





KE5205 TEXT MINING 2018

INFORMATION EXTRACTION

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Outline for this Module



- What is Information Extraction (IE)?
- How does an IE System work?
 - Rule based methods for IE
 - Statistics based methods for IE
 - Coreference Resolution
- Practical Information Extraction
 - Example on how to extract concepts using a tool
 - Analysis of the tool and the results







FROM WORDS TO CONCEPTS





- A word is a bunch of characters separated by spaces or punctuation
 - "The cat is sitting on the mat" has 7 words
- How about "The cat's sitting on the mat" how many words?
- How about "新加坡" (Chinese) how many words? Characters?
- How about "memperkuasakan" (Malay) how many words?







- "Cardinal Jorge Mario Bergoglio of Buenos Aires was elected the 266th pope of the Catholic Church on 13 May 2013, taking the name Pope Francis"
- Cardinal: a concept = a prince of the Catholic Church
- Jorge Mario Bergoglio: a concept = Cardinal of Buenos Aires
- Buenos Aires: a concept = city in Argentina
- Pope: a concept = head of Catholic Church
- Catholic Church: a concept = religious organization
- 13 May 2013: a concept = a date
- Pope Francis: a concept = current Pope







How do you "automatically" create concepts

- Start with nouns or names or simple named entities (NEs)
 - Identify NEs using dictionaries = lists = gazetter = ...
 - Eg, name lists: boy names, girl names, world names,...
 - Eg, days, months, cities, countries, ... (can be multiwords)
 - Proper nouns = words that start with Capitals (in the middle of sentence)
- If nouns or names are <u>contiguous</u>, then aggregate them
 - Jorge + Mario + Bergoglio = "Jorge Mario Bergoglio"
 - Pope + Francis = "Pope Francis"
 - Catholic + Church = "Catholic Church"
- Some lexical and syntactic markup can help
 - Cardinal [title] + Jorge Mario Bergoglio[name] + of[prep] + Buenos Aires[place]
 - with the right rule: "Cardinal Jorge Mario Bergoglio of Buenos Aires" is a concept



Concept vs. Named Entity vs. Information



- Name Entity = lowest level of recognition by an IE system
 - Normally recognized by dictionaries or rules
- Concept = rule or heuristic to create an abstraction
 - Sometimes called a "natural class" = different people at different times and in different places would refer to the same referent with that concept
 - "president of the United States" vs. "president of the United Kingdom"
- Information = words, named entities, concepts which fulfill a need
 - So if you have a question, and a phrase answers that question, then that phrase is an example of information
 - Information is often regular, i.e., with a pattern
 - Eg, information about a person = name, age, sex, address, hp#, ...
 - Information about a company = name, address, stock symbol, Chairman,

. . .













The National University of Singapore and The Hebrew University of Jerusalem (HUJ) are launching a Joint Doctor of Philosophy degree programme in biomedical science from August 2013. Professor Tan Eng Chye, NUS Deputy President and Provost, and Professor Menahem Ben-Sasson, President of HUJ signed the joint degree agreement at NUS, in the presence of Ambassador of Israel to Singapore Her Excellency Amira Arnon and about 30 invited guests.

Identify concepts from the text above, and categorize the concepts. Follow the examples below:

#	Concept	Category	#	Concept	Category
1	National University of Singapore	Place	7		
2	Tan Eng Chye	Name	8		
3			9		
4			10		
5			11		
6			12		







- Professor Tan Eng Chye, NUS Deputy President and Provost, and Professor Menahem Ben-Sasson, President of HUJ signed the joint degree agreement at NUS, in the presence of Ambassador of Israel to Singapore Her Excellency Amira Arnon and about 30 invited guests.
- The following is a simple lexical rule to recognize names of universities in the paragraph above:

"The" + word_[capitalized] + "University" + "of" + Place

 Write a rule similar in structure to recognize the names of persons mentioned in the paragraph above. You may assume reasonable dictionaries (lists) exist and that suitable parts of speech and meaningful markup has been done:

10









🖶 Contrast with a search engine





IMAGES N

EWS

MOR

oi∩g∵∣i

how tall is the eiffel tower



4,850,000 RESULTS

Narrow by language ▼

Narrow by region ▼

How Tall is the Eiffel Tower? - Buzzle

www.buzzle.com/articles/how-tall-is-the-eiffel-tower.html >

How Tall is the Eiffel Tower? Eiffel Tower is the tallest man-made marvel in France. This massive latticework structure is owned by the city of Paris and is a ...

