

coreference between KILL and

ASSASSINATE types:

Identical?

Sub-event?

Conceptual and Practical Steps in

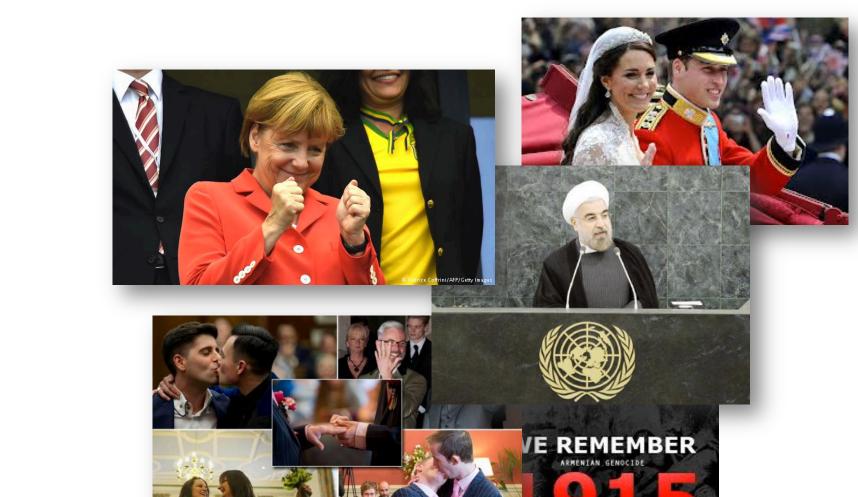
Event Coreference Analysis of Large-scale Data

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Unrelated mentions should be filtered out according to the attribute values:

> People, Time, Location,

Mention 1

"President Kennedy was killed three days before he was to make these amendments public."

Mention 2

"Shortly after noon on November 22, 1963, President John F. Kennedy was assassinated as he rode in a motorcade through Dealey Plaza. "

Mentions 3

"Lushan, China (CNN) -- A strong earthquake that struck the southwestern Chinese province of Sichuan this weekend has killed 186 people, sent nearly 8,200 to hospitals and created a dire dearth of drinking water, Chinese state-run Xinhua reported Sunday."

A Two-step Event Clustering System

Semantic type identification

m = input event mention identified by ClearTK

2. for every synset in S add all hypernym and lexico-semantically related

Similarity-based clustering

1. for every candidate cluster c calculate similarity with m and find the most

similar cluster scored sim_{max} (sim is calculated by applying Jaccard index

3. nominate clusters that include some instances of the types in *S or R*

NE = named entities in the context of m,

of the NE, TE, and S sets belonging to m and c)

2. if $sim_{max} > \theta$, assign m to the most similar cluster

3. else if $sim_{max} < \theta$, create a singleton cluster

4. update the type index w.r.t. S

TE =temporal expressions in the context of m

 $S = \{\}, R = \{\}$ //sets of synsets,

mlem = head lemma of m,

 $mpos = PoS ext{ of } m$

synsets to R

1. add all *WordNet* synsets for (*mlem,mpos*) to *S*

Causally related event?

Identification of Events in the e-Identity Project

The e-Identity project: political scientists want to track interesting events in news and obtain collective information about them, i.e., the identity of every real-world event.

- Event instances or event identities are:
 - Finer-grained than topics
 - Coarser-grained than individual mentions in text
 - Units representative of human understanding of real-world eventualities (particular time, place and participants)
- Data:
- English news 1990 2012
- Avg. 1200 articles per month
- Avg. 100 event mentions per article
- Performance requirement: recall is as important as precision
- Efficiency requirement: single-visit of mentions in the text is desired

