} |
| \text{RETAIN} | \text{ROUGH SHIFT} |

\(^1\) Whenever the word \text{topic} is used in this article, we mean the general notion of topic as the subject of a discourse unit, not the specialized meaning of topic/comment.
and will assume that these clauses form a flat, linear sequence of discourse units, such that the centering output of the first clause in the sentence is the input to the next, and so on, in the spirit of Kameyama (1998) and Suri and McCoy (1994).

Because the notion of backward-looking center will be critical to the discussion that follows, we look at this notion in greater detail here. According to Grosz, Joshi, and Weinstein (1986, quoted in Walker, Joshi, and Prince [1998]), a center is realized in an utterance $U$ if it “is an element of the situation described by $U$ or the semantic interpretation of some subpart of $U$” (page 4). As Walker, Joshi, and Prince (1998) point out, this covers “pronouns, zero pronouns, explicitly realized discourse entities, and... entities inferable from the discourse situation” (page 4). The definition proposed by Grosz, Joshi, and Weinstein allows inferable entities, that is, entities that are not expressed at the surface level of the utterance or immediately recoverable from the subcategorization properties of the verb (as, for example, zero pronouns are) to constitute centers of an utterance. However, the theory does not make explicit the parameters within which to characterize the class of permissible inferable elements or the constraints on doing so. We will return to this difficulty later.

3. The Corpus

3.1 The Nature of the Corpus

The data examined consist of a collection of 32 e-mail messages exchanged among five employees of a Japanese company from June 5 to June 16, 1995. The messages constitute a collective attempt to schedule a sports-watching outing convenient for and interesting to all five in the group. Thus, the tone is usually casual. The authors combine standard aspects of written text with various strategies for encoding speech-like information in the messages: nonstandard uses of punctuation, katakana (the syllabary for writing foreign words), and English; nonstandard spelling; emoticons; discourse markers, sentence-final particles, tense, and formality typical of speech; and fillers (Fais 2001; Fais and Yamura-Takei 2003). Quantitative information for the corpus is given in Table 2.

| Authors | Messages | Paragraphs | Sentences | Clauses | Characters |
|---------|----------|------------|-----------|---------|------------|
| H       | 5        | 9          | 20        | 60      | 880        |
| I       | 10       | 46         | 50        | 122     | 1,904      |
| M       | 5        | 27         | 39        | 84      | 1,268      |
| R       | 5        | 29         | 56        | 125     | 1,805      |
| U       | 7        | 35         | 60        | 127     | 2,015      |
| Total   | 32       | 146        | 225       | 518     | 7,872      |

2 Kameyama (1998) also mentions two cases in which a hierarchical interpretation is warranted: reported speech and nonreport complements. Our corpus does not contain examples of these kinds of utterances. On the other hand, it does contain tensed clauses acting as “relative clauses” and as verbal complements. We have not separated these from their heads or matrix clauses. This has no effect on the analysis presented in this article, except for the fact that had we separated these clauses, it would have made the problematic situation we describe later even more marked. The status of these clauses vis-à-vis centering is a topic in need of extensive investigation.
3.2 Inferables and Transition States in the Corpus

Example (1) illustrates how centering mechanisms apply to the beginning of one of the messages and how the CONTINUE and RETAIN transitions can model coherence:

1a. 私は18日のヤクルト一横浜戦は watashi wa 18nichi no yakuluto-yokohamasen wa
CONTINUE
me TOP 18th on Swallows vs. Baystars game TOP

Cb = (beginning)  
Cf = Swallows vs. Baystars game

CONTINUE

1b. OKです。 Ø OK desu  
Ø OK is

“As for me, since I’m really up for the Swallows vs. Baystars game on 18th, that is OK with me.”

Cb = SvsB game Ø  
Cf = SvsB game Ø

RETAIN

1c. もともと乗り気だったので motomoto noriki datta node

Cb = it = Swallows vs. Baystars game  
M-san wa koredeindeshouka
M-san TOP I’m wondering if it is ok

M-san TOP I’m wondering if it is ok for U-san and M-san, who have shown a keen interest in taekwondo.

M-san wa koredeindeshouka

Cb = it = Swallows vs. Baystars game

M-san, M-san, it, interest, taekwondo

Clause (1a) has no Cb, since it occurs at the beginning of the message. The Cp of (1a) is Swallows vs. Baystars game, by virtue of the fact that it is topic-marked. The subject of the copular desu in (1b) is omitted; rule 1 implies that the referent for this zero argument is Swallows vs. Baystars game, which is a correct assignment. A similar process resolves the referent for kore ‘this’ in (1c). Note that the CONTINUE and RETAIN transitions do, in fact, capture the intuition that this segment of the message is coherent and is “about” the Swallows vs. Baystars game.

Example (2) is much more typical of the messages in the corpus and is not as well behaved as (1). Notice that all the Cbs in this example are inferable “from the discourse situation” of $U_{i-1}$. The preponderance of inferable Cbs is typical; out of 330 Cbs in the corpus, 250 (more than 75%) are entities other than pronouns, zero

3 We will ignore first-person arguments in this analysis, since centering is intended to handle only third-person arguments. How topic-marked first-person arguments affect attentional states in discourse is an interesting question, though beyond the scope of this article.

We use the following abbreviations for Japanese case markers in this article: TOP, topic; SUBJ, subject; OBJ, object.

4 There is no consensus as to the ordering of the arguments in a Japanese A no B construction (roughly equivalent to possessives; Tetreault 2001; Matsui 1999; but see Fais 2002). We have listed the arguments in the order suggested in Fais (2002); this has little effect on the present discussion.
arguments, or explicitly realized entities. This raises a question undiscussed in the centering literature: How do we interpret inferable Cbs in the context of assigning transition types? Example (2) illustrates the problems involved:

2a. ところで  宴席  なん  ですか。
      tokorode  enseki nan desuga
      by the way  restaurant  is

2b. 千駄ヶ谷 近辺 に 酒 を 飲める
      sendagaya kinpen ni sake wo nomeru
      sendagaya neighborhood  in alcohol OBJ can drink

umlai sobaya ga aruto miminishimashita
good soba shop SUBJ is  have heard
“By the way, about the restaurant, I’ve heard there is a good soba shop, which also serves alcohol, around Sendagaya.”

2c. 閉店 時刻 が 早い そうなので、
      heiten jikoku ga hayaisunanode
      shop closing time SUBJ early seems because

2d. ちょっと  まずい  かもしれない  が
      ∅  chotto mazui kamoshiremasen ga
      ∅  a little  bad  might be  but

2e. I-さんあたり  ご存じでは  ないでしょうか？
      I-sanatari gozonjidewa naideshouka
      I-san  information  have

“Choosing this restaurant may not be good because it closes early. I-san, do you have information about this restaurant?”

The Cb of (2a) is null, since that clause is the first in the discourse segment. Clause (2b) has three centers, listed in the Cf list. All three of these entities are inferable from the discourse context of (2a). We mentioned previously that there is no principled way to determine the list of inferables of an utterance; it is even more difficult, then, to

5 In the subsequent discussion, for the sake of simplicity, we will refer to this group as “explicit” centers, realizing that zero arguments are not precisely “explicit,” but setting them off in this way from inferable centers.
Table 3
Transitions occurring in the corpus, by type.

| Transition type             | Number | Percentage |
|----------------------------|--------|------------|
| CONTINUE                   | 54     | 16.4       |
| RETAIN                     | 22     | 6.7        |
| SMOOTH SHIFT               | 2      | 0.6        |
| ROUGH SHIFT                | 2      | 0.6        |
| Transition to inferable Cb | 138    | 41.8       |
| NULL                       | 112    | 33.9       |
| Total                      | 330    |            |

determine the order in which inferables should be listed on a Cf list. Therefore, we cannot say which of the Cfs in (2b) is the “highest ranked” and thus the Cb for (2b). Since we cannot determine which is the Cb, we likewise cannot determine whether the Cb of (2b) is the same as its Cp, and so we cannot label the transition at all.

The fact that there is only one Cf for (2c) simplifies the problem somewhat. That Cf, which is inferable from (2b), must also be the Cb of (2c), but since we could not ascertain the Cb of (2b), we do not know if the transition from (2b) to (2c) is a CONTINUE or a SMOOTH SHIFT. Again, the presence of only one Cf in (2d) makes matters easier; we are able to label the transition from (2c) to (2d) a SMOOTH SHIFT.

