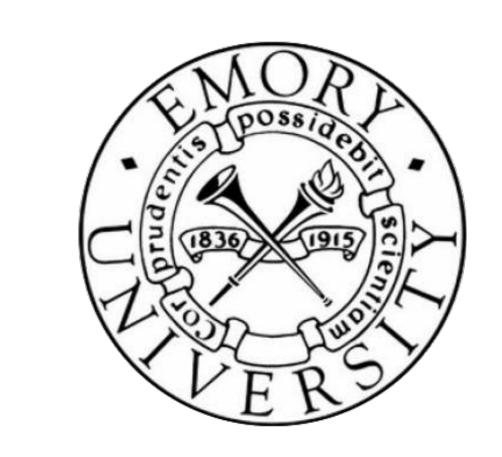


Classifying Non-Referential It for Question Answer Pairs



Timothy Lee, Alex Lutz, and Jinho D. Choi Department of Mathematics and Computer Science, Emory University

Introduction

- Classifying non-referential it is an important task in coreference resolution
- But no such attempts in this classification has been done for question answer pairs
- The style of English used in question answer documents differs greatly from standard documents such as the Wall Street Journal
- Our task is to classify non-referential it for question answer pairs and introduce the dataset

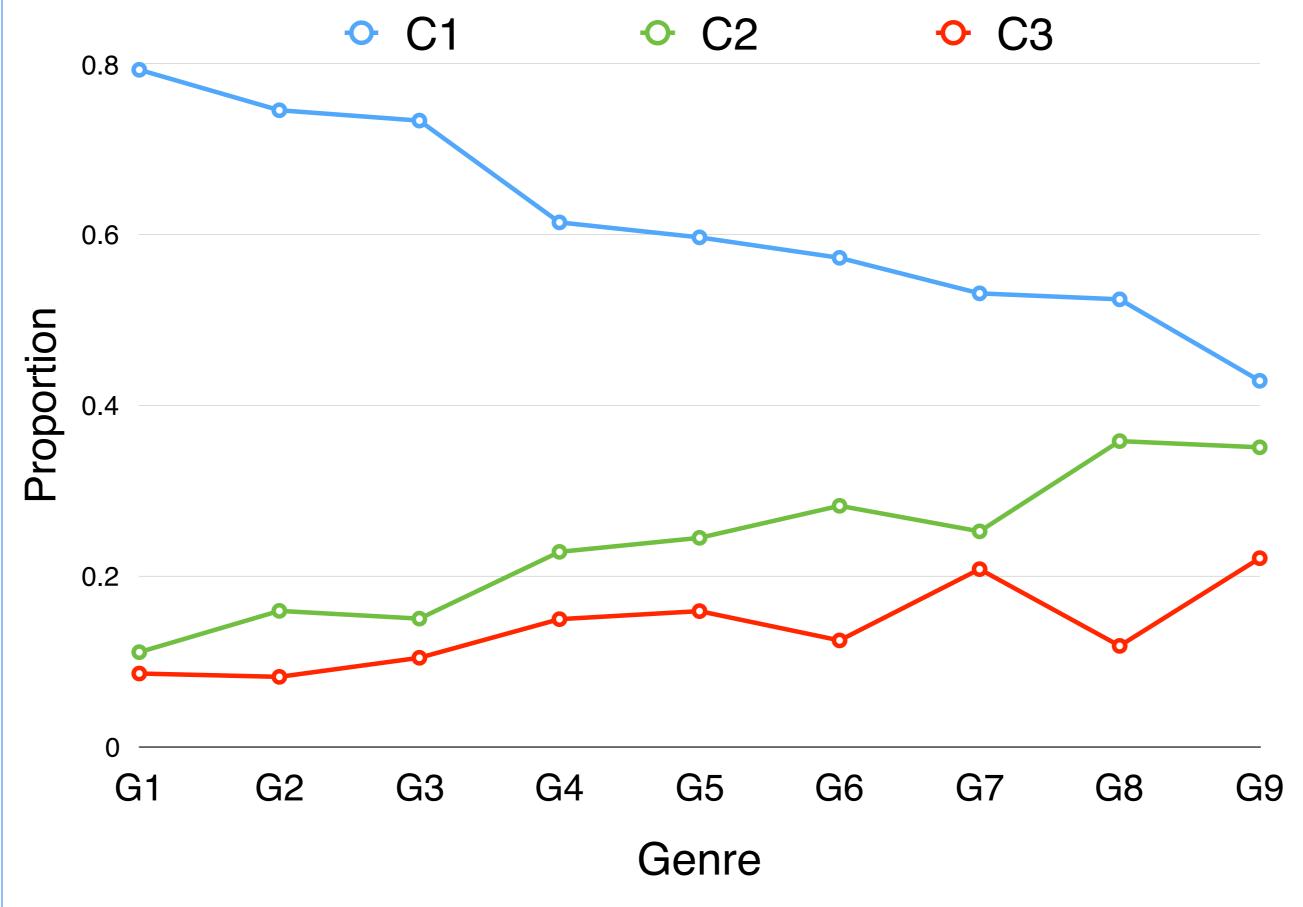
Evans (2001)		Boyd (2005)		Bergsma (2008)		Li (2009)		This Work (2016)		
Cataphoric				Referential - Noun		Referential - Nominal		1. Referential - Nominal		
Proaction		Referential		. 1. Nominal Anaphoric	,	1. Nominal Anaphoric	,	1. Nominal Anaphoric		
Discourse Topic]]>			Non-Referential 1. Cataphoric 2. Proaction 3. Discourse Topic 4. Clause Anaphoric 5. Pleonastic 6. Idiomatic/ Stereotypic		Referential - Clause		2. Referential - Others 1. Proaction		
Nominal Anaphoric						Clause Anaphoric Non-Referential		2. Discourse Topic 3. Clause Anaphoric		
Clause Anaphoric		Non-Referential 1. Extra positional 2. Cleft 3. Weather/Time				1. Cataphoric	***************************************	3. Non-Referential		
Pleonastic	2. 3.					2. Proaction3. Discourse Topic4. Pleonastic		1. Cataphoric 2. Pleonastic		
Idiomatic / Stereotypic		Condition/Place 4. Idiomatic				5. Idiomatic/ Stereotypic		3. Idiomatic/Stereotypic 4. Errors		

