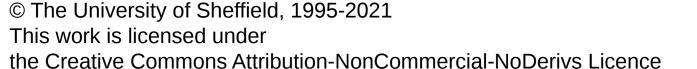


Module 4: GATE and Social Media Part 4. Named entities





Named Entity Recognition



- Texts frequently focus on particular entities
- To discover what documents say about them, we can:
 - Recognise entity mentions
 - Disambiguate entities to external vocabularies
 - Find opinions that authors have about the entities
- Important because:
 - Enables IE over tweets
 - Critical for event extraction (actors, events)
 - Describes the topic of the tweet
- Tough because:
 - ANNIE doesn't do well around 50% F1
 - Stanford's leading tool does even worse around 40% F1!



Goal is to find mentions of entities

Newswire

London Fashion Week grows up – but mustn't take itself too seriously. Once a launching pad for new designers, it is fast becoming the main event. But LFW mustn't let the luxury and money crush its sense of silliness.

Social media Gotta dress up for london fashion week and party in style!!!



Named entity recognition: example



Person mentions in news

Left context	Match	Right context
in dicated Atef, including	Douglas Feith	, the United States defence
, the group that killed	President Sadat	in 1981 as retribution for
. The current leader,	President Olusegun Obasanjo	, who recently came to
Kuwait, whose information minister	Sheikh Ahmed Fahed al-Sabah	met editors of local newspapers
The current defence minister,	Theophilus Danjuma	, has also been threatened
The three right-wing MPs,	Andrew Rosindell	(Romford), Andrew
Late on Wednesday night,	Justice Oputa	, who chairs the commission
the militarily-manoeuvred civilian elec	President Obasanjo	in 1999 and is widely
after the mysterious death of	General Sani Abacha	in 1998.
have learnt that one of	Bin Laden	's closest and most senior
evidence confirms the involvement of	Osama bin Laden	in those attacks."
. He is one of	Bin Laden	's two most senior associates
for future civilian office.	General Buhari	took power in a 1983
\$5m price on	Atef	's head and prosecutors have
Afghanistan. He was once	Bin Laden	's chief media adviser and
thinking in the Tory party	lain Duncan Smith	has ordered three Tory MPs
club and the party,	David Maclean	, the Tory Chief Whip
Centre and the Pentagon.	Mohammed Atef	, who is thought to
are still very powerful.	General Babangida	supported the militarily-manoeuvred civ
sexual orientation or religion.	Mr Duncan Smith	's purge of the Monday
," he said.	Atef	, who is reported variously
of the late singer,	Fela Kuti	♦ which took place while
field in Penn sylvania.	President Bush	included Atef in an order
. It is believed that	Mr Duncan Smith	intended to launch his crackdown

Named entity recognition: example



Person mentions in tweets

Left context	Match	Right context
i was your age ,	spencer	from iCarly was Crazy Steve
iCarly was Crazy Steve ,	Carly	was Megan and Josh was
bath , shut up ,	sam	's coming tomorrow and steve
. All are welcome ,	joe	included
. All are welcome ,	joe	included
teachers , chinese takeaways ,	gatt holly	, phil collins , the
takeaways , gatt holly ,	phil collins	, the skin of a
@GdnPolitics : RT AlJahom :	Blair	: " I'm gonna
Empls of the Month:	Deborah L	#Speech #Pathologist-Childrens
be the next Pope "	Brown	: " I won't
(via POPSUGAR)	Sarah Jessica Parker	and Gwen Stefani Wrap Up
and is smexy !!;)And	Chelsea Handler	is hilarious ! Finally got
him befnrjustthen about	kenny	signing his book but it
three kinds of reactions after	Ayodhya	verdict.
, Carly was Megan and	Josh	was fat . #damnteenquotes
sam 's coming tomorrow and	steve	and tanya will be round
coming tomorrow and steve and	tanya	will be round at 10am
photo caption contest- Nadal and	Novak	in the tub http://ow.ly/2G3Jh
) Sarah Jessica Parker and	Gwen Stefani	Wrap Up Another Successful New
#Pathologist-Childrens Rehab and	Patricia M	#Referral/#Auth #
Just casually stalking Cheryl AND	Dermot	tomorrow NO BIGGIE
did tweet him befnr	justthen	about kenny signing his book
Test : We just congratulated	Lindsay	an hour ago on h
the funny photo caption contest-	Nadal	and Novak in the tub

Named entity recognition: issues



Genre differences in entity type

	News	Tweets		
PER	Politicians, business leaders, journalists, celebrities	Sportsmen, actors, TV personalities, celebrities, names of friends		
LOC	Countries, cities, rivers, and other places related to current affairs	Restaurants, bars, local landmarks/areas, cities, rarely countries		
ORG	Public and private companies, government organisations	Bands, internet companies, sports clubs		

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Tweet Capitalisation: an NER nightmare.

