

Data-Driven and Ontological Analysis of FrameNet for Natural Language Reasoning

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Introduction

- ▶ Lexical-**semantic** knowledge for reasoning
 - ▶ WordNet [*Morato et al., 2004*]
 - search
 - information extraction
 - ...
 - ▶ **FrameNet**
 - question answering [*Shen and Lapata, 2007*]
 - recognizing textual entailment [*Burchardt et al., 2009*]
 - ...

Introduction

- ▶ Shortcomings of FrameNet with regard to NL reasoning
 - ▶ low coverage [*Shen and Lapata, 2007; Cao et al., 2008*]
 - ▶ **conceptual inconsistency and lack of axiomatization**

Our focus

methodology for improving the conceptual structure of FrameNet for the goals of NL reasoning

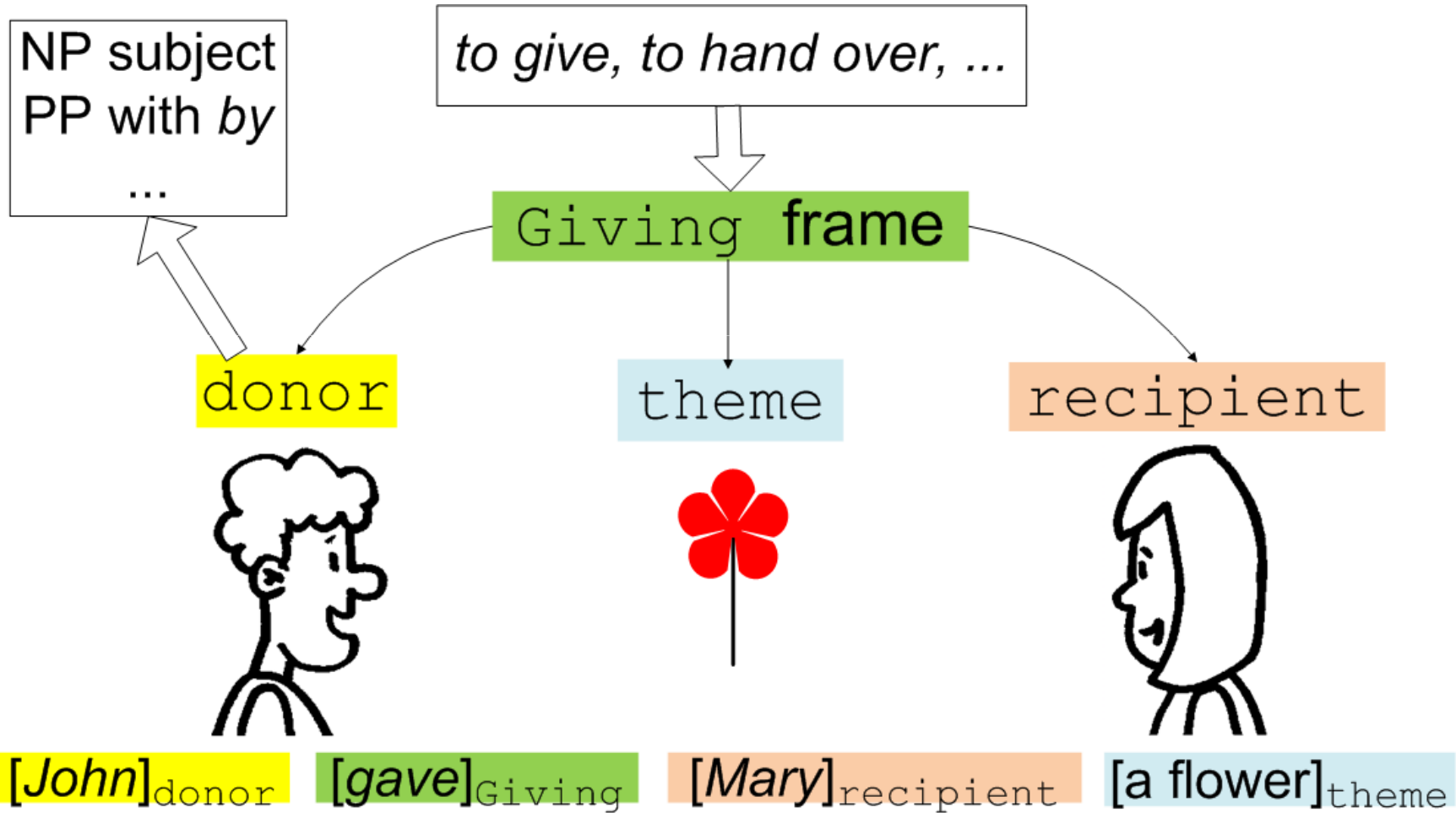
Outline

1. FrameNet for Reasoning
2. Proposed Methodology
3. Conceptual Problems
4. Data-Driven Analysis
5. Ontological Analysis
6. Case Study
7. Conclusion

