Information Extraction (IE)

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CS4705
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Information Extraction (IE) -- Task

 Idea: 'extract' or tag particular types of information from arbitrary text or transcribed speech

Named Entity Tagger

- Identify types and boundaries of named entity
- For example:
 - Alexander Mackenzie , (January 28, 1822 April 17, 1892), a building contractor and writer, was the second Prime Minister of Canada from
 - → <PERSON>Alexander Mackenzie</PERSON> , (<TIMEX >January 28, 1822 <TIMEX> <TIMEX>April 17, 1892</TIMEX>), a building contractor and writer, was the second Prime Minister of <GPE>Canada</GPE> from

IE for Template Filling Relation Detection

- Given a set of documents and a domain of interest, fill a table of required fields.
- For example:
 - Number of car accidents per vehicle type and number of casualty in the accidents.

Vehicle Type	# accidents	# casualties	Weather
SUV	1200	190	Rainy
Trucks	200	20	Sunny

IE for Question Answering

- Q: When was Gandhi born?
- A: October 2, 1869
- Q: Where was Bill Clinton educated?
- A: Georgetown University in Washington, D.C.
- Q: What was the education of Yassir Arafat?
- A: Civil Engineering
- Q: What is the religion of Noam Chomsky?
- A: Jewish

Approaches

- 1. Statistical Sequence Labeling
- 2. Supervised
- 3. Semi-Supervised and Bootstrapping

Approach for NER

- <PERSON>Alexander Mackenzie</PERSON> , (<TIMEX >January 28, 1822 <TIMEX> -</ti>
 <TIMEX>April 17, 1892</TIMEX>), a building contractor and writer, was the second Prime Minister of <GPE>Canada</GPE> from
- Statistical sequence-labeling techniques approach can be used – similar to POS tagging.
 - Word-by-word sequence labeling
 - Example of Features:
 - POS tags
 - Syntactic constituents
 - Shape features
 - Presence in a named entity list

Supervised Approach for relation detection

- Given a corpus of annotated relations between entities, train two classifiers:
 - 1. A binary classifier:
 - Given a span of text and two entities
 - Decide if there is a relationship between these two entities.
 - 2. A classifier is trained to determine the types of relations exist between the entities

Features:

- Types of two named entities
- Bag-of-words
- **–** ...

• Example:

- A rented SUV went out of control on Sunday, causing the death of seven people in Brooklyn
- Relation: Type=Accident, Vehicle Type= SUV, causality = 7, weather = ?
- Pros and Cons?

Pattern Matching for Relation Detection

Patterns:

- "[CAR TYPE] went out of control on [TIMEX], causing the death of [NUM] people"
- "[PERSON] was born in [GPE]"
- "[PERSON] was graduated from [FAC]"
- "[PERSON] was killed by <X>"

Matching Techniques

- Exact matching
 - Pros and Cons?
- Flexible matching (e.g., [X] was .* killed .* by [Y])
 - Pros and Cons?

Pattern Matching

- How can we come up with these patterns?
- Manually?
 - Task and domain specific -- tedious, time consuming, and not scalable.

Semi-Supervised Approach

AutoSlog-TS (Riloff, 1996)