We surmise that I-san may not be inferable from the discourse context, though information (about the restaurant) is, and thus the latter becomes the Cb of (2e). Because this Cb is neither the same as the Cb of (2d) nor the same as its own Cp, the transition from (2d) to (2e) is a ROUGH SHIFT.

Using the standard centering definitions, supplemented with the notion of NULL transitions to label transitions to utterances that contain no Cb, we hand-tabulated the number of transitions of each type occurring in the corpus (Table 3). The figures for CONTINUE, RETAIN, SMOOTH SHIFT, and ROUGH SHIFT are those for transitions to utterances containing either explicit, pronominal, or zero-argument Cbs. Given the difficulties in accurately labeling transitions to utterances containing inferable Cbs, we grouped these latter together separately from transitions to utterances with explicit, pronominal, or zero-argument Cbs.

There are two points of special interest in Table 3. First, of course, is the particularly high number of transitions to utterances containing inferable Cbs and the number of NULL transitions. At over 40%, the utterances involved in transitions to utterances containing inferable Cbs make up a substantial portion of the corpus. Second are the relative proportions of CONTINUE, RETAIN, SMOOTH SHIFT, and ROUGH SHIFT transitions. Note that, considering just these transitions to utterances containing explicitly realized centers, these proportions are roughly what we expect of coherent text: Most of the shifts are CONTINUE, followed by a respectable number of RETAINS, and a very few SHIFTS.

Of course, in the absence of any principled way to determine this, this is merely conjecture. In fact, because I-san is a part of the ongoing e-mail exchange, he could possibly be considered part of the discourse context. The difficulty is, of course, the lack of any rigorous way to determine what the elements in the discourse context are.
4. Lexical Cohesion and Discourse Coherence: A Revision of Standard Transition Types

4.1 Inferable Centers in Centering Theory
We saw in Section 3.2 that the introduction of inferable centers into a centering analysis leads to indeterminate transition identification. We also saw that transitions to utterances with inferable centers make up a large proportion (over 40%) of the transitions in this corpus. It is important, then, if a centering analysis is to represent the nature of coherence in this corpus accurately, that we make principled provisions for the notion captured by inferable centers in the theory.

Only rarely are inferable entities actually listed in analyses in the literature, and then usually when their presence is supported by previous explicit mention and by a clear, tight semantic or syntactic relationship between the entities involved, as in (3), taken from (17) in Kameyama (1998):

3a. It is the apparent intention of the Republican Party to campaign on the carcass of what they call Eisenhower Republicanism.

3b. but the heart stopped beating

3c. and the lifeblood congealed

Note that (3) avoids at least one of the difficulties we encountered in (2); when there is only one center to deal with, as there is in (2c) and (2d) and (3b) and (3c), the choice of Cb is trivial. However, in cases such as (2a) and (2e), or in (3b) if it had been something like *but the heart and lifeblood stopped pumping*, we cannot determine what the Cb should be. Further, in these cases, it is impossible to identify the appropriate transition to the following utterance; when the Cb of $U_{i-1}$ is undetermined (as it is for (2b)), the transition to $U_i$ (in this case, (2c)) is undeterminable.

4.2 Logical Difficulties with Inferable Centers
There are further problems resulting from taking the use of inferable centers to its logical extreme. In our analysis so far, we have been concerned with explicit entities in $U_i$ that realize centers in $U_{i-1}$ that are inferable from the discourse context. To be accurate, those inferable centers need to be listed in the Cf list for $U_{i-1}$. However, if we process discourse incrementally, this leads to the conclusion that since we do not know which inferable entity from $U_{i-1}$ will be evoked in $U_i$, we need to list every inferable entity in the Cf list of $U_{i-1}$. This is both computationally untenable and, in view of the lack of any parameters for determining what constitutes an allowable inferable Cb, impossible. Even if it were possible and desirable, how could we define for inferable entities the type of grammatical-role information essential to determining the placement of these entities on a Cf list?

Again taking the definition of inferable centers to its extreme, we note another problem. Every utterance, not just $U_{i-1}$, evokes inferable centers. There is nothing in the theory to preclude a situation such as that shown in (4) (a version of (2) simplified for illustrative purposes) (implicit inferable centers are given in italics):
4a. 千駄ヶ谷 近辺 に
sendagaya kinenben ni
sendagaya neighborhood in

旨い 蕎麦屋 が あると 耳にしました。
umai sobaya ga aruto miminishimashita
good soba shop SBJ is have heard

“I’ve heard there is a good soba shop around Sendagaya.”

4b. でも 閉店 時刻 が 早いそなんです。
demo heiten jikoku ga hayaisounandesu
but shop closing time SUBJ is early

“But the shop closes early.”

In (4), an inferable center from the Cf list of (4b) matches an inferable center from the Cf list of (4a) and is chosen as the Cb. Of course, there could be numerous identical inferable centers on the Cf lists of \( U_i \) and \( U_{i-1} \), all “vying” for Cb of \( U_i \). Although this is obviously an absurd extension of the inclusion of inferable centers in centering theory, there is nothing, unfortunately, in the theory itself to rule it out.

4.3 Possible Solutions to the Problems of Inferable Centers

4.3.1 Inferable Centers as Bridging References. Inferable centers are similar to bridging references (Clark 1977); they have a conceptual relationship to entities in a previous utterance. There is a sense that bridging references should participate in the creation of coherence in a discourse (Hahn, Markert, and Strube 1996). But the work on bridging references characterizes this relationship as referential or anaphoric; this can be seen in the various terms under which this phenomenon is discussed: bridging references, indirect anaphora, functional anaphora, and partial anaphora. Bridging references, however, unlike the usual case of anaphora, may be mediated not only by a strict identity condition, but also by any number of other semantic relationships (is-a, has-a, made-of, at-time, etc.).

Unfortunately, the establishment of the semantic relationship between an anchor and its bridging reference is notoriously difficult. Poesio et al. (2000), even after severely restricting the types of relationships to be labeled, had extremely poor interlabeler reliability on a first pass. Every account in which bridging references are addressed restricts allowable relationships to a small, relatively well-defined set (Vieira and Poesio 2001; Poesio et al. 2000; Murata, Isahara, and Nagao 1999; Strube and Hahn 1999). Cote (1998) proposes the use of lexical-conceptual primitives instead of grammatical relations in Cf templates and suggests that the conceptual information that this approach provides might be rich enough to supply part-whole information necessary to the resolution of bridging references. She points out as well, however, that a number of other types of semantic relationships manifested in bridging references would not be identifiable from lexical-conceptual information. Thus, although work on bridging references has attempted to provide a characterization of the possible semantic relationships involved, what success has been achieved is limited to a small subset of cases.
4.3.2 Restrictions on the Notion of Inferable Center. One possible way around the logical problems introduced by inferable centers is to take into account only inferable centers that are explicitly realized in an utterance when determining the Cb of that utterance. This would mean, for example, considering *soba shop, Sendagaya neighborhood,* and *alcohol* from (2b) in choosing the Cb for (2c). But how do we make the choice among these possibilities? Any of these three different, explicit, inferable centers could be chosen to be the Cb. But this is exactly the difficulty: We have no principled way to make such a choice. We have no way of knowing which of these inferable centers is ranked highest in the Cf list for (2a) so that we can select that Cf to be the Cb of (2b).

A second possibility is to allow only explicitly realized (i.e., noninferable) centers. This seems to be the approach taken by Passonneau (1998); her definition of a null Cb seems to imply that a Cb must be (noninferable and) explicitly realized, and her null Cbs constitute the cases in which there is no explicit Cb. In her examination of the Pear Stories (recordings of people describing to another person a movie they had seen; Chafe 1980), NULL transitions (transitions to an utterance with a null Cb) represent the majority of transitions. Although her concern is discourse segmentation, Passonneau does note that the patterning of transition types does not accurately reflect the coherence of the stories.

