

# Aligning Discourse and Argumentation Structures using Subtrees and Redescription Mining

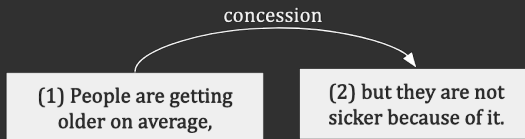
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ArgMining  
August 1, 2019

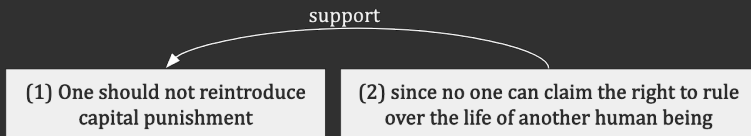
# Discourse structure

- ▶ Semantic and pragmatic relations between text segments (*reason, cause, concession ...*)
- ▶ Rhetorical Structure Theory [Mann and Thompson, 1988] (*RST*)
- ▶ Distinction between nucleus and satellite



# Argumentation Structure

- ▶ Argumentation relations between text segments (*support, attack, ...*)
- ▶ Macro-structure of argumentation [Freeman, 2011]
- ▶ Distinction between premisses and conclusion



# So what?

**Goal:** Understand the similarities between discourse and argumentation structures.

- ▶ Building bridges between theories
- ▶ Improve Argument Mining systems

# Corpus

- ▶ ArgMicroTexts corpus [Peldszus and Stede, 2015] \*
- ▶ 112 short argumentative texts
- ▶ 18 controversial questions

## ***"Should Germany introduce the death penalty?"***

- 1: The death penalty is a legal means that as such is not practicable in Germany.
- 2: For one thing, inviolable human dignity is anchored in our constitution,
- 3: and furthermore no one may have the right to adjudicate upon the death of another human being.
- 4: Even if many people think that a murderer has already decided on the life or death of another person,
- 5: this is precisely the crime that we should not repay with the same.

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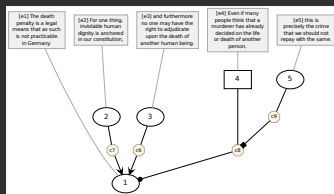
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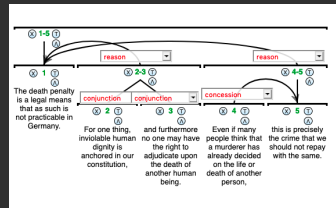


# Corpus

- ▶ Macro-structure of argumentation [Peldszus and Stede, 2016]
- ▶ RST
- ▶ (SDRT [Lascarides and Asher, 2007])



(a) ARG annotation



(b) RST annotation

# Overview of the approach

**Goal:** can we align ARG and RST at the subtree level ?

1. Representing ARG and RST structures as trees
2. Building two descriptions of each text
  - ▶ ARG and RST descriptions
  - ▶ A description is a set of subtrees
3. Aligning set of subtrees that describe almost the same set of texts

# Representing ARG and RST structures as trees

**Goal:** Unify and anonymise the structures.

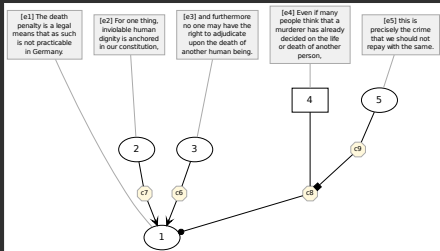
- ▶ Transform *ARG* and *RST* structures into labeled trees
- ▶ Keep only structure, no text

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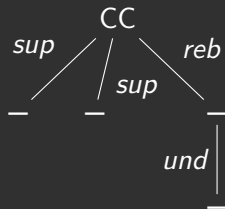
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# Representing ARG and RST structures as trees : ARG