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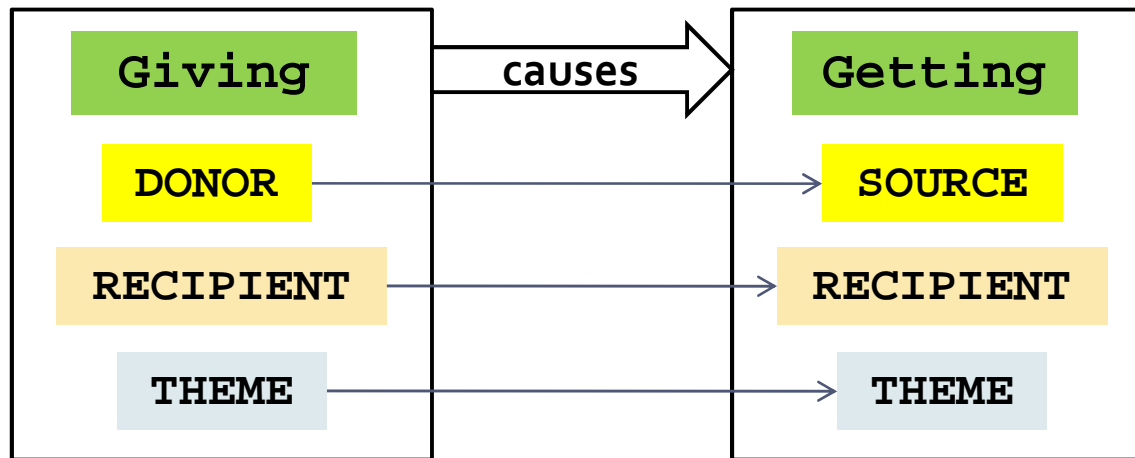
FrameNet for reasoning



FrameNet for reasoning

(a) [*John*]_{DONOR} [*gave*]_{Giving} [*Mary*]_{RECIPIENT} [*a flower*]_{THEME}

(b) [*Mary*]_{RECIPIENT} [*got*]_{Getting} [*a flower*]_{THEME} [*from John*]_{SOURCE}



Frame relations

1. Inheritance: 441
 - ▶ **Vehicle - Arfitact, Motion_directional - Motion**
2. Precedence: 55
 - ▶ **Being_awake - Fall_asleep**
3. Perspective: 43
 - ▶ **Buy, Sell - Goods_transfer**
4. Causation: 49
 - ▶ **Giving - Getting**
5. Subframe: 87
 - ▶ **Trial, Sentencing - Criminal_process**
6. Using: 426
 - ▶ **Recovery - Medical_conditions**
7. See_also: 669
 - ▶ **Scrunity - Seeking**

Research goals

1. Axiomatizing frame relations
2. Finding missing frame relations
3. Cleaning up frame relations
4. Applying frame relations to NL reasoning

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Proposed improvement methodology

1. Conceptual problems in FrameNet :
Frame-Annotated Corpus for Textual Entailment (FATE)
2. Clustering frames
3. Ontological analysis of frames and frame relations
 - ▶ axiomatizing frame relations
 - ▶ constraints on frame relations
4. Evaluation: enriched, axiomatized and cleaned up
frame relations in RTE

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Frame-Annotated Corpus for Textual Entailment

FATE [*Burchardt & Pennacchiotti, 2008*]

- ▶ 800 *T-H* entailment pairs annotated with FrameNet frames and roles
- ▶ we have analyzed cases when *T* was known to entail *H* (400 pairs) applying a frame matching strategy

FATE analysis results

- ▶ **170** pairs: matching is possible
- ▶ **131** pairs: this approach does not work
 - ▶ annotation disagreements
 - ▶ different conceptualizations of *T* and *H*
- ▶ **99 pairs**: the same facts in *T* and *H* are represented by different frames which are related semantically and could be mapped on each other with the help of reasoning
 - ▶ FrameNet enables inferences only for **17** pairs

Discovered problems

1. missing relations

(t) ...X [*survived*]_{Surviving} Sars...

(h) ...X [*recovered*]_{Recovery} from Sars...

