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Word Sense Acquisition from Bilingual Comparable Corpora

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Overview of the talk

1. Introduction
   Motivation, goal and related work
2. Unsupervised word sense disambiguation using bilingual comparable corpora
3. Proposed method for clustering translation equivalents of a polysemous word
4. Experiment using Wall Street Journal and *Nihon Keizai Shimbun* corpora
5. Discussion
   Advantages and limitations
6. Conclusion

1.1 Motivation

- Word sense disambiguation
  - A subtask necessary for most NLP tasks, esp. MT and IR
  - Great deal of research has been done over the past decade.
- Word sense acquisition
  - Human activity
  - Inventories of word senses have been constructed by lexicographers based on their intuition.
  - Problems with manual construction
    - High cost
    - Arbitrary division of word senses
    - Mismatch to application domains

1.2 Goal

- Unsupervised word sense disambiguation using bilingual comparable corpora (Kaji and Morimoto, COLING2002)
- Automatic word sense acquisition
  - Cluster translation equivalents of a polysemous word to divide and define the senses of that word.

[Example]

1.3 Related work

- Overlapping distributional word clustering—define word senses using sets of synonyms
  - Fukumoto and Tsujii, COLING 1994
    - Cluster synonyms of each target polysemous verb.
    - Each cluster represents a sense of the target word.
  - Pantel and Lin, KDD 2002
    - Cluster all nouns with occurrence frequencies larger than a threshold.
    - A polysemous word is assigned to multiple clusters, each of which represents one of its senses.
- Word sense discrimination
  - Schuetze, Computational Linguistics 1998
    - Cluster documents containing each target polysemous word.
    - Each document cluster corresponds to a sense of the target word. However it is not labeled.

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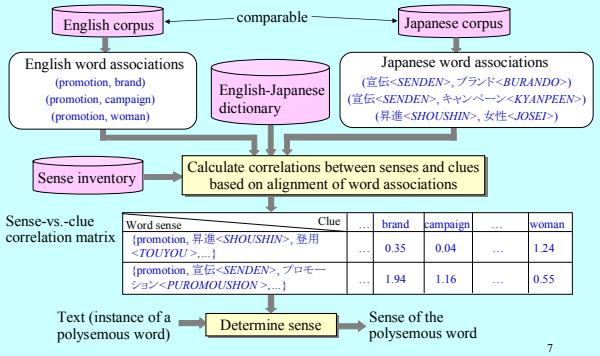
Unsupervised WSD using bilingual comparable corpora

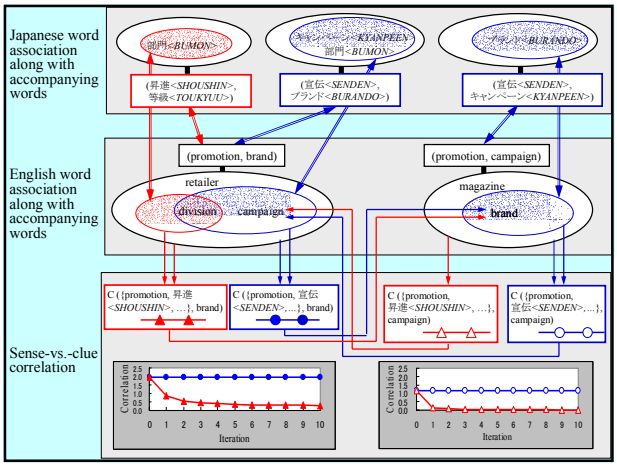


7

Iterative calculation of sense-vs.-clue correlations

- Problems in WSD using bilingual comparable corpora
  - Ambiguity in alignment of word associations
  - Failure in alignment of word associations caused by
    - Disparity of topical coverage between the texts of different languages
    - Incomplete coverage of the bilingual dictionary
- Solution
  - Define correlation between a sense of the target word and a clue as the mutual information of the target word and the clue multiplied by the maximum plausibility of alignments suggesting the sense-clue pair.
  - Calculate the correlations iteratively based on
    - Assumption 1: Plausibility of an alignment of word associations depends on plausibility of alignments between words accompanying those word associations.
    - Assumption 2: The correlation between a sense and a clue depends on the correlations between that sense and clues accompanying that clue.