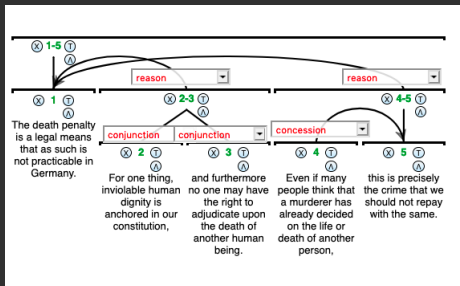
ARG annotation

- ▶ **Root:** central claim
- ▶ **Parent:** conclusion
- ▶ **Child:** premiss

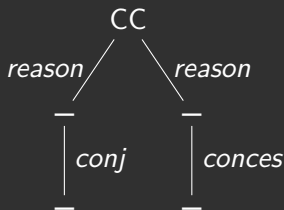


ARG tree derivation

# Representing ARG and RST structures as trees : *RST*



RST annotation



RST tree derivation

- ▶ **Root:** most central nucleus
- ▶ **Parent:** nucleus
- ▶ **Child:** satellite

# Building two descriptions of the corpus

**Goal:** Produce 2 descriptions of each texts in term of subtrees

1. Extract all subtrees of ARG
2. Extract all subtrees of RST

Frequent subgraph mining: gSpan [Yan and Han, 2002]

# Building two descriptions of the corpus

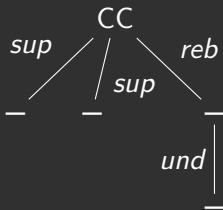
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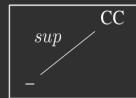
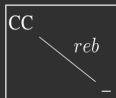
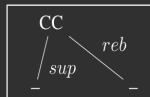
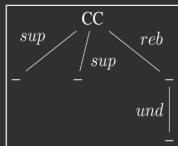
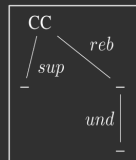
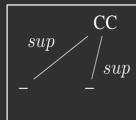
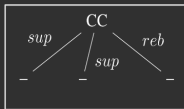
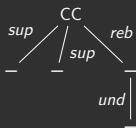
Frequent subgraph mining: gSpan [Yan and Han, 2002]



# Building two descriptions of the corpus: subtrees extraction

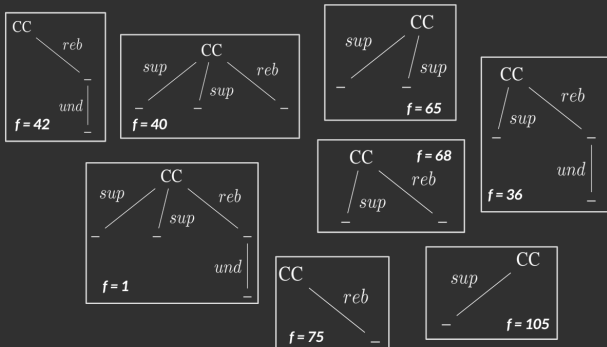


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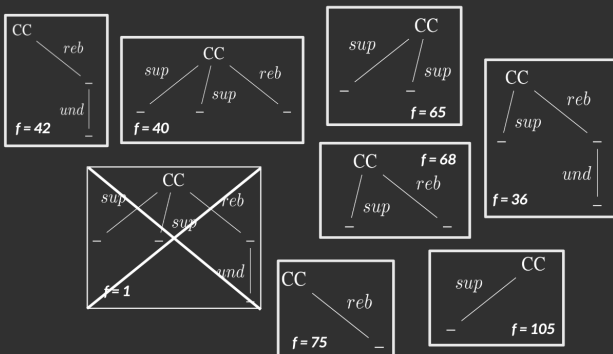
# Building two descriptions of the corpus: subtrees extraction

- ▶  $f$  is the frequency of occurrence of subtrees in the corpus



# Building two descriptions of the corpus: subtrees extraction

- ▶ keep subtrees with  $f \geq 2$



# Redescription mining

**Goal:** Find an ARG description and a RST description that characterize almost the same set of objects

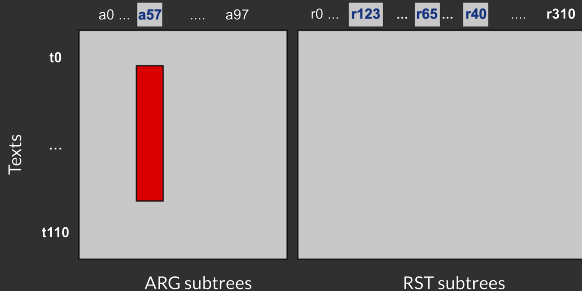
- ▶ Two different descriptions of the each text
  - ▶  $ARG = \{a_0, a_1, \dots, a_{98}\}$
  - ▶  $RST = \{r_0, r_1, \dots, r_{311}\}$
- ▶ A set of objects: a set of texts from the corpus
- ▶ A text  $t_i$  is described by
  - ▶ a subset of  $ARG$
  - ▶ a subset of  $RST$

