

Natural Language Processing
SoSe 2014



Relation Extraction

Dr. Mariana Neves

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(based on the slides of Dr. Saeedeh Momtazi)

Outline

- Introduction
- Task
- Pattern Extraction
- Supervised Learning
- Semi-supervised Learning

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Information Extraction

- Named entity recognition
- Relation Extraction

Named Entity Recognition

- HPI is affiliated to the Potsdam University and located in Potsdam near Berlin. It was founded in 1998 by Hasso Plattner, one of the co-founders of the European software company, SAP AG.
 - HPI (ORG)
 - Potsdam University (ORG)
 - Potsdam (LOC)
 - Berlin (LOC)
 - 1998 (DATE)
 - Hasso Plattner (PER)
 - SAP AG (ORG)

Relation Extraction

- HPI is affiliated to the Potsdam University and located in Potsdam near Berlin. It was founded in 1998 by Hasso Plattner, one of the co-founders of the European software company, SAP AG.
 - HPI – Potsdam: located (ORG-LOC)
 - HPI – Berlin: near (ORG-LOC)
 - Potsdam – Berlin: near (LOC-LOC)
 - HPI – 1998: founded (ORG-DATE)
 - HPI - Hasso Plattner: founder (ORG-PER)
 - SAP AG - Hasso Plattner: co-founder (ORG-PER)

Information Extraction

Potsdam



View of the rebuilt Potsdam City Palace with St. Nicholas' Church in the background



Flag



Coat of arms



Coordinates:  52°24'N 13°4'E

Country	Germany
State	Brandenburg
District	Urban district
Government	
• Lord Mayor	Jann Jakobs (SPD)
Area	
• Total	187.28 km ² (72.31 sq mi)
Elevation	32 m (105 ft)
Population (2012-12-31) ^[1]	
• Total	159,456
• Density	850/km ² (2,200/sq mi)
Time zone	CET/CEST (UTC+1/+2)
Postal codes	14467–14482
Dialling codes	0331
Vehicle registration	P
Website	www.potsdam.de 

Hasso Plattner Institute
 Hasso-Plattner-Institut für
 Softwaresystemtechnik GmbH


Hasso
Plattner
Institut

IT Systems Engineering | Universität Potsdam

Motto	Design IT. Create Knowledge.
Established	1998 ^[1]
Type	Private university institute
Director	Prof. Dr. Christoph Meinel
Admin. staff	60 ^[2]
Students	about 480 ^[2]
Location	Potsdam, Germany
Campus	Griebnitzsee
Colors	 Orange  Vivid orange  Dark pink
Affiliations	University of Potsdam
Website	www.hpi.uni-potsdam.de 

Motivation

- Creating new structured data sources (knowledge bases)
 - DBpedia
 - Freebase
 - Yago
- Answering complex questions using multiple sources
 - Which soccer player married a Spice Girls star?
 - ("?x" is-a "soccer player")
 - ("?x" married "?y")
 - ("?y" member "Spice Girls")

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Relation Representation

- Representing data as triples
 - (Argument1 RelationType Argument2)
 - (Subject Predicate Object)

- Resource Description Framework (RDF)

Relation Types

- Having various relation types based on the type of arguments
 - PER-PER: Spouse, Parent, Child, Friendship, Colleague, ...
 - PER-LOC: Place of birth, Lives in, Place of death, Buried in, ...
 - PER-ORG: Founder, Co-founder, Owner, Employee, Student/Alum, Professor, ...
 - ORG-LOC: Located, Near, Founded-location, Headquarter, ...
 - PER-DATE: Date of Birth, Date of Marriage, Date of Death, ...

Approaches

- Manually created patterns
- Supervised machine learning
- Semi-supervised learning

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Pattern Extraction

- What are the potential words to express a relation type?
 - (PER Member ORG)
 - ("?x" Member "?y")

 - x is a member of y.
 - x is an employee of y.
 - x works at y.
 - x is a staff of y.

 - x is (a|an) (member|employee|staff) of y.
 - x (works) at y.

