

# Information Extraction

Lecture 10 – Ontological and Open IE

CIS, LMU München

Winter Semester 2015-2016

Dr. Alexander Fraser, CIS

# Administravia

- Suggested Klausur date is in the last week of the Vorlesung (the week before Fasching)
  - Klausur: February 3rd
  - There will be a review for the Klausur on Wed January 27th
- NEW: there is a conflict with a different course, I will look into this

- Before I start on Ontological IE, two topics I wanted to briefly talk about today:
  - Semantic Role Labeling
  - Wikification

# Syntactic Parsing and Relation Extraction

- We saw in the previous two lectures that syntactic features are useful for relation extraction (and event extraction)
- For instance...

# Parse Features for Relation Extraction

*American Airlines*, a unit of AMR, immediately matched the move, spokesman *Tim Wagner* said

Mention 1

Mention 2

- Base syntactic chunk sequence from one to the other

NP NP PP VP NP NP

- Constituent path through the tree from one to the other

NP ↑ NP ↑ S ↑ S ↓ NP

- Dependency path

Airlines matched Wagner said

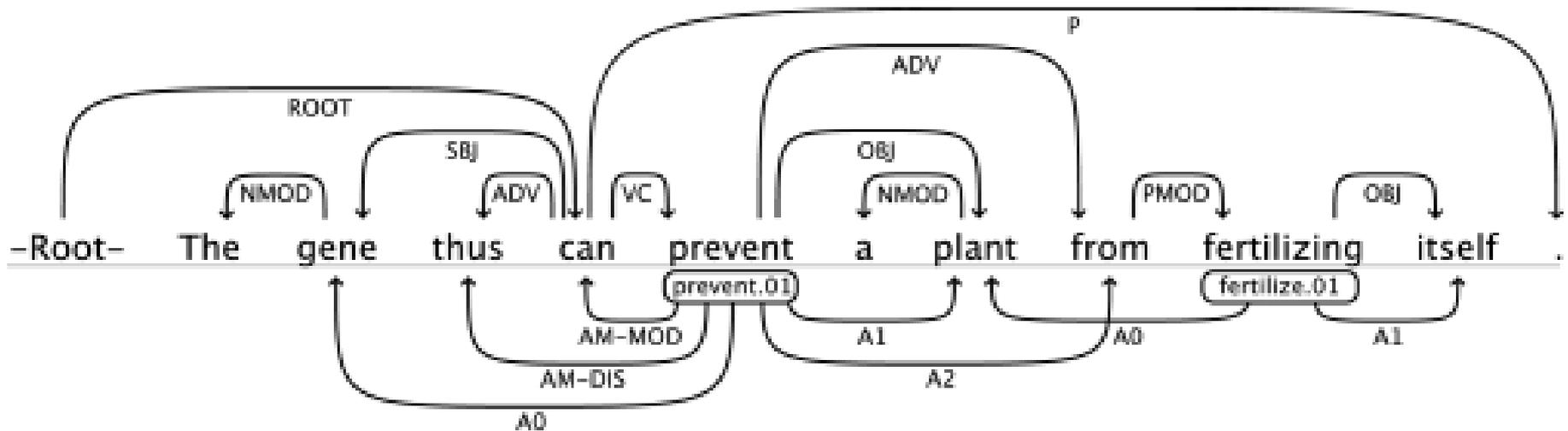
# Semantic Role Labeling

- A generalization beyond syntactic parsing is Semantic Role Labeling (often abbreviated to SRL)
- Here the idea is to identify the arguments to a verb
  - So this can capture the same information as, e.g., a dependency parse
  - It should be clear that this will be useful in IE
- But the difference is that the arguments are captured in terms of their semantic function rather than their syntactic function

# Subcategorization Frame

- Consider the sentences:
  - The man was bitten by the dog
  - The dog bit the man
- In terms of the verb and the subcategorized arguments, there is no difference here
- In Semantic Role Labeling, these will have the same representation!
- Consider also:
  - The man was bitten.

