04 – Named Entity Recognition IA161 Advanced Techniques of Natural Language Processing

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Washington: Ben Carson said Wednesday he's pulling in lots of money amid all the backlash he's received for remarks he made regarding Muslims in politics. The retired neurosurgeon said he raised \$1 million within 24 hours following the CNN debate on Sept. 16, and that donations have poured in after remarks he made over the weekend about Islam and the presidency. "The money has been coming in so fast, it's hard to even keep up with it," he said Wednesday morning on Fox News, when asked about whether his comments had affected his donations. "I remember the day of the last debate, within 24 hours we raised \$1 million. And it's coming in at least at that rate if not quite a bit faster." CNN will not be able to verify fundraising totals with the Federal Election Commission until after the quarter ends Sept 30.

Outline

- Named Entity Recognition
- Named Entity Classification
- Methods for NER
 - Gazetteer Methods for NER
 - Semi-supervised methods for NER
 - Supervised methods for NER
- 4 Evaluation of NER systems

Named Entity Recognition (NER)

NER aims to recognize and classify names of people, locations, organizations, products, artworks, domain names, phone numbers, dates, money, measurements (numbers with units), law or patent numbers etc.

Named entities (NEs) can be one word or multi word. [overlap with multi word expression (MWE) processing]

| Example | | |
|----------|----|-----|
| | NE | MWE |
| Brno | 1 | X |
| a priori | X | ✓ |
| New York | 1 | ✓ |

Named Entity Recognition (NER)

NER is vital for information extraction (IE).

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Example
MIT Press published a book by Patrick Hanks with the title
Lexical Analysis: Norms and Exploitations. MIT Press published a book
by Patrick Hanks with the title
Lexical Analysis: Norms and Exploitations .
MIT Press published a book by Randy Thornhill and Craig T. Palmer
entitled A Natural History of Rape: Biological Bases of Sexual Coercion
MIT Press | published a book by | Randy Thornhill | and | Craig T. Palmer
entitled A Natural History of Rape: Biological Bases of Sexual Coercion
                                   Title
 Authors
 Patrick Hanks
                                   Lexical Analysis: Norms and
                                   Exploitations
 Randy
                  Craig T.
                                   A Natural History of Rape:
```

Named Entity Recognition (NER)

Treating the whole multiword NE as one entity can improve advanced natural language processing:

Example

NER: recognizing boundaries

Example

Masaryk University in Brno

Masaryk University in Brno

Masaryk University in Brno

Example

The Picture of Dorian Gray

The Picture of Dorian Gray

Franz Válek



Nová opera Vladimíra

Franze Válka s mloky . . .

Named Entity Classification

Common classes: PERSON, ORGANIZATION, LOCATION Less common classes: MONEY, PERCENT, DATE, TIME

Rare classes: ARTWORK, PRODUCT, ROLE

Example

The White House Othello PERSON? ARTWORK? PRODUCT?

Motorola ORGANIZATION? PRODUCT?

The Pope PERSON? ROLE?

two years ago DATE? nothing?

The main problem is with metonymy.

Methods for NER

- gazetteer methods (list of NEs)
- semi-supervised machine learning (bootstrapping)
- supervised machine learning (training)

Gazetteer Methods for NER

lists of NEs + substring search algorithms:

- list of names
- list of company names
- list of place names

search all occurences of substrings S_k, \ldots, S_l from lists of pattern strings P_1, \ldots, P_p in a target string $T[1 \ldots m]$ Example algorithms:

- naïve multi-pass: O(p(m-n+1))
- improvements: Rabin-Karp, Boyer-Moore, Knuth-Morris-Pratt
- single-pass: Aho-Corasick: O(m + k)

where p is the number of patterns, m is the target (searchable) string length, n is the average pattern length, k is the total number of occurrences of the pattern strings in the text

Gazetteer Methods for NER

Problems: disambiguation + fixedness

Example

May the force be with you!

I was born on May.

Karel May is my favorite writer.

Example

Google was bought by Brand New So-far-unknown Company Inc.

Semi-supervised methods for NER

bootstrapping = a small degree of supervision typically requires a small set of *seeds*

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Example
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seeds: John, James, Steve
search patterns in contexts:
Peter, David, Michael . . .
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Example

```
[Capitalized words and letters], the CEO of [Capitalized words and non-capitalized stop words],
```

```
Richard Rosenblatt, the CEO of Demand Media,

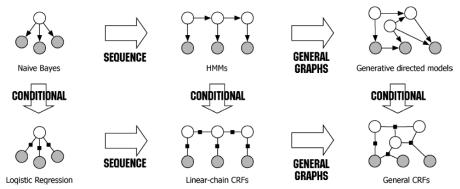
Michael Close, the CEO of Enterprise Training Centre,
```

Semi-supervised methods for NER

Supervised methods for NER

manually annotated training set manually annotated test set (the golden standard) + optionally the gazetteer

discriminative vs. generative methods



Evaluation of NER systems

precision, recall, F1-score separate precision, recall, F1-score measurements for different classes the less difficult classes are: DATE, MONEY, PERCENT the most difficult classes are: PERSON, ORGANIZATION

Error analysis:

- errors in boundaries detection
- errors in class labeling

What is preferred: high precision (and low recall) or high recall (and more false positives)?

... see also [8]

Current state-of-the-art results

| Language | System | F1 |
|----------|---------------------------------------|--------|
| English | MUC-7 ¹ , baseline | 58.89% |
| English | MUC-7 human annotation | 97.60% |
| English | MUC-7 best result [9] | 93.39% |
| English | CONLL-2003 best result [3] | 88.76% |
| English | CONLL-2003 [6] | 90.10% |
| German | GermEval 2014 best result [5] | 77.14% |
| Russian | [4] | 75.05% |
| Czech | [11] | 82.82% |
| Czech | [7] | 83.24% |
| Arabic | [1] | 65.76% |
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¹Message Understanding Conference

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