Eiffel Tower - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Eiffel_Tower >

History · Design of the tower · Tourism · Attempted relocation · Economics

The **tower** stands 320 metres (1,050 ft) **tall**, about the same height as an 81-story building. During its construction, **the Eiffel Tower** surpassed the Washington Monument to ...

How tall is The Eiffel Tower? | Height of the Celebrities

howtallis.info/eiffel-tower.html >

Want to know **How tall is The Eiffel Tower**? Ok, here we are. Find the answer of this question. **Eiffel Tower** height is 324 metres (1,063 ft)

How tall is the Eiffel tower? - Yahoo! Answers India

in.answers.yahoo.com/question/index?qid=20071125233320AAnuotX $\,\,^{\star}$

Resolved · 7 total answers

26/11/2007 · Best Answer: According to The Oxford Dictionary of Phrase and Fable: "Eiffel Tower a wrought-iron structure erected in Paris for the World Exhibition of ...

Eiffel Tower Height, How Tall Is The Eiffel Tower?

myeiffeltower.com/eiffel-tower-height *

Find out the exact Eiffel Tower height here! Discover how tall is the Eiffel Tower - the most visited monument in the world!

How tall is the Eiffel Tower? - Yahoo! Answers

answers.yahoo.com/question/index?qid=20070216082426AA7wNpQ >

Resolved · 3 total answers

16/2/2007 · Best Answer: Eiffel Tower (French: La Tour Eiffel, /tuռ ɛfɛl/) is an iron tower built on the Champ de Mars beside the River Seine in Paris, France. It ...

How tall is the Eiffel Tower?

- A search engine returns a list of documents which (hopefully) includes the answer
- The user needs to open/read the document to get the answer



Information extraction version







how tall is the eiffel tower





Mun Kew Leong





Search

About 2,420,000 results (0.26 seconds)

Web

Images

Maps

Videos

News

Shopping

More

Show search tools

324 m

Eiffel Tower, Height Hide details

How tall is the Eiffel Tower - answers.com

The Eiffel tower is located in Paris, France. It is 324 m ...

Eiffel Tower Height, How ... - myeiffeltower.com

Eiffel Tower height - 324 m. of iron parts! In 1889 ...

How tall is the eifel tower? - Yahoo! Answers - yahoo.com ... the structure is 324 m (1058 ft) ...

.. alo oa dolalo lo **02**4 III (1000 lt) ...

Is this accurate? Yes - No

Eiffel Tower - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Eiffel_Tower

Jump to <u>Lattice towers taller than the Eiffel Tower</u>: The **Eiffel Tower** is a puddled iron lattice tower located on the Champ de Mars in Paris, named ...

History - Design of the tower - Tourism - Attempted relocation

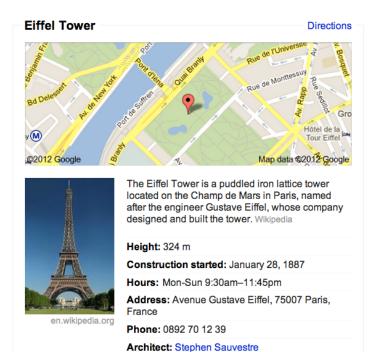
How tall is the Eiffel Tower

wiki.answers.com > ... > France > Paris > Eiffel Tower

The **Eiffel Tower** is 100 meters **tall**. **How tall is the Eiffel tower** and which city is it in? The **Eiffel tower** is located in Paris, France. It is **324 m** (1,063 ft) **tall**.

Would Shard critics once have slammed the **Eiffel Tower?** | Daniel ... www.guardian.co.uk/commentisfree/.../shard-critics-**eiffel-tow**... Share

5 Jul 2012 – But while the main body of the **Eiffel Tower** is only 300 metres **tall**, it stands at 324 metres once you add the base and television antenna.