Allowing only explicit centers would mean, for the corpus studied in this article, that inferable Cbs become null Cbs and the proportion of NULL transitions becomes 75.7%. Under this assumption, this corpus would be characterized as extremely incoherent, a claim belied both by native-speaker intuitions (an acceptable level of coherence for these texts was confirmed by three native speakers) and also by the fact that the task that was the central concern of these messages was successfully completed; the group exchanged a number of opinions and pieces of information and came to a consensus regarding their sports outing, with no message showing confusion about information contained in previous messages. Thus, the solution of allowing only explicit centers does not yield an accurate characterization of the coherence of this corpus.

Hurewitz (1998) chose to define allowable inferable Cbs fairly narrowly in her English data, requiring functional dependency or a poset relationship to hold in order for a Cb to be recognized. Even with this definition, which is more constrained than we have taken “inferable” to be in our previous discussion, she finds that 21% of the Brown corpus (a variety of written texts) and 28% of the switchboard corpus (taped telephone conversations) consist of what she calls a no-Cb condition. Poesio et al. (2000) report a similarly high proportion of nonexplicit Cbs in their English text corpora. They test a number of configurations of parameters of centering theory to attempt to minimize the number of null Cbs (reasoning that the best configuration of parameters would result in the fewest violations of the constraints of the theory, in this case, the constraint that all utterances in the discourse except for the first have at least one Cb). One way in which they are able to improve their results significantly is by allowing a restricted set of three types of nonidentity relationships between centers, that is, by recognizing three types of well-defined inferables. However, simply limiting the type of inferables allowed still does not address the issue of the indeterminacy of transitions to utterances containing inferable Cbs. And the central question raised by the high number of nonexplicit Cbs found in naturally occurring texts remains unaddressed: How can we characterize coherence in a text in which Cbs are so often inferable and thus in which transition types are often indeterminate?

The crux of the problem lies in the application of standard centering processes to inferable centers. In a nonproblematic case, a Cb in $U_i$ is recognized by virtue of its
identity to a Cf in \(U_{i-1}\). This is (relatively) straightforward in the case of explicit centers. But now apply this process to inferable centers. In a standard centering approach, a Cb in \(U_i\) can also be recognized by virtue of its identity to an inferable center in \(U_{i-1}\). This implies that we must somehow make available all the possible inferable centers in \(U_{i-1}\) in order to recognize (possibly) one of them as the Cb for \(U_i\). We have already noted that this position is untenable. Even if we recognize the inferable centers of \(U_{i-1}\) a posteriori by considering only those that appear in \(U_i\), it is still impossible to select which of these centers is highest ranked in \(U_i\). What we need, instead, is a way to recognize a Cb in \(U_i\) not by virtue of its identity with a preestablished list of (explicit and inferable) centers, but by virtue of a relationship, other than identity, with the explicit centers of \(U_{i-1}\). We propose the relationship of lexical cohesion to fill this function. The recognition of a lexically cohesive relationship, then, admits inferable centers without allowing the virtually uncontrollable proliferation of hard-to-define inferable centers in the Cf lists for utterances. We propose a principled way to define this relationship that not only avoids the problems discussed above but also more accurately characterizes the coherence of this corpus.

### 4.4 Coherence and Cohesion

Halliday (1994) characterizes cohesion in text as the establishment of “relations within the text that are not subject to [grammatical] limitations; relations that may involve elements of any extent, both smaller and larger than clauses, from single words to lengthy passages of text; and that may hold across gaps of any extent... without regard to the nature of whatever intervenes” (page 309). Cohesion is that aspect “whereby the flow of meaning is channelled into a traceable current of discourse instead of spilling out formlessly in every possible direction” (page 311). It is this “traceable current of discourse” that centering is meant to model.

Lexical cohesion contributes to textual coherence. In other words, strong semantic and structural relationships among words in a text help to make that piece of text “make sense.” Coherence is a property of discourse; cohesion is a property of discourse elements. Centering models coherence by characterizing relationships between elements of discourse. We claim that it is not only the continuation of identical explicit discourse elements that creates coherence, but also strong cohesion among discourse elements.8

Halliday identifies four features of text that create cohesion among discourse elements: conjunction, reference, ellipsis, and lexical cohesion. Insights concerning conjunction types and their interactions with the processes of referent resolution have been elaborated in a number of works (Nariyama 2000; Nakaiwa and Shirai 1996; Kuno 1973) but have not been well integrated into the centering approach. Reference and ellipsis are, of course, some of the mainstays of centering research. Lexical cohesion, however, is an aspect of text coherence that has had only a trivial application in a centering approach, although it has been incorporated into other aspects of natural language processing (see subsequent discussion). “Lexical cohesion,” according to Halliday, “comes about through the selection of items that are related in some way to those that have gone before” (page 330). In centering, that relationship has been

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7 We follow Halliday in assuming that “[f]or a text to be coherent, it must be cohesive; but it must be more besides.” He characterizes the “more” as being socially, semantically, and structurally appropriate. We will not deal with these elements here but rather will limit ourselves to the contribution of lexical cohesion to coherence.

8 Of course it is possible to have cohesion without coherence and vice versa; Morris and Hirst (1991) give some nice examples. However, as they assert, “most sentences that relate coherently do exhibit cohesion as well” (page 26).
assumed to be one of identity of centers; an element in $U_i$ is required to be the same as an element in $U_{i-1}$ (or in its discourse context) in order for it to be considered a Cb. However, earlier we saw the difficulties of admitting inferables and the inability of a centering approach to characterize coherence for inferable centers. Applying the notion of lexical relatedness to cases involving inferables allows us to capture what seems intuitively to constitute the relationship between clauses containing inferable centers.

A number of other accounts provide relevant information concerning lexical cohesion in text. These accounts are based upon the characterization of semantic relations among discourse elements by reference to semantic information contained in WordNet (Harabagiu 1998, 1999), thesauruses (Harabagiu 1998; Morris and Hirst 1991; Okumura and Honda 1994), or dictionaries (Kasahara et al. 1996; Kozima 1993; Kozima and Furugori 1993). In all of these approaches, the semantic distance or similarity between (or among) words is computed, and in most of these accounts, the results are applied to the segmentation of discourse.

The intuition behind the importance of lexical relatedness has been applied to a number of other tasks in natural language processing and analysis as well. Lotfiour-Saedi (1997) uses lexical cohesion to develop a rigorous notion of “translation equivalence”; Boguraev and Neff (2000) to improve document summarization techniques; Sack (1999) to create “diagrams of social cohesion” for newsgroup postings; and Okumura and Honda (1994) to disambiguate word senses.

Halliday asserts that “this interaction between lexical cohesion and reference is the principal means for tracking a participant through the discourse” (page 332), that is, for modeling focus. Centering has provided us with a principled way of characterizing the tracking of reference; the addition of the notion of lexical cohesion allows centering to function in an even more empirically comprehensive way, by making possible the principled inclusion of what have been called inferable centers. In the next section, we outline how lexical cohesion can be incorporated into a centering theory.

4.5 Cohesive Transition and Complete Shift

With the use of the sorts of techniques to establish semantic distance described in the works cited earlier, it is possible to be precise about the notion of lexical cohesion. In this section, we discuss how semantic distance is established using one of these techniques, a semantic similarity measure derived from the Gainen Base (‘Concept Database’) (Kasahara et al. 1996). We then indicate how semantic distance can be used to define the notion of lexical cohesion as a crucial element in the creation of two new types of transition: COHESIVE and COMPLETE SHIFT, which allow us to adequately characterize coherence in a corpus containing a high proportion of nonexplicit Cbs.