	000000000000				. 4	_	
	2651 True https://si0.twimg.com/profile_images/58439629/petepassport_normal.PNG 88dbf4					_	Lookup
	se 3b3b3b 2408043 False NYC / SF False 972651 -28800 37706 Breaking social media, tech and				3		Lookup
digital news and analysis from Mashable.com, the top resource and guide for all things web. Updates					3	ν'	Organization
from @mashable staff. 2269 https://si0.twimg.com/profile_background_images/208575865/mashable_main_twitter_bk_v3.png 0f78c2 http://a1.twimg.com/profile_images/58439629/petepassport_normal.PNG False False fffef0				3		Sentence	
						SpaceToken	
http://a2.twimg.com/profile_background_images/208575865/mashable_main_twitter_bk_v3.png mashable			3		Token		
en False 0 Pete Cashmore http://mashable.com Mon Mar 12 01:28:01 +0000 2007 False Pacific Time (US & Canada) 5ea7db False 78468 False False Nokia Posts Huge Quarterly Loss, Sees Better Times Ahead -			2000		Tweet		
http://on.mash.to/nCSh4i Thu Jul 21 13:12:30 +0000 2011 False 59 83 http://on.mash.to/nCSh4i			2000		URL		
94031999962071040 HootSuite 0 94031999962071040			2000		Unknown		
			2000	١	Original markups		
<u> </u>			4 8	•	PreProcess		
Type Se	t Start	End Id	d	Features			
Organization	736	741 50	80	{orgType=company, rule1=GazOrganization, rule2=OrgFinal}			Lookup
Organization	769	786 50	9	{rule1=TheOrgXKey, rule2=OrgFinal}	3		Sentence

#WiredBizCon #nike vp said when @Apple saw what http://nikeplus.com did, #SteveJobs was like wow I didn't expect this at all

...And hashtag semantics is yet another...



Case-Insensitive matching

- This would seem the ideal solution, especially for gazetteer lookup, when people don't use case information as expected
- However, setting all PRs to be case-insensitive can have undesired consequences
- POS tagging becomes unreliable (e.g. "May" vs "may")
- Back-off strategies may fail, e.g. unknown words beginning with a capital letter are normally assumed to be proper nouns
- BUT this doesn't work on tweets anyway!
- Gazetteer entries quickly become ambiguous (e.g. many place names and first names are ambiguous with common words)
- Solutions include selective use of case insensitivity, removal of ambiguous terms from lists, additional verification (e.g. use of the text of any contained URLs)



More flexible matching techniques

- In GATE, as well as the standard gazetteers, we have options for modified versions which allow for more flexible matching
 - Extended Gazetteer: has a number of parameters for matching prefixes, suffixes, initial capitalisation and so on

Hands-on: NER evaluation



Let's measure ANNIE performance on social media text

- Open the Ritter-dev corpus from the datastore saved in corpora/r-tweets
- Change all the annotationSetName, inputAS and outputAS parameters in your ANNIE application to ANNIE
- Run your ANNIE pipeline on this corpus
- Have a look at the entities annotated. Can you find any mistakes?
- If so, why do you think this mistake has been made?

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Now let's try with TwitlE



- Remove your Twitter application from GATE (to avoid confusion)
- Load the TwitIE application from the "Ready-made Applications"
- Add ANNIE in the setsToKeep parameter of the Document Reset
- Run TwitlE

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Compare ANNIE and TwitlE



- Open the corpus and click the "Corpus Quality Assurance" tab
- We can now compare 3 annotation sets: Original Markups (the gold standard set) with both TwitlE and ANNIE results
- Pick 2 of these sets to compare (TwitlE results are now in the default set)
- Select annotation types Location, Organization, and Person
- Pick an evaluation measure
- How does it do? What kinds of errors are most prevalent, missed or spurious?
- You can also pick individual documents and see which single annotations are picked up or missed

Named entity recognition summary



Named entity recognition in tweets is hard Three major classes of Tweet NER approach:

- **Sequence labelling** like Stanford CRF chunker
 - Problem: tweets aren't well-formed enough
 - Problem: lack of training data
- **Lookup-based** using local grammar and string matching
 - Problem: strings are often misspelled
 - Problem: entity mentions not in gazetteers (Eisenstein 2013, Plank 2014)
 - Advantage: cuts through linguistic noise, agnostic to many style variation
- **Grounding to vocabulary** (e.g. Dbpedia)
 - Problem: insufficient context to disambiguate
 - Problem: entities often appear in social media before the resource

Overall solutions to twitter noise



Normalisation

- Convert twitter text to "well-formed" text; e.g. slang resolution
- Some success using noisy channel model (Han 2011)
- Techniques include: edit distance; double metaphone with threshold
- Issues: false positives can change meanings, e.g. reversing sentiment (apolitical)

Domain adaptation

- Treat twitter as its own genre, and create customised tools and techniques
- Some success in language ID (Carter 2013), PoS tagging (Gimpel 2011), NER (Ritter 2011)