# Semantic Role Labeling



Example from Kozhevnikov and Titov

List of SRL tools (see also the comments):

<http://www.kenvanharen.com/2012/11/comparison-of-semantic-role-labelers.html>

# Last Word: Training Data

- The critical problem for statistical approaches is labeled training data
- There are two mature data sets for training semantic role labelers for English
  - **Framenet** is the one that may be more useful for many IE purposes (but **Propbank** is also interesting)
- There has been some work on projecting these two resources to other languages using machine translation techniques
  - E.g., for German, the "Salsa" project at Uni SB

# Wikification

- Wikification is the problem of automatically annotating entities in free text with their (English) Wikipedia page
- Let's start with motivation...

# Wikification: The Reference Problem

Blumenthal (D) is a candidate for the U.S. Senate seat now held by Christopher Dodd (D), and he has held a commanding lead in the race since he entered it. But the Times report has the potential to fundamentally reshape the contest in the Nutmeg State.



Richard Blumenthal

From Wikipedia, the free encyclopedia

Democratic Party (United States)

From Wikipedia, the free encyclopedia

United States Senate

From Wikipedia, the free encyclopedia

[Blumenthal](#) ([D](#)) is a candidate for the [U.S. Senate](#) seat now held by [Christopher Dodd](#) ([D](#)), and he has held a commanding lead in the race since he entered it. But the [Times](#) report has the potential to fundamentally reshape the contest in [the Nutmeg State](#).

Chris Dodd

From Wikipedia, the free encyclopedia

*The New York Times*

From Wikipedia, the free encyclopedia

Connecticut

From Wikipedia, the free encyclopedia

# Wikification: Motivation

- Dealing with **Ambiguity** of Natural Language
  - Mentions of entities and concepts could have multiple meanings
- Dealing with **Variability** of Natural Language
  - A given concept could be expressed in many ways
- **Wikification** addresses these two issues in a specific way:
- The Reference Problem
  - What is meant by this concept? (WSD + Grounding)
  - More than just co-reference (within and across documents)

# Ontological IE

- In the last two lectures, we discussed how to extract relations and events from text
  - We looked in detail at relations expressed in a single sentence
  - Event extraction captures relations which are often expressed at either the sentence or at the document level (i.e., in multiple sentences)
    - Consider the CMU Seminar task – the task is to extract events (seminars), with speaker, location, start time and end time
- Today we will discuss updating a knowledge base with the extracted relations or events
  - This is called "Ontological IE"

# Ontologies

An **ontology** is a consistent knowledge base without redundancy

Person	Nationality
Angela Merkel	German
Merkel	Germany
A. Merkel	French



Entity	Relation	Entity
Angela Merkel	citizenOf	Germany



- Every entity appears only with exactly the same name
- There are no semantic contradictions

# Ontological IE

**Ontological Information Extraction** (IE) aims to create or extend an ontology.

Entity	Relation	Entity
Angela Merkel	citizenOf	Germany



Angela Merkel is the German chancellor....  
...Merkel was born in Germany...

...A. Merkel has French nationality...