2. problems in the relational structure

...[*parts*]_{Part_whole} [*of Aceh province*]_{WHOLE}...

Part_whole → **Part_piece**, **WHOLE** → **SUBSTANCE**

3. missing axiomatization of the relations

(t) ...X [*recovered*]_{Recovery} from Sars...

(h) ...X [*was ill*]_{Medical_conditions}...

Recovery uses Medical_conditions

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Clustering frames

For every two frames f_1 and f_2 we apply similarity measures based on [Pennacchiotti & Wirth, 2009] :

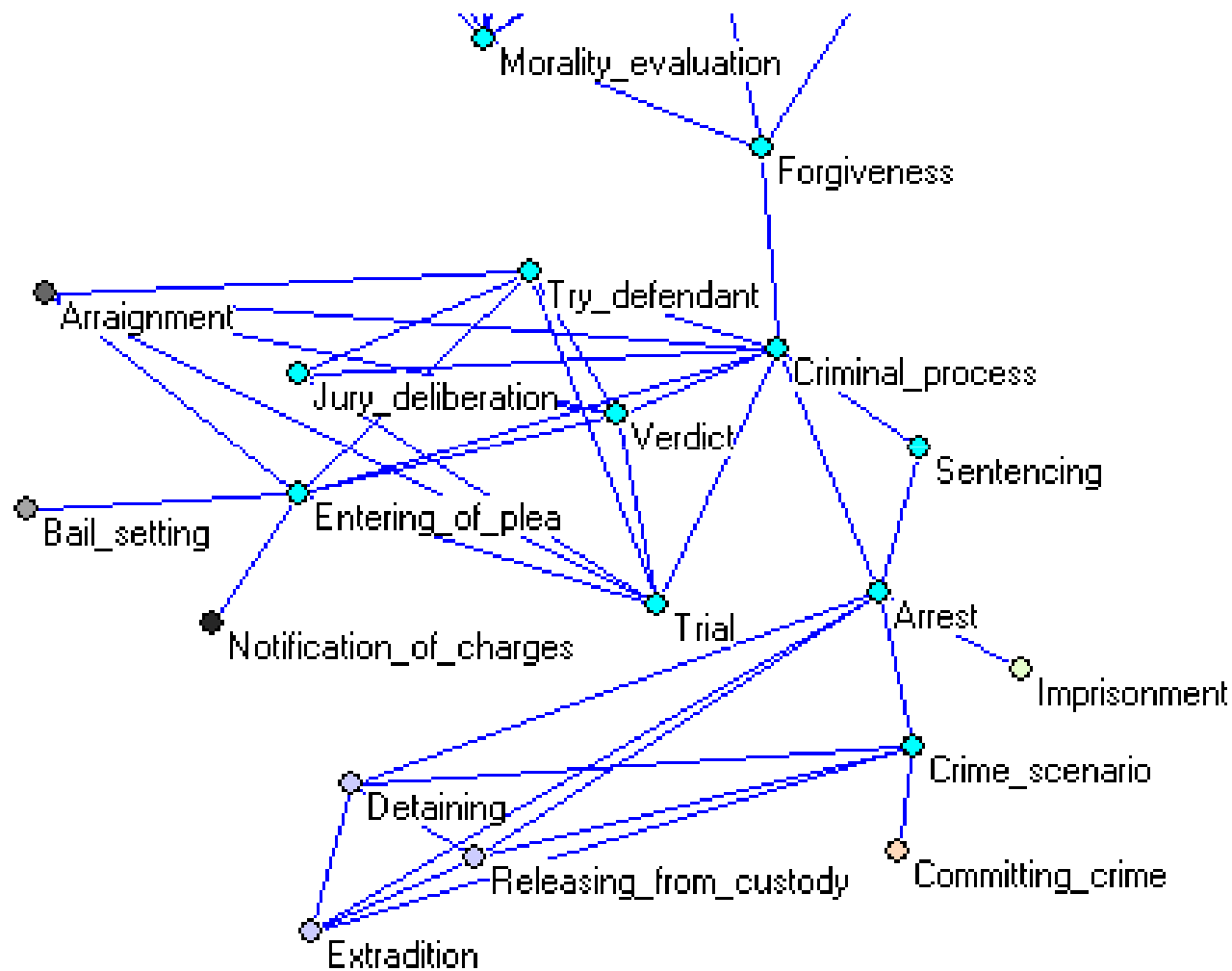
1. overlapping frame elements in f_1 and f_2
2. co-occurrence of lexemes evoking f_1 and f_2 in corpora (pmi)

Clustering results

1. Clusters based on overlapping frame elements
 - ▶ 228 clusters in total
 - ▶ 1497 relations not contained in FrameNet
 - ▶ 73 clusters from 100 random contain semantically related frames (2 experts, agreement 0.85)
2. Clusters based on co-occurrence of lexemes evoking frames
 - ▶ 113 clusters in total
 - ▶ 1149 relations not contained in FrameNet
 - ▶ 65 clusters from 100 random contain semantically related frames (2 experts, agreement 0.85)

Frame clusters: visualization

(Pajek tool)



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Frames and situations

What do frames describe?

