Week 4: Named Entity Linking

Nhung Nguyen slides courtesy of Phong Le

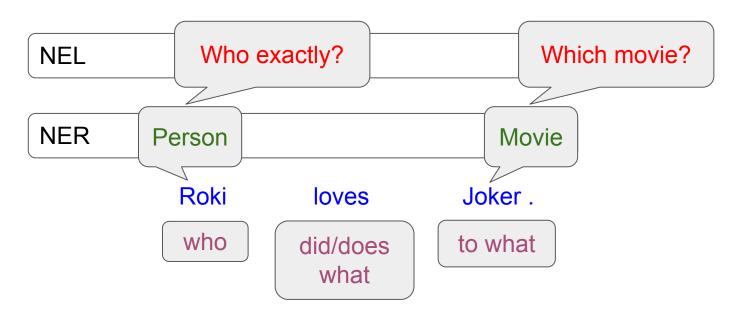
Intended learning outcomes

- Understand the importance of Named Entity Linking (NEL)
- Know several knowledge bases used in the task
- Know basic steps of NEL and their approaches

Materials

- Shen et al., Entity Linking with a Knowledge Base: Issues, Techniques, and Solutions. TKDE14
- Ganea and Hofmann, Deep Joint Entity Disambiguation with Local Neural Attention, EMNLP 2017

Recap: Named entity recognition



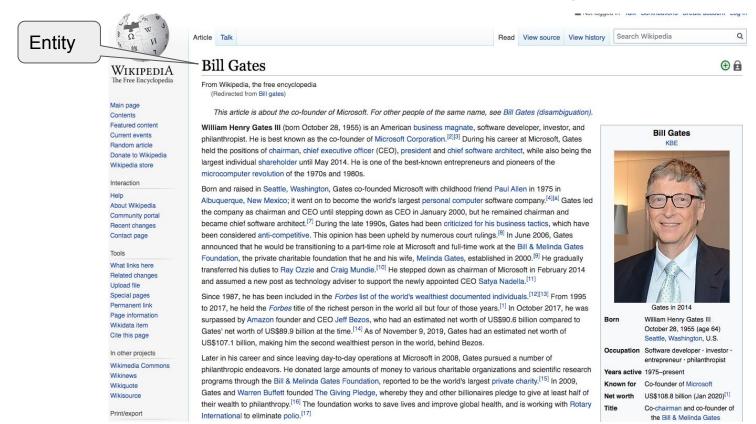
- NER: identify <u>named entity mentions</u> and their <u>types</u>
- NER however doesn't tell us <u>who/what exactly the entities are</u>

Which entities?

- We need a knowledge base to store
 - entries for entities in interest (called entities in short)
 - relations between entities
- → depending on domains / tasks
 - News/open domain:
 - Wikipedia (each Wikipedia article represents an entity)
 - Freebase, a set of triplets <subject, relation, object> (subjects and objects are entities)
 - Wikidata, DBpedia, ...
 - Biomedical
 - o <u>ICD-9</u>, <u>MedDRA</u>
 - o <u>UMLS</u>
 - SNOMED

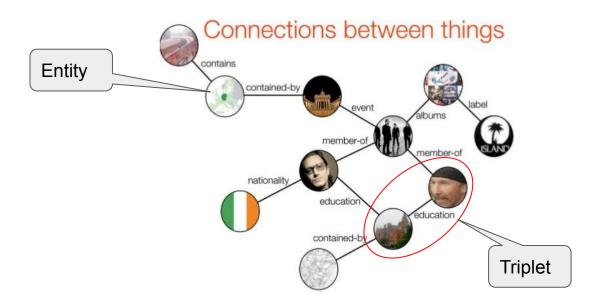
Wikipedia

about 5 million entities, each entities is described by an article



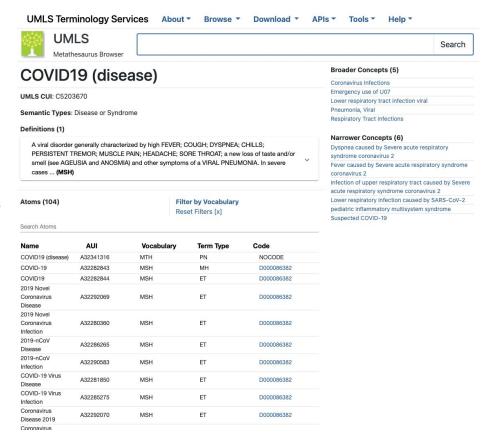
Freebase

• Freebase is a knowledge graph (44 million nodes, 2.5 billion edges, Jan 2014)