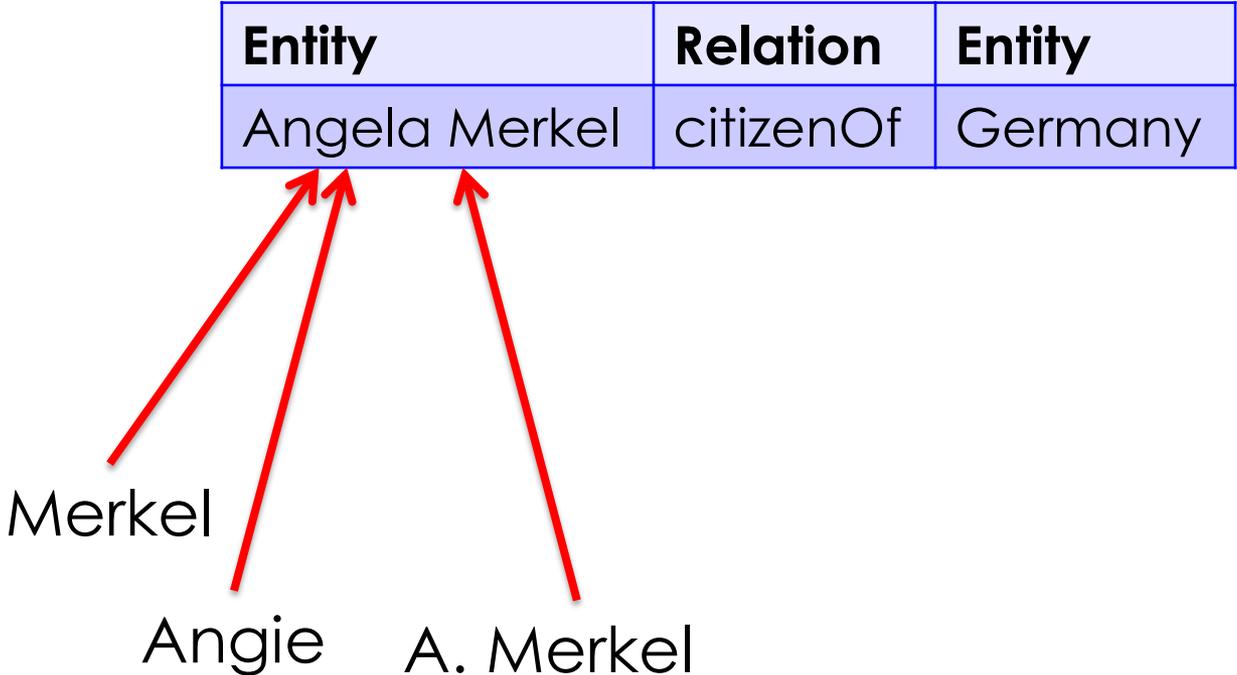


<del>Person</del>	<del>Nationality</del>
<del>Angela Merkel</del>	<del>German</del>
<del>Merkel</del>	<del>Germany</del>
<del>A. Merkel</del>	<del>French</del>

# Ontological IE Challenges

## Challenge 1:

Map names to names that are already known



# Ontological IE Challenges

## Challenge 2:

Be sure to map the names to the right known names

Entity	Relation	Entity
Angela Merkel	citizenOf	Germany
Una Merkel	citizenOf	USA



?

Merkel is great!

# Ontological IE Challenges

## Challenge 3:

Map to known relationships

Entity	Relation	Entity
Angela Merkel	citizenOf	Germany



... has nationality ...  
... has citizenship ...  
... is citizen of ...

# Ontological IE Challenges

Challenge 4:

Take care of consistency

Entity	Relation	Entity
Angela Merkel	citizenOf	Germany



Angela Merkel is  
French...

# Triples

A **triple** (in the sense of ontologies) is a tuple of an entity, a relation name and another entity:

Entity	Relation	Entity
Angela Merkel	citizenOf	Germany

Most ontological IE approaches produce triples as output. This decreases the variance in schema.

Person	Country
Angela	Germany

Citizen	Nationality
Angela	Germany

Person	Birthdate	Country
Angela	1980	Germany

# Triples

A triple can be represented in multiple forms:

Entity	Relation	Entity
Angela Merkel	citizenOf	Germany

=

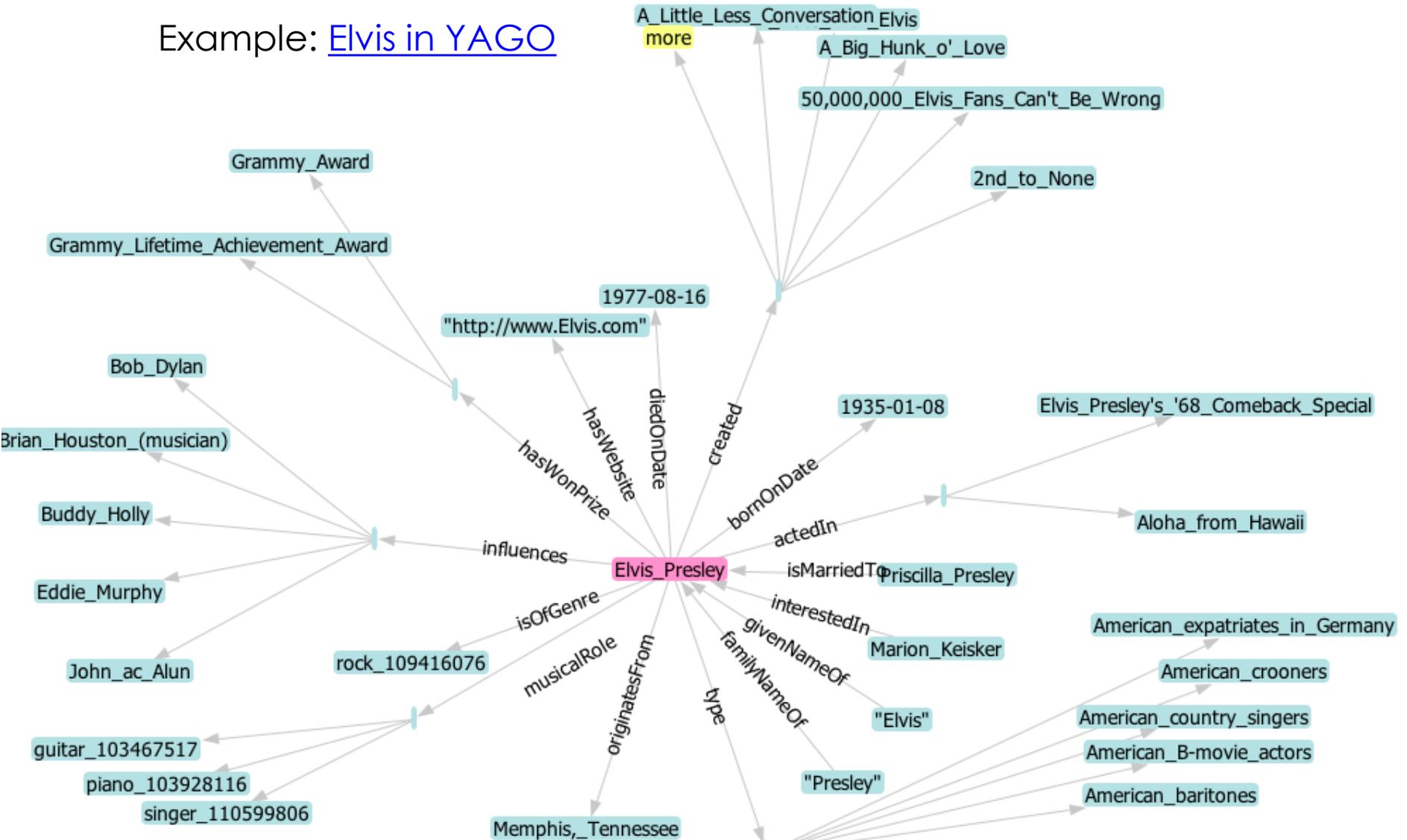


=

<Angela Merkel, citizenOf, Germany>

# YAGO

Example: [Elvis in YAGO](#)



- Let's talk about ontological IE using extraction from Wikipedia as an example
- Then we will go on to open IE, which uses similar ideas to extract from all the text on the web!

# Wikipedia



- Wikipedia is a free online encyclopedia
- 3.4 million articles in English
  - 16 million articles in dozens of languages

Why is Wikipedia good for information extraction?

- It is a huge, but homogenous resource  
(more homogenous than the Web)
- It is considered authoritative  
(more authoritative than a random Web page)
- It is well-structured with infoboxes and categories
- It provides a wealth of meta information  
(inter article links, inter language links, user discussion,...)