People also search for



How tall is the eiffel tower?



324 m

Answer

Eiffel Tower, Height Hide details

How tall is the Eiffel Tower - answers.com
The Eiffel tower is located in Paris, France. It is 324 m ...

Eiffel Tower Height, How ... - myeiffeltower.com Eiffel Tower height - 324 m. of iron parts! In 1889 ...

How tall is the eifel tower? - Yahoo! Answers - yahoo.com ... the structure is **324 m** (1058 ft) ...

Is this accurate? Yes - No

Sources (evidence)



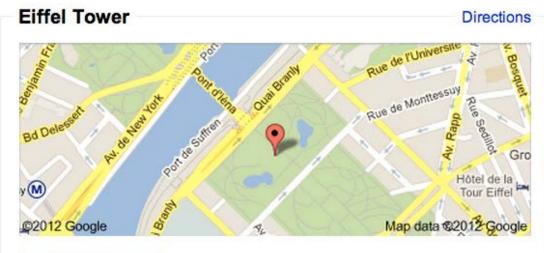
Template version





Automatic template filling

- Different fields for different places
- Depends on sources; not fixed template
- Information Extraction
- High degree of accuracy
- Human checking?





en.wikipedia.org

The Eiffel Tower is a puddled iron lattice tower located on the Champ de Mars in Paris, named after the engineer Gustave Eiffel, whose company designed and built the tower. Wikipedia

Height: 324 m

Construction started: January 28, 1887

Hours: Mon-Sun 9:30am-11:45pm

Address: Avenue Gustave Eiffel, 75007 Paris,

France

Phone: 0892 70 12 39

Architect: Stephen Sauvestre





en.wikipedia.org

The Lincoln Memorial is an American national monument built to honor the 16th President of the United States, Abraham Lincoln. It is located on the National Mall in Washington, D.C. across from the Washington Monument. Wikipedia

Height: 30 m

Area: 66 ha

Address: 2 Lincoln Memorial Cir NW, Washington,

DC 20037

Hours: Mon-Sun Open 24 hours

Architectural styles: Beaux-Arts architecture,

Doric order

Architect: Henry Bacon





plus.google.com

The White House is the official residence and principal workplace of the President of the United States. Wikipedia

Construction started: 1792

Address: Alexander Hamilton Place Northwest,

Washington, DC 20005

Phone: (202) 208-1631

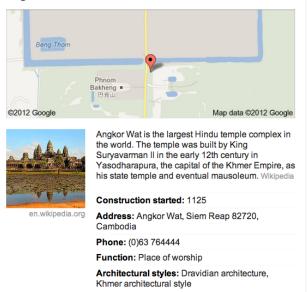
Hours: Mon-Sun 7:30am-4pm

Architectural styles: Neoclassicism, Palladian

architecture, Neoclassical architecture

Architect: James Hoban

Angkor Wat









en.wikipedia.org

Machu Picchu is a pre-Columbian 15th-century Inca site located 2,430 metres above sea level. Machu Picchu is located in the Cusco Region of Peru, South America. Wikipedia

Address: Muro de la Ciudad, Peru Management: Government of Peru

Phone: (0)1 610 0400



What is information extraction?



- The automatic extraction of (possibly prespecified) information from natural language documents
 - Facts about types of <u>entities</u>, <u>events</u>, <u>relationships</u>
- The automatic population of a structured information source (template) from natural language documents (i.e., create a table!)
 - Documents may be semi-structured (eg., patents), unstructured (e.g., websites) or free text (e.g., documents)





- IE does not specify the documents to be read; rather it extracts the salient information from the documents and presents just those information to the user
- Normally maintains links from facts to source documents
 - Allows evidence trail, context, further discovery
- Accuracy is variable both in completeness and reliability
- Certain information (e.g., person names; organizations) have high reliability. Other information (e.g., sentiment) have very poor accuracy.



Drilling down on information...



Entity/Concept:

an object of interest (date, person, building,...)