- ▶ Frames abstract from natural language expressions (predicates with their arguments)
- ▶ Natural language expressions describe situations
- ▶ Frames can be seen as abstractions from situations

Types of situations

From which types of situations do frames abstract?

- ▶ categories from the DOLCE ontology [*Masolo et al., 2002*] for describing types of situations

Types of situations:

1. „Event“ situation
 - e.g. **Motion** (*John is running in the park*)
2. „Object“ situation
 - e.g. **People** (*A man*)
3. „Quality“ situation
 - e.g. **Color** (*This rose is red*)
4. „Relation“ situation
 - e.g. **Part_whole** (*This park is a part of the town*)

Situations and time

1. Situations having temporal qualities

- ▶ *John is running in the park, a clerk, This rose is red, John is next to Mary*
- ▶ can participate in temporal relations (*precedence, temporal inclusion* etc.)

2. Non-temporal situations

- ▶ *A man, The war lasted four years, Einstein's birth preceded my birth*
- ▶ cannot participate in temporal relation

Causation: f_1 is causative of f_2

$$\forall s_1 (f_1(s_1) \rightarrow \exists s_2 (f_2(s_2) \wedge \text{causes}(s_1, s_2)))$$

$$\forall s_1 s_2 (\text{causes}(s_1, s_2) \rightarrow \neg \text{starts_before}(s_2, s_1))$$

Subframe: f_1 is subframe of f_2

1. Subframe of "Events"

$$\forall s_1 s_2 (sub_ev(s_1, s_2) \rightarrow (strict_temp_inc(s_2, s_1) \wedge spatially_inc(s_2, s_1)))$$

- part presupposes whole

$$\forall s_1 (f_1(s_1) \rightarrow \exists s_2 (f_2(s_2) \wedge sub_ev(s_1, s_2)))$$

- whole presupposes part

$$\forall s_2 (f_2(s_2) \rightarrow \exists s_1 (f_1(s_1) \wedge sub_ev(s_1, s_2)))$$

2. Subframe of "Objects"

- part presupposes whole

$$\forall s_1 en_1 (f_1(s_1) \wedge FE_1(s_1, en_1) \rightarrow \exists s_2 en_2 (f_2(s_2) \wedge FE_2(s_2, en_2) \wedge part_of(en_1, en_2)))$$

- whole presupposes part

$$\forall s_2 en_2 (f_2(s_2) \wedge FE_2(s_2, en_2) \rightarrow \exists s_1 en_1 (f_1(s_1) \wedge FE_1(s_1, en_1) \wedge part_of(en_1, en_2)))$$

Using and See_also

- ▶ the most frequent relations in FN
- ▶ sometimes can be represented in terms of other axiomatized relations

- ▶ otherwise

$$\forall s_1 (f_1(s_1) \rightarrow \exists s_2 (f_2(s_2) \wedge \textit{depends}(s_1, s_2)))$$

- ▶ often represent typical rather than necessary dependence (e.g. **Medical_professionals-Cure**)

Mapping frame elements

If f_1 is related to f_2 with a relation in FN then

$$\forall s_1 s_2 ((f_1(s_1) \wedge f_2(s_2)) \rightarrow \\ (rel(s_1, s_2) \leftrightarrow \forall x (FE_1(x, s_1) \leftrightarrow FE_2(x, s_2))),$$

where FE_1 in f_1 is mapped to FE_2 in f_2 .