# Ontological IE from Wikipedia



- Wikipedia is a free online encyclopedia
- 3.4 million articles in English
  - 16 million articles in dozens of languages

Every article is (should be) unique  
=> We get a set of unique entities  
that cover numerous areas of interest



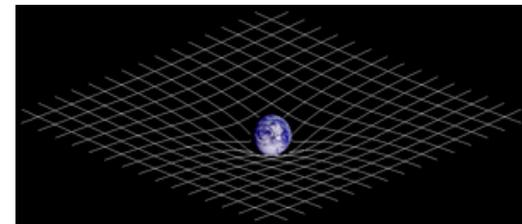
Angela\_Merkel



Una\_Merkel



Germany



Theory\_of\_Relativity

# Wikipedia Source

Example: [Elvis on Wikipedia](#)

## Background information

<b>Birth name</b>	Elvis Aaron Presley
<b>Born</b>	January 8, 1935 <a href="#">Tupelo, Mississippi</a> , United States
<b>Died</b>	August 16, 1977 (aged 42) <a href="#">Memphis, Tennessee</a> , United States
<b>Genres</b>	<a href="#">Rock and roll</a> , <a href="#">pop</a> , <a href="#">rockabilly</a> , <a href="#">country</a> , <a href="#">blues</a> , <a href="#">gospel</a> , <a href="#">R&amp;B</a>
<b>Occupations</b>	Musician, actor
<b>Instruments</b>	Vocals, guitar, piano
<b>Years active</b>	1954–77
<b>Labels</b>	<a href="#">Sun</a> , <a href="#">RCA Victor</a>
<b>Associated acts</b>	<a href="#">The Blue Moon Boys</a> , <a href="#">The Jordanaires</a> , <a href="#">The Imperials</a>
<b>Website</b>	<a href="http://www.elvis.com">www.elvis.com</a> 

```
| Birth_name = Elvis Aaron Presley  
| Born = {{Birth date | 1935 | 1 | 8}}<br />  
[[Tupelo, Mississippi | Tupelo]]
```

# IE from Wikipedia

bornOnDate = 1935  
(hello regexes!)

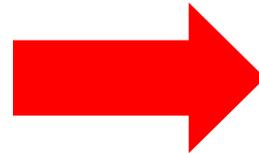


Elvis Presley

Blah blah blub  
fasel (do not  
read this, better  
listen to the talk)  
blah blah Elvis  
blub (you are still  
reading this) blah  
Elvis blah blub  
later became  
astronaut blah

Categories: Rock singers

~Infobox~  
Born: 1935  
...



born

1935

Exploit Infoboxes

# IE from Wikipedia

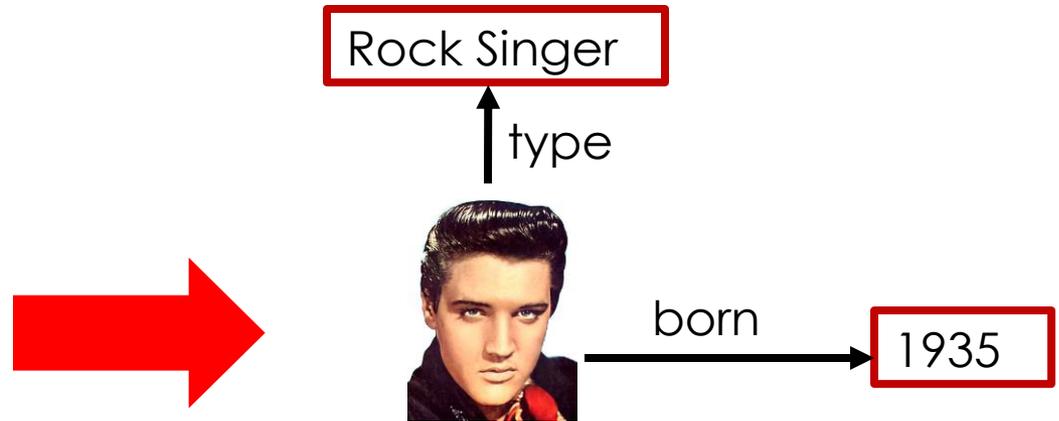


Elvis Presley

Blah blah blub  
fasel (do not  
read this, better  
listen to the talk)  
blah blah Elvis  
blub (you are still  
reading this) blah  
Elvis blah blub  
later became  
astronaut blah