Attribute:

property of an entity (name, height, type)

Fact:

- a predicate about an entity, e.g., tall(X)
- a relationship between two or more entities, e.g., married(X,Y)

Event

 an abstraction over a possibly sequential relationship of several entities e.g., terrorist attack, birthday party, etc.

Example (Entities and Facts)

- Fletcher Maddox, former Dean of the UCSD Business School, announced the formation of La Jolla Genomatics together with his two sons. La Jolla Genomatics will release its product Geninfo in June 1999. Geninfo is a turnkey system to assist biotechnology researchers in keeping up with the voluminous literature in all aspects of their field.
- Dr. Maddox will be the firm's CEO. His son, Oliver, is the Chief Scientist and holds patents on many of the algorithms used in Geninfo. Oliver's brother, Ambrose, follows more in his father's footsteps and will be the CFO of L.J.G. headquartered in the Maddox family's hometown of La Jolla, CA.

Persons:	Organizations:	Locations:	Artifacts:	Dates:	
Fletcher Maddox	UCSD Business School	La Jolla	Geninfo	June 1	ŝ
Dr. Maddox	La Jolla Genomatics	CA	Geninfo		
Oliver	La Jolla Genomatics				
Oliver	LJ.G.				
Ambrose					
Maddox					

PERSON	Employee_of	ORGANIZATION
Fletcher	Employee_of	UCSD Business
Maddox	Employee_of	School
Fletcher		La Jolla Genomatics
Maddox	Employee_of	La Jolla Genomatics
Oliver		La Jolla Genomatics
Ambrose		
ARTIFACT	Product_of	ORGANIZATION
Geninfo	Product_of	La Jolla Genomatics
LOCATION	Location_of	ORGANIZATION
La Jolla	Location of	La Jolla Genomatics
La Jona	Excauou_or	La Jona Genomanes

From: Ron Feldman, Information Extraction: Theory and Practice, http://cs.fit.edu/~pkc/icdm03/printing/tutorials/extraction/extraction.tutorial.pdf

Events and Attributes

		NAME:	Fletcher Maddox
			Maddox
		DESCRIPTOR:	former Dean of the UCSD
			Business School
COMPANY-F	ORMATION_EVENT:	_	his father
COMPANY:	La Jolla Genomatics		the firm's CEO
	Fletcher Maddox	CATEGORY:	PERSON
PKINCIPALS	1	NAME:	Oliver
	Oliver	DESCRIPTOR:	
	Ambrose		Chief Scientist
DATE:		CATEGORY:	PERSON
		NAME:	Ambrose
CAPITAL:		DESCRIPTOR:	
			the CFO of L.J.G.
RELEASE-EV	TENT-	CATEGORY:	PERSON
KELLINGE-EV	1241.	NAME:	UCSD Business School
		DESCRIPTOR:	
COMPANY:	La Jolla Genomatics	CATEGORY:	ORGANIZATION
PRODUCT:	Geninfo	NAME:	La Jolla Genomatics
			L.J.G.
DATE:	June 1999	DESCRIPTOR:	
COST:		CATEGORY:	ORGANIZATION
		NAME:	Geninfo
		DESCRIPTOR:	its product
		CATEGORY:	ARTIFACT
		NAME:	La Jolla
		DESCRIPTOR:	the Maddox family's hometown
١		CATEGORY:	LOCATION
`		NAME:	CA
•		DESCRIPTION.	

From: Ron Feldman, Information Extraction: Theory and Practice, http://cs.fit.edu/~pkc/icdm03/printing/tutorials/extraction/extraction.tutorial.pdf







Unstructured Text

POLICE ARE INVESTIGATING A ROBBERY THAT (
ELEVEN STORE LOCATED AT 2545 LITTLE RIVE
LINCOLNIA AREA ABOUT 12:30 AM FRIDAY. A 2
ALEXANDRIA AREA EMPLOYEE WAS APPROAC
DEMANDED MONEY. SHE RELINQUISHED AN U
OF CASH AND THE MEN LEFT. NO ONE WAS IN
DESCRIBED AS BLACK, IN THEIR MID TWENTIE:
FEET NINE INCHES TALL, WITH MEDIUM BUILDS
CLEAN SHAVEN. THEY WERE BOTH WEARING
BLACK COATS. ANYONE WITH INFORMATION A
OR THE SUSPECTS INVOLVED IS ASKED TO CA