Pattern Extraction

- Advantages
 - Having high precision results
- Disadvantages
 - Having low recall
 - Finding all possible patterns is labor intensive
 - Covering all relations is very difficult

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- Semi-supervised Learning

Supervised Classification

- Training data
 - Defining a fix set of relation types
 - Choosing the corresponding named entities
 - Selecting a set of texts as training data
 - Recognizing the named entities in the text
 - Labeling the relations between named entities manually

Task

- Input
 - A pair of entities (NER)
 - A context in which this pair appears
 - Possible relation types
- Output
 - Type of relation between two entities, if there exist any
- „Thomas Edison died on October 18, 1931, in New Jersey due to complications of diabetes.“
 - PER-LOC (Thomas Edison, New Jersey)
 - Place of birth, **Place of death**, Buried in

Feature Selection

- „Thomas Edison died on October 18, 1931, in New Jersey due to complications of diabetes.“
- The target entities
 - T1: Thomas Edison
 - T2: New Jersey
- Surrounding words of target entities
 - $T1_{+1}$: died
 - $T2_{-1}$: in
 - $T2_{+1}$: due
- All words between the target entities (bag-of-word)
 - 1931 October died 18 , on , in

Feature Selection

- The named entity label of the target words
 - NE(T1): PER
 - NE(T2): LOC
- The syntactic structure of the sentence
 - Shortest dependency path
 - nsubj-prep-pobj

```

1 ➔ nn(Edison-2, Thomas-1)
   nsubj(died-3, Edison-2)
   root(ROOT-0, died-3)
   prep(died-3, on-4)
   pobj(on-4, October-5)
   num(October-5, 18-6)
   num(October-5, 1931-8)
2 ➔ prep(died-3, in-10)
   nn(Jersey-12, New-11)
3 ➔ pobj(in-10, Jersey-12)
   amod(Jersey-12, due-13)
   prep(due-13, to-14)
   pobj(to-14, complications-15)
   prep(complications-15, of-16)
   pobj(of-16, diabetes-17)

```

<http://nlp.stanford.edu:8080/parser/index.jsp>

Classification Algorithm

- Applying any of the classifiers
 - K Nearest Neighbor
 - Support Vector Machines
 - Naïve Bayes
 - Maximum Entropy
 - Logistic Regression
 - ...

Supervised Classification

- Advantages
 - Very good performance if
 - Having enough training data
 - Having test data similar to training data
- Disadvantages
 - Manual labeling of training data is labor expensive
 - Difficult to get good results for other domains

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- **Semi-supervised Learning**

Semi-supervised Learning

- Having no large training data
 - but a large collection of documents
- Producing a small training data (seed data)
 - A set of triples
- Bootstrapping
 - Using the seed data to find further entity pairs with the same relation

Bootstrapping

- Using the collected seed data
- Finding sentences which contain at least one entity pair
- Extracting the common contexts of the pair
- Creating patterns (or models) from the extracted context
- Using the pattern (or model) to get more pairs and add them to seed data

Bootstrapping

- Using the collected seed data
 - (Thomas Edison Spouse Mina Mille)

Bootstrapping

- Finding sentences which contain at least one entity pair (normalized entities)
- Thomas Edison married Mina Mille.
- Edison married a young woman named Mina Mille.
- In 1871, Thomas Edison married Mina Mille.
- Thomas Edison marries Mina Mille on December 25.

Bootstrapping

- Extracting the common contexts of the pair
- Creating patterns (or models) from the extracted context
- Thomas Edison married Mina Mille.
- Edison married a young woman named Mina Mille.
- In 1871, Thomas Edison married Mina Mille.
- Thomas Edison marries Mina Mille on December 25.

Bootstrapping

- Using the pattern (or model) to get more pairs and add them to seed data
 - (Albert Einstein Spouse “?”)
- Einstein marries his cousin Elsa Löwenthal on June 2.
- Einstein married Elsa Löwenthal in Berlin.
- Einstein married Elsa Löwenthal on 2 June 1919.
- After their divorce in 1919, Einstein married Elsa Löwenthal in the same year.
- Albert Einstein was married to Elsa Löwenthal for 17 years.
- Einstein marries Elsa Löwenthal.
- In the same year Albert Einstein married Elsa Löwenthal.

⇒ (Albert Einstein Spouse Elsa Löwenthal)

Bootstrapping

- Using the collected seed data
 - (Thomas Edison Spouse Mina Mille)
 - (Albert Eistein Spouse Elsa Löwenthal)

Bootstrapping

- Using the collected seed data
- Finding sentences which contain at least one entity pairs
- Extracting the common contexts of the pair
- Creating patterns (or models) from the extracted context

- **Albert Einstein's wife, Elsa Löwenthal**, was his first cousin.
- **Elsa Löwenthal** was the **wife** of **Albert Einstein**.
- **Einstein's wife** was named **Elsa Löwenthal**.

Further Reading

- Speech and Language Processing
 - Chapters 22.2