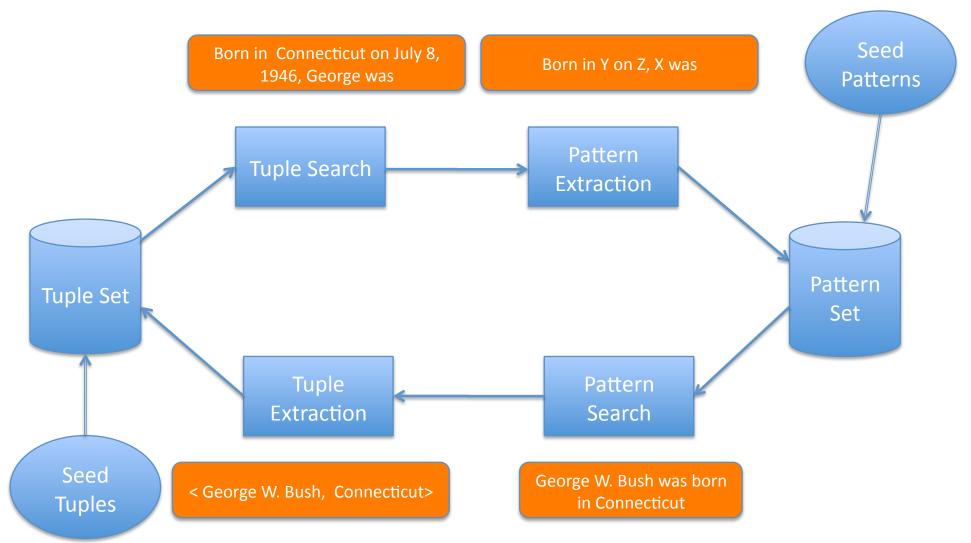
- MUC-4 task: extract information about terrorist events in Latin America.
- Two corpora:
 - 1) Domain-dependent corpus that contains relevant information
 - 2) A set of irrelevant documents
- Algorithm:
 - 1. Using some heuristic rules, all patterns are extracted from both corpora. For example:

Rule: <Subj> passive-verb

- <Subj> was murdered
- → <Subj> was called
- 2. Pattern Ranking: The output patterns are then ranked by frequency of their occurrences in corpus1 / corpus2.
- 3. Filter out the patterns by hand

Bootstrapping

X was born in Y



TASK 12: (DARPA – GALE year 2) PRODUCE A BIOGRAPHY OF [PERON].

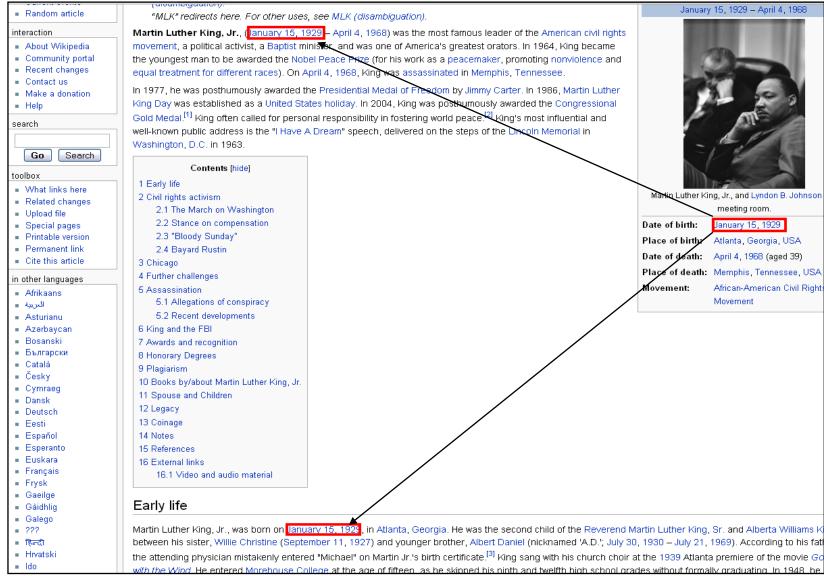
- 1. Name(s), aliases:
- *Date of Birth or Current Age:
- 3. *Date of Death:
- 4. *Place of Birth:
- 5. *Place of Death:
- 6. Cause of Death:
- 7. Religion (Affiliations):
- 8. Known locations and dates:
- 9. Last known address:
- 10. Previous domiciles:
- 11. Ethnic or tribal affiliations:
- 12. Immediate family members
- 13. Native Language spoken:
- 14. Secondary Languages spoken:
- 15. Physical Characteristics
- 16. Passport number and country of issue:
- 17. Professional positions:
- 18. Education
- 19. Party or other organization affiliations:
- 20. Publications (titles and dates):

