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Investigating the characteristics of the causal relations in Japanese text

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Background & research goal (1/2)

Background It rains hard → Flooding occurs

- Some applications (e.g. dialog system and QA system) need inference rules or **knowledge of causal relation**.
- A large amount of text documents are used as resources in causal knowledge acquisition system

[Girju02, Torisawa03, Inui04].

Problem

- The characteristics of causal relation expressions in text is **unclear**.
- Low knowledge acquisition performance

Background & research goal (2/2)

Research goal

- Investigating the characteristics of causal relation expressions in text

To do this,

- We created a corpus tagged with causal relation information.
- We used Japanese text.

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- Background & research goal
- Corpus construction
- Investigation
 - Suggestion for developing knowledge acquisition methods for causal relation

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Target expressions

- We treat a variety of expressions with wide coverage.

Our target expressions

- explicit (e.g. [Altenberg 84])
 - Flooding occurs **because** it rains hard.

VP explicit cue phrase marker VP
- implicit
 - It rains hard, and flooding occurs.
- noun phrase
 - Flooding occurs because of **heavy rains**.

NP
- between sentences
 - S1**: It rains hard. **S2**: Flooding occurs.

New targets

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- **Corpus construction**
 - ◆ Target expressions
 - ◆ **Tags**
 - ◆ Annotation criteria
- Investigation
 - ◆ Suggestion for developing knowledge acquisition methods for causal relation

Tags (1/3)

- We use 4 tags
 - ◆ For events: *mod* *head*
(modifier)
 - ◆ For causal relations between two events: *causal_rel*
 - ◆ For cue phrase markers: *marker*
- Tagging to *bunsetsu*-phrase (BP) chunk units
BP: fundamental units in Japanese

Ex. *ooame-ga* *fu-tta* BP
heavy rain-NOM fall-PAST

Tags (2/3)

- Annotation scheme for events
 - ◆ Similar to PropBank [Palmer 05]
 - ◆ An event = one head BP + some modification BPs

Ex. *ooame-ga* *fu-tta*
heavy rain-NOM fall-PAST
It rains hard.

Tags (3/3)

- Annotation scheme for causal relations

Ex.
ooame-ga fu-tta tame kawa-ga zousui-sita
heavy rain-NOM fall-PAST because river-NOM rise-PAST
cause effect
Flooding occurs because it rains hard.
effect cause

Tags (3/3)

- Annotation scheme for causal relations

Ex. *ooame-ga* *fu-tta* *tame* *kawa-ga* *zousui-sita*
heavy rain-NOM fall-PAST because river-NOM rise-PAST
cause effect
Flooding occurs because it rains hard.
effect cause

causal_rel

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Annotation criteria (1/2)

- Our criteria are based on linguistic templates.
 - ◆ The templates have **two slots** and **a explicit cue phrase marker**.

slot

Example of linguistic templates

The judgements process

1. Embedding two events in the **[cause]** and **[effect]** slots.
2. If the template is syntactically and semantically correct, a causal relation is supposed to hold between two events.

Annotation criteria (2/2)

- The alternative templates
 - ◆ One adverb (**often, usually, always**) is included.
 - ◆ Introducing an attribute (**necessity / chance**) to causal relations
- The alternative template → **necessity** relation
The basic template → **chance** relation

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- Investigation

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- Background & research goal
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- **Investigation**

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- Background & research goal
- Corpus construction
- **Investigation**
 - ◆ Summary of our corpus
 - ◆ Point 1: Cue phrase markers
 - ◆ Point 2: Syntactic categories
 - ◆ Point 3: The positions of events

Summary of our corpus (1/2)

- 750 newspaper article documents in [Mainichi95]
- Three annotators
- Reading docs, detecting causal relation and tagging

annotator	causal relation instances
A	2014
B	1587
C	1048

◆ The difference is mainly caused by missing judgements.
◆ In particular, for the implicit expressions, annotators tend to miss the judgements.

