

Corpus-based Question Answering for *why*-Questions

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Overview

- **Corpus-based approach for whyQA**
 - Use relation-annotated corpora to automatically acquire causal expression patterns
 - Use a corpus of why-questions and answers to train an answer-candidate ranker
- Implemented system: **NAZEQA**
- Experimental results

WhyQA

- Answering generic “Why X?” questions by extracting causes from a text archive
 - E.g., Why do ostriches run fast?
- Few implemented systems
- Previous approach (Fukumoto, 2007)
 - Uses hand-crafted patterns to extract cause-bearing passages
 - Patterns rely on explicit causal cue words e.g., **tame**, **node**, **de**, **kara** (because/by/from)

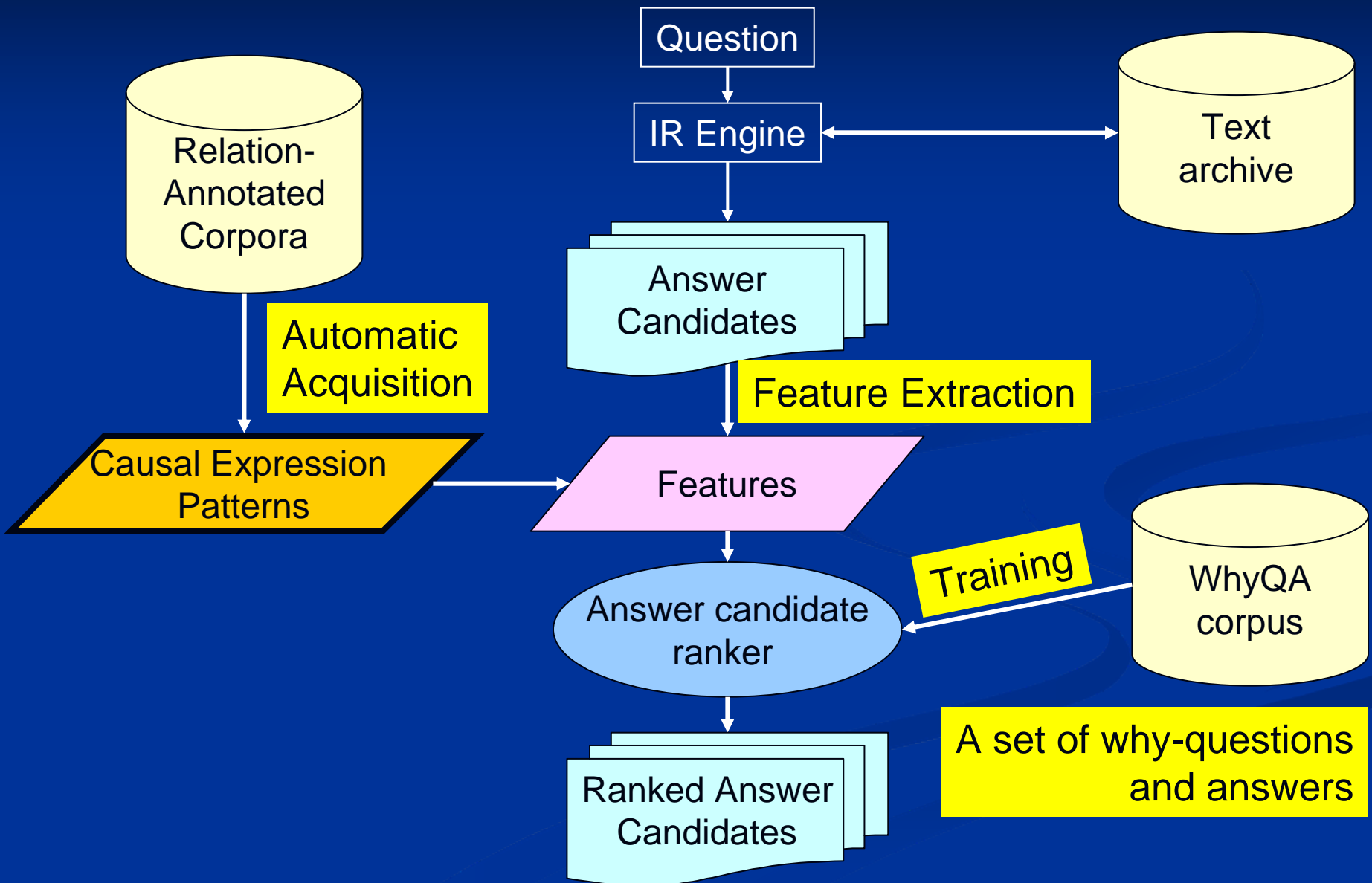
Problem

- **Hard to cover causal expressions by hand**
 - Causes are expressed by a wide variety of expressions
 - Half of causes are not marked by cue words (Inui and Okumura, 2005)
 - Cue words are not always reliable
 - Only 6-7 % of “de” (by) trigger causes (Abekawa and Okumura, 2004)
- **Hand-crafted patterns are costly to make**

Approach

- Automatically acquire causal expression patterns from relation-annotated corpora
 - FrameNet, PropBank, EDR Corpus
 - Typically annotated with a causal relation
- Use the acquired patterns to create features to represent answer candidates
- Train an answer-candidate ranker that ranks answer candidates on the basis of the features