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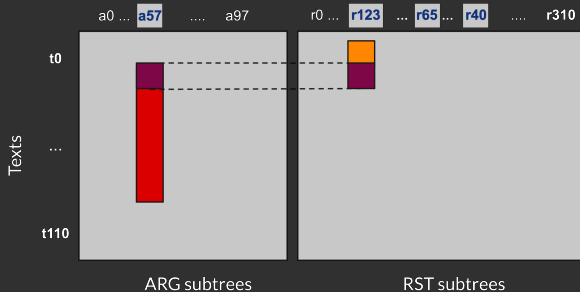
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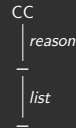
$$Rd1 : a_{57} \leftrightarrow \emptyset$$



# Redescription mining

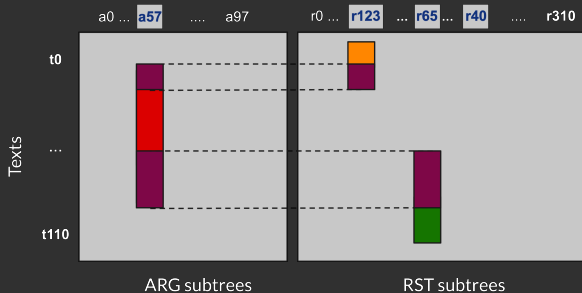


$Rd1 : a57 \longleftrightarrow r123$

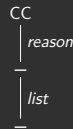




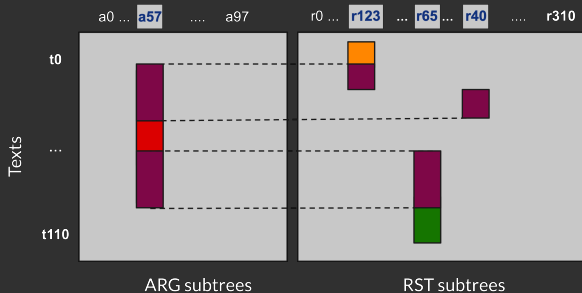
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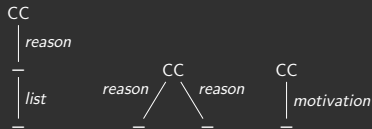
$$Rd1 : a57 \longleftrightarrow r123 \vee r65$$



# Redescription mining



$$Rd1 : a57 \longleftrightarrow r123 \vee r65 \vee r40$$



# Redescription mining

- ▶ A redescription is pair of queries
  - ▶  $qArg$  a logical formulae over the  $Arg$  subtrees
  - ▶  $qRst$  a logical formulae over the  $Rst$  subtrees
- ▶  $qArg$  and  $qRst$  should describe **almost** the same set of texts
- ▶ "Almost": given a similarity threshold calculated with Jaccard index

$$Jacc(qArg, qRst) = \frac{supp(qArg \wedge qRst)}{supp(qArg \vee qRst)}$$

# Experiment setup

- ▶ Algorithm: ReRemi
- ▶ Conjunctions and disjunctions allowed
- ▶ Length of the query limited to 4
- ▶ Output: 35 redescriptions

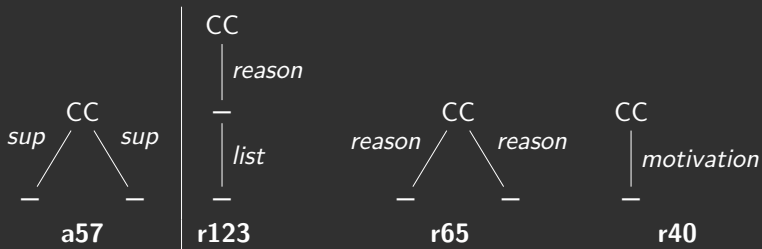
# Results

id	q1	q2	$J(q1,q2)$	# texts
<i>Rd1</i>	a57	r123 $\vee$ r65 $\vee$ r40	0.691	54
<i>Rd2</i>	a58	r61 $\vee$ r119 $\vee$ r125	0.351	13
<i>Rd3</i>	a23 $\vee$ a59	r125	0.3	8

3 over 35 obtained redescriptions  
aX and rX correspond to *ARG* and *RST* subtrees respectively.