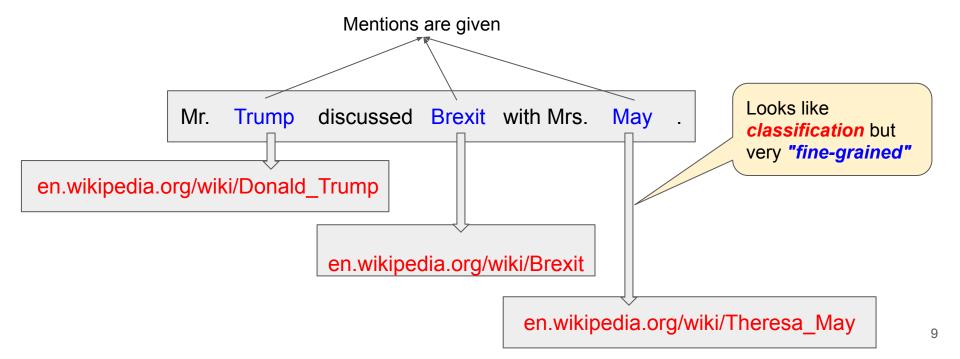
UMLS - Unified Medical Language System

- Combine many biomedical vocabularies
- UMLS Metathesaurus:
 - Entities: more than 4 millions
 - Number of mentions: more than 15 millions
 - Number of distinct mentions: 13 millions



Named entity linking (NEL)

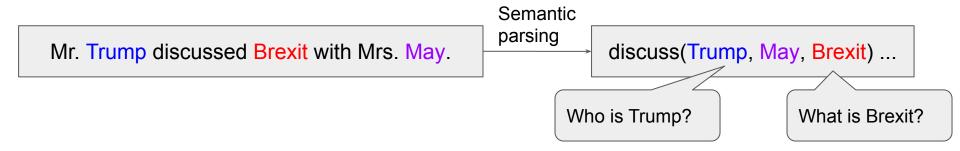
Named entity linking (NEL) is the task of linking a mention to the corresponding entity in a knowledge base (e.g., Wikipedia)



NEL is difficult

- A knowledge base can have a very large number of entities
 - Wikipedia: 5 millions
 - Freebase: 44 millions
 - UMLS: 13 millions
- An entity can have very different surface forms
 - Donald Trump: Trump, President, Snowflake-in-Chief
- A surface form can be linked to several entities
 - Trump: Donald Trump, Ivanka Trump
- Knowledge base can't cover all entities or forms of entities in texts

NEL is important for natural language understanding



Question: Who discussed Brexit with Theresa May?

Answer:

- Trump X
- US President Donald Trump



Entity embeddings

- We can learn entity embeddings in a similar way we learn word embeddings
- Using Wikipedia for entity annotations
 - We use pre-trained word embeddings (word2vec or glove)
 - Learn entity embeddings using (entity, context) pairs from Wikipedia
 - (We can jointly train entity embeddings and word embeddings)



...but partially blames the

antitrust

litigation during the time...

...but partially blames the wiki/United States antitrust law litigation during the time...

Wikipedia2vec (demo)

General approach

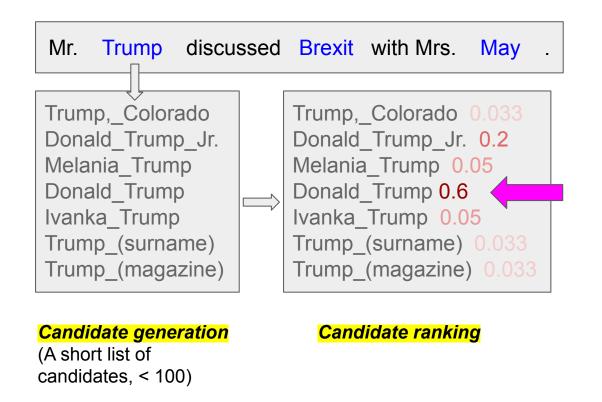
Generate candidates

- Dictionary-based
- Rule-based
- Similarity
- o Information retrieval techniques, etc.

Rank candidates

- Classification
- Neural-based

Example



Step 1: Candidate generation

Extracting an entity-alias dictionary from Wikipedia

Computer

From Wikipedia, the free encyclopedia

For other uses, see Computer (disambiguation).

A **computer** is a machine that can be instructed to carry out sequences of arithmetic or logical operations automatically via computer programming. Modern computers have the ability to follow generalized sets of operations, called *programs*. These programs enable computers to perform an extremely wide range of tasks. A "complete" computer including the hardware, the operating system (main software), and peripheral equipment required and used for "full" operation can be referred to as a **computer system**. This term may as

Boolean algebra

From Wikipedia, the free encyclopedia

For other uses, see Boolean algebra (disambiguation).