The Gainen Base is a knowledge base built from machine-readable dictionaries of Japanese. Each word in the knowledge base is defined by a list of weighted keywords extracted from the dictionary definition of the word. The number of times a keyword appears in the word’s definitions determines the weight for the keyword. Keywords are standardized to take into account the presence of semantically similar words in the definitions, and their weights are normalized to take into account the differing lengths of definitions in the dictionaries. The semantic distance between two words is calculated as a function of the nearness of the two words in a vector space.9 For

9 Each Word, is defined by a list of standardized, weighted keywords from which is generated a vectorized-Word. More specifically, then, semantic similarity is measured by a function $R$ that satisfies
Table 4
Transition definitions.

| Cb(Ui) = Cb(Ui−1) or Cb(Ui−1) = ? and Cb(Ui−1) ≠ ? | Cb(Ui) ≠ Cb(Ui−1) | Cb(Ui) = ? |
|---------------------------------------------------|---------------------|----------|
| Cb(Ui) = Cp(Ui) CONTINUE SMOOTH SHIFT COHESIVE | Cp(Ui) CONTINUE SMOOTH SHIFT COHESIVE | CONTINUE SMOOTH SHIFT COHESIVE |
| Cb(Ui) ≠ Cp(Ui) RETAIN ROUGH SHIFT COMPLETE SHIFT | Cp(Ui−1) ≈ Cp(Ui−1) RETAIN ROUGH SHIFT COMPLETE SHIFT | Cp(Ui−1) ≈ Cp(Ui−1) RETAIN ROUGH SHIFT COMPLETE SHIFT |

Our present purposes, then, we say that there is lexical cohesion between Ui and U_i−1 when Cf(Ui) is semantically close to Cf(U_i−1) as determined using a well-defined semantic similarity measure over the Gainen Base.10 We will examine in more detail in Section 4.6 whether semantic similarity is best viewed as holding between the sets of Cfs of utterances or between individual discourse entities in those utterances. For now, we will talk equally of lexical cohesion between utterances and lexical cohesion among the Cfs that participate in defining cohesion for those utterances. We define the relation ≈ as indicating strong lexical cohesion, with a lexical cohesion factor of one indicating identity.

We supplement the standard table of transition states, shown in the left side of Table 4, with transitions defined in the right side. The sense of this table is as follows. We assume all centers to be explicit, that is, pronouns, zero arguments, or explicitly realized entities. The left portion of Table 4 allows us to model transition states in well-behaved explicit contexts, tracking the focus of the discourse in a specific, local way. It includes the cases in which the Cb of U_i−1 might be “?,” that is, that U_i−1 might be the discourse-initial utterance or might simply have no Cb, while Ui does have (an explicit) Cb. However, where there is not strict identity between any (explicit) element in Ui and any (explicit) element in U_i−1, we have the situation in which Cb(Ui) = ?. We propose a new interpretation for these cases, described in the right portion of Table 4.

Table 4 defines two new types of shift to utterances that do not contain an explicit Cb. If there is at least one Cf in Ui that has a high lexical cohesion value with some Cf(s) in U_i−1, then the transition from U_i−1 to Ui is a COHESIVE transition. This situation is illustrated in clauses (2b), (2c), (2d), and (2e) of example (2). The transitions to these clauses, under the present proposal, are reanalyzed as COHESIVE, as shown in a reanalyzed version of example (2) (entities claimed to bear close semantic relationships to one another are shown in boldface here and in subsequent examples):

\[0 \leq R(\text{Word}_a, \text{Word}_b) \leq 1\]
\[R(\text{Word}_a, \text{Word}_b) = R(\text{Word}_b, \text{Word}_a)\]
\[R(\text{Word}_a, \text{Word}_a) = 1\]

Word_a is more similar to Word_b than to Word_c if R(Word_a, Word_b) ≤ R(Word_c, Word_b)

The function \( R = \cos \alpha \), where \( \alpha \) is the angle defined by the vectors for \( \text{Word}_a \) and \( \text{Word}_b \), satisfies these conditions and is therefore chosen as the function to define the similarity between \( \text{Word}_a \) and \( \text{Word}_b \).

We refer the reader to Kasahara et al. (1996) for a full discussion of the algorithms used to weight and normalize keywords and to calculate semantic distance.

10 What constitutes "semantically close" is an issue we will discuss briefly in Section 5.4; exactly how semantic distance is measured is elaborated in Section 4.6.
Fais Transitions and Coherence

2a. ところでお席なんですが。
tokorode enekin desuga
by the way restaurant is

2b. 千駄ヶ谷近辺に酒を飲める
sendagaya kinenben ni sake wo nomeru
sendagaya neighborhood in alcohol OBJ can drink

旨い蕎麦屋があると耳にしました。
umai sobaya ga aruto miminoshimashita
good soba shop subj is have heard
“By the way, about the restaurant, I’ve heard there is a good soba shop, which also serves alcohol, around Sendagaya.”

2c. 閉店時刻が早いようなので、
heiten jikoku ga hayaiso nanode
shop closing time subj early seems because

2d. ちょっとまずいかもしれないが
∅ chotto mazui kamoshiremasen ga
∅ a little bad might be but

2e. I-さんあたりご存じですか？
I-sanatari gozonjide wa naideshouka
I-san information have

“Choosing this restaurant may not be good because it closes early. I-san, do you know about this restaurant?”

Each COHESIVE transition in the modified version of (2) is justified by the presence of a strong semantic relation between at least one Cf in $U_i$ and at least one Cf in the previous utterance. For example, soba shop in (2b) is semantically related to restaurant in (2a), as is alcohol, probably to a lesser extent, and Sendagaya (if our database includes the information that this is the name of a restaurant). Note that the transition from (2d) to (2e) in this interpretation is a COHESIVE one, rather than a ROUGH SHIFT. Certainly the Cp changes from restaurant to I-san, but since restaurant is still present in the Cf list for (2e), the transition is by no means as abrupt as the designation ROUGH SHIFT implies.

The presence of a null Cb in and of itself, then, is not necessarily indicative of incoherence. The level of coherence is captured instead by the proportions of the various transition states present in a corpus, including COHESIVE transitions.

Not all utterances containing null Cbs are felt to be cohesive with previous utterances, of course. If there is no explicit Cb and no Cf in $U_i$ such that it has strong lexical cohesion with a Cf in $U_{i-1}$, then the shift is considered COMPLETE. This is illustrated in example (5), which shows the continuation of example (1):
5a. でも、テコンドーに興味津々だったU-さんや
demo teknondoh ni kyominishinshin datta U-san ya
but taekwondo in keen interest was U-san and

M-さん は これでいいんですか？
M-san wa koredeindoishouka
“But I’m wondering if it is OK for U-san and M-san, who have
shown a keen interest in taekwondo.”

5b. どうやら 梅雨 入りしたらしいという宣言も出て
douyara tsuyuiri shitarashii toiu senenmodeta
somehow rainy season has come declaration

ようで、今日もしっかり雨が降っています。
youde kyou mo shikkari ame ga futteimasu
with today also heavily rain SUBJ is raining
“With the declaration that the rainy season has come, it is
raining heavily today.”

5c. 18日が雨という確率も高そうですよね。
18nichi ga ame toiu kakuritsu mo taka soudesuyone
18th SUBJ rain probability high seems
“There seems to be a high probability that it will rain on the 18th.”

There is very low cohesion between the Cfs of (5b), that is, today, rain, declaration, and
beginning of rainy season, and those of (5a), namely, U-san, M-san, game, taekwondo, and
interest. Thus, we designate the transition from (5a) to (5b) as a COMPLETE SHIFT, which
matches our intuition that, in fact, the topic of the message has changed. This is further
corroborated by the fact that (5c) RETAINS the Cp of (5b) as its Cb; in other words, it
go on to develop the new topic begun in (5b).

Since the two sides of Table 4 are complementary, it is perfectly possible for co-
herence to be characterized by various combinations of the transitions in the table.
Example (6) illustrates the interaction between the two sides of Table 4:

6a. このうち 神宮の方は、17日と18日
konouchi jinguu no hou wa 17 nichi to 18 nichi
of these Jingu TOP 17th and 18th

CONTINUE, from above
Cb = these
Cf = these, Jingu
Stadium,
Yokohama
"Of these, for Jingu Stadium I suppose the game versus Yokohama on the 17th or the 18th would be good. As for J-league, since I don’t have further information such as the match-ups, please let me know if you have any information.”

“Since there are no good boxing match-ups in this month, I would like to skip it this time.”

The transition from (6a) to (6b) is COHESIVE, since (6b) has no Cb and yet the Cfs of the two utterances are semantically related. (6c) RETAINS match-up information from (6b), which is then COHESIVE with match-ups and boxing in (6d). Since (6e) does in fact have an explicit Cb identical both to the Cb of (6d) and to its own Cp, the transition to (6e) is CONTINUE. This latter transition is an example of a non-discourse-initial utterance

It is a departure from the standard Cf template for Japanese to designate these as the Cb instead of the topic-marked Jingu Stadium; however, this pronominal form comes immediately after the mention of two options in the previous clause and so seems to be the Cb regardless of its lack of marking. This choice of Cb has no bearing on the point of this example.
without a Cb that licenses a CONTINUE transition by virtue of the second part of the OR disjunction in the second column of Table 4, namely, that the Cb(Ui−1) = ? and Cb(Ui) ≠ ?. (Note that (5) also demonstrates the interaction between the two sides of Table 4, with a RETAIN transition following a COMPLETE SHIFT.)