Structured	(Desired)	Information

Crime	Address	Town	Time	Day
ABDUCTION	8700 BLOCK OF LITTLE RIVER TURNPIKE,	ANNANDAL E	11:30 PM	SUNDA Y
ROBBERY	7- ELEVEN STORE LOCATED AT 2545 LITTLE RIVER TURNPIKE,	LINCOLNIA	12:45 AM	FRIDAY
ROBBERY	7- ELEVEN STORE LOCATED AT 5624 OX ROAD,	FAIRFAX	3:00 AM	FRIDAY

From: Ron Feldman, Information Extraction: Theory and Practice, http://cs.fit.edu/~pkc/icdm03/printing/tutorials/extraction/extraction.tutorial.pdf







Read this text:

During the Boston Marathon on April 15, 2013, two pressure cooker bombs exploded at 2:49 pm EDT killing 3 people and injuring 264 others. The bombs exploded about 13 seconds and 210 yards (190 m) apart, near the finish line on Boylston Street. The Federal Bureau of Investigation (FBI) took over the investigation, and on April 18, released photographs and surveillance video of two suspects. The suspects were identified later that day as the Chechen brothers Dzhokhar and Tamerlan Tsarnaev.

(from: https://en.wikipedia.org/wiki/Boston_Marathon_bombings)

Fill up this template:

Attribute	Value
Occurrence	
Event	
Location	
Start Date	
Start Time	
End Date	
End Time	
Method	
Incident_1	
Incident_2	
Incident_3	
Perpetrators	
Caught (Y/N)	■







- Switzerland, Sweden and Singapore are the three most innovative countries in the world, according to the Global Innovation Index (GII) 2012, a ranking of 141 countries co-produced by INSEAD and the World Intellectual Property Organization (WIPO, a specialised agency of the United Nations) published in Geneva on July 3, 2012. Currently in its fifth year, the GII measures the degree to which countries and economies integrate innovation into their political business and social spheres. Knowledge partners for the GII 2012 are Alcatel-Lucent, Booz & Company and the Confederation of Indian Industry (CII).
- This is the second year running that Switzerland, Sweden and Singapore have been in the top three
 positions. The rest of the top ten this year are: Finland, the United Kingdom, the Netherlands, Denmark,
 Hong Kong (China), Ireland and the United States. Canada dropped out of the top ten this year, while
 the U.S. fell to tenth position from number seven last year, changes which the report attributes to
 cutbacks in spending on, and support of, education and research and development.

(from: http://knowledge.insead.edu/innovation/global-innovation-index-2012-481)

Read the text above and fill in the table below:

Country	2012 Rank	2011 Rank	Country	2012 Rank	2011 Rank
	1			6	
	2			7	
	3			8	
	4			9	
	5			10	











Rule-based Systems

- Hand-coded rules
 - Coded by linguists, with domain input
 - Iterative method based on document inspection
 - Slow but very good results
- Induced (machine learning) rules
 - Fully machine learning
 - Given an annotated corpus, derive a basis set of rules that cover a pre-determined % of the annotated examples (and only the annotated examples)
 - Heuristic approach: one rule at a time!
 - Hybrid systems machine learning to fine-tune the rules



Approaches to building an IE system





Statistics-based Systems

- Start with a well-annotated corpus
- Depending on the method (e.g., Hidden markov models), derive statistical rules to create a model that generates the examples
- Advantages compared to Rule based systems
 - Language independent (within representational limits)
 - No linguistic or domain knowledge needed in the team
 - Relatively small effort in creating the models