Example

$$\forall s_1 (\mathbf{Giving}(s_1) \rightarrow \exists s_2 (\mathbf{Getting}(s_2) \wedge \mathit{causes}(s_1, s_2)))$$

$$\forall s_1 s_2 ((\mathbf{Giving}(s_1) \wedge \mathbf{Getting}(s_2)) \rightarrow \\ (\mathit{causes}(s_1, s_2) \leftrightarrow \forall x (\mathbf{DONOR}(x, s_1) \leftrightarrow \mathbf{SOURCE}(x, s_2))))$$

Cleaning up constraints

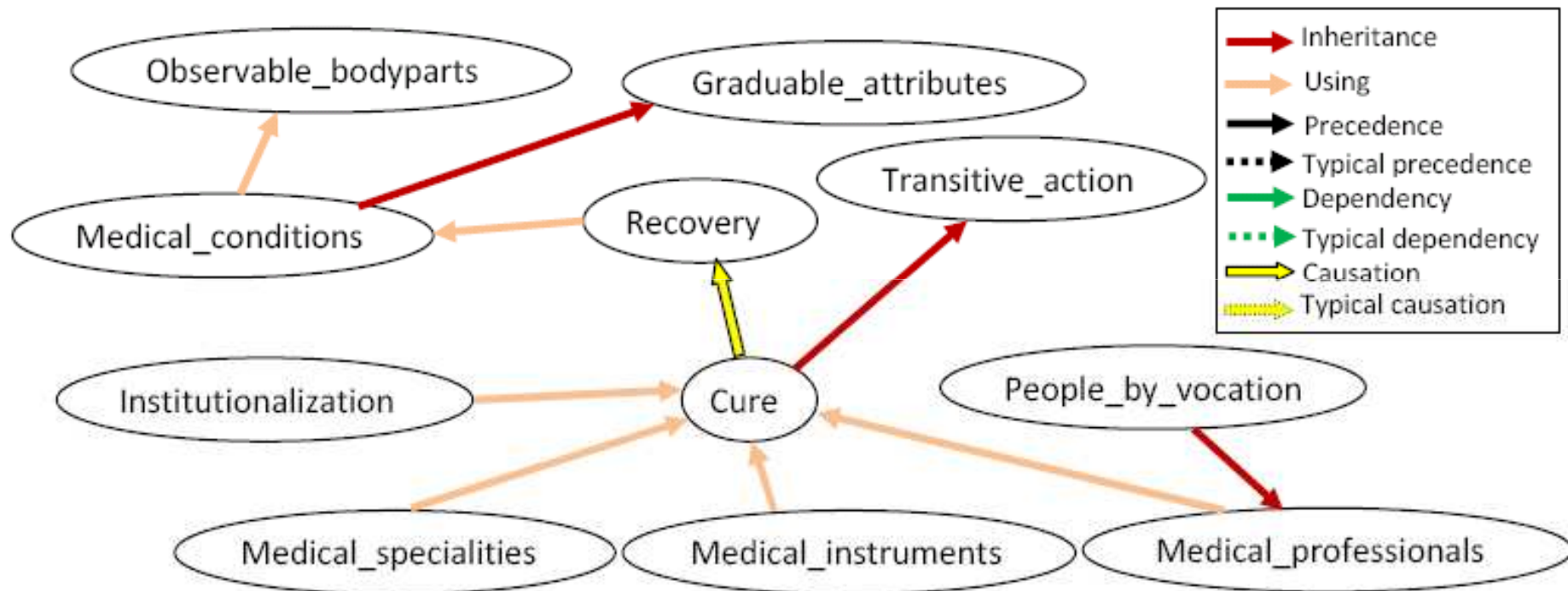
Given frames f_1 and f_2 connected with a relation r

1. define the types of situations that instantiate f_1 and f_2
2. if r is a temporal relation, make sure that both f_1 and f_2 refer to „temporal“ situations
3. define whether r has a typical or a necessary character
4. check whether the frame relation axioms apply to all instantiations of f_1 and f_2