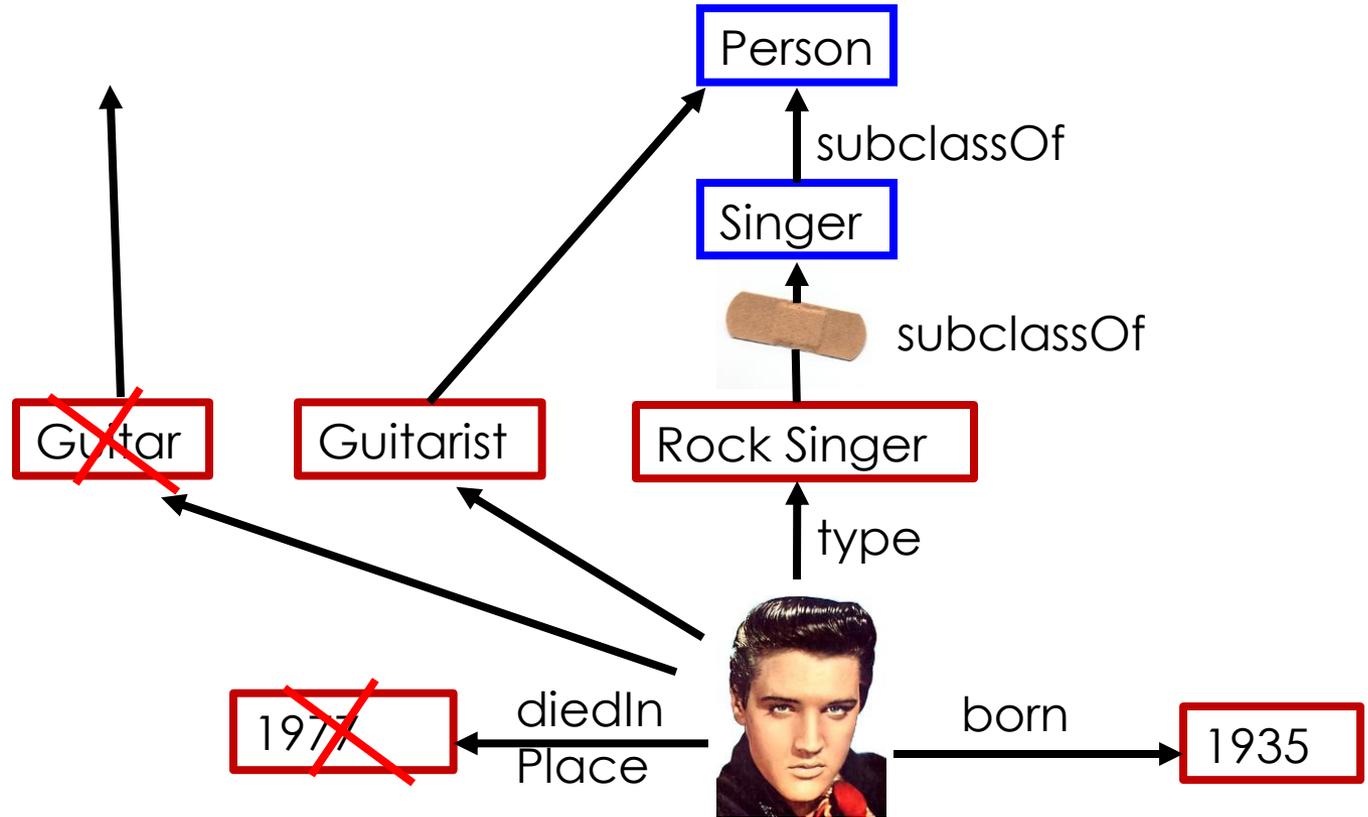
~Infobox~  
Born: 1935  
...

