# Textual Entailment & It's Application in QA

### Partha Pakray

### My Education Career

- Bachelor of Engineering, Computer Science & Engineering, Jalpaiguri Govt. Engineering College, 2000-2004, INDIA
- Master of Engineering, Computer Science & Engineering, Jadavpur University, 2005-2007, INDIA
   Maters Thesis: "MULTILINGUAL RESTRICTED DOMAIN QA SYSTEM WITH DIALOGUE MANAGEMENT (BENGALI AND TELUGU AS A CASE STUDY)"
- PhD(Engineering), Computer Science & Engineering Dept., Jadavpur University, 2009-2013, INDIA
   Supervisors: Prof Sivaji Bandyopadhyay and Prof Alexander Gelbukh
   Doctoral Thesis: "Answer Validation through Textual Entailment"

### My Work Career

- Research Intern, Semantics & Parsem Group, <u>Xerox Research Center</u> <u>Europe</u> (XRCE), Grenoble, FRANCE. [May- Oct, 2012]
- Worked as a Senior Research Engineer at <u>Jadavpur University</u> in Indian Languages Search Engine project "Cross Lingual Information Access" sponsored by Govt. of India, INDIA. [2008-2012]
- Worked as a Research Engineer in <u>National Polytechnic Institute (IPN),</u> <u>Mexico</u> in Project "Answer Validation through Textual Entailment" as a Bilateral Programme between Govt. of India and Conacyt, Mexico and worked in IPN, MEXICO. [Nov-Dec, 2011 and June-July, 2013]. (Under Prof Alexander Gelbukh)
- Post Doctoral Fellow, Computer and Information system, <u>Norwegian</u> <u>University of Science and Technology</u>, (ERCIM Marie-curie Fellowship ABCDE Programme August, 2013 - July, 2014). NORWAY.
- Now, I am here as a Post Doctoral Researcher.

#### Research Interests

- Textual Entailment
- Question Answering
- Answer Validation
- Case Based Reasoning
- Transliteration (e.g. Indian Languages)
- Information Retrieval
- Causal Sentence Identification



## Outline

- Textual Entailment
- Lexical Entailment
- Syntactic Entailment
- Semantic Entailment
- Entailment Classification
- Application: Answer Validation
- Application: Question Answering

#### Textual Entailment



## **Textual Entailment**

Recognizing Textual Entailment (RTE) is one of the recent challenges of Natural Language Processing (NLP).

**Textual Entailment**: directional relationship between two texts, denoted by the entailing "Text" (T) and the entailed "Hypothesis" (H).

T entails H if the meaning of H can be inferred from the meaning of T.

For Entailment cases: H is simply a paraphrase of all or part of T.

#### **Examples**

Positive Entailment (Entailment Class: "True")Text: iTunes software has seen strong sales in Europe.Hypothesis: Strong sales for iTunes in Europe.

Negative Entailment (Entailment Class: "False")
Text: After the deal closes, Teva will earn about \$7 billion a year, the company said.
Hypothesis: Teva earns \$7.5 billion a year.

#### **Application of Textual Entailment**

- Question Answering (QA)
- Paraphrases (PP)
- Summarization (SUM)
- Machine Translation (MT)

#### Approaches

- Lexical Entailment
- Syntactic Entailment
- Semantic Entailment





*Lexical: Smith bought a book.*  $\Longrightarrow$  *Smith purchased a book.* 

*Syntactic: Ram was singing and dancing.*  $\Longrightarrow$  *Ram was dancing.* 

*Semantic: Eric married Mary.*  $\Longrightarrow$  *Mary married Eric.* 

## Outline

- Textual Entailment
- Lexical Entailment
- Syntactic Entailment
- Semantic Entailment
- Entailment Classification
- Application: Answer Validation
- Application: Question Answering

## **Lexical Entailment**

Weight= Number of common [#] between text and hypothesis

Number of [#] in hypothesis

- 1. Unigram Matching
- 2. Unigram with WordNet Matching
- **3.** Bigram Matching
- 4. Stemmed Unigram Matching
- 5. Named Entity Matching
- 6. Lexical Distance Measure

## **Unigram Matching**

Text (T)	Hypothesis (H)
A whale that became stranded in the River Thames	A whale died in the River
has died after a massive rescue attempt to save its life.	Thames.
The 18ft (5m) northern bottle-nosed whale was first	
spotted in the river on Friday and rescuers began an	
attempt to save it on Saturday morning. But the whale	
died at about 1900 GMT on Saturday as rescuers	
transported it on a barge towards deeper water in the	
Thames Estuary.	