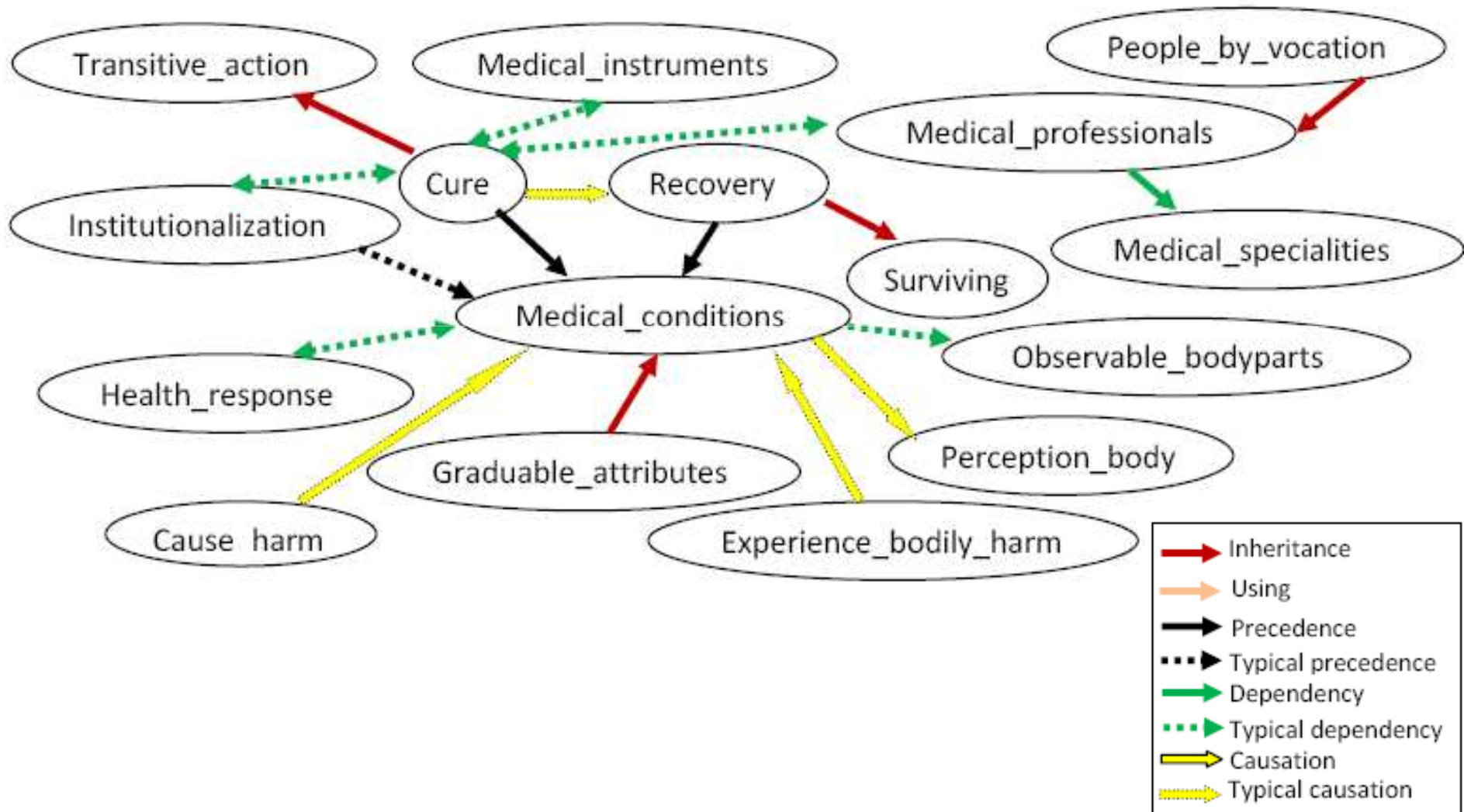
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Case Study: „medical cluster“



Enriched and cleaned up „medical“ cluster



„Medical“ cluster in RTE

- ▶ 39 *T-H* entailment pairs (18 true entailments) annotated in FATE with „medical“ frames
- ▶ TE computed by the Nutcracker system [*Bos&Markert,2006*]

	NFA	FA	FA&A	FA&CA
Correct proofs	1	4	7	10
Wrong proofs	1	1	1	1
Overall accuracy	0.56	0.5	0.61	0.78

- ▶ Problems:
 - Incompleteness of the FATE annotation: 8
 - Nutcracker processing errors: 5
 - Lack of general non-definitional knowledge: 7

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Conclusion

1. Presented
 - i. Conceptual problems in FN
 - ii. Methodology for improvement
 - data-driven analysis
 - ontological analysis
 - iii. Case study

2. Lessons learned
 - I. Many usefull relations can be aquired automatically
 - II. Axiomatization helps
 - III. RTE is still a difficult task
 - difficult to create an appropriate annotation
 - difficult to provide necessary knowledge
 - difficult to find a proof

Ongoing and future work

1. **Automatic relation extraction**
 - ▶ automatic mapping of frame roles
 - ▶ detecting type of the relation
2. **Ontological analysis**
 - ▶ applying OntoClean to FN hierachy of frames and roles
3. **FrameNet in RTE**
 - ▶ applying frame relations to a full RTE set
 - ▶ using frame similarity measures for weighting axioms
 - ▶ using probabalistic reasoning

Thank you!
Any questions?