8



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10

Translingual distributional clustering of translation equivalents

Initial set of senses: {promotion, 売り込み<*URIKOMI*>}, {promotion, 昇進<*SHOUSHIN*>}, {promotion, 宣伝<*SENDEN*>}, {promotion, 促進<*SOKUSHIN*>}, {promotion, 登用<*TOUYOU*>}, {promotion, プロモーション<*PUROMOUSHON*>}, ...

Calculate sense-vs.-clue correlations

Sense-vs.-clue correlation matrix

Word sense \ Clue	...	brand	...	campaign	...	woman
{promotion, 売り込み<*URIKOMI*>}	...	...	...	...	...	...
{promotion, 昇進<*SHOUSHIN*>}	...	...	...	...	...	...
{promotion, 宣伝<*SENDEN*>}	...	...	...	...	...	...
:	...	...	...	...	...	...

Calculate similarities between senses

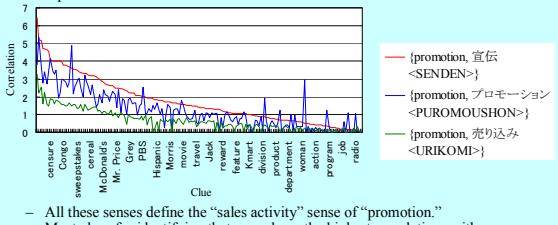
Merge similar senses

Updated set of senses: {promotion, 売り込み<*URIKOMI*>}, {promotion, 昇進<*SHOUSHIN*>, 登用<*TOUYOU*>}, {promotion, 宣伝<*SENDEN*>, プロモーション<*PUROMOUSHON*>}, {promotion, 促進<*SOKUSHIN*>}, ...

11

Improved similarity between senses using subordinate distribution patterns (1/2)

- Weakness of translingual distribution patterns:
  A clue has always high correlation with only one sense, and therefore translation equivalents representing the same sense do not necessarily have very similar patterns.
- Example



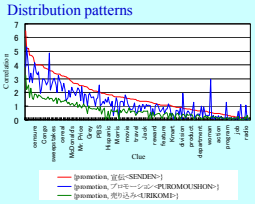
- All these senses define the "sales activity" sense of "promotion."
- Most clues for identifying that sense have the highest correlations with {promotion, **宣伝**<*SENDEN*>}, which is the most dominant translation equivalent of "promotion" in the corpus.

12

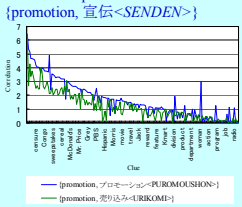
Improved similarity between senses using subordinate distribution patterns (2/2)

- Distribution pattern subordinate to S_1 :
  Distribution pattern resulting from the sense-vs.-clue correlation matrix for the set of senses excluding sense S_1
- Example

Distribution patterns



Distribution patterns subordinate to {promotion, 宣伝<*SENDEN*>}



- Similarity of sense S_2 to sense S_1:
  Similarity between distribution pattern for S_2 subordinate to S_1 and distribution pattern for S_1

13

Merger of restricted pairs of senses

- Merge senses S_1 and S_2
  if and only if
  S_1 is an active sense, i.e., the ratio of clues with which S_1 has the highest correlation exceeds a predetermined threshold,
  and
  S_2 has the mutually highest similarity to S_1.
- Exclude corpus-irrelevant translation equivalents from the clustering results.

14

Comparison with an alternative method

- Second-language monolingual distributional clustering
  - Characterize translation equivalents by their distribution patterns in the second language.
  - Problems
    - A polysemous translation equivalent is characterized by mixture of distribution patterns for the sense relevant to the target word and for those irrelevant to the target word.
    - Sparseness of co-occurrence data
- Translingual distributional clustering
  - Characterize translation equivalents by distribution patterns in the first language for the sense they represents.
  - Advantages
    - Even if it is polysemous, a translation equivalent is characterized by a distribution pattern for the sense relevant to the target word.
    - The iterative algorithm for calculating sense-vs.-clue correlations smoothes out the sparse data.

15

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16

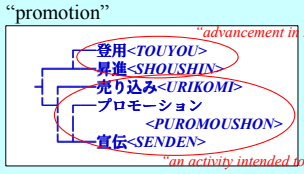
Experimental settings

- Training comparable corpus
  - Wall Street Journal (July, 1994 to December, 1995; 189 MB)
  - *Nihon Keizai Shimbun* (December, 1993 to November, 1994; 275 MB)
- Bilingual dictionary
  - The EDR (Japan Electronic Dictionary Research Institute) bilingual dictionary containing 633,000 pairs of 269,000 English nouns and 276,000 Japanese nouns
- Extraction of word associations
  - Nouns co-occurring in a window of 25 words excluding function words
  - Pairs of nouns with mutual information larger than 0.0
- Clustering of translation equivalents
  - Translation equivalents that occur more than 10 times in the training corpus

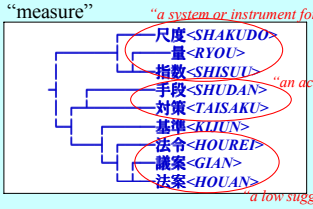
17

Examples of clustering

- "promotion"



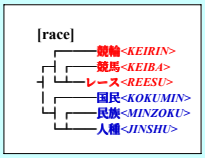
- "measure"

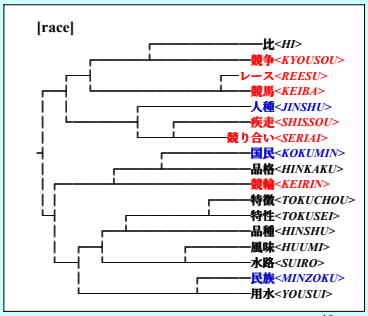


18

Clustering with proposed method and an alternative method

- Proposed method
- Monolingual distributional clustering





[Note]
Red words represent the "competition" sense of "race."
Blue words represent the "group of people" sense of "race."

19

Evaluation

- Difficulty in evaluating word sense acquisition methods
  - Prepare a standard sense inventory
    Translation equivalents used in the standard sense inventory do not always occur frequently in the training corpus.
  - Establish the complete set of senses appearing in the training corpus.
- Our evaluative measure
  - Recall of senses
    Judge a sense as acquired when the output dendrogram includes at least one translation equivalent defining that sense.
  - Accuracy of sense definitions
    Regard a set of senses as a set of pairs of translation equivalents defining the same senses, and define the accuracy of sense definitions as the maximum *F*-measure in all cycles of clustering.
- Evaluation results for 60 test English polysemous words
  - Recall of senses: 87% for senses whose ratios in the corpus were not less than 5%.
  - Accuracy of sense definitions: 77%

20

5. Discussion

- Advantages of our method
  - Corpus-dependent division and definition of word senses
  - Unify word sense acquisition with word sense disambiguation—acquire senses of distinguishable granularity.
  - Effective for translation equivalents with moderate occurrence frequencies
  - Moderate computational load—35 seconds per target polysemous word on a Windows 2000 server (CPU; Pentium 4 (1.9 GHz), memory: 2 GB)
- Limitations and directions for extension
  - Difficulty in determining how many senses are appropriate for each target word
    Avoid merging senses having complementary distribution patterns.
  - Effective for topical senses but ineffective for generic senses
    Use syntactic co-occurrence.

21

6. Conclusion

- Unified approach to word sense acquisition and word sense disambiguation using a bilingual comparable corpus and a bilingual dictionary
- An experiment using Wall Street Journal and *Nihon Keizai Shimbun* corpora and the EDR bilingual dictionary demonstrated the effectiveness of the method.

22