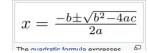
In mathematics and mathematical logic, **Boolean algebra** is the branch of algebra in which the values of the variables are the truth values *true* and *false*, usually denoted 1 and 0 respectively. Instead of elementary algebra where the values of the variables are numbers, and the prime operations are addition and multiplication, the main operations of Boolean

Algebra

From Wikipedia, the free encyclopedia

For the kind of algebraic structure, see Algebra over a field. For other uses, see Algebra (disambiguation).

Algebra (from Arabic: الجبر (al-jabr, meaning "reunion of broken parts" [1] and "bonesetting" [2]) is one of the broad parts of mathematics, together with number theory, geometry and analysis. In its most general form, algebra is the study of mathematical symbols and the rules for manipulating these symbols, [3] it is a



Entity: https://en.wikipedia.org/wiki/Boolean_algebra

Alias: logical

Entity: https://en.wikipedia.org/wiki/Computer_program

Alias: programs



entity	alias	count
wiki/Computer_program	programs	120
wiki/boolean_algebra	logical	15
-	boolean	9

Step 1: Candidate generation (cont.)

Extracting an entity-alias dictionary from Wikipedia

entity	alias	count
wiki/Computer_program	programs	120
wiki/boolean_algebra	logical	15
-	boolean	9

An entity-alias dictionary gives us Pr(entity=e | alias/mention=m)

$$Pr(e|m) = \frac{count(e,m)}{\sum_{e'} count(e',m)}$$
 smoothing?

Step 1: Candidate generation (cont.)

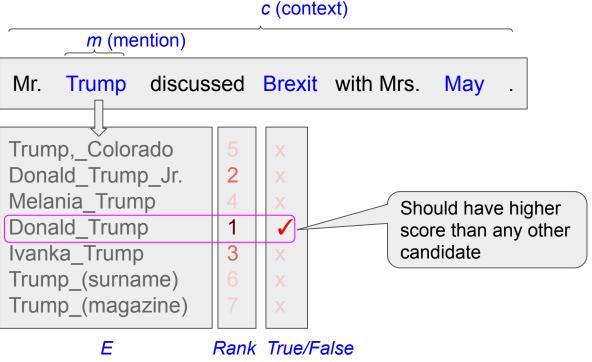
Selecting n (< 100) candidates by their Pr(entity|mention)



Using word/entity embeddings?

Step 2: Candidate ranking

Learning to rank



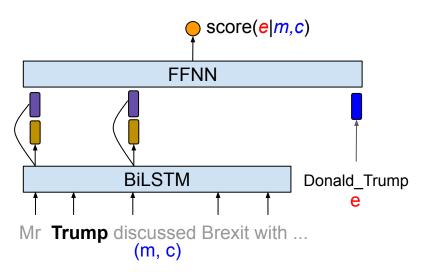
Like NER, there are two approaches: <u>local</u> and <u>global</u>

Local approach

- For each mention m in context c, we have a set E of candidates $(e_1, ..., e_{|E|})$
- We want the correct entity e has the higher score among the candidates

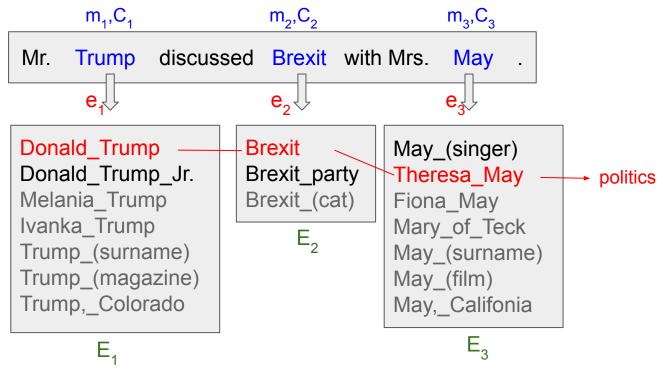
score(e|m,c) > score(e'|m,c) for all $e' \in E$, $e' \neq e$

Using BiLSTM



Global approach

 Coherency hypothesis: entities appearing together in a document should be coherent.



Global approach (cont.)

- $E = (e_1,...,e_n), M = (m_1,...,m_n), C = (c_1,...,c_n)$
- Fully connected graph CRF (Ganea & Hofmann, 2017): Score function of all entities at onces

Downstream applications

- Information retrieval
 - Linking ambiguous entity mentions in query to improve the search results
- Content analysis
 - Linking named entity mentions with a knowledge base across documents
- Question answering
 - exploit the entity linking technique to predict the types of questions and candidate answers, and obtain promising results
- Knowledge base (KB) construction/completion
 - Extract new facts/knowledge from texts, link them to existing KB

Summary

- NEL is an important task in language understanding and useful for many downstream applications
- NEL is challenging
- Two main steps of NEL: candidate generation and ranking
- For candidate ranking, we can implement local or global approach using CRF and/or neural networks.