Issues

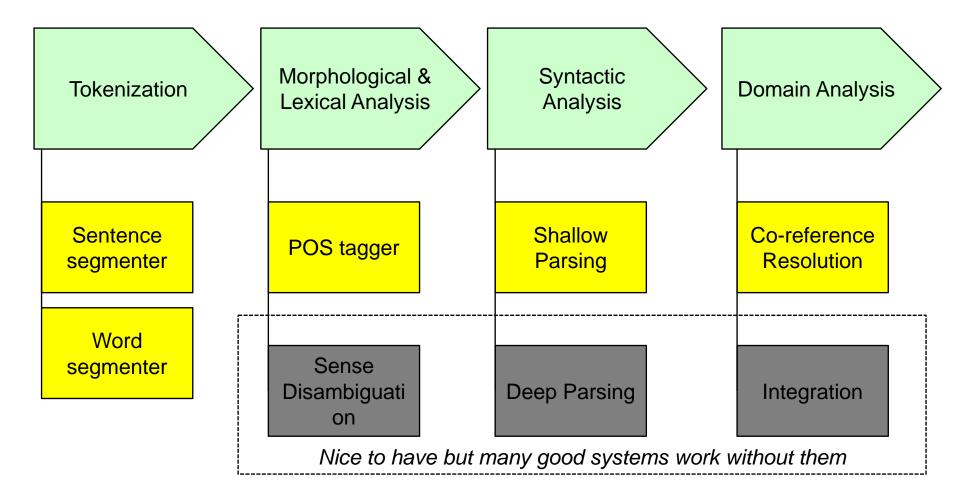
- The complexity moves to the corpus must be well annotated and must cover the full space of possibilities
- Requires <u>very large number of training examples</u> to get good results







Main components of an IE system

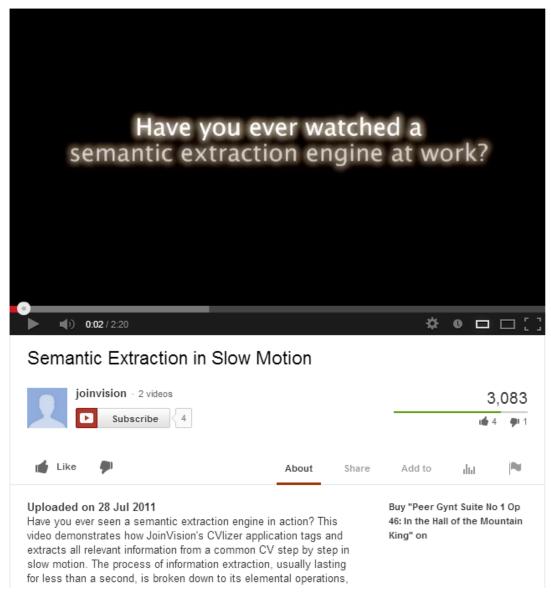


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Upload





From: http://www.youtube.com/watch?v=gsAATmT0qEc



Issues in IE - Tokenization





Language Considerations

- Morphology is easy in English, much harder in German, Hebrew
- Word boundary easy in roman script, ambiguous in character scripts
- English is one-dimensional, other languages have more than one

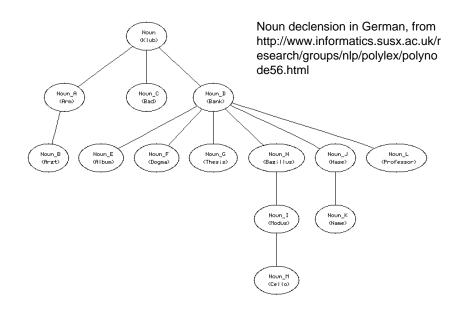
original, un-segmented text

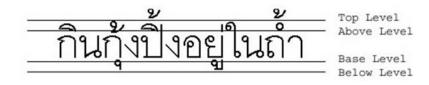
再往远些看,随着汉字识别和语音识别技术的发展, 中文计算机用户将跨越语言差异的鸿沟, 在录入上走向中西文求同的道路。

separated word entities after segmentation

再 往 远 些 看 , 随着 汉字 识别 和 语音 识别 技术 的 发展 , 中文 计算机 用户 将 跨越 语言 差异 的 鸿沟 , 在 录入 上 走 向 中 西文 求 同 的 道路 。

Word segmentation in Chinese, from http://blog.faroo.com/2009/06/09/lightweight-chinese-wordsegmentation/





Word composition in Thai, from http://www-01.ibm.com/software/globalization/topics/thai/character.html



