



Textual Entailment Part 2: Classes of Strategies and Learning

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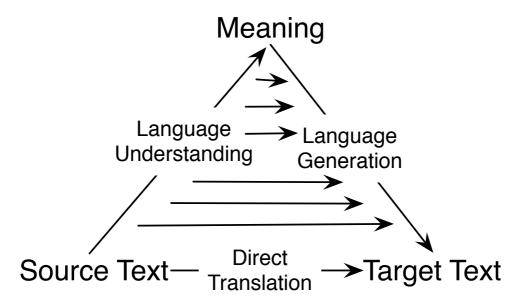
Structure of the Tutorial

- Part 1 [SP]: Introduction and Basics
- Part 2 [RW]: Classes of Strategies and Learning
 - * BRFAK*
- Part 3 [SP]: Knowledge and Knowledge Acquisition
- Part 4 [SP]: Applications
- Part 5 [RW]: Multilingual, Component-based System Building





MT Triangle

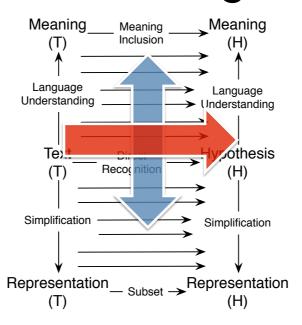


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RTE Rectangle







Architecture

- Linguistic analysis pipeline (LAP)
- Entailment decision algorithm (EDA)
 - Classification-based
 - Transformation-based
- Knowledge base (KB) (next section)

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Architecture

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Overview of LAPs

- Tokenization (Word Segmentation)
- Part-of-Speech (POS) Tagging
- Lemmatization
- Named-Entity Recognition
- Syntactic Parsing
 - Constituent Parsing
 - Dependency Parsing
- Semantic Role Labeling
- Coreference Resolution
- ...

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Token-Level Processing

- Tokenization
 - Word segmentation
- Lemmatization
 - Morphological analysis
- POS Tagging
- Lexical Semantics
 - WordNet, distributional similarity, etc.

Performance >97%





word	pos	lemma
The	DT	the
TreeTagger	NP	TreeTagger
is	VBZ	be
easy	JJ	easy
to	то	to
use	VB	use
	SENT	

From TreeTagger website

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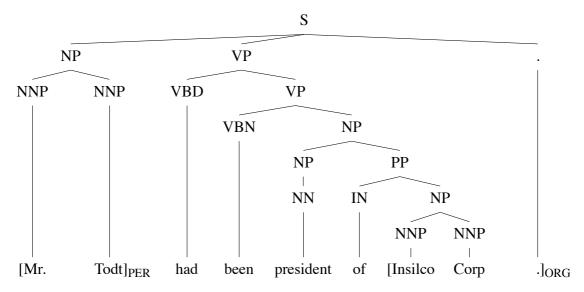
Constituents

- Chunking
- Named-Entity Recognition
- Constituent Parsing

NER: 70~90%







From Stanford NER (Finkel and Manning, 2009)

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Dependency

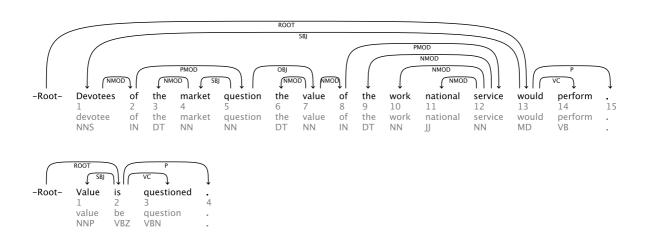
- Syntactic Dependency Parsing
- Semantic Dependency Parsing
 - Semantic Role Labeling
 - Predicate-Argument Structure

Syn: 80~90% Sem: 75~85%

Logic Form Composition







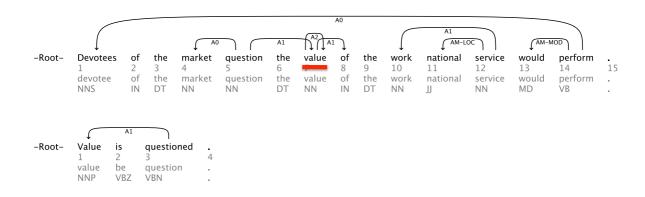
From MSTParser (McDonald et al., 2005); Visualized by https://code.google.com/p/whatswrong/

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An Example (cont.)



