

Natural Language Inference

NatLog

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- 1** Entailment Relations
 - ▶ Introduce entailment relations
 - ▶ Joining multiple relations
- 2** Compositional Entailment
 - ▶ Monotonicity
 - ▶ Projectivity
 - ▶ Implicature
- 3** NatLog
 - ▶ Entailment System
 - ▶ Evaluation

Recap: Entailment Relations

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Compositional
Entailment

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FraCaS Test Suite

RTE3 Test Suite

Symbol	Name	Example	Set Theoretic Definition
$x \equiv y$	equivalence	couch \equiv sofa	$x = y$
$x \sqsubset y$	forward entailment	crow \sqsubset bird	$x \subset y$
$x \supset y$	reverse entailment	bird \supset crow	$x \supset y$
$x \wedge y$	negation	able \wedge unable	$x \cap y = \emptyset \wedge x \cup y = U$
$x y$	alternation	cat dog	$x \cap y = \emptyset \wedge x \cup y \neq U$
$x \smile y$	cover	animal \smile non-ape	$x \cap y \neq \emptyset \wedge x \cup y = U$
$x \# y$	independence	hungry $\#$ hippo	(all other cases)

Entailment joins

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	\boxtimes	\equiv	\sqsubset	\sqsupset	\wedge	\vee	\sim	$\#$
	\equiv	\equiv	\sqsubset	\sqsupset	\wedge	\vee	\sim	$\#$
	\sqsubset	\sqsubset	\sqsubset	$\equiv \sqsubset \sqsupset \mid \#$	\mid	\mid	$\sqsubset \wedge \mid \sim \#$	$\sqsubset \mid \#$
	\sqsupset	\sqsupset	$\equiv \sqsubset \sqsupset \sim \#$	\sqsupset	\sim	$\sqsupset \wedge \mid \sim \#$	\sim	$\sqsupset \sim \#$
	\wedge	\wedge	\sim	\mid	\equiv	\sqsupset	\sqsubset	$\#$
	\mid	\mid	$\sqsubset \wedge \mid \sim \#$	\mid	\sqsubset	$\equiv \sqsubset \sqsupset \mid \#$	\sqsubset	$\sqsubset \mid \#$
	\sim	\sim	\sim	$\sqsupset \wedge \mid \sim \#$	\sqsupset	\sqsupset	$\equiv \sqsubset \sqsupset \sim \#$	$\sqsupset \sim \#$
	$\#$	$\#$	$\sqsubset \sim \#$	$\sqsupset \mid \#$	$\#$	$\sqsupset \mid \#$	$\sqsubset \sim \#$	$\equiv \sqsubset \sqsupset \wedge \mid \sim \#$

Recap: Compositional Entailment

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i	e_i	$x_i = e_i(x_{i-1})$	$\beta(e_i)$	$\beta(x_{i-1}, e_i)$	$\beta(x_0, x_i)$
		Stimpy is a cat.			
1	SUB(<i>cat</i> , <i>dog</i>)				
		Stimpy is a dog.			
2	INS(<i>not</i>)		^	^	□
		Stimpy is not a dog.			
3	SUB(<i>dog</i> , <i>poodle</i>)		□	□	□
		Stimpy is not a poodle			

NatLog Pipeline

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- 1 Linguistic analysis
- 2 Alignment
- 3 Entailment classification
 - 1 Lexical entailment classification
 - 2 Entailment projection
 - 3 Entailment joining

Linguistic Analysis

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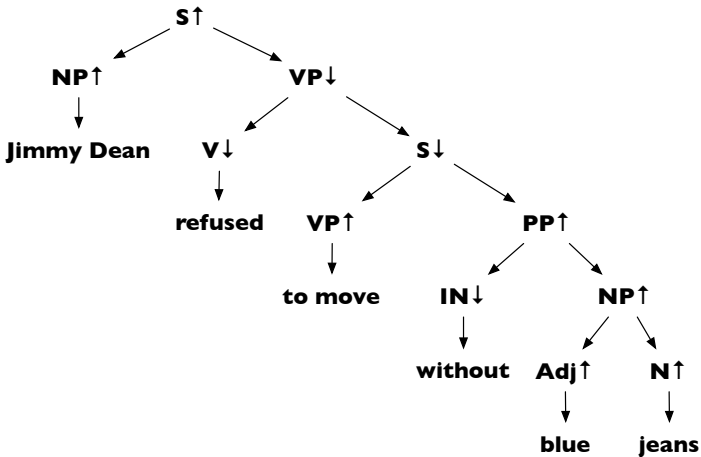
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- 1 Tokenisation
- 2 POS tagging
- 3 Syntactic phrase-structure parsing
⇒ Stanford parser
- 4 Lemmatisation
⇒ Finite-State Analyser (*Minnen et al.*)
- 5 Monotonicity marking

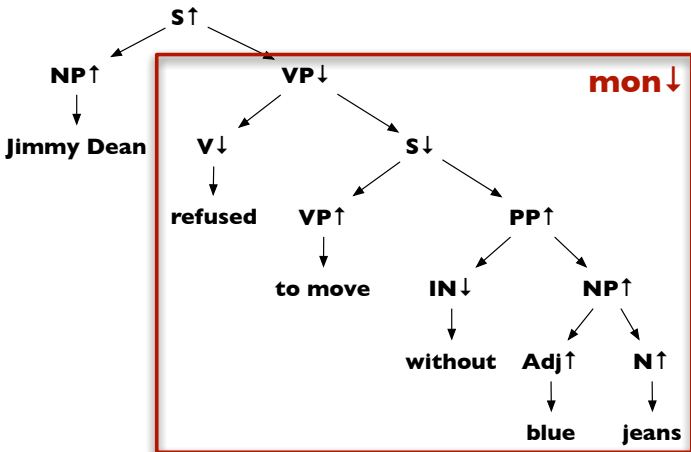
Tasks not performed

- ▶ Named entity recognition
- ▶ Coreference resolution
- ▶ Dependency parsing
- ▶ Semantic role labeling

Monotonicity marking



Monotonicity marking



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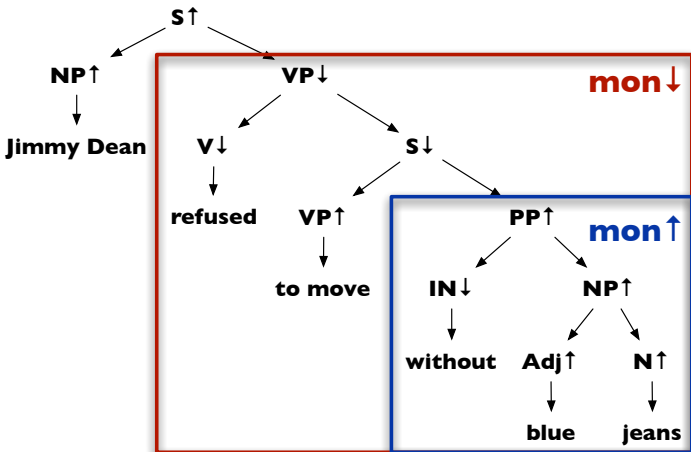
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Monotonicity marking



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Alignment

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Jimmy Dean refused					to move without blue jeans			
	SUB	DEL	INS	INS	SUB	EQ	DEL	SUB
James Dean			did	n't	dance	without		pants

Sequence of atomic edits

- 1 Generate alignments
 - ▶ Atomic Edits: EQ, SUB, DEL, INS
 - ⇒ Stanford RTE aligner, string edit distance
- 2 Heuristic order of edits
 - ▶ DEL
 - ▶ SUB
 - ▶ All downward/non-monotone edits
 - ▶ INS

Lexical entailment classification

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- 1 Linguistic analysis
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- 3 **Entailment classification**
 - 1 **Lexical entailment classification**
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Lexical entailment classification

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Jimmy Dean	refused				to move	without	blue jeans	
	SUB	DEL	INS	INS	SUB	EQ	DEL	SUB
James Dean			did	n't	dance	without		pants

Predicting entailment relations

- 1 Find lexical features
 - ▶ **WordNet** Synonymy, hypernymy, ...
 - ▶ **String similarity** Levenshtein, ...
 - ▶ **Relatedness** Semantic, derivational, distributional
 - ▶ **Lexical category** Prepositions, light, quantifiers,...
 - ▶ **Misc. features** Number mismatches, special cases
- 2 Look up in decision tree
 - ⇒ Weka

Entailment projection

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Entailment joining

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- ▶ Manually designed simple sentences
- ▶ Premise and hypothesis quite similar
- ▶ Designed to reflect various phenomena
- ▶ Answer types:
 - ▶ YES (entailment) (59%) ($\equiv \square$)
 - ▶ NO (contradiction) (28%) (\wedge)
 - ▶ UNK (compatibility) (10%) ($\square \sim \#$)
- ▶ NatLog ignores multi-premise problems
 - ▶ Remaining: 192 problems (55%)

p No delegate finished the report.

h Some delegate finished the report on time. **NO**

p Either Smith, Jones or Anderson signed the contract.

h Jones signed the contract. **UNK**

p John wrote a report, and Bill said Peter did too.

h Bill said Peter wrote a report. **YES**

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System	P (YES)	R (YES)	Acc
Baseline	55.7%	100.0%	55.7%
Bag of words	59.7%	87.2%	57.4%
NatLog	89.3%	65.7%	70.5%

Section	#	P (YES)	R (YES)	Acc
Quantifiers	44	95.2%	100.0%	97.7%
Plurals	24	90.0%	64.3%	75.0%
Anaphora	6	100.0%	60.0%	50.0%
Elipsis	25	100.0%	5.3%	24.0%
All sections	183	89.3%	65.7%	70.5%

RTE3 test suite

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- ▶ RTE3 = 3rd Recognizing Textual Entailment Challenge
- ▶ Real premises, hand-written hypotheses
- ▶ Premises a lot longer than hypotheses (30 vs 7 words)
- ▶ Binary answers:
 - ▶ YES (Entailment) (50%) ($\equiv \square$)
 - ▶ NO (Non-entailment) (50%) ($\wedge \sqsupset \sim \#$)
- ▶ Development & test set (800 problems each)
- ▶ Potential problems for NatLog:
 - ▶ Other types of inference required
 - ▶ Large edit distance

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p As leaders gather in Argentina ahead of this weekend's regional talks, Hugo Chávez, Venezuela's populist president is using an energy windfall to win friends and promote his vision of 21st-century socialism.

h Hugo Chávez acts as Venezuela's president.

▶ **YES**

p Mr. Fitzgerald revealed he was one of several top officials who told Mr. Libby in June 2003 that Valerie Plame, wife of the former ambassador Joseph Wilson, worked for the CIA.

h Joseph Wilson worked for CIA.

▶ **NO**

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Section	Data	YES	P (YES)	R (YES)	Acc
NatLog	dev	22.5%	73.9%	32.4%	59.3%
	test	26.4%	70.1%	36.1%	59.4%
Stanford	dev	50.3%	68.7%	67.0%	67.3%
	test	50.0%	61.8%	60.2%	60.5%
Hybrid	dev	56.0%	69.2%	75.2%	70.0%
	test	54.5%	64.5%	68.5%	64.5%

- ▶ **Stanford system** predicts via *inference score*
- ▶ **Hybrid system** biasses Stanford towards NatLog decision

NatLog

- ▶ Performs well for certain problems.
- ▶ Not yet fit for real world application.
- ▶ Low noise resistance.
- ▶ Strong bias towards independence relation.
- ▶ Potential as a component of a larger system