Issues in IE – Language in use (pragmatics)





Examples of difficulties:

- Sarcasm
 - John is really smart…like a coconut!
- Negation and ambiguity
 - "All that glitters is <u>not</u> gold"
 - *For all (x), if x glitters, then x is (not gold)
 - It is not the case that (for all (x), if x glitters, then x is gold)

Prosody

- John and Anne are married? I thought the wedding was next year!
- John and Anne are married? I thought they were just going out!
- John and Anne are married? I thought John was interested in Mary!
- John and Anne are married? I thought Anne was going out with Tom!
- John and Anne are married? I thought they were both single!

Colloquialisms

"Referee kayu! Blind like donkey, is it?! So obvious ball out one leh!"





- Genres ("styles" of documents)
 - Newspapers, journal articles, patents
 - Semi-structured there is a pattern to how artifacts are created
 - Emails, tweets, SMSs
 - Semi-structured, well defined fields
 - Short or very short "sentences"
 - Speech transcripts (e.g., court documents, Hansard, etc.)
 - Natural speech is rarely fluent
 - Interjections, repetitions, hesitations, malapropisms, indexicals, etc.

Structure

- Information in tables, in captions
- Precedence (implied order)







RULE-BASED IE METHODS



When to use rule-based systems?



Rule-based systems can and do work well

- Corpus is relatively static (in terms of vocabulary, language structure, etc.)
- Can be fast especially in well-defined limited domains (compared to annotating training examples)

A typical rule-based system comprises

- Set of rules
- Policies to control when and how (multiple) rules are applied, e.g., order, looping.

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What does a rule look like?



• Form:

Match(pattern) then Do(action)

```
Rule: Company1 from gate.ac.uk

(({Token.orthography == upperInitial})+
{Lookup.kind == companyDesignator}
):match
-->
:match.NamedEntity = { kind=company, rule="Company1" }
```

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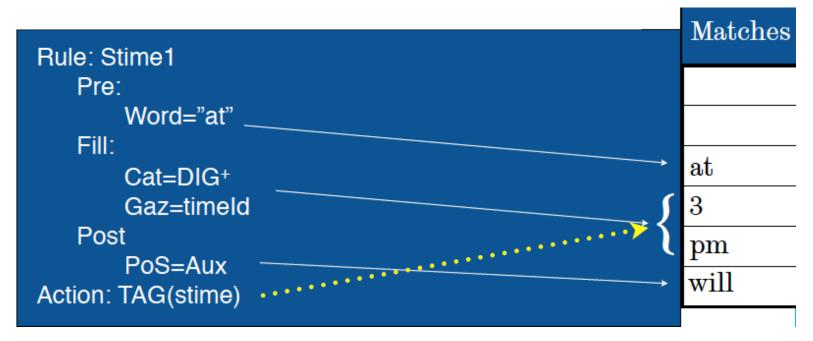
Types of IE recognition rules





Lexical pattern matching – whole entity patterns

- Classic rule: (left context + filler + right context) patterns
- Simplistic: entities are considered independent
- More natural for hand-coded rules.





Types of IE recognition rules





Lexical pattern matching – boundary patterns

- Start boundary is distinct from end boundary
- Different rules are required to recognize the entity
- Developed primarily for machine learning of rules

```
Rule: Stime1
Pre:
Word="seminar":
Word="at":
Post:
Cat=DIG+
Gaz=timeId

Action: TAG(<stime>)

Matches

The
seminar
at
3
pm
```



Types of IE recognition rules





- Structural pattern recognition
 - Typically used only for highly structured text
 - Can recognize more than one entity at a time

```
Example:
Capitol Hill- 1 br twnhme. D/W W/D. Pkg incl
```

\$675. 3 BR upper flr no gar. \$995. (206)999-9999

Rule:

ID:7

Pattern: * ('Capitol Hill') * (Digit) * '\$' (Number)

Output: Rental (Neighborhood \$1) {Bedrooms \$2} {Price \$3}

Rule from: STEPHEN SODERLAND:

Learning Information Extraction Rules for Semi-structured and Free Text,

Machine Learning 1, 440









When to use statistics based systems?