The proposal to include COHESIVE and COMPLETE SHIFT in centering theory is motivated by concerns about having to specify two identical entities in adjacent utterances in order for those utterances to exhibit coherence. The requirement of identity leads to the need to allow inferable entities to play a part. However, it is often impossible to characterize transition states to utterances containing inferable Cbs. By including the notion of COHESIVE transitions, we capture the relatedness of two entities without the need to invoke inferable centers, and we can far better characterize the apparent coherence in this corpus.

4.5.1 Transition States in the Corpus, Revisited. Given the inclusion of COHESIVE and COMPLETE SHIFT transitions, we reinterpreted the labeling of transitions in the corpus (as represented in Table 3) and hand-tabulated a revised distribution of transition types (Table 5). The decision to designate transitions to inferable Cbs and NULL transitions as COHESIVE or COMPLETE SHIFT was made on the basis of an intuitive assessment of the possible semantic relations between entities in adjacent utterances. However, a possible method for automatic determination is described in the next section.

4.6 A Preliminary Implementation

4.6.1 Preparation of the Corpus. In order to explore the computational feasibility of COHESIVE and COMPLETE SHIFT, we subjected the corpus to an analysis of semantic distance using the Gainen Base. Before implementing the analysis, we processed the corpus so that it contained only the canonical forms of the words in each utterance, the forms accepted by the Gainen Base algorithms. The first step in this procedure was to run the corpus through a morphological analyzer, ALT-JAWS (Nippon Telegraph and Telephone Corporation 1996), which rendered all word forms in their canonical (usually kanji [Chinese character]) form. This is a necessary and standard preprocessing step for most computational analyses of Japanese text, which contains no spaces between word forms to indicate their morphological structure. In addition, writers of e-mail may use hiragana or katakana (the two syllabaries used for writing primarily function words and foreign words, respectively) even for words that have standard kanji forms (Fais and Yamura-Takei 2003). These words need to be rendered in kanji as well. The results of this analysis were checked and corrected by native speakers of Japanese in order to eliminate cases in which the analyzer chose the incorrect kanji form for a homophonous hiragana representation.

| Table 5 |
|------------------|--------|--------|
| Transition type  | Number | Percentage |
| CONTINUE         | 54     | 16.4    |
| RETAIN           | 22     | 6.7     |
| SMOOTH SHIFT     | 2      | 0.6     |
| ROUGH SHIFT      | 2      | 0.6     |
| COHESIVE         | 138    | 41.8    |
| COMPLETE SHIFT   | 112    | 33.9    |
| Total            | 330    |         |
The next step was to delete all but noun forms from the corpus. This step substantiates the notion that the central factors in the centering approach are the discourse entities of utterances (we will return to this in Section 5.3). Further, the antecedents of all pronominal and zero-argument references were made explicit in the text. This was done by inserting the kanji forms for these antecedents, as they appeared in the text, into the utterance in which the pronominal or zero-argument form appeared. Antecedents were determined by a native speaker of Japanese. The same native speaker assisted in dividing the text into clauses. Both of these steps, identifying antecedents and parsing sentences into clauses, can be completed, in theory, by automatic, computational methods (Huls, Bos, and Claassen 1995; Nakaiwa and Shirai 1996; Paul and Sumita 2001; Yamura-Takei et al. 2002), but the success rate of these approaches is not high enough to rely on them for completely accurate analyses of this type of corpus at this time. Because our intent is to examine the effectiveness of the use of the Gaien Base in determining lexical cohesion, and not to implement a fully automatic process, we did not attempt to use entirely automatic methods in the preprocessing of the text.

4.6.2 Determining Semantic Similarity with the Gaien Base: The Problem of Coverage. Of the 670 types of nouns present in the e-mail corpus, only 235 are found in the machine-readable dictionaries with which the Gaien Base was constructed. What makes this ratio even more problematic is that a number of these missing words (e.g., supo-tsu kansen, ‘sports-watching event,’ rakurosu, ‘lacrosse,’ and the names of sports teams) are high-frequency words in this corpus.

The problem of coverage in automatic language-processing systems is a common one (Hutchins 1995; Sag et al. 2002; Fujita and Bond 2002). At the level of coverage provided by the Gaien Base, however, we cannot usefully assess how well semantic similarity characterizes coherence in this corpus as a whole. However, we can make this assessment for those clauses in which every noun can be found in the Gaien Base. In order to get an accurate assessment of lexical cohesion between adjacent clauses that fall into this subset of the corpus, we included those cases in which all the nouns in the clause itself as well as those in the clause before it were found in the Gaien Base. Out of an original 443 clauses, there are 66 clauses that meet these two criteria.

We measured semantic similarity between adjacent clauses in two ways. In the first, we measured the semantic distance between the group of nouns in \( U_{i-1} \) and the group of nouns in \( U_i \). In the second, we measured the semantic distances between each individual noun in \( U_i \) and the group of nouns in \( U_{i-1} \). The first method is far “lighter” computationally, but the second method gives us useful information about the contribution of each noun in \( U_i \) to the lexical cohesion between utterances. We will compare the information derived from these two approaches hereafter.

4.6.3 Evaluation of the Use of Semantic Similarity. We assessed in three ways how well semantic similarity can define COHESIVE transitions and thus contribute to the characterization of coherence in the text. First, in the 66 clauses with full coverage by the Gaien Base, we examined the 18 instances of what we had, on the basis of intuitive human judgment, designated COHESIVE transitions. In particular, we focused
on the noun(s) we had singled out in our revised hand-tabulation of the corpus (Table 5) as providing the lexical cohesion in these transitions. Using results from the individual measures of semantic similarity (the second method described earlier), we determined the noun in $U_i$ with the highest level of semantic similarity to the nouns in $U_{i-1}$. We then compared the nouns picked out by human judgment as providing lexical cohesion with those determined computationally to see if they matched. We eliminated three cases in which there was only one noun in $U_i$ and thus only one possible choice for the lexically cohering entity, which would, by default, have had the highest semantic similarity to the preceding utterance. Out of the 15 remaining examples of COHESIVE transitions, in 13 cases, or 87% of the time, the human judgments and the computationally determined choices matched.

The other two assessments of the results are based on centering claims for the relative ease of processing of the different transitions, claims captured in rule 2: CONTINUE transitions impose the lowest inferential load on processors, ROUGH SHIFTS the highest. Since COHESIVE transitions act like CONTINUE transitions but replace the identity condition on Cbs with a similarity condition, we conjecture that they place only a slightly higher load on processing than CONTINUE transitions. Likewise, since COMPLETE SHIFTS represent an even greater discontinuity than ROUGH SHIFTS (there being not only no Cb in the utterance, but also no entity even similar to entities in the previous utterance), we conjecture that COMPLETE SHIFTS impose a higher processing load than ROUGH SHIFTS.

We reason that greater semantic similarity corresponds to a lower processing load. Granted this assumption, then, utterances that are connected by CONTINUE transitions have the highest semantic similarity, since CONTINUE represents the lowest processing load; those connected by COMPLETE SHIFTS, the lowest semantic similarity, since COMPLETE SHIFT has the highest processing load. Although the particular ranking of RETAIN, SMOOTH, ROUGH, and COHESIVE transitions is problematic, we are saved from having to make an exact determination by the fact that, among the 66 clauses with full coverage, only CONTINUE, COHESIVE, and COMPLETE SHIFT transitions are well-represented (we simply note the results for the two RETAIN transitions, since this is hardly a large enough sample to be meaningful). Our reasoning then predicts that CONTINUE transitions have the highest semantic similarity measures, followed by COHESIVE SHIFTS, followed by COMPLETE SHIFTS.

In our first assessment of this prediction, we averaged semantic measures for each transition type over the 66 clauses with full coverage by the Gainen Base. Table 6 gives the averages obtained for both the groupwise and individual analyses. These results support the ranking of transitions for processing load predicted on the basis of similarity measures. However, averaging over all messages provides only a gross approximation of the values for each transition type. To get a more detailed look at this claim, we examined the semantic distances between entities involved in each type of transition within messages.

