

OntoGUM: Evaluating Contextualized SOTA Coreference Resolution on 12 More Genres

Yilun Zhu¹, Sameer Pradhan^{2,3}, Amir Zeldes¹







1 GEORGETOWN UNIVERSITY

The Coreference Resolution Task

"I voted for Mary because Mary was most aligned with my values", John said.

Problems of Existing Coreference Datasets

Datasets	Out-of ON domain	Scheme compatibility	Multi-genre	Multi-coreference types	Singletons
OntoNotes (Pradhan et al., 2013)	×	\checkmark	\checkmark	\checkmark	×
WikiCoref (Ghaddar and Langlais, 2016)	\checkmark	\checkmark	×	\checkmark	×
GAP (Webster et al., 2018)	\checkmark	\checkmark	×	×	×
GUM (Zeldes, 2017)	\checkmark	×	\checkmark	\checkmark	\checkmark
ARRAU (Poesio et al., 2018)	\checkmark	×	\checkmark	\checkmark	\checkmark
PreCo (Chen et al., 2018)	\checkmark	×	\checkmark	\checkmark	\checkmark

Problems of Existing Out-of-domain Evaluation

• No study has investigated if contextualized embeddings encounter the same overfitting problem identified by Moosavi and Strube (2017)

- Previous work may underestimate the performance degradation on WikiCoref
 - embeddings were also trained on Wikipedia themselves (Moosavi and Strube, 2018)
 - -> higher coreference scores on Wikipedia texts

OntoGUM Dataset

- Conversion from GUM using gold standard syntax trees
- Statistics
 - 168 documents with 12 genres, ~150K tokens
 - 19,378 mentions, 4,471 clusters
- Genres
 - Text: News / Fiction / Bio / Academic / Forum / Travel / How-to / Textbook
 - Speech: Interview / Political / Vlog / Conversation



Dataset Conversion

- OntoNotes ⊆ GUM
 - Don't need human annotation to recognize additional mentions in the conversion process
- Annotation layers used in the conversion
 - Coreference layer
 - Gold syntax trees

GUM:	Kim visited Seoul The visit
OntoGUM:	Kim visited Seoul The visit



- Annotation agreement
 - Agreement study on 3 docs (2,500 tokens, 371 mentions), 8/371 errors
 - Span detection: ~0.96 CoNLL coreference score: ~0.92

Gold speaker information (fiction, reddit and spoken data)

Experiments & Results 1/3

Models	OntoNotes	OntoGUM
dcoref (Manning et al. 2014, CoreNLP)	57.8	
e2e + SpanBERT (Joshi et al., 2019, SOTA)	79.6	

Experiments & Results 1/3

Models	OntoNotes	OntoGUM
dcoref (Manning et al. 2014, CoreNLP)	57.8	39.7
e2e + SpanBERT (Joshi et al., 2019, SOTA)	79.6	64.6

• Both systems encounter a substantial degradation on OntoGUM

Experiments & Results 2/3

- Genre disparity does not guarantee low performance (e.g., vlog), and errors occur readily even in overlapping genres (e.g., news)
- Performance is correlated with the proportions of pronouns

		PRON (R)	Other (R)	Total	CoNLL	Span
	vl	600 (.66)	309 (.34)	909	1	1
	it	1223 (.45)	1485 (.55)	2708	2	6
	cn	729 (.61)	323 (.39)	1052	3	2
	sp	245 (.40)	364 (.60)	609	4	4
	bi	796 (.34)	1529 (.66)	2325	5	3
	fc	1700 (.61)	1091 (.39)	2791	6	5
	ac	262 (.21)	997 (.79)	1259	7	10
	vy	300 (.22)	1053 (.78)	1353	8	7
	rd	1337 (.55)	1077 (.45)	2414	9	8
	nw	340 (.19)	1483 (.81)	1823	10	9
	wh	1001 (.47)	1129 (.53)	2130	11	11
	tx	165 (.34)	315 (.66)	480	12	12

Table 1: Genre-breakdown Statistics of OntoGUM

Experiments & Results 3/3

- Genre disparity does not guarantee low performance (e.g., vlog), and errors occur readily even in overlapping genres (e.g., news)
- Performance is correlated with the proportions of pronouns or gold speaker information

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Conclusion

- We release the largest open, gold, coreference dataset with new genres following the OntoNotes scheme
- Results showed a lack of generalizability of existing systems, especially in genres low in pronouns and lacking speaker information
- A genre-by-genre analysis reveals relative strengths and weaknesses of current approaches

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We hope people can use OntoGUM as an out-ofdomain benchmark for systems developed using OntoNotes!