Common Unigrams	whale, died, River, Thames	

### Unigram with WordNet Matching

Text (T)	Hypothesis (H)
Smith bought a book.	Smith purchased a book.

Common Unigrams	Smith, book
WordNet Matching	bought=purchase

WordNet: Synonym Dictionary for English.

## Lexical Distance Matching

- Vector Space Measures: Euclidean distance, Block distance, Minkowsky distance, Cosine similarity
- Set-based Similarities: Dice, Jaccard, Overlap
- Edit Distance Measures: Levenshtein distance, Smith-Waterman distance, Jaro distance

Lexical distance measurement libraries: Soft Cardinality, SimMetrics, SimPack and SecondString.

## Outline

Textual Entailment Lexical Entailment Syntactic Entailment Semantic Entailment Entailment Classification Application: Answer Validation Application: Question Answering **Lexical textual entailment** = Lexical level **Syntactic textual entailment** = Syntactic level e.g. structural paraphrasing between two texts.

For example:

- 1. active passive changes between two texts.
- 2. Auxiliary Reduction: Example, "Ahmedinejad was attacked by the US" => "the US attacked Ahmedinejad"
- **3.** Copula Reduction: Example, "Microsoft Corp. is a partner of Intel Corp." => "Microsoft Corp., a partner of Intel Corp."

### Syntactic Textual Entailment

**Input**: Pairs of dependency relations of text snippets (text and hypothesis)

**Output**: "YES" if the text entails the hypothesis and "NO" otherwise.

**Parser generally used**: Stanford Dependency Parser and C&C Combinatory Categorical Grammar (CCG) Parser.

#### **Output of Stanford Dependency Parser**

Sentence	Nigeria seizes 80 tonnes of drugs
Output	[ <b>nsubj</b> (seizes, Nigeria), <b>num</b> (tonnes, 80), <b>dobj</b> (seizes, tonnes), <b>prep_of</b> (tonnes, drugs) ]



Text	Hypothesis
One species of ambiguity tries to	One species tries to baffle.
baffle by interweaving repetition.	
num species One	num species One
nsubj tries species	nsubj tries species
prep_of species ambiguity	aux baffle to
aux baffle to	xcomp tries baffle
xcomp tries baffle	
prepc_by baffle interweaving	
dobj interweaving repetition	

#### **Comparison of Dependency Relations**

- i. Comparison of Subject-Subject
- ii. Comparison of WordNet Based Subject-Verb
- iii. Comparison of Object-Verb
- iv. Comparison of WordNet Based Object-Verb
- v. Comparison of Cross Subject-Object
- vi. Comparison of Number
- vii. Comparison of Noun
- viii. Comparison of Prepositional Phrase

#### System Architecture



## Outline

Textual Entailment Lexical Entailment Syntactic Entailment Semantic Entailment Entailment Classification Application: Answer Validation Application: Question Answering

#### Semantic Textual Entailment

*Universal Networking Language (UNL)* Knowledge Representation in the form of semantic network with nodes.

**Enconversion**: Natural language sentences to UNL graphs. **Deconversion**: UNL graphs to natural language sentence.

**UNL format**:

[Relation Name] [Relation Scope ID] {[UW1][UW1 Scope id], [UW2][UW2 Scope id]}

**UNL Features** : Relations, Concept words (Universal Words) and Attributes.

## UNL Expression

- Smith is reading a novel.
- UNL Hypergraph



UNL Expression [UNL] agt(read(icl>do) @entry.@present.@progress, Smith(iof>person)) obj(read(icl>do) @entry.@present.@progress, novel(icl>book)) [/UNL]

### UNL Example



#### **EnConverter Module**

[S:00] {org:en} **Pfizer is accused of murdering 11 children** {/org} {unl}

**obj**(accuse(icl>do,equ>charge,cob>abstract\_thing,agt>person,obj>person).@entry.@prese nt,pfizer.@topic)

```
qua:01(child(icl>juvenile>thing).@pl,11)
```

obj:01(murder(icl>kill>do,agt>thing,obj>living\_thing).@entry,child(icl>juvenile>thing).@pl)

**cob**(accuse(icl>do,equ>charge,cob>abstract\_thing,agt>person,obj>person).@entry.@prese nt,:01)

{/unl}

[/S]



#### **Pre-processing Module**

Separation Module: Extraction individual UWs from T and H.