The 66 clauses with full coverage include six messages with only one clause and four messages with only one transition type represented; this resulted in the elimina-
Table 6
Average similarity measures for each transition type, for both groupwise and individual analyses.

| Transition type | Number | Groupwise analysis | Individual analysis |
|-----------------|--------|--------------------|---------------------|
|                 |        | Average similarity | Average similarity  |
| CONTINUE        | 10     | 0.605              | 0.639               |
| RETAIN          | 2      | 0.530              | 0.496               |
| COHESIVE        | 18     | 0.204              | 0.231               |
| COMPLETE SHIFT  | 36     | 0.068              | 0.069               |
| Total           | 66     |                    |                     |

tion of 19 clauses (since it is impossible to determine relative values of transitions in either of these cases), leaving us with a total of 47 clauses grouped into 12 messages. We sorted the types of transitions to these clauses within each message by similarity measure. For each transition, we ascertained whether it fulfilled the prediction for ranking transitions by inferential load that was made on the basis of our earlier assumptions concerning semantic distance. We assigned two scores for each transition: whether it was appropriately positioned with respect to the preceding transition and with respect to the following transition in the sort. Those transitions with the highest and the lowest semantic measures were scored only with respect to the following and the preceding transitions, respectively, and thus received just one score. Two consecutive identical transitions were scored “correct.”

Table 7 gives an example of how this scoring was performed for the clauses from one representative message that have full coverage in the Gainen Base. It lists the type of transition to each clause, the semantic distance to the previous clause, and the scores designating whether that transition is appropriately positioned with respect to the previous and following transitions. (Recall that the predicted order is CONTINUE, COHESIVE, COMPLETE.) So, for example, the CONTINUE transition to clause 235 fulfills the ranking prediction with respect to both the previous and the following clauses; its semantic measure is lower than that of only one other clause, which is also a CONTINUE, and is higher than that of a COHESIVE transition. The COMPLETE SHIFT to clause 227, on the other hand, fulfills the prediction vis-à-vis the previous transition on the list.

Table 7
Ranking transitions in a representative message by similarity measure.

| Clause | Transition to clause | Semantic measure (semantic distance to previous clause) | Score |
|--------|----------------------|--------------------------------------------------------|-------|
| 220    | CONTINUE             | 0.44113                                                | Correct v |
| 235    | CONTINUE w/?         | 0.13974                                                | Correct ↑ |
| 228    | COHESIVE             | 0.06226                                                | Correct ↑ |
| 227    | COMPLETE             | 0.03922                                                | Correct ↑ |
| 216    | COHESIVE             | 0.01370                                                | Incorrect ↑ |
| 234    | COMPLETE             | 0.00000                                                | Correct ↑ |
Table 8
Evaluation of relative similarity measures for transition types.

| Transition type | Groupwise analysis | Individual analysis |
|-----------------|--------------------|---------------------|
|                 | Number  | Percentage | Number  | Percentage |
| CONTINUE        | 12/12   | 100        | 11/12   | 92         |
| RETAIN          | Ranked before cohesive shift | Ranked before cohesive shift |
| COHESIVE        | 26/29   | 90         | 25/29   | 86         |
| COMPLETE SHIFT  | 25/28   | 89         | 24/28   | 89         |
| Total           | 63/69   | 91         | 60/69   | 87         |

*Since we make no claim as to whether the placement of RETAIN is correct or not, we do not count it in these totals.

(i.e., its semantic measure is lower than that of a COHESIVE transition) but violates the prediction with respect to the following transition (i.e., its semantic measure is higher than that of another COHESIVE transition). For 47 clauses over 12 messages, then, the total number of such scores was 70: 24 scores for the transitions having the highest and lowest measures in each message, and two each for the remaining 23 transitions, one for their positions relative to the transitions above, and the other for their positions relative to the transitions below them.

Recall that we determined similarity in two different ways: first, for the group of nouns in $U_i - 1$ and the group of nouns in $U_i$, and second, for the group of nouns in $U_i - 1$ and each individual noun in $U_i$. Table 8 reports the total number and percentage of correct and incorrect scores for each transition type in both the groupwise and the individual analyses (the measure taken for $U_i$ in the latter case is the maximum similarity measure out of the measures for all the nouns in $U_i$). The results in this table suggest that similarity scores can accurately represent relative coherence as characterized by the transitions in this small sample. That is, the similarity scores we have examined here reflect the relative load on processing imposed by each type of transition with between 86% and 100% accuracy, with groupwise scores being slightly more accurate than those based on the highest individual score. In addition, similarity measures never predict a CONTINUE transition with a higher processing load (i.e., lower similarity score) than a COMPLETE SHIFT. That is, the relative positions of CONTINUE and COMPLETE SHIFT are always correct.

We examined the cases in which similarity scores make an incorrect prediction about relative placement in the scale of processing load. The one incorrect prediction concerning a CONTINUE (in the individual-analysis method) involves an example in which $U_i - 1$ contains eight nouns; the similarity of each individual noun in $U_i$ to the group of nouns in $U_i - 1$ is “diluted” by the high number of nouns in $U_i$ in this case (the average number of nouns per clause in the corpus is 2.3).

The remaining incorrect predictions involve the assignment of lower similarity scores to COHESIVE transitions than to COMPLETE SHIFTS. Some of these lower scores for COHESIVE transitions are the result of the fact that world knowledge is necessary to infer a connection between the two clauses. This is the case, for example, in (7):

17 Table 7 reported groupwise scores.
7a. 誰も東京に帰って来たところで
    toode demo Tokyo ni modotte kitatokorode
    at a distance though Tokyo to come back

7b. 宴会タイムには正度いいですよね。
    enkai taimu niwa choudo iidesuyone
    drinking party time for TOP exact right

"Even if it's far we can come back to Tokyo just the right
time for a drinking party."

It requires world knowledge to understand that there is a connection between how far
away something is and the time that it will take to travel there. Although humans can
make this inference intuitively, that understanding is not represented in the Gainen
Base.18

Other incorrect judgments are the result of how the database handles determining
the similarity scores, “quirks” that don’t seem to match our intuitive judgments, as in
(8):

8a. 時間的にはちょっとタイミングはいいだろうし,
    jikanteki niwa choudo taimingu wa iidasuyone
    timewise TOP exact timing TOP seems good and

8b. 地理的にても遠くはないんで。
    chiritekini mo tookuwa nainde
    geographically too far TOP not

"Timewise, the timing seems good, and the night game is not far
away either."

The Gainen Base yielded a relatively low score for the similarity between jiken, ‘time,’
and yoru, ‘night,’ despite our strong sense that these two words should be closely
related.19

Overall, however, similarity scores seem to provide a fairly accurate measure of
the relative coherence of this subset of the corpus. This result, coupled with the high

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18 Recall that in 13 out of 15 cases, the Cf judged by humans to license a COHESIVE transition and the Cf
picked out by the similarity measure matched. (7a) is one of the two clauses in which the
human-chosen Cf did not match that chosen by the Gainen Base (the second is given in (8)). The
previous clause is if it is a day game. The discourse entity tode, ‘at a distance,’ was the Cf chosen by
human judgment to license the COHESIVE transition, since a human can make the connection that it is
the day game that is at a distance (as evidenced in the translation). However, the use of the Gainen Base
determined Tokyo to be more semantically similar to day game, with a score of 0.105 as compared to
0.005 for tode.

19 This is the second of the two cases in which human judgment and computational choice for lexically
cohering entity did not match.
level of correlation discussed earlier between lexically cohesive entities designated by human judgment and those determined by semantic similarity, supports our proposal that lexical cohesion, as measured by semantic distance, can feasibly be included as a well-defined notion to capture crucial aspects of text coherence.

4.7 Exploring the Implications of the COHESIVE Transition and COMPLETE SHIFT

4.7.1 The Need to Identify Relation Type. The usefulness of relaxing the notion of identity for Cbs has been recognized by Poesio et al. (2000) and others (Hahn, Markert, and Strube 1996; Murata, Isahara, and Nagao 1999). Poesio et al. supplemented the identity relation with three different possible semantic relations between Cfs in the utterances: set membership, subset, and “generalized possession.” Murata, Ishara, and Nagao induced a number of possible relations between bridging reference and anchor using a verb case frame dictionary and a corpus of Japanese A no B expressions (where A and B are nominal arguments and the A no B construction encodes a wide variety of semantic relations between the two nominal arguments [Shimazu, Naito, and Nomura 1987]).