Approach (cont'd)



Implemented System: NAZEQA

- A Japanese whyQA System
- Uses the EDR corpus as a relation annotated corpus to acquire causal expression patterns
- Derives three types of features for answer candidates
- Uses Ranking SVM for ranker training
- Uses a manually constructed WhyQA corpus (WHYQA Collection)

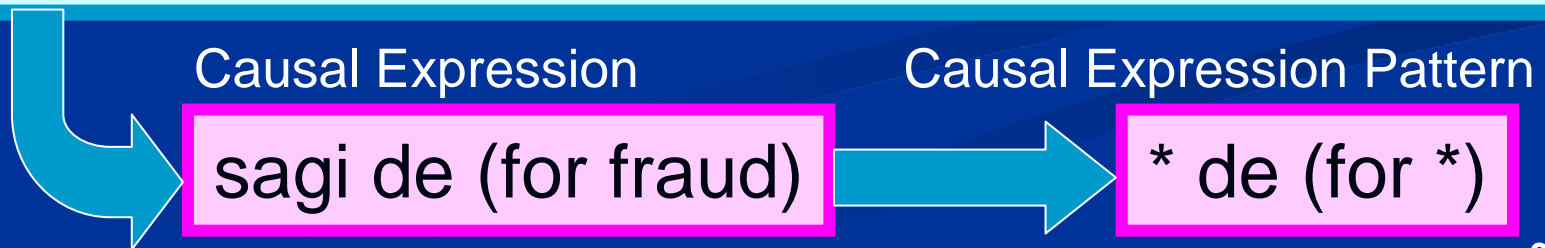
Acquiring Causal Expression Patterns from the EDR Corpus

- The EDR Corpus
 - A collection of Japanese sentences from various sources
 - Provides a semantic representation for each sentence
 - Specifies relations of content words
 - Relations include a causal relation
 - **8,474 text spans annotated with a causal relation** in 207,802 sentences

Acquiring Causal Expression Patterns from the EDR Corpus

- Extract text spans annotated with a causal relation as causal expressions
- Convert the expressions into patterns by
 - Leaving only functional words
 - Auxiliary verbs, case, aspect, tense markers
 - Replacing others with wildcards “*”

Kare ha **sagi de** taiho sareta (He was arrested **for fraud**)




Acquired Causal Expression Patterns from the EDR Corpus

- Obtained 394 distinct causal expression patterns
- Examples
 - de (by/for)
 - tame (because)
 - niyoru, niyotte (because of)
 - no (-GEN) * wa (topic marker)
 - de (by/for) * wo (-ACC) * teshimai (-PERF)
 - koto niyotte (by the fact that)