Event Mentions

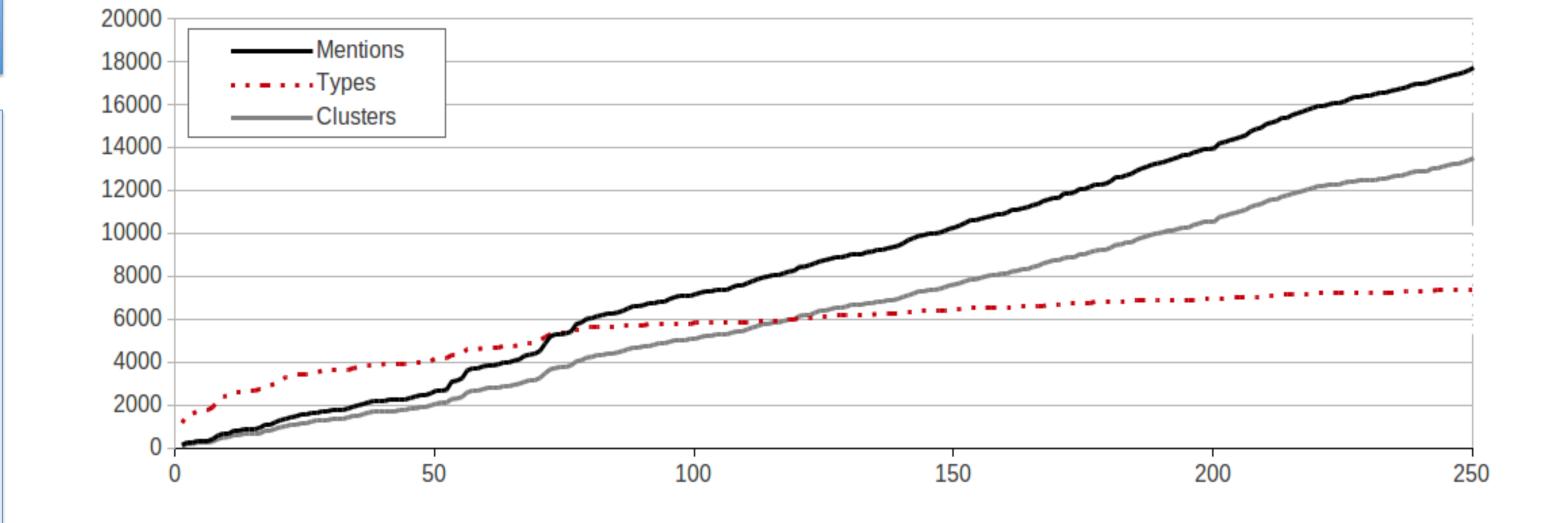
<what we find in text>

Real Event Instances *what we are interested in*

Annotating Event-pairs from TimeML AQUAINT

Event Types /the formalism/

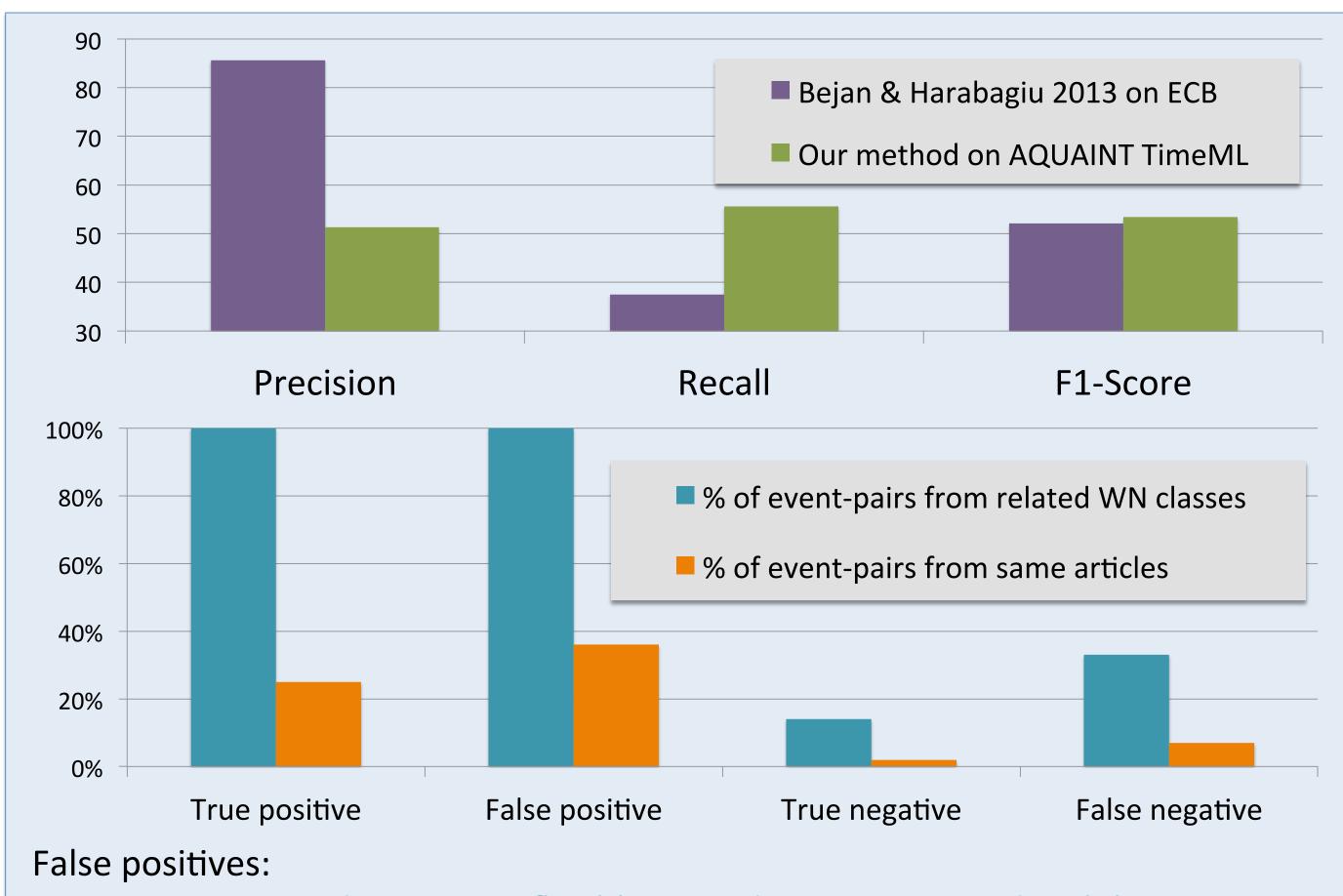
Growth of Types & Clusters (250 articles in a 2-week window)



- Source corpus: AQUAINT TimeML
- Semi-random selection of 100 mentions, i.e., 4950 pairs
- Cross-document and cross-topic decisions
- Result: 36 corefrential and 4914 non-coreferential

http://www.coli.uni-saarland.de/~fatemeh/resources.htm

Error Analysis of the Annotated Corpus



System considers a more flexible type identity compared with human

"the immigration service decided the boy should go home."

"they made a reasonable decision Wednesday in ruling that"

• Similar context (same article, paragraph, or sentence)

"some people are born rich, some are born poor."

False negatives:

- Event types not connected in WordNet
- Sparse context (NEs and/or Timex not resolved)

"the Clinton administration has pushed for the boy's return." "his son said he didn't want to go."

Conclusion

- Named entities and timestamps are useful features (76.5% BLANC).
- Human decisions are more conservative than our system.
- False positives should be fixed by considering linguistic features.
- False negatives should be fixed by extending the semantic layer (e.g., more WordNet links or longer paths to be allowed).
- Temporal expressions are not very helpful in real data clustering.