As noted in Section 4.3.1, however, in all of these approaches, only a small subset of examples of bridging relations can be handled, because they all attempt to identify some particular relation existing between two elements. This is a necessary move for resolving bridging references and building text understanding systems, but neither of those is our aim here. We merely need to identify the level of semantic closeness or similarity between Cfs in Ui and Cfs in Ui−1. Utilizing instead the more general semantic-distance measure proposed here, then, has several advantages over the explicit choice of particular relation labels. First, it avoids the need to make choices about which relations can or cannot, should or should not be included, as well as the difficulties with interlabeler reliability that Poesio et al. note, since there is no labeling in our approach. This is actually closely tied to another advantage: There is no need to limit the types of semantic relationships into which the Cfs can enter. Thus, there is no need to restrict our analysis to a subset of the phenomenon; our account will handle ineriable centers having any kind of semantic relationship to the centers in the previous utterance.

4.7.2 Overestimation of COMPLETE SHIFTS. In examining Table 5, we see a high number of COMPLETE SHIFTS. Is this number an accurate estimation of the (in)coherence in this corpus? Of course, as we saw in (5), when a message makes a shift in topic, we expect a COMPLETE SHIFT to occur. The writers of the messages in this corpus made 146 paragraphs (see Table 2); although we know that there is no guarantee that writers’ paragraphing will coincide with shifts in cohesion, this number at least gives us a general estimation of a possible maximal number of COMPLETE SHIFTS (we would expect writers to err on the side of more paragraphs than topics rather than on the side of more topics than paragraphs). The number of COMPLETE SHIFTS is comfortably within that range.

But there are two confounding factors that make this number higher than is actually appropriate for the nature of the corpus. The first is the inability of a centering approach to handle event deixis (Fais and Yamura-Teke 2003). Consider (9):

9a. 急ぎょ U-さんの 仕事の 都合 上 17日 の Cb = ?
   kyukyo U-sanno shigotonotsugou jou 17nichi no Cb = U-san, business,
   urgent U-san business because of 17th on 17th, match-up
Because of U-san’s urgent business, she would like to change to the same match-up on the 17th.

“That’s OK with me.”

“How is it for everyone else?”

The zero argument translated as *that* in (9b) is an example of event deixis. When we examine the Cf list for (9a) to ascertain the Cb of (9b), we do not encounter *that* or its referent. In fact, we cannot encounter *that*. It is not possible for the discourse element represented by *that* to have appeared in the Cf list in (9a); clauses do not contain self-referential discourse elements. Thus, it is impossible to recognize, within the theory, that the discourse element *that* in (9b) is functioning as a strong cohesive element in the discourse. This is a problem to be resolved within centering theory regardless of whether COHESIVE transitions and COMPLETE SHIFTS are countenanced.

The second confounding factor is actually simply a byproduct of the nature of the corpus. Example (10), which is an entire message, illustrates:

“All of I-san’s proposals are OK with me.”

“I don’t mind the first-base side or the third-base side.”

“If it’s Sendagaya, we don’t have to care even if the party goes late.”
The COMPLETE shifts in this passage are perfectly appropriate; there is no cohesion among any of the discourse entities in this message. Why would anyone send a message that was completely, by this account, incoherent? In fact, the statements in (10) refer to the outcomes of various discussions held in the course of exchanging the messages in this corpus. This message occurs toward the end of the exchange of messages, as resolution of the questions of what sports event to go to, where to sit, and what restaurant to go to afterward is in sight. This writer is simply adding his opinions on each of these apparently unrelated topics. The relationship among them all holds only in the understanding of the coparticipants in the message exchange. Although we might be able to imagine the sorts of mechanisms required to model this level of understanding, we are a very long way from realizing them.

4.7.3 Lexical Cohesion and Text Understanding. The incorporation of a COHESIVE transition into our centering account gives us a flexibility that is important for full-text understanding of the discourse. Consider example (11):

11a. わたし も テレビ で しか見たことはない ので、
I too TV only have seen since

11b. 生で 観戦 したい気もする のです が、
live game watching feel like because but

11c. 2 5 日 だと
25th is if

11d. M-さん、 H-さん が 参加 できない ですね
M-san, H-san SUBJ join cannot isn’t it

“Since I have seen it only on TV, I feel like watching a live game, but if it is on the 25th, M-san and H-san cannot join us for the game, can they?”

Lacrosse in (11a), (some) live game watching in (11b) and (11c), and (a) game (on the 25th) in (11d) are actually semantically distinct elements, and recognizing their distinctiveness is important for full-text understanding or summarizing. However, maintaining these distinctions in a standard account means characterizing the transitions between the utterances containing them as NULL SHIFTS. (It is not clear that we would even want to say that (a) game (on the 25th) was an inferable center from (some) live game watching.) Being able to designate these transitions as COHESIVE allows us both to maintain the

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20 The message that this is taken from constitutes a sort of lesson on lacrosse from one of the authors to all the others. The ∅ in (11a) refers to this global topic; however, neither lacrosse nor TV appears in the preceding utterance, and so (11a) has no Cb.
semantic distinctiveness of the discourse elements and to capture the coherence in this portion of the discourse.

4.7.4 Lexical Cohesion and Discourse Segmentation. As we saw earlier, the use of some notion of lexical cohesion to delineate discourse structure is fairly well researched. In the lexical chain approach (Morris and Hirst 1991), a new discourse segment is hypothesized where the chain “breaks,” that is, where subsequent entities do not bear a semantic relationship to previous entities that would allow them to be added to the chain. Kozima (1993) provides an algorithm for determining where, on the graph of semantic cohesion values of words in a text, likely topic breaks occur and validates that determination against human judgment. We would say, then, that once a chain breaks or a significant dip in the semantic cohesion value graph occurs, the utterance following such a break is considered the first utterance of a new discourse segment.

In terms of semantic distance as determined by the Gainen Base, we suggest that a sufficiently low similarity measure might characterize both COMPLETE SHIFTS and the beginning of a new discourse segment. Once again, “sufficiently” must be defined; in light of the preliminary results we saw previously, we conjecture that the definition of “low” will be relative to the measures for similarity in the message under scrutiny and not an absolute value (see Section 5.4).

In addition, examination of the particular entities contributing to high levels of semantic similarity might also allow us to characterize the notion of “global topic,” albeit in a differentiated way. That is, if we determine not just one semantic distance measure for the sets of entities in two adjacent utterances, but the individual distances for each combination of those entities (as briefly described in note 13), we can determine those entities that are contributing the greatest amount of semantic similarity to the measure and identify a cluster of entities that can be taken to represent a global topic.

This concept is worth examining more closely, since it bears on some of the very foundations of centering theory. The Cb of an utterance “represents the discourse entity that the utterance $U_i$ most centrally concerns, similar to what is elsewhere called the ‘topic’” (Walker, Joshi, and Prince 1998, page 3). The presence of a Cb is taken to be both a necessary and a sufficient condition for topic coherence. However, in our account, utterances that have a coherent relationship to the immediate context of discourse may nonetheless have no Cb; that is, the presence of a Cb is, in our approach, only a sufficient condition. Our claim, then, is that this more accurately reflects the nature of how coherence is maintained in discourse: not only through the explicit repetition of a central entity, but also through the successive use of entities that are closely related semantically. In our account, Cbs are recognized and function just as in standard centering theory, but their absence, a common situation in at least some kinds of discourse, does not signal a breakdown in coherence. Coherence may be maintained as well by semantically similar entities, which can become Cbs in their own right, as match-up information and boxing do in (6). Using an individuated approach to determining semantic similarity, we can identify these particular entities.