- Initially Evans classified it into seven categories which effectively captures all types of it
- We propose a more efficient set of rules to classify it for coreference specifically for question answer pairs

\mathbf{Type}	Description
C1	Anaphoric instances of it which refer to nouns, noun phrases, or gerunds e.g. John bought a book. It was about space.
C2	Anaphoric instances of it which do not refer to nominals such as proaction, clause anaphoras, and discourse topic. e.g. Always use tools correctly. If it feels very awkward, stop.
C3	Contains the most common instances of pleonastic it including extraposition, cleft, atmospheric, and idiomatic e.g. It is sunny outside.
C4	Non-pronoun forms of it including disfluencies, abbreviations, and misspellings e.g. Why did my email move it self?

Corpus Analytics $\mathbf{C}_1 \mid \mathbf{C}_2 \mid \mathbf{C}_3 \mid \mathbf{C}_4 \parallel \mathbf{C}_*$ Genre Sen \mathbf{Tok} 1. Computers and Internet $11,\!586$ 2. Science and Mathematics $11,\!589$ 1,027 11,803 3. Yahoo! Products 4. Education and Reference 11,520 $11,\!267$ 5. Business and Finance 11,656 6. Entertainment and Music $11,\!589$ 7. Society and Culture 11,305 8. Health 11,482 9. Politics and Government 900 | 7,986 | 103,797 | 1,348 | 517 | 300 | 18 | 2,183 Total

- "Politics and Government" and "Society and Culture" had the highest proportion of non-referential instances due to their abstract ideas
- "Computers and Internet" and "Science and Mathematics" had the most referential-nominal cases because these dealt with tangible objects



- One large problem while annotating was resolving ambiguous references of *it*
- If it were \$1,700.00 ... and let *it* go but for \$170,000...
- Here, it can be either idiomatic, or refer to the "post dated cheque" or the "process of receiving the post dated cheque"
- Contextual information proved vital to solve this ambiguity
- Q: Regarding *IT*, what are the fastest ways of getting superich? A: with maintenece or service of systems or with old programming languages.
- Since Yahoo! Answers isn't standard english, *IT* could have been capitalized for emphasis or to mean Information Technology. With the help of contextual information, such ambiguity could be avoided

Results

Experimental Setup

- Used stochastic adaptive gradient descent with mini-batch and L1 regularization
- Tested new sets of features including brown clusters, word embeddings, and dependency derived relationships

The following features were used to classify instances of it:

- POS and dependency of current word
- POS and lemma for dependency head of current word
- POS and lemma for succeeding token and POS of 2nd succeeding token
- POS of succeeding token with lemma of 2nd succeeding token
- POS of 1st and 2nd succeeding token with lemma of 3rd succeeding token

Model	Development Set					Evaluation Set				
	ACC	C_1	C_2	C_3	C_4	ACC	C_1	C_2	C_3	C_4
$\overline{\mathrm{M}_{0}}$	72.73	82.43	35.48	57.14	0.00	74.05	82.65	49.20	71.07	0.00
M_1	73.21	82.56	50.00	62.50	$\mid 0.00 \mid$	74.68	82.93	53.14	73.33	0.00
M_2	73.08	82.56	49.41	60.00	_	75.21	83.39	51.23	73.95	
$\overline{\mathrm{M}_{3}}$	76.44	82.31	64.75		-	77.14	82.26	67.87		-
${ m M}_4$	76.92	83.45	61.90		-	78.21	83.39	68.32		_

- m0 only used the baseline features
- m1 uses additional features based on the relative position of it, the relative distance from preceding noun, and relative position of sentence within document
- m2 discard the annotations for errors
- m3 merges referential-nominal and referential-other during the training set
- m4 merges referential-nominal and referential-other during the evaluation set

Conclusion

- We introduced a new corpus from Yahoo! Answers which classified instances of it into four categories
- Using a mixture of old and new features, we were able to achieve promising results despite a challenging dataset
- In the future, we plan to increase the size of our dataset by adding more genres from Yahoo! Answers
- We plan to use a recurrent neural network with our dataset in the future to see if that will yield better results