Categories: Rock singers



Exploit Infoboxes  
Exploit conceptual categories

# Consistency Checks



- Check uniqueness of functional arguments
- Check domains and ranges of relations
- Check type coherence

# Ontological IE from Wikipedia



## YAGO

- 3m entities, 28m facts
- focus on precision 95%  
(automatic checking of facts)

<http://yago-knowledge.org>



## DBpedia

- 3.4m entities
- 1b facts (also from non-English Wikipedia)
- large community

<http://dbpedia.org>

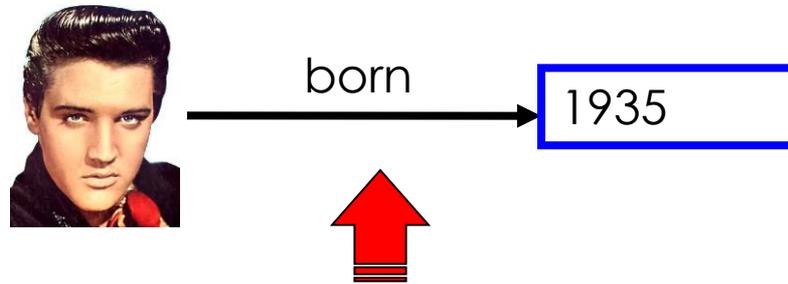


Community project on top of Wikipedia  
(bought by Google, but still open)

<http://freebase.com>

--- Now integrated into **Wikidata!!!**

# Ontological IE by Reasoning



Elvis was born in 1935

Recap: The challenges:

- deliver canonic relations
- deliver canonic entities
- deliver consistent facts

died in, was killed in

Elvis, Elvis Presley, The King

born (Elvis, 1970)  
born (Elvis, 1935)

Idea: These problems are interleaved,  
solve all of them together.

# Using Reasoning

Ontology



First Order Logic

```
type(Elvis_Presley,singer)
subclassof(singer,person)
...

appears("Elvis","was born in",
"1935")

...
means("Elvis",Elvis_Presley,0.8)
means("Elvis",Elvis_Costello,0.2)
...

born(X,Y) & died(X,Z) => Y<Z
appears(A,P,B) & R(A,B)
=> expresses(P,R)
appears(A,P,B) & expresses(P,R)
=> R(A,B)
...
```

Documents

Elvis was born in 1935

Consistency

Rules

birthdate<deathdate



born

1935

SOFIE  
system

# Ontological IE by Reasoning

Reasoning-based approaches use logical rules to extract knowledge from natural language documents.

Current approaches use either

- Weighted MAX SAT
- or Datalog
- or Markov Logic

Input:

- often an ontology
- manually designed rules

Condition:

- homogeneous corpus helps

# Ontological IE Summary

**Ontological Information Extraction** (IE) tries to create or extend an ontology through information extraction.



Current hot approaches:

- extraction from Wikipedia
- reasoning-based approaches
- integrating uncertainty

# Open Information Extraction

## **Open Information Extraction/Machine Reading**

aims at information extraction from the entire Web.

Vision of Open Information Extraction:

- the system runs perpetually, constantly gathering new information
- the system creates meaning on its own from the gathered data
- the system learns and becomes more intelligent, i.e. better at gathering information

# Open Information Extraction

## **Open Information Extraction/Machine Reading**

aims at information extraction from the entire Web.

Rationale for Open Information Extraction:

- We do not need to care for every single sentence, but just for the ones we understand
- The size of the Web generates redundancy
- The size of the Web can generate synergies

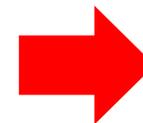
# KnowItAll & Co

KnowItAll, KnowItNow and TextRunner are projects at the University of Washington (in Seattle, WA).

*gyptian  
mplex.* more than the question of how the  
Egyptians built the pyramids was,  
he says, "how the pyramids built



Subject	Verb	Object	Count
Egyptians	built	pyramids	400
Americans	built	pyramids	20
...	...	...	...



Valuable  
common sense  
knowledge  
(if filtered)

# KnowItAll & Co

TextRunner took .80 seconds.