5. Future Work

5.1 Refinement of the COHESIVE Transition
There are a number of aspects of the proposal that need further scrutiny. Note that we have defined the COHESIVE transition by reference to any Cf in $U_i$. It might be the case, in fact, that there is motivation to define two different COHESIVE transitions: a
“CONTINUE COHESIVE,” in which the Cp(Ui) has the highest similarity measure with respect to Cf(Ui−1), and a “RETAIN COHESIVE,” in which some other Cf has the highest similarity to Cf(Ui−1). In our preliminary implementation with this corpus, the entity in Ui with the maximum semantic similarity to Ui−1 was the Cp of Ui 55% of the time. Whether such a distinction is a necessary or meaningful one is a completely open question.

In a similar vein, we might suggest that the Cf in Ui with the highest similarity to the Cfs in Ui−1 be designated as a type of Cb (say, Cb′). So, for example, if soba shop in (2b) had the highest level of similarity to restaurant in (2a), soba shop would be the Cb′ of that utterance. What would be the ramifications of this move for the theory?

Certainly there would be far fewer utterances with Cb = ?. In terms of tracking focus in the discourse, one of the major aims of the centering approach, this is a positive result. However, if the Cb′ is chosen simply on the basis of semantic similarity, we lose another major insight of centering theory, namely, that focus is not dependent upon semantics (Hudson-D’Zmura and Tanenhaus 1998) or upon word order (Gordon et al. 1999; but see also Gordon, Grosz, and Gilliom 1993), but upon the grammatical roles played by the Cfs. It is possible that for any given language, speakers tend to place Cfs that have strong semantic ties to the previous utterance in grammatical roles high on the Cf template, but this is an empirical question that we do not have the data to answer here.

5.2 Scalability
The incorporation of scalable considerations into a centering account provides enormous flexibility. It remains to be seen how best to use this added capability. For example, some COHESIVE transitions seem more cohesive than others. Compare the COHESIVE transition in (12) with the COHESIVE transitions in (13):

12a. どうやら梅雨入りしたらしい という
douyara tsuyuiri shitarashii toiu
rainy season beginning has happened

宣言 も出た ようで、
sengenmodeta youde
declaration with

COHESIVE

12b. 今日 も しっかり雨が降っています。
kyou mo shikkari ame ga futteimasu
today heavily rain SUBJ is raining

“With the declaration that the rainy season has come, it is raining heavily today.”

21 We can’t corroborate this, because sobya is not in the Gainen Base.
We have the intuition that the COHESIVE transition in (12) can be characterized as “more” COHESIVE than the ones in (13). Both of these examples come from the same message, and the semantic similarity measures for each of these examples confirms our intuitions: The measure for the COHESIVE transition between (12a) and (12b) is 0.583; for that between (13a) and (13b), 0.166; and for that between (13b) and (13c), 0.011.\(^2\) It is not clear to what use we might put this more detailed information about the varying levels of strength of connection among utterances; however, it parallels the observation in Fais (2001) that some e-mail authors make hierarchical distinctions in marking paragraphs in their messages, using line breaks to separate utterance clusters having some semantic connection and full spaces to separate utterance clusters having weak or no semantic connection.

Recall, too, that lexical cohesion, as measured by semantic distance, is only one aspect of coherence. Merging this measure with information provided by conjunctions,

\(^2\) We chose examples from the same message because comparison of raw measures across messages may not be informative (see Section 5.4). However, we expect that measures within messages can be compared with one another usefully.

The very low measure for the transition between (13b) and (13c) is consistent with the observation that the cohesion between these two clauses actually may not reside in the similarity between person and impressions, but in our intuitive understanding of a relationship between game and impressions (of the game).
referential form, and other aspects of discourse would allow a more comprehensive account of coherence.

5.3 Beyond Simplex Nominals
The ramifications of exploiting lexical cohesion within a centering theory are exciting. Up to this point, we have considered only the cohesion exhibited by discourse entities. But not only (simplex) nominal arguments enter into cohesive relationships. Consider (14), in which the recognition of the contribution of the verb to the cohesion of the excerpt could allow us to account more accurately for the coherence in the passage:

14a. watashi wa asuno doyoubi wa shusshashinai node
    I TOP tomorrow Saturday TOP won’t go to work since

14b. chiketto ya shuugoubasho nado wa
    tickets and meeting place and so on TOP

14c. hataraku hito gomennasai
    working people sorry

“Since I won’t go to the office tomorrow (Saturday), I’d like to discuss tickets and a meeting place on Monday (sorry to those who are working).”

The first transition shown is COHESIVE by virtue of the Cfs tomorrow and Saturday in (14a) and Monday in (14b). But considering only the discourse entities appearing in the Cf lists, working people in (14c) is not cohesive with any of the other elements in this passage, although the inclusion of (14c) in this portion of the message seems quite coherent and natural. However, if we allow verbal elements to contribute to cohesion, working people then shows cohesion with shusshashinai, ‘won’t go to work,’ and we can explain why the passage seems coherent.

In a similar vein, the incorporation of an adequate analysis for complex nominals would allow us to refine our measurement of cohesion as well. Recall (2), in which lexical cohesion exists among the entities enseki ‘restaurant,’ sake ‘alcohol,’ and sobaya, ‘soba shop.’ While the recognition of this cohesion allows us to characterize the coherence in the message, in fact, we probably underestimate the cohesion present if we base our measures on the individual lexical items listed. If we could calculate lexical cohesion not just for simplex nominals, but for complex nominals as well, we would calculate cohesion for enseki and for the entire phrase sake wo nomeru umai sobaya, ‘a
good soba shop, which also serves alcohol,’ possibly a much stronger cohesive connection than that among the individual items.23

5.4 Improving Implementation
We demonstrated in Section 4.6 that semantic distance as measured by the Gainen Base provides a feasible basis for a rigorous definition of lexical cohesion. In order for this or any similar implementation of semantic distance measure to be fully effective, however, two major areas need to be addressed. The first is coverage; while the lack of complete coverage of the corpus by the Gainen Base does not prevent us from making an assessment of the Gainen Base’s effectiveness over a subset of the data, it does make it impossible to characterize cohesion over the corpus as a whole. The second area that needs to be addressed has to do with definitions of distance for each transition type. Throughout our discussion we have examined semantic distances as relative strengths within messages; it would be useful to determine empirically whether it is possible to set definitive, independent levels for each transition type. These levels might be absolute (e.g., SMOOTH SHIFTS are those transitions having a semantic measure of 0.15–0.3) or, what seems more likely, relative, such that a portion of the range of semantic distance in a particular message is defined for each transition type (e.g., in a message in which semantic distances measure from 0.005 to 0.875, SMOOTH SHIFTS correspond to distances falling between 15% and 25% of the total range of 0.87, that is, from 0.13 to 0.22, for that message).

6. Conclusion
Upon subjecting a corpus of Japanese e-mail data to a centering analysis, it became clear that a centering description of these messages had to rely heavily on inferable centers and was not adequate to capture the coherence this corpus displays. We couched the notion of connectedness that inferable centers were intended to capture in terms of the more principled and explicit relation of lexical relatedness. We used this relation, then, to supplement the standard inventory of transitions with well-defined transition types that more accurately characterize the nature of coherence in this corpus and demonstrated the computational feasibility of this approach in a preliminary implementation. The proposed transition types provide a characterization of a previously unaccounted-for situation in centering theory, namely, the coherence of sequences of utterances containing inferable Cbs. This is a crucial improvement over the standard model because of the high number of nonexplicit Cbs in this corpus (and others; see Passonneau 1998). The inclusion of a COHESIVE transition and COMPLETE SHIFT allows us to characterize the 76% of the corpus in which nonexplicit Cbs play a part, a portion of the corpus undescribed in the standard approach, while maintaining the standard operation of the usual transition states of centering theory in cases in which explicit Cbs are present.

It may be the case that lexical cohesion and the notion of a COHESIVE transition can make other contributions to discourse analysis as well, such as allowing us to characterize coherence in a discourse while still recognizing referentially distinct discourse elements. Further, these notions can augment the definition of the topic of a discourse segment by virtue of the semantic information contained in a cohesion analysis. Finally, such an analysis can provide clues to discourse segment boundaries.

23 I would like to thank an anonymous reviewer for pointing this out.
The incorporation of the notion of lexical cohesion has been shown to be crucial to the characterization of the coherence in this corpus. In addition, it can make a variety of further contributions to discourse analysis. And finally, it opens the door to a number of useful related strategies that can allow us to come closer to understanding and comprehensively modeling coherence in discourse.

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