Biography – two approaches

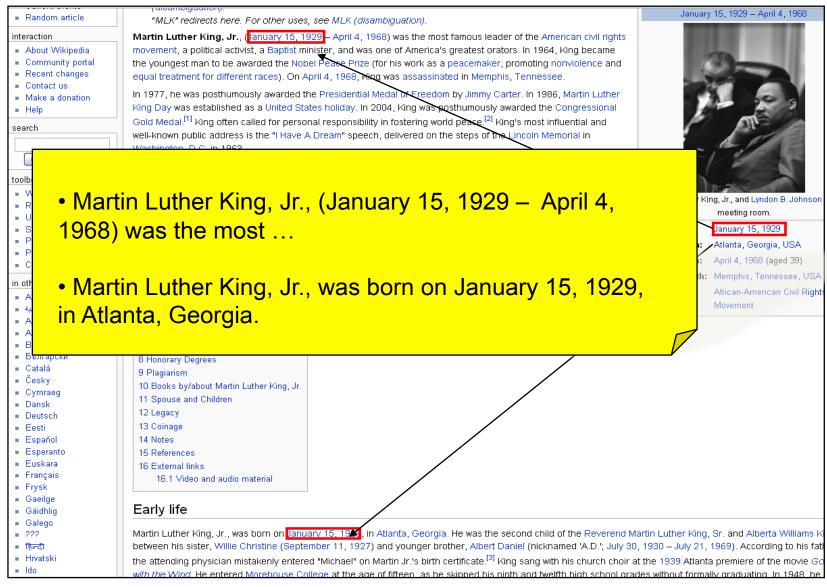
 To obtain high precision, we handle each slot independently using bootstrapping to learn IE patterns.

• To improve the recall, we utilize a biographical-sentence classifier.

Biography patterns from Wikipedia



Biography patterns from Wikipedia



Run NER on these sentences

- <Person> Martin Luther King, Jr. </Person>, (<Date>January 15, 1929</Date> <Date> April 4, 1968</Date>) was the most...
- <Person> Martin Luther King, Jr. </Person>, was born on <Date>
 January 15, 1929 </Date>, in <GPE> Atlanta, Georgia </GPE>.
- Take the token sequence that includes the tags of interest + some context (2 tokens before and 2 tokens after)

Convert to Patterns:

- <Target_Person> (<Target_Date> <Date>) was the
- <Target_Person> , was born on <Target_Date>, in
- Remove more specific patterns if there is a pattern that contains other, take the smallest > k tokens.
- Target_Person>, was born on <Target_Date>
- Target_Person> (<Target_Date> <Date>)
- Finally, verify the patterns manually to remove irrelevant patterns.

Examples of Patterns:

• 502 distinct place-of-birth patterns:

```
- 600 <Target_Person> was born in <Target_GPE>
- 169 <Target_Person> (born <Date> in <Target_GPE> )
- 44 Born in <Target_GPE> , <Target_Person>
- 10 <Target_Person> was a native <Target_GPE>
- 10 <Target_Person> 's hometown of <Target_GPE>
- 1 <Target_Person> was baptized in <Target_GPE>
```

291 distinct date-of-death patterns:

Biography as an IE task

 This approach is good for the consistently annotated fields in Wikipedia: place of birth, date of birth, place of death, date of death

 Not all fields of interests are annotated, a different approach is needed to cover the rest of the slots

Bouncing between Wikipedia and Google

- Use one seed tuple only:
 - <Target Person> and <Target field>
 - Google: "Arafat" "civil engineering", we get:



Web Images Video News Maps more »

Arafat "civil engineering"

Search

Advanced Search Preferences

Web

Yasser Arafat

By 1956 Arafat graduated with a bachelor's degree in civil engineering and served as a second lieutenant in the Egyptian Army during the Suez Crisis. ... www.jewishvirtuallibrary.org/jsource/biography/arafat.html - 61k - Cached - Similar pages - Note this

Yasser Arafat: Biography and Much More from Answers.com

In the 1950s, **Arafat** studied at Fu'ad I University in Cairo (now Cairo University), majoring in **civil engineering** He was reportedly a member of the Muslim ...

www.answers.com/topic/yasser-**arafat** - 89k - Cached - Similar pages - Note this

Engology.com, Engineer Yasser Arafat, Nobel Piece Prize Winner ...