Retrieved **391** results for Predicate containing "**built**" and Argument 2 containing "**pyramids**"

Grouping results by predicate. Group by: [argument 2](#) | [argument 1](#)

**built** - 159 results

Egyptians (297), aliens (71), Pharaohs (40), **85 more...** **built** the **pyramids**

Egyptians (26), Khufu (18), Maya (9), **30 more...** **built** the Great **Pyramid**

Imhotep (8), Pharaoh Zoser (4), Egyptians (2), King Djoser (2) **built** the Step **Pyramid**

two symbols of life (4), 6th dynasty kings (3), King Sneferu (3), Snefru (3) **built** two large **Pyramids**

Egyptians (8) **built** the Great **Pyramids**

ancient Egyptians (6) **built** more than 90 royal **pyramids**

colonial silver city of Taxco (3), Explore (2) **built** the gigantic **pyramids** of the Sun

Central America (2), part of Mexico (2) **built** great cities , temples and **pyramids**

# Read the Web

“Read the Web” is a project at the Carnegie Mellon University in Pittsburgh, PA.

Initial Ontology

Table Extractor

Krzewski	Blue Angels
Miller	Red Angels

Natural Language  
Pattern Extractor

Krzewski coaches  
the Blue Devils.

Mutual exclusion

sports coach != scientist

Type Check

If I coach, am I a coach?

<http://rtw.ml.cmu.edu/rtw/>

# Open IE: Read the Web

NELL Know

CMU Read the Web

- arthropod (100.0%)

- Seed

- CPL @156 (100.0%) on 30-sep-2010 [ "hind wings of \_ "invertebrates , such as \_ "  
" \_ swarm from" "other insects , including \_ " \_ marching home" "honeydew produce  
like \_ " "other insects , such as \_ " \_ do not eat wood" "many legs as \_ " \_ produce s  
have complete metamorphosis" "I do n't see anymore \_ " "ants , so \_ " "insecticide fo  
"such insects as \_ " \_ are the only insects" "red imported \_ " "insects like \_ " "social in  
, such as \_ " "arthropods include \_ " "insect pests including \_ " "meaty foods like \_ " \_  
pests , such as \_ " "other insects such as \_ " "insects , in particular \_ " \_ release a ph  
like \_ " "many insects , including \_ " \_ are social insects" "insect pests such as \_ " \_ a  
pests , including \_ " "arthropods , including \_ " \_ are beneficial insects" " \_ are comm  
"arthropods , such as \_ " ]

- SEAL @151 (50.0%) on 26-sep-2010 [ 1 ]

kateretes (Seed)

mosquito (Seed)

peppered moth (Seed)

sap beetle (Seed)

tettigoniidae (Seed)

triatoma protracta (Seed)

honeylocust spider mite

grape flea beetle

blueberry leaf beetle

sugarcane moth borer

psychoda moth flies

bagworm moth

carpenterworm moths

leafcurl plum aphid

merchant grain beetle

- fung
- plan
- arch
- bact
- politica
- color
- language
- programminglanguage
- dateliteral
- gamescore
- nonnegativeinteger
- politicsissue
- llcoordinate
- agent
  - animal
    - invertebrate
      - arthropod
        - arachnid
        - insect
        - crustacean
      - mollusk
    - vertebrate
      - amphibian
      - bird
      - fish

<http://rtw.ml.cmu.edu/rtw/>

# Open Information Extraction

## **Open Information Extraction/Machine Reading**

aims at information extraction from the entire Web.

Main hot projects

- TextRunner (University of Washington)
- Read the Web (Carnegie Mellon)
- Prospera/SOFIE (Max-Planck Informatics Saarbrücken)

Input

- The Web
- Read the Web: Manual rules
- Read the Web: initial ontology

Conditions

- none

- Slide sources
  - Many of the slides today on Ontological IE and Open IE are from Fabian Suchanek (Télécom ParisTech)
  - See the web page I mentioned for a list of semantic role labelers
  - Some of the Wikification slides are from Dan Roth's tutorial, this is highly recommended

- Thank you for your attention!