





Cross-document Coreference Resolution over Predicted Mentions

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Findings of ACL 2021





Cross-document Coreference Resolution

U.S President nominates new surgeon general: MacArthur "genius grant" fellow Regina Benjamin on the July 13, 2009.

Obama emphasize his decision in "her extensive and distinguished career in medicine".

President Obama will name Dr. Regina Benjamin as U.S. Surgeon General on Monday in a Rose Garden announcement, later this morning.

News that Barack Obama may name Dr. Sanjay Gupta of Emory University and CNN as his Surgeon General has caused a spasm of celebrity reporting. CNN's management confirmed that Dr. Gupta had been approached by the Obama team on March 2009. The chief medical correspondent has declined comment.

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There are two main limitations in previous work:

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- Not efficient