After the war Arafat studied civil engineering at the University of Cairo. He headed the Palestinian Students League and, by the time he graduated, ... www.engology.com/engpg5eyasserarafat.htm - 7k - Cached - Similar pages - Note this

Yasser Arafat and the Palestine Liberation Organization

It was there that Yasser **Arafat**, a **Civil Engineering** student and his coterie, including Salah Khalaf (Abu Iyad), later to become **Arafat**'s second in command ... www.palestinefacts.org/pf_1948to1967_plo_**arafat**.php - 14k - Cached - Similar pages - Note this

A Life in Retrospect: Yasser Arafat | TIME

Here's one thing we know for sure: Yasser **Arafat** was a grand ... at King Fuad I University (now Cairo University), where he studied civil engineering... www.time.com/time/world/article/0,8599,781566-1,00.html - 39k - Cached - Similar pages - Note this

Yassir Arafat's Biography

Yasser **Arafat** was born in 1929 in Jerusalem. His full name is: Mohammed Abad Arouf **Arafat**. He studied **civil engineering** at CairoUniversity. ... www.eretzyisroel.org/~jkatz/arafatbio.html - 72k - Cached - Similar pages - Note this

Biographical and other information on Yasser Arafat who is in bad ...

In 1951, at the age of 21, **Arafat** got military training with the Egyptian army. — In 1956, **Arafat** earned a degree in **civil engineering** at the University of ...

www.freemuslims.org/news/article.php?article=198 - 14k -

Cached - Similar pages - Note this

Bouncing between Wikipedia and Google

- Use one seed tuple only:
 - Google: "Arafat" "civil engineering", we get:
 - ⇒ Arafat graduated with a bachelor's degree in civil engineering
 - ⇒ Arafat studied civil engineering
 - \Rightarrow Arafat, a civil engineering student

⇒...

 Using these snippets, corresponding patterns are created, then filtered out.

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- Using these snippets, corresponding patterns are created, then filtered out manually
- Due to time limitation the automatic filter was not completed.
- To get more seed tuples, go to Wikipedia biography pages only and search for:
 - "graduated with a bachelor's degree in"
 - We get:



Web Images: Video News Maps more »

site:en.wikipedia.org "graduated with a bache | Search

Advanced Search Preferences

Web Results 1

Burnie Thompson - Wikipedia, the free encyclopedia

In 2000, he graduated with a bachelor's degree in political science from California State University, Fullerton, Two years later he graduated from The ... en.wikipedia.org/wiki/Burnie Thompson - 19k - Cached - Similar pages - Note this

Roscoe Lee Browne - Wikipedia, the free encyclopedia

Born in Woodbury, New Jersey, Browne first attended historically black Lincoln University in Pennsylvania, and graduated with a bachelor's degree in 1946. ... en.wikipedia.org/wiki/Roscoe Lee Browne - 38k - Cached - Similar pages - Note this

Henry Luke Orombi - Wikipedia, the free encyclopedia

Robert has graduated with a Bachelor's Degree in Environment Studies from Makerere University and Daniel, a gifted musician like his father, is working on ... en.wikipedia.org/wiki/Henry Luke Orombi - 25k - Cached - Similar pages - Note this

Gustave Eiffel - Wikipedia, the free encyclopedia

Eiffel's study habits improved and he graduated with a bachelor's degree in both science and humanities. Eiffel went on to attend college at Sainte Barbe ... en.wikipedia.org/wiki/Gustave Eiffel - 52k - Cached - Similar pages - Note this

Erin Crocker - Wikipedia, the free encyclopedia

... New York, where she graduated with a bachelor's degree in industrial and management engineering in 2003. In 2002, Crocker signed with Woodring Racing to ... en.wikipedia.org/wiki/Erin Crocker - 30k - Cached - Similar pages - Note this

Jim Boeheim - Wikipedia, the free encyclopedia

Boeheim enrolled in Syracuse University as a student in 1963 and graduated with a bachelor's degree in social science in 1969(SU Athletics). ... en.wikipedia.org/wiki/Jim Boeheim - 30k - Cached - Similar pages - Note this

Denise Bode - Wikipedia, the free encyclopedia

She graduated with a bachelor's degree in political science from the University of Oklahoma where she chaired the University of Oklahoma Student Congress. ...