Current top performing systems are statistics based

- Machine learning (ML) on very large corpora is state-of-theart
- Beyond the scope of this course

Annotation based corpora for training

- You have a well annotated corpora with many features
- Various ML techniques from simple to sophisticated
- Relatively homogeneous real data (not training data) in any given domain. Note that models don't transfer well across domains
- You don't have domain or language resources in that area



Simple model is at token level



- Text is a linear sequence of tokens (such as words)
- Token boundaries can be fairly easily derived in some languages, e.g., space & punctuation for English, but much harder for others, e.g., Chinese
- Simple tokenization
 - Dictionary based
 - Colocation frequencies (see next page)
- Alternatives
 - Ignore multi-unit tokens
 - Bi-grams, tri-grams, multi-grams



Colocation frequencies identify boundaries



...worked at Coca Cola for 27 years...

Coca Cola reported its quarterly earnings...

"Coca" and other words

co-located less
frequently

"Coca" and "Cola" co-located more frequently

implying that "Coca Cola" is a single token

- Colocation frequencies (given sufficient examples) can very accurately identify token boundaries.
- Greedy algorithm the longer the token, the better
- Works on non-English languages, e.g., Chinese





Apple Inc	(AAPL)	reported	its	earnings	for	Q3	2010
Organization	Stock	Null	Null	Null	Null	Quarter	Year
	symbol						

Neil	Armstrong	died	August	25	,	2012	in	Cincinnati
Null	Null	Null	dateStart	dateContinue	DateContinue	DateEnd	Null	Null

- The IE task is to assign a <u>pre-defined</u> entity label to each token in the stream
- Two examples above different methods
- Treated as a generative model
 "Null" is the output when the model fails





- **Hidden Markov Models (HMM)**
 - Simple, joint probability
- **Conditional Random Fields (CRF)**
 - Conditional probability
 - Considers features of current token, and of preceding *n* tokens (window=n)
- Similarity algorithms
 - Measure distance of group of words to a dictionary list
 - Works especially well for jargon and other terminology
- **Support Vector Machines (SVM)**
 - Training method for standard perceptron
 - Optimize the points to determine the hyperplane dividing the positive training samples from the negative ones











Coreference resolution

- Determine relationship between entities which are related
 - Identity relation (morning star vs. evening star)
 - Whole-part relation
- Simple version
 - Determine entities which have the same referent
 - Anaphora (Pronouns)
 - Proper names, proper nouns, noun phrases,...
 - Definite descriptions (may be time dependent)
 - Usain Bolt & "the fastest man in the world"



Examples & Discussion





Coreference Examples

What do you need to know to do accurate co-reference?

Anaphora

- The <u>elephant</u> stepped on the <u>rabbit</u> and <u>it</u> died.
- The <u>elephant</u> stepped on the landmine and it died.

Proper nouns

 John Smith and Mary Brown were married this morning. <u>The</u> <u>groom</u> was dressed in a white tuxedo while <u>the bride</u> was...

• Definite descriptions

 Usain Bolt has won the Olympic 100m gold medal. The fastest man in the world successfully defended his title last night.









Making the right choice



- Buy an IE package
- Economies of scale
- Expensive, but much cheaper than building your own
- One size fits all (mostly)
- Use out-of-the-box
 - Start fast; explore; prototype
- Limited customization
 - Still need skills

Build an IE package

- Needs special skills
- Needs a lot of time
- You have training data
- Get exactly what you want
- When the out-of-the-box system is the limiting factor
- When it's for sustainable strategic advantage



Reference & Resources



- Ron Feldman, Information Extraction: Theory and Practice, http://cs.fit.edu/~pkc/icdm03/printing/tutorials/extraction/extraction.tutorial.pdf
- Fabio Ciravegna, Tutorial on Information Extraction from Text, http://www.isweb.unikoblenz.de/files/ssms09/SSMS_Slides/ciravegna-IE text.pdf
- Chris Manning & Hinrich Schutze, Foundations of Statistical Natural Language Processing, MIT Press, 1999
- NLP resources: http://nlp.stanford.edu/links/statnlp.html