Summary of our corpus (2/2)

- Inter-annotator agreement

	#agreement	necessity	chance
no-agreement	one (A, B, C)	1253(0.64)	997(0.76)
	two (AB,BC,AC)	429(0.22)	250(0.19)
agreement	three (ABC)	270(0.14)	64(0.05)
	two & three	699(0.36)	314(0.24)

Summary of our corpus (2/2)

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■ The **necessity** causal instances tend to be in *agreement*. (statistically significant)

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 - ◆ Point 1: Cue phrase markers
 - ◆ Point 2: Syntactic categories
 - ◆ Point 3: The positions of events

Cue phrase markers

- Proportion of instances attached with the **marker** tag

Flooding occurs **because** it rains hard.

marker

	freq.
with marker (explicit)	219
without marker (implicit)	480
Total	699

- 70% instances do **not** have a explicit cue phrase marker.

Syntactic categories

- Proportion of the syntactic categories of the **cause** event and the **effect** event in the causal instances.

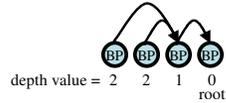
category	cause	effect
VP	365	412
NP	322	269
others	12	18

- More than half of the events are classified as VP.
- # of NPs is comparable to # of VPs.

- ### The positions of events (1/6)
- We investigated two types.
1. Positions of both the **cause** event and the **effect** event
 2. Relative positions between the **cause** event and the **effect** event in the same causal instances

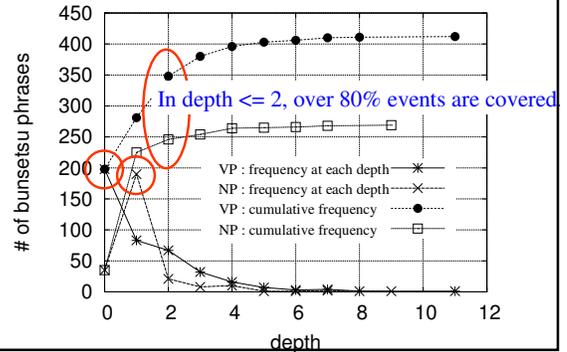
The positions of events (2/6)

1. Positions of both the **cause** event and the **effect** event

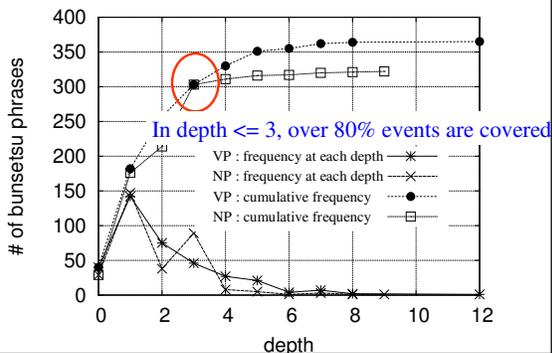


- Dependency tree structures between *bunsetsu*-phrases (BPs)
 - ◆ The end of BP in a sentence is the root node, depth =0.
- We consider the depth of a BP including the *head* tag as the position of an event.

The positions of events: **effect** (3/6)



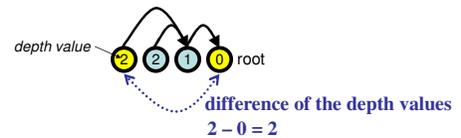
The positions of events: **cause** (4/6)



The positions of events (5/6)

2. Relative positions between the **cause** event and the **effect** event in the causal instances

- Difference of the depth values between the **cause** event and the **effect** event



The positions of events (6/6)

2. Relative positions between the **cause** event and the **effect** event in the causal instances

		freq.
intra-sentential	difference = 1	259
	= 2	152
	> 2	33
	no dependency	72
inter-sentential		141

Summary of investigation

- 70% of instances are expressed without markers.
- Over 50% of instances are formed by VP, but # of NPs is comparable to # of VPs.
- **cause**; depth=1, **effect**; VP: depth=0 / NP: depth=1.
- The **cause** events directly depend on the **effect** events.



- Existing knowledge acquisition methods
 - ◆ "VP marker VP"

Only some limited patterns of expressions have been treated.

Summary

- We constructed a Japanese corpus tagged with causal relation information.
- Using this corpus, we investigated the characteristics of causal relation expressions from some viewpoints.
- We obtain some useful suggestions for developing knowledge acquisition methods for causal relation.

Thank you!