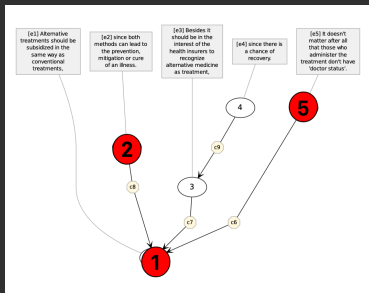
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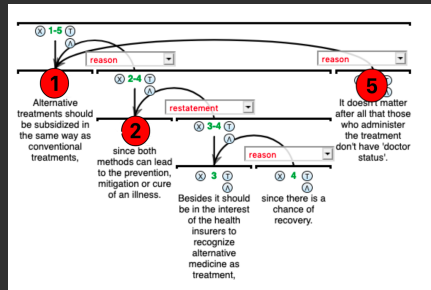


**RST is more fine grained than ARG**

# Well captured information



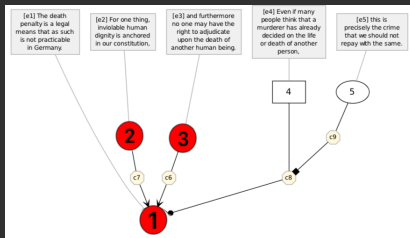
(a) ARG annotation



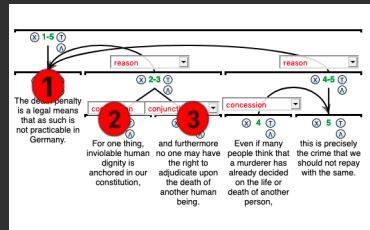
(b) RST annotation



# Anonymization lead to wrong captured patterns



(a) ARG annotation



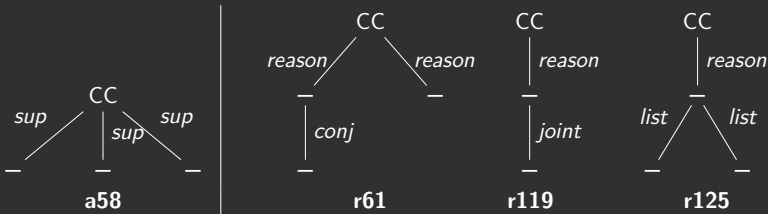
(b) RST annotation





# Results

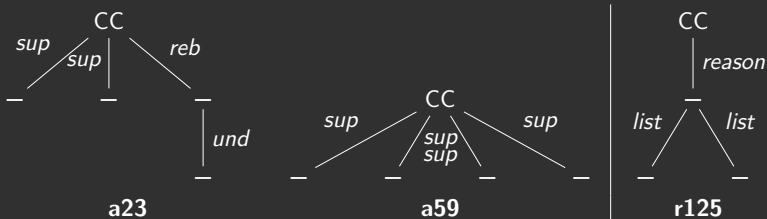
$Rd2 : a58 \longleftrightarrow r61 \vee r119 \vee 125$



**Rd2 is a specialization of Rd1**

# Results

$Rd3 : a23 \vee a59 \longleftrightarrow r125$



**2  $\neq$  ARG representations of the one RST subtree**

# Conclusion

- ▶ Turn a linguistic problem into a Data Mining problem
- ▶ Systematic, generic and automatic comparison
- ▶ Understand the links between  $\neq$  theories

# Future work

- ▶ Take segments into account
- ▶ Play with parameters of ReReMi
- ▶ Propose an exhaustive analysis of the redescrptions
- ▶ Investigate other Data Mining formalisms  
(*e.g. FCA, association rules*)
- ▶ Extend to other formalisms  
(*e.g. SDRT*)

# Thank you!




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