Bouncing between Wikipedia and Google

New seed tuples:

- "Burnie Thompson" "political science"
- "Henrey Luke" "Environment Studies"
- "Erin Crocker" "industrial and management engineering"
- "Denise Bode" "political science"
- **—** ...
- Go back to Google and repeat the process to get more seed patterns!

Bouncing between Wikipedia and Google

- This approach worked well for a few fields such as: education, publication, Immediate family members, and Party or other organization affiliations
- Did not provide good patterns for some of the fields, such as: Religion, Ethnic or tribal affiliations, and Previous domiciles), we got a lot of noise
- Why the bouncing idea is better than using only one corpus?
- Non of the patterns match? Back-off strategy...

Biographical-Sentence Classifier

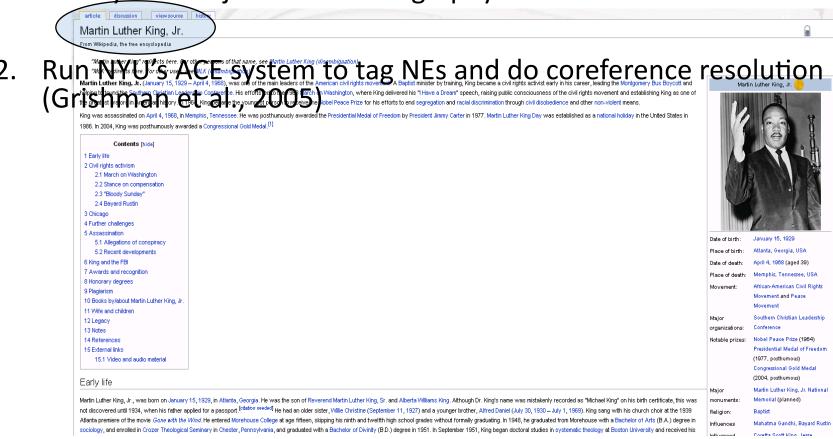
(Biadsy, et al., 2008)

- Train a binary classifier to identify biographical sentences
- Manually annotating a large corpus of biographical and non-biographical information (e.g., Zhou et al., 2004) is labor intensive
- Our approach: collect biographical and nonbiographical corpora automatically

Training Data – Biographical Corpus from Wikipedia

- Utilize Wikipedia biographies
- Extract 17K biographies from the xml version of Wikipedia
- Apply simple text processing techniques to clean up the text

1. Identify the subject of each biography



3. Replace each **NE** by its tag type and subtype

In September 1951, King began his doctoral studies In theology at Boston University.

In [TIMEX], [PER_ Individual] began [TARGET_HIS] doctoral studies In theology at [ORG_Educational].

- 3. Replace each NE by its tag type and subtype
- 4. Non-pronominal referring expression that is coreferential with the target person is replaced by [TARGET_PER]

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- 6. Sentences containing no reference to the target person are removed

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In [TIMEX], [TARGET_PER] began [TARGET_HIS] doctoral studies In theology at [ORG_Educational].

- English newswire articles in TDT4 used to represent nonbiographical sentences
- 1. Run NYU's ACE system on each article
- 2. Select a PERSON NE mention at **random** from all NEs in article to represent the target person
- 3. Exclude sentences with no reference to this target
- 4. Replace referring expressions and NEs as in biography corpus

Biographical-Sentence Classifier

- Train a classifier on the biographical and non-biographical corpora
 - Biographical corpus:
 - 30,002 sentences from Wikipedia
 - 2,108 sentences held out for testing
 - Non-Biographical corpus:
 - 23,424 sentences from TDT4
 - 2,108 sentences held out for testing

Biographical-Sentence Classifier

- Features:
 - Frequency of 1-2-3 grams of class-based/lexical, e.g.:
 - [TARGET_PER] was born
 - [TARGET_HER] husband was
 - [TARGET_PER] said
 - Frequency of 1-2 grams of POS
- Chi-square for feature selection

Classification Results

• Experimented with three types of classifiers:

Classifier	Accuracy	F-Meassure
SVM	87.6%	0.87
M. Naïve Bayes (MNB)	84.1%	0.84
C4.5	81.8%	0.82

 Note: Classifiers provide a confidence score for each classified sample

Thank you