Feature Extraction

- For an answer candidate to be the correct answer, it should
 - have a causal expression
 - be similar to the question in content
 - show some causal relation to the question
- Three types of features to represent how each candidate satisfies each condition
 - **Causal Expression Features**
 - **Content Similarity Features**
 - **Causal Relation Features**



Based on the acquired patterns

Causal Expression Features

- **AUTO-Causal Expression Features**
 - 394 binary features representing if each acquired pattern matches the answer candidate
- **MAN-Causal Expression Feature**
 - A binary feature representing if the answer candidate is matched by existing hand-crafted patterns (Fukumoto, 2007)

Content Similarity Features

- **Question-Cand Cosine Similarity feature**
 - Cosine similarity of word frequency vectors
- **Document-Question Relevance feature**
 - Similarity between the question and the document where the candidate is found
 - Inverse rank of the retrieved document
- **Synonym Pair feature**
 - A binary feature indicating if synonyms are found in the question and answer candidate

Causal Relation Features

■ Cause-Effect Pair feature

- A binary feature representing if a cause-effect word pair is found in the answer candidate and the question
- Cause-effect word pairs
 - Explode → Die
 - Murder → Arrest
 - Fraud → Arrest
- Generated from the EDR concept dictionary

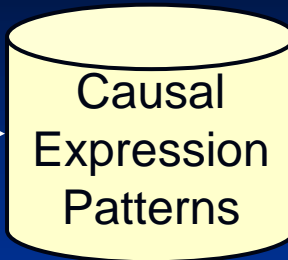
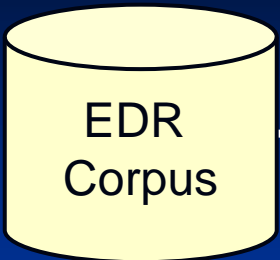
Q: Why was John arrested?

Cand: John was arrested for fraud.

Cause-Effect pair feature = 1

Q: Why was John arrested?

Cand: John was arrested for suspicion of fraud.



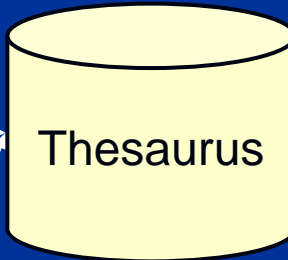
Causal Expression Features

Pattern1 for * of
Pattern2 by * of
Pattern3 due to
Pattern394 because

Hand-crafted patterns
(Fukumoto, 2007)

Matches pattern1 1
Matches pattern2 0
Matches pattern3 0
Matches pattern394 0

Matches hand-crafted
patterns 0



Question-Candidate Cosine Similarity

Document-Question Relevance

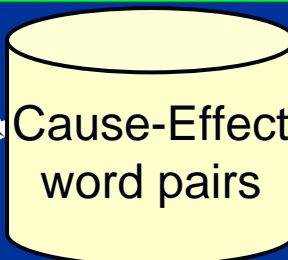
Examples
arrest ⇔ apprehend
arrest ⇔ capture
arrest ⇔ nip

0.3

0.1

Q and Cand have
synonyms 1

Content Similarity Features



Examples
explode → die
murder → arrest
fraud → arrest

Causal Relation Feature

Q and Cand have
A cause-effect word
pair 1

WHYQA Collection

- 1,000 sets of why-questions and answers
- Procedure
 - Create a why-question
 - Retrieve top-20 documents for the question
 - Extract all sentences in the top-20 documents
 - Create sentence-level answers by selecting sentences that contain answers
 - Create paragraph-level answers by selecting paragraphs that contain answer sentences

Training an Answer-candidate Ranker

- **Ranking SVM** (Joachims, 2002)
 - Learns ranking by minimizing pairwise ranking error
- Training data
 - WHYQA Collection
 - Train a ranker so that answers (e.g., sentences selected as answers) get ranked higher than non-answers
- Trained answer-candidate rankers for sentence and paragraph-levels

Experiment

- Compare NAZEQA with two baselines
 - **Baseline-1 (COS)**
 - Uses cosine similarity to rank answer candidates
 - **Baseline-2 (FK) (Fukumoto, 2007)**
 - Selects answer candidates by hand-crafted patterns
 - Ranks the candidates by cosine similarity
- NAZEQA and baselines process the same answer candidates
 - All sentences/paragraphs in top-20 documents retrieved by an IR engine