[Relation Name] [Relation Scope ID] {[UW1][UW1 Scope id], [UW2][UW2 Scope id]}

#### **Relation Grouping Module**: [46 Relations in UNL]

Group Name	Relations
Agent	agt, cag, aoj, cao, ptn
Object	obj, cob, opl, ben
Place	plc, plf, plt
Instrument	ins, met
State	src, gol, via
Time	tim, tmf, tmt, dur
Manner	man, bas
Logical	and, or
Concept	equ, icl,iof
Cause	con, pur, rsn
Sequence	coo, seq, cnt, mod, nam, per, pof, pos, qua

#### **Scoring Module**

The rules are

- i. Relation grouping rule
- ii. UW Rule
- iii. Name Entity Rule

## **Dataset used for Experiments**

Recognizing Textual Entailment (RTE) competitions

- i. RTE-1 in 2005
- ii. RTE-2 in 2006
- iii. RTE-3 in 2007
- iv. RTE-4 in 2008
- v. RTE-5 in 2009
- vi. RTE-6 in 2010
- vii. RTE-7 in 2011

#### Experiment Results

ρτε υλτλ	<b>F-SCORE</b>
<b>NIL DAIA</b>	(%)
RTE-1 Development Set	0.61
RTE-1 Test Set	0.63
RTE-2 Development Set	0.56
RTE-2 Test Set	0.61
RTE-3 Development Set	0.59
RTE-3 Test Set	0.68
RTE-4 Test Set	0.69
RTE-5 Development Set	0.58
RTE-5 Test Set	0.65
# Outline

Textual Entailment Lexical Entailment Syntactic Entailment Semantic Entailment Entailment Classification Application: Answer Validation Application: Question Answering

# **Entailment Class**

#### Before:

Yes / No

#### Now :

Entailment Class	Termed As	Definition
Forward Entailment	F	t1 entails t2 AND t2 does not entail t1
Reverse Entailment	R	t2 entails t1 AND t1 does not entail t2
Bidirectional Entailment	В	t1 entails t2 AND t2 entails t1
Contradiction	С	t1 and t2 contradict or cannot be true at
		the same time

### **Before:**

Monolingual: Text and Hypothesis is in Same language., i.e. English

#### Now:

Multilingual / Cross-lingual: Text is in one Language i.e. in Spanish and Hypothesis is in another language i.e. English

# Example

Entailment Class: Bidirectional

Text (t1): Mozart nació en la ciudad de Salzburgo.

[Google Trns: Mozart was born in Salzburg.] Text (t2): Mozart was born in Salzburg.

Entailment Class: Forward

Text (t1): Mozart nació el 27 de enero de 1756 en Salzburgo.

[Google Trns: Mozart was born on January 27, 1756 in Salzburg.] Text (t2): Mozart was born in 1756 in the city of Salzburg.

Entailment Class: Backward

Text (t1): Mozart nació en la ciudad de Salzburgo. [Google Trns: Mozart was born in Salzburg.]

Text (t2): Mozart was born on 27th January 1756 in Salzburg.





# Dataset@Evaluation Track

- NTCIR-9 Recognizing Inference in TExt (RITE) Tasks in 2011 and 2012
- Cross-lingual Textual Entailment for Content Synchronization in SemEval-2012 and SemEval-2013

# Outline

Textual Entailment Lexical Entailment Syntactic Entailment Semantic Entailment Entailment Classification Application: Answer Validation Application: Question Answering

# **Application: Answer Validation**



# Find the answer to a question in a large collection of documents

# **Answer Validation**

#### Validate the correctness of real systems answers



Answer is not correct or not enough evidence



If the text entails the hypothesis, then the answer is expected to be correct.



# **Answer Validation System**

- Input: <Question, Candidate Answer, Supporting Text>
- Output: "Boolean Value" indicating if the Answer is correct for the Question according to the Supporting Text or not.

