## Information Extraction - Introduction

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# Goal: "machine reading"

#### Goal

Acquire structured information knowledge from unstructured text



### Information Extraction (IE) Systems

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#### Goals

- Organize information so that it is useful to people
- Put information in a semantically precise form that allows further inferences to be made by computer algorithms

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- The headquarters of BHP Billiton Limited, and the global headquarters of the combined BHP Billiton Group, are located in Melbourne, Australia.
- headquarters("BHP Biliton Limited", "Melbourne, Australia")

#### Example

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We can extract the following information,

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Such information can be used by search engines and database management systems to provide better services to end users.

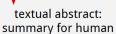
# Applications in Biomedical domain

#### Biomedical domain

- A large amount of scientific publications
- Need to look for discoveries related to particular genes, proteins or other biomedical entities
- Biomedical entities often have synonyms and ambiguous names
- Critical task: automatically identify mentions of biomedical entities in text and link them to their corresponding entries in existing knowledge bases.

## Biomedical domain







Subject	Relation	Object
p53	is_a	protein
Bax	is_a	protein
p53	has_function	apoptosis
Bax	has_function	induction
apoptosis	involved_in	cell_death
Bax	is_in	mitochondrial outer membrane
Bax	is_in	cytoplasm
apoptosis	related_to	caspase activation

structured knowledge extraction: summary for machine

### Relation Extraction

CHICAGO (AP) — Citing high fuel prices, United Airlines said Friday it has increased fares by \$6 per round trip on flights to some cities also served by lower-cost carriers. American Airlines, a unit of AMR, immediately matched the move, spokesman Tim Wagner said.

United, a unit of UAL, said the increase took effect Thursday night and applies to most routes where it competes against discount carriers, such as Chicago to Dallas and Atlanta and Denver to San Francisco, Los Angeles and New York.

Subject	Relation	Object
American Airlines	subsidiary	AMR
Tim Wagner	employee	American Airlines
United Airlines	subsidiary	UAL

# Relation types

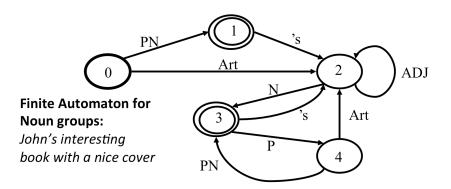
### For generic news text ...

Relations		Examples	Types
Affiliations			
	Personal	married to, mother of	$\mathtt{PER} \to \mathtt{PER}$
	Organizational Artifactual	spokesman for, president of owns, invented, produces	$\begin{array}{c} PER \to ORG \\ (PER \mid ORG) \to ART \end{array}$
Geospatial		•	
_	Proximity	near, on outskirts	$LOC \to LOC$
	Directional	southeast of	$\mathtt{LOC} \to \mathtt{LOC}$
Part-Of			
	Organizational	a unit of, parent of	$ORG \to ORG$
	Political	annexed, acquired	$\mathtt{GPE} \to \mathtt{GPE}$

# Relation extraction: 5 easy methods

- Hand-built patterns
- Bootstrapping methods
- Supervised methods
- Distant supervision
- Unsupervised methods

## Hand-written Information Extraction: use regex



Determining which person holds what position in what organization

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[person], [position] of [org]

Vuk Draskovic, leader of the Serbian Renewal Movement

Determining which person holds what position in what organization

[person], [position] of [org]

Vuk Draskovic, leader of the Serbian Renewal Movement

[org] (named, appointed, etc.) [person] Prep [office]

NATO appointed Wesley Clark as Commander in Chief

Determining where an organization is located

Determining where an organization is located

[org] in [loc]

NATO headquarters in Brussels

Determining where an organization is located

[org] in [loc]

NATO headquarters in Brussels

[org] [loc] (division, branch, headquarters, etc.)

KFOR Kosovo headquarters

# Patterns for learning hyponyms

### Intuition from Hearst (1992)

Agar is a substance prepared from a mixture of red algae, such as Gelidium, for laboratory or industrial use.

What is Gelidium?

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- How do you know?

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# Hearst's lexico-syntactic patterns

### Automatic Acquisition of Hyponyms

- Y such as X((,X)\*(, and/or) X)
- such Y as X
- X or other Y
- X and other Y
- Y including X
- Y, especially X

# Examples of Hearst patterns

Hearst pattern	Example occurrences	
X and other Ytemples, treasuries, and other important civic building		
X or other Y	bruises, wounds, broken bones or other injuries	
Y such as X	The bow lute, such as the Bambara ndang	
such Y as X	such authors as Herrick, Goldsmith, and Shakespeare.	
Y including X	common-law countries, including Canada and England	
Y, especially X	European countries, especially France, England, and Spain	

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#### Berland and Charniak's patterns

 Selected initial patterns by finding all sentences in a corpus containing basement and building

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```
whole NN[-PL] 's POS part NN[-PL]
part NN[-PL] of PREP {the|a} DET mods [JJ|NN]* whole NN
part NN in PREP {the|a} DET mods [JJ|NN]* whole NN
parts NN-PL of PREP wholes NN-PL
parts NN-PL in PREP wholes NN-PL
```

```
... building's basement ...
... basement of a building ...
... basement in a building ...
... basements of buildings ...
... basements in buildings ...
```

## Problems with hand-built patterns

- Requires hand-building patterns for each relation!
  - hard to write; hard to maintain
  - there are zillions of them
  - domain-dependent
- Don't want to do this for all possible relations!
- Plus, we'd like better accuracy
  - ► Hearst: 66% accuracy on hyponym extraction
  - Berland and Charniak: 55% accuracy on meronyms