Experiment (cont'd)

- Question set
 - All questions in the WHYQA Collection
 - 10-fold cross validation to evaluate NAZEQA
- Evaluation criteria
 - Mean Reciprocal Rank (MRR)
 - Coverage
(Rate of questions correctly answered by top-N answers)

Results

	MRR			Coverage		
Top-N	COS	FK	NAZEQA	COS	FK	NAZEQA
Sentences as answer candidates						
Top-1	0.036	0.091++	0.113	3.6%	9.1%	11.3%
Top-5	0.086	0.139++	0.196**	19.1%	23.1%	35.4%
Top-10	0.102	0.149++	0.216**	31.3%	30.7%	50.4%
Top-20	0.115	0.152	0.227**	51.4%	35.5%	66.6%
Paragraphs as answer candidates						
Top-1	0.065	0.152++	0.186*	6.5%	15.2%	18.6%
Top-5	0.140	0.245++	0.305**	29.2%	41.6%	53.1%
Top-10	0.166	0.257++	0.328**	48.8%	50.5%	70.3%
Top-20	0.181	0.262++	0.339**	70.7%	56.4%	85.6%

** (p<0.01) * (p<0.05) NAZEQA's statistical significance over FK
++ (p<0.01) FK's statistical significance over COS

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NAZEQA significantly outperforms FK

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High coverage achieved by NAZEQA

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- Low coverage of FK
- Possible limitation of hand-crafted patterns

High coverage achieved by NAZEQA

Impact of features

Feature Set	Sentence (Top-5 MRR)	Paragraph (Top-5 MRR)
All features	0.181	0.287
w/o AUTO-Causal Expression	0.138**	0.217**
w/o MAN-Causal Expression	0.179	0.286
w/o Question-Candidate Cosine Similarity	0.131**	0.188**
w/o Document-Question Relevance	0.161	0.275
w/o Synonym Pair	0.180	0.282
w/o Cause-Effect Pair	0.184	0.287

MRR drops significantly without AUTO-Causal expression features → **Verifies effectiveness of automatically acquired causal expression patterns**

Weights of features assigned by Ranking SVM

Rank	Feature Name	Weight
1	Question-Candidate Cosine Similarity	4.66
2	Exp.[<i>de</i> (by) * <i>wo</i> (-ACC) * <i>teshimai</i> (-PERF)]	1.86
3	Exp.[<i>no</i> (of) * <i>niyote wa</i> (according to)]	1.44
4	Exp.[<i>no</i> (of) * <i>na</i> (AUX) * <i>no</i> (of) * <i>de</i> (by)]	1.42
5	Exp.[<i>no</i> (of) * <i>ya</i> (or) * <i>niyotte</i> (by)]	1.35
6	Exp.[<i>no</i> (of) * <i>ya</i> (or) * <i>no</i> (of) * <i>de</i> (by)]	1.30
7	Exp.[<i>na</i> (AUX) * <i>niyotte</i> (by)]	1.23
8	Exp.[<i>koto niyotte</i> (by the fact that)]	1.22
9	Exp.[<i>to</i> (and) * <i>no</i> (of) * <i>niyotte</i> (by)]	1.20
10	Document-Question Relevance	0.89
	⋮	
27	Synonym Pair	0.40
102	MAN-Causal Expression	0.16
127	Cause-Effect Pair	0.15

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10	Document-Question Relevance	0.89

- Many complex patterns were found to be important
- Difficult to cover such patterns by hand
 - ➔ shows effectiveness of an automatic approach

Conclusion

- **Corpus-based approach for whyQA**
 - Use relation-annotated corpora to acquire causal expression patterns
 - Train an answer-candidate ranker using a corpus of why-questions and answers
- **Experimental results show validity of our approach**
- **Future work**
 - Use other relations (e.g., purpose) to increase coverage of causal expressions
 - Incorporate syntactic and semantic features