Question	In what date was the first tennis championship at			
	Wimbledon?			
Supporting Text	The first championships at Wimbledon, in London were played in			
(i.e., T)	1877.			
Answer	1877			
Expected	The first tennis championship at Wimbledon was in 1877.			
Answer (i.e., H)				

< q id="0061" lang="EN">

<q\_str>Where was Joseph Fourier born?</q\_str>
<a id="0061 1" value="REJECTED">

<a\_str>*Paris*</a\_str>

<t\_str doc="Joseph Fourier">Joseph Fourier Joseph Fourier Born March 21, 1768 Auxerre, Yonne, France Died May 16, 1830 Paris, France Residence France Nationality French Field Mathematician, physicist, and historian Institution École Normale École Polytechnique Alma Mater École Normale Doctoral Advisor Joseph Lagrange Doctoral Students Gustav Dirichlet Giovanni Plana Known for Fourier transform Religion Roman Catholic Jean Baptiste Joseph Fourier (March 21, 1768 - May 16, 1830) was a French mathematician and physicist who is best known for initiating the investigation of Fourier series and their application to problems of heat flow.

</a>

<a id="0061\_2" value="UNKNOWN">

<a\_str>France</a\_str>

<t\_str doc="Joseph Fourier">Life Fourier was born at Auxerre in the Yonne
département of France, the son of a tailor.</t\_str>

</a>

<a id="0061\_3" value="REJECTED">

<a\_str>*Grenoble*</a\_str>

<t\_str doc="Joseph Fourier University">Joseph Fourier University Joseph
Fourier University Université Joseph Fourier (Joseph Fourier University) is a French university situated in the
city of Grenoble and focused on the fields of sciences, technologies and health.</t\_str>

</a>

<a id="0061\_8" value="REJECTED">

<a\_str>*Auxerre*</a\_str>

<t\_str doc="Auxerre.html">Auxerre.</t\_str>

</a>

# System Architecture



# Answer Pattern generation

Convert

Question Assertive sentence (e.g. Expected Hypothesis)







	Text (T)	Joseph Fourier Joseph Fourier Joseph Fourier Jean Baptiste Joseph Fourier Born March			
		21, 1768 Auxerre, Yonne, France Died May 16, 1830 Paris, France Residence France			
		Nationality French Field Mathematician, physicist, and historian Institution École Normale			
		École Polytechnique Alma Mater École Normale Doctoral Advisor Joseph Lagrange Doctora			
		Students Gustav Dirichlet Giovanni Plana Known for Fourier transform Religion Roman			
Pair 1		Catholic Jean Baptiste Joseph Fourier (March 21, 1768 - May 16, 1830) was a French			
		mathematician and physicist who is best known for initiating the investigation of Fourier			
		series and their application to problems of heat flow.			
	Hypothesis (H)	Joseph Fourier was born Paris.			
	Text (T)	Life Fourier was born at Auxerre in the Yonne département of France, the son of a tailor.			
Pair 2					
	Hypothesis (H)	Joseph Fourier was born France.			
	Text (T)	Joseph Fourier University Joseph Fourier University Université Joseph Fourier (Joseph			
		Fourier University) is a French university situated in the city of Grenoble and focused on the			
Pair 3		fields of sciences, technologies and health			
	Hypothesis (H)	Joseph Fourier was born Grenoble.			
Pair 4	Text (T)	Auxerre			
	Hypothesis (H)	Joseph Fourier was born Auxerre.			

## System Architecture



## Features

- Lexical similarity features: WordNet based Unigram match, Bigram match, Longest Common Subsequence match, Skip-gram and stemming.
- Lexical distance features: Vector Space Measures (Euclidean distance, Manhattan distance, Minkowsky distance, Cosine similarity, Matching coefficient), Set-based Similarities (Dice, Jaccard, Overlap, Cosine, Harmonic), Soft-Cardinality, Q-Grams Distance, Edit Distance Measures (Levenshtein distance, Smith-Waterman Distance, Jaro).
- Syntactic features: Subject-subject comparison, WordNet Based Subject-Verb Comparison, Subject-Subject Comparison, Object-Verb Comparison, WordNet Based Object-Verb Comparison, Cross Subject-Object Comparison, Number Comparison, Noun Comparison, Prepositional Phrase Comparison, Determiner Comparison and Other relation Comparison.

#### Semantic Features from UNL

# Machine Learning

**Tool generally used**: LIBSVM (Chang and Lin, 2001)

**Training Data sets**: RTE-1 development and test set, RTE-2 development and annotated test set, RTE-3 development and annotated test set and RTE-4 annotated test set. 3967 text-hypothesis pairs have been used.

Test Data sets: AVE 2008 data set.

**Output**: Entailment score with entailment decisions (i.e., "YES" or "VALIDATED" / "NO" or "REJECTED").

# **Question Answer Type Module**

Factoid Question Type	<b>Expected Answer</b>
Who	PERSON
When	DATE / TIME
Where	LOCATION
What	OBJECT
How	MEASURE

Question	Where was the Volkswagen Polo Playa built?			
Target Answer	LOCATION			
Pair ID	Answer	NER	Target Answer	
0108_1	South Africa	Location	VALIDATED	
0108_2	Successor	Х	REJECTED	
0108_3	Pointer	Х	REJECTED	
0108_9	Tom Ennis	Person	REJECTED	
0108_11	Ibiza	Х	REJECTED	
0108_12	Playa	Х	REJECTED	
0108_13	Europe	Location	VALIDATED	



# CLEF 2008 Answer Validation Exercise track

# **Evaluation Result**

	AVE Development Data	AVE Test Data
	Set	Set
"VALIDATED" in the Development Set	21	79
(Gold Standard)		
"VALIDATED" the proposed my AV	32	75
system output		
"VALIDATED" match	18	54
Precision	0.56	0.72
Recall	0.85	0.68
F-score	0.68	0.69

# **Application: Question Answering**

# Participated Evaluation Track

- ResPubliQA @ CLEF 2010
- QA4MRE @ CLEF 2011
- QA4MRE @ CLEF 2012
- QA4MRE @ CLEF 2013

# QA4MRE: Question Answering System

### A Hybrid system based on

- A system based on Information Retrieval and Answer Validation
- QA based Machine Reading system

Goal:

An Answer Validation System (AV) based on Textual Entailment and Question Answering.

# Method

- Hypothesis Generation: Combine the question and the answer into Hypothesis (H) and each retrieved sentence from passage is considered as the Text (T) to identify the entailment relation and give each pair as an entailment score.
- The important features in the AV system are Lexical Textual Entailment, Named Entity Recognition, Question-Answer type analysis, Chunk boundary module and Syntactic similarity module.
- The proposed AV system is rule based.
- Textual Entailment module

# Basic Model: QA4MRE





Validate Factor Generator

# Module

## Document Processing Module

- XML parser
- Anaphora Resolution
- Validate Factor Generator
  - Answer Pattern Generation
  - Answer Validation
- Answering Module
  - Scoring Assignment
  - Answer Selection

Track Name	C@1	Technique	Position
ResPubliQA @ CLEF 2010	0.50	IR, NGram	7 <sup>th</sup> out of 9
QA4MRE @CLEF 2011	0.57	IR, Answer Validation, Textual Entailment	1 <sup>st</sup> out of 25
QA4MRE @ CLEF 2012	0.53	IR, Answer Validation, Textual Entailment, Anaphora Resolution	1 <sup>st</sup> out of 21
QA4MRE @ CLEF 2013	0.59	IR, Answer Validation, Textual Entailment, Anaphora Resolution, World Knowledge	1 <sup>st</sup> out of 19
## Resources & Tools

- Parsers: Stanford Dependency, CCG Parser
- Named Entity: Stanford NER, LT-TTT2
- Stanford POS Tagger, Stanford Core NLP
- NLTK
- Text Similarity tool: SimMetrics, SimPack and SecondString, DISCO, SEMILAR, Soft Cardinality
- Wikipedia Knowledge
- Anaphora Resolution GuiTAR, JavaRAP
- Machine Learning: Weka tools (e.g. SVM, J48, etc. algorithm)
- Universal Networking Language (UNL)

- No. of Journal: 6
- No. of International Conference: 34
- No. of Book: 1

www.parthapakray.com

## Thank you