From Laputa SRL (Zhang et al., 2008); Visualized by https://code.google.com/p/whatswrong/

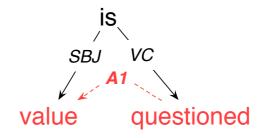
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An Example (cont.)

- **H**: Value is questioned.
- Syntactic dependency
 - <is, SBJ, value>
 - <is, VC, questioned>



- Semantic dependency
 - <questioned, A1, value>

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Semantic Roles

- PropBank (Palmer et al., 2005) and NomBank (Meyers et al., 2004)
- Core arguments: A0-A5
 - different semantics for each verb
 - specified in the PropBank Frame files
- 13 types of adjuncts labeled as AM-arg
 - where arg specifies the adjunct type





Discourse

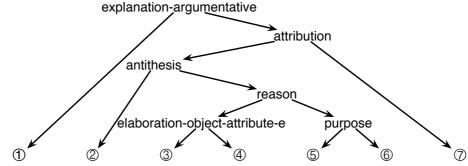
- Coreference Resolution
- Event Structure
- Discourse Parsing

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An Example



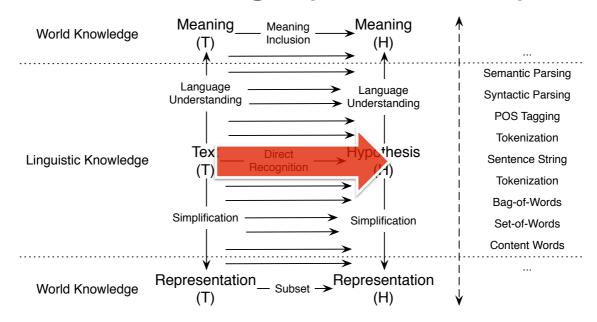
[① Ford Motor Co. and Chrysler Corp. representatives criticized Mr. Tonkin's plan as unworkable.] [② It "is going to sound neat to the dealer] [③ except when his 15-day car supply doesn't include the bright red one] [④ that the lady wants to buy] [⑤ and she goes up the street] [⑥ to buy one,"] [⑦ a Chrysler spokesman said.]

From RST Discourse Treebank (Carlson et al., 2002)





RTE Rectangle (more details)



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Overview of EDAs

- Classification-based
 - Score / Threshold
 - Structure / Alignment
- Transformation-based
 - Edit distance
 - (Knowledge) rule application
- Meta-EDA





Classification (RTE Style)



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Popular Classifiers

Model	Perceptron/SVM	Naïve Bayes	Logistic Regression
Туре	Discriminative	Generative	Discriminative
Distribution	N/A	P(X, Y)	P(Y X)
Independence	None	Strong	None
Features	Ex/Impilicit	Explicit	Explicit
Speed	Fast/Slow	Fast	Intermediate

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Kernel-Based Methods

- Kernel Function
 - Mapping between spaces
 - Cross-combination of features (implicitly!)
 - Intro-pair features → cross-pair features
- Subsequence Kernel (Lodhi et al., 2002; Wang and Neumann, 2007a)
- Tree Kernel (Collins and Duffy, 2001; Zanzotto et al., 2007)

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Linguistic Features

- Measure somethingsimilarity between t and h:
 - Lexical overlap (unigram, N-gram, subsequence)
 - Assisted by lexical resources like WordNet
 - Syntactic matching
 - Lexical-syntactic variations ("paraphrases")
 - Semantic role matching
 - Global similarity parameters (e.g. negation, modality)
- Detect mismatch (for non-entailment)





Data Structures

- String-to-String rewriting
 - String edit distance (MacCartney and Manning, 2007)
 - Tree skeleton difference (Wang and Neumann, 2007a)
- Tree-to-Tree editing
 - Tree edit distance (Kouylekov and Magnini, 2005)
- · Graph-to-Graph mapping
 - Graph matching (Haghighi et al., 2005)

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Word Overlap

- |T|: number of words in T
- |H|: number of words in H
- $E_1 = |T \wedge H| / |H|$
- $E_2 = |T \wedge H| / |T|$
- $E_3 = (2 * E_1 * E_2) / E_1 + E_2$

57.2 on average

- Content words only
- Lemmatization

From (Mehdad and Magnini, 2009)





Dependencies

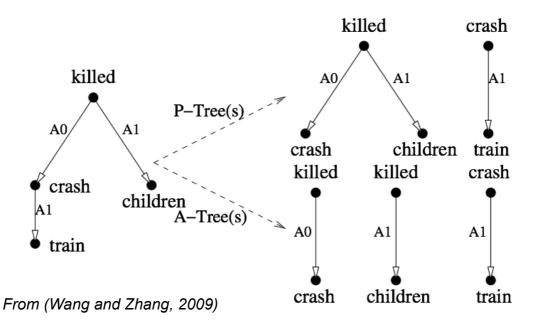
- · Syntactic dependency trees
 - Dependency triples < Node, Relation, Head>
 - Bag of such triples
- E₁' = |Triple(T) ∧ Triple (H) | / |Triple(H) |

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Dependencies (cont.)







Results (RTE-5)

- DFKI1: BoW and syntactic dependency
- DFKI2: BoW, syntactic, and semantic dependency
- DFKI3: BoW and joint syntactic and semantic representation

Runs	Main	Main -VO	Main -WN	Main -VO-WN
DFKI1	62.5%	62.5%	62.7%	62.5%
DFKI2	66.8%	66.5%	66.7%	66.3%
DFKI3	68.5%	68.3%	68.3%	68.3%

From (Wang et al., 2009)

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Larger Sub-Structures

- Dependency paths
 - Common sub-paths
- Subtrees
- E₁ ' ' = |Subtree(T) ∧ Subtree(H)| / |Subtree(H)|





Subtrees

$T_1 \Rightarrow H_1$	$T_2 \Rightarrow H_2$

T₁ "Farmers feed cows animal extracts"

H₁ "Cows eat animal extracts"

	2	
T ₂ '	They fee	ed dolphins fish"
H ₂ '	'Fish eat	dolphins"

feed
$$\boxed{x} \boxed{y} \rightarrow \boxed{x} \text{ eat } \boxed{y}$$

 $T_3 \Rightarrow H_3$

T₃ "Mothers feed babies milk"

"Babies eat milk"

From (Zanzotto and Dell'Arciprete, 2009)

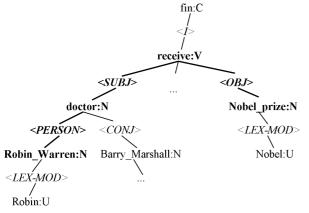
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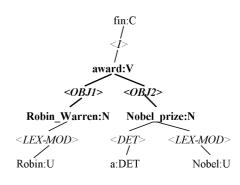




Tree Skeletons

- T: Doctor Robin Warren and Barry Marshall received Nobel awarded a Nobel Prize. Prize ...
- H: Robin Warren was





From (Wang and Neumann, 2007)





Results

• RTE-2





• RTE-3

Exp2BT&Exp2BL: Training on the RTE-3 Dev Set and Testing on the Test Set							
Systems IE IR QA SUM ALL							
BoW	54.5%	66.5%	76.5%	56.0%	63.4%		
TSM	54.5%	62.5%	66.0%	54.5%	59.4%		
SK+BS (Mi+SP+Task) - run1	59.5%	70.5%	75.5%	60.5%	65.5%		
SK+BS (Mi+Length) - run2	58.5%	70.5%	79.5%	59.0%	66.9%		

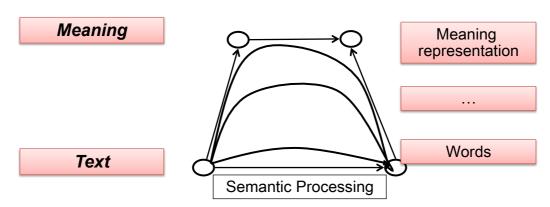
From (Wang, 2007)

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The RIGHT level



- Trade-offs between
 - Competence of the knowledge (deeper)
 - Performance of the processing (shallower)

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Alignment-Based Approaches

- Word alignment (Glickman et al., 2006)
- Phrase alignment (chambers et al., 2007; MacCartney et al., 2008)
- Relation alignment (Sammons et al., 2009)

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Matching vs. Transformations

- Direct matching (so far, no chaining)
- Sequence of transformations (A proof)

$$T = T_0 \rightarrow T_1 \rightarrow T_2 \rightarrow ... \rightarrow T_n = H$$

- Tree-Edits
- Knowledge based Entailment Rules

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Edit Distance

- (Limited) pre-defined operators
 - Insertion
 - Deletion
 - Substition
- String-to-String
- Tree-to-Tree

Weakly linguistically motivated!

- The EDITS system (Kouylekov and Negri, 2010)
 - Estimate confidence in each operation
- Wang and Manning (2010), Heilman and Smith (2010), etc.





Knowledge-Based Rules

- Rule application
 - Arbitrary knowledge-based transformations
 - Formalize many types of knowledge
- BIUTEE (Stern and Dagan, 2011)
 - On-the-fly operations
 - Cost model
 - Search for the best inference

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An Example

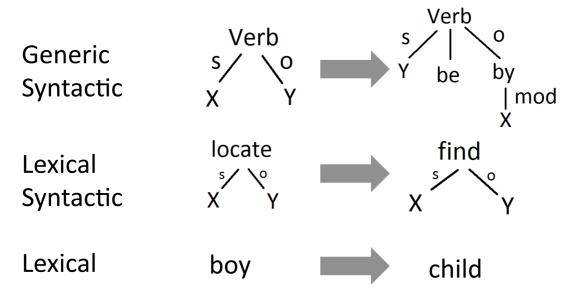
Id	Operation	Generated Text
0	-	He received the letter from the secretary.
1	Coreference substitution	The employee received the letter from the secretary.
2	X received Y from Z → Y was sent to X by Z	The letter was sent to the employee by the secretary.
3	Y [verb-passive] by X → X [verb-active] Y	The secretary sent the letter to the employee.
4	X send Y → X deliver Y	The secretary delivered the letter to the employee.
5	letter → message	The secretary delivered the message to the employee.

From (Stern et al., 2012)





Entailment Rules



From (Bar-Haim et al., 2007)

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Cost Based Model

- Define operation cost
 - Represent each operation as a feature vector
 - Cost is linear combination of feature values
- Define proof cost as the sum of the operations' costs



Learn

Variant of (Raina et al., 2005)





Search the Best Proof

 $T \rightarrow H$ $T \rightarrow H$ Proof #1 $T \rightarrow H \times Proof #1 T \rightarrow H \times Proof #1 T \rightarrow H \times H$

Proof #2 $T \longrightarrow H \checkmark$ Proof #2 $T \longrightarrow H \checkmark$ Proof #3 $T \longrightarrow H \checkmark$ Proof #3 $T \longrightarrow H \checkmark$ Proof #4 $T \longrightarrow H \checkmark$ Proof #4 $T \longrightarrow H \checkmark$

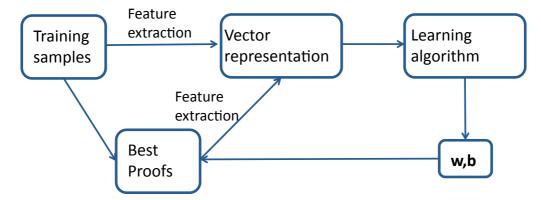
- "Best Proof" = proof with lowest cost
- Search space exponential AI-style search (Stern et al., 2012)
 - Gradient-based evaluation function
 - Local look ahead for "complex" operations

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Inference vs. Learning

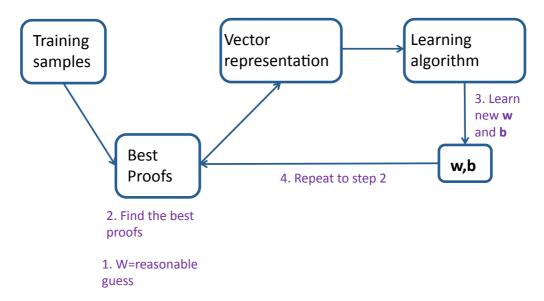


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Iterative Learning Scheme



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Performance (Classification)

Text:

Hypothesis:

Text:

Hypothesis:

Text:

Hypothesis:

System	RTE-1	RTE-2	RTE-3	RTE-5
Raina et al. 2005	57.0			
Harmeling, 2009		56.39	57.88	
Wang and Manning, 2010		63.0	61.10	
Bar-Haim et al., 2007			61.12	63.80
Mehdad and Magnini, 2009	58.62	59.87	62.4	60.2
BIUTEE (2011)	57.13	61.63	67.13	63.50





Performance (Search)

I draw a dot in the middle of a square and call that dot the self, the essence. In acting, everything must pass through that dot. The wildest style, the most absurd, the natural, the "be yourself," all must pass through. It takes ngor and constancy. Good actors work this way by inclination and training.

Acting is a paradox. The lie a good actor tells: What's Hecuba to him...) is catharsis. It's a cleansing. It can't happen unless the actor passes the lie through that dot of self, of reality.



RTE 6 (F1%)	
Base line (Use IR top-5 relevance)	34.63
Median (2010)	36.14
Best (2010)	48.01
BIUTEE (2012)	49.54

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- **T**: Bush used his weekly radio address to try to build support for his plan to allow workers to divert part of their Social Security payroll taxes into private investment accounts.
- **H**: Mr. Bush is proposing that workers be allowed to divert their payroll taxes into private accounts.

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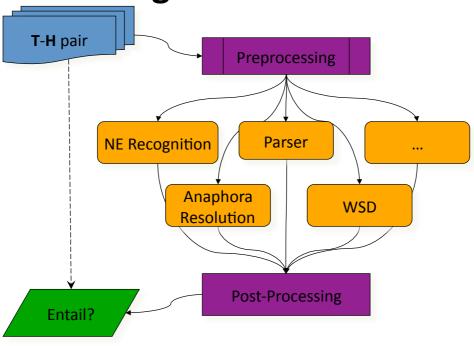
An Example

- T: Bush used his weekly radio address to try to build support for his plan to allow workers to divert part of their Social Security payroll taxes into private investment accounts.
- H: Mr. Bush is proposing that workers be allowed to divert their payroll taxes into private accounts.





Bag-of-Features

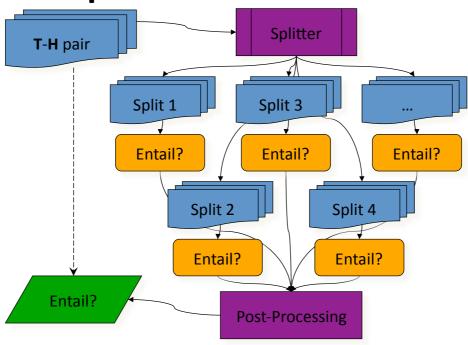






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Specialized Modules



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Divide-and-Conquer

• A specialized RTE module

A good targetA good tackleAnchoring Tree Skeleton

Matching Named-Entity
Matching

Results on RTE-4

Modules	TAC-M	TS-M	NE-M	BoW-BM	Tri-BM	Overall
Accuracy	80.6%	74.6%	54.3%	56.5%	52.8%	70.6%
Coverage	3.1%	34.6%	47.7%	100%	100%	100%

From (Wang and Neumann, 2009)

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Summary

- · Linguistic analysis pipeline
 - Various linguistic processing



implemented

- Entailment decision algorithm
 - Classification & feature space
 - Transformation & knowledge bases (upcoming)
- Overall Strategy
 - Specialized modules







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