Error Analysis of Noun Phrase Anaphora Resolution using a Machine Learning-based Approach

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Background

- ■There is a huge amount of text data on the Web
- Natural Language Processing (NLP) techniques are a required component for ubiquitous computing
 - Machine Translation, Information Extraction and Question Answering
- ■Inter-sentential anaphora resolution is a major obstacle to their progress

Anaphora resolution

■ Anaphora resolution is the process of determining whether two expressions in natural language refer to the same entity in the world

A federal judge in Pittsburgh issued a temporary restraining order preventing Trans World Airlines from buying additional shares of USAir Group Inc. antecedent anaphor

The order, requested in a suit filed by USAir, dealt another blow to TWA's bid to buy the company for \$52 a share.

- Anaphora resolution is classified into two processes 1. Anaphoricity determination
 - 2 Antecedent identification

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- Anaphora resolution is classified into two processes
- non-anaphor
- **1.Anaphoricity determination** is the task of classifying whether a given noun phrase (NP) is *anaphoric* or *non-anaphoric*
- 2.Antecedent identification is the identification of the antecedent of a given an aphoric NP

Anaphora resolution

Anaphora resolution is the process of determining whether two expressions in natural language refer to the same entity in the world



- ■Anaphora resolution is classified into two processes
 - **1.Anaphoricity determination** is the task of classifying whether a given noun phrase (NP) is *anaphoric* or *non-anaphoric*
 - **2.Antecedent identification** is process of identifying the antecedent of a given anaphoric NP

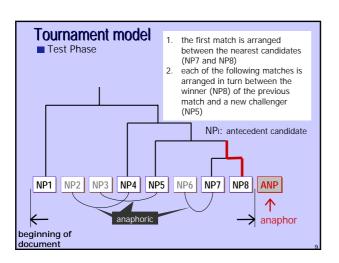
Talk outlines

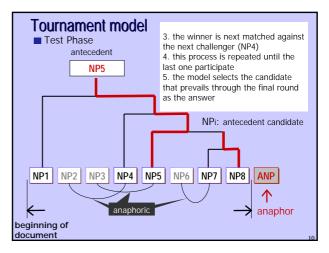
- 1. Our previous work on anaphora resolution
- 2. Error analysis of Japanese NP anaphora resolution
- 3. Discussion and future work

Our previous work on anaphora resolution

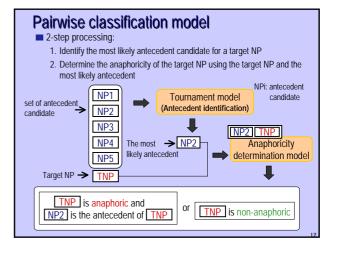
- Antecedent identification model (tournament model)
- Anaphoricity determination model (pairwise classification model using the most likely antecedents)

Antecedent identification model ■Tournament model (lida , 03) ■ A model which makes pair-wise comparisons between candidates to capture the preferences of the antecedents





Anaphoricity determination model ■ Determining anaphoricity using the most likely antecedents



Current performance of our anaphora resolution model

Antecedent identification for given anaphors:

■ Precision: **87.4%** (773/884)

■Anaphoricity determination

■Recall: 67.1 % (593/884)
■Precision: 79.5 % (593/746)

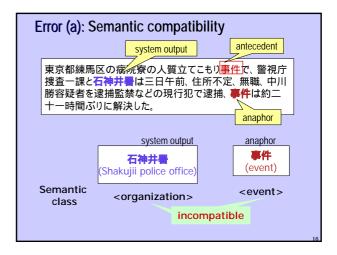
■Overall performance

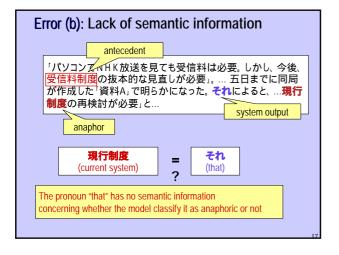
■ Recall: 65.3% (577/884) ■ Precision: 77.3% (577/746)

Error analysis of Japanese NP anaphora resolution

- · Investigating the source of errors manually
- 1. Antecedent identification
 - 111 anaphors
- 2. Anaphoricity determination
 - · Sampled 100 anaphors
 - Sampled 100 non-anaphors

Error analysis: Antecedent identification ■ Investigating examples that our system could not identify the correct antecedent for a given anaphor Main source of errors Percentage (#) (a) Semantic compatibility 36.9 % (41/111) (b) Lack of semantic information 5.4 % (6/111) (c) Referring to set expressions 7.2 % (8/111) (d) Relationship between entities 9.0 % (10/111) 7.2 % (e) Annotation error (8/111)(f) Others 35.1 % (39/111) (Each example is assigned to more than one source of errors)

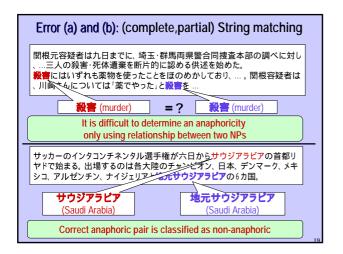


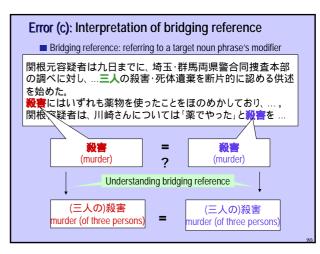


Error analysis: Anaphoricity determination (1)

 Sampling 100 anaphors classified as non-anaphoric for given correct anaphors

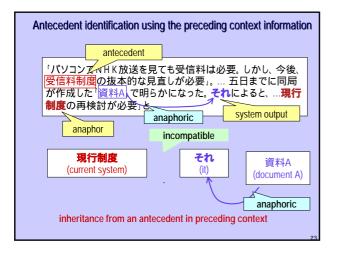
	Main source of errors	Percentage (#)
	(a) Complete string matching	71% (71/100)
	(b) Partial string matching	23% (23/100)
	(c) Interpretation of bridging reference	49% (49/100)
	(d) Lack of semantic information	7% (7/100)
	(e) Annotation error	1% (1/100)
(Each example is assigned to more than one source of errors)		





Error analysis: Anaphoricity determination (2) ■ Sampling 100 non-anaphors classified as anaphoric Main source of errors Percentage (#) (a) Relationship between entities 12% (12/100) (b) Complete string matching 43% (43/100) (c) Partial string matching 26% (26/100) (d) Interpretation of bridging reference 42% (42/100) (e) Lack of semantic information 2% (2/100) (f) Annotation error 14% (14/100) (Each example is assigned to more than one source of errors)

Future work Improvement of semantic compatibility Constructing Fine-grained semantic classes by combining existing linguistic resources Antecedent identification using the preceding context information Resolving bridging reference



Resolving bridging reference

- Creating a corpus annotated bridging reference
 - Defining bridging reference is difficult
 - inter-annotator agreement is very low
 - → We need to set specifications for annotating corpora
- Developing a bridging reference resolution model
 - utilizing A of B (A-no-B) co-occurrence information extracted from large corpora
- Noun phrases anaphora resolution utilizing resolved bridging reference

Summary

- Explaining our framework of Noun phrases anaphora resolution
- ■Main source of errors:
 - Antecedent identification
 - Anaphoricity determination
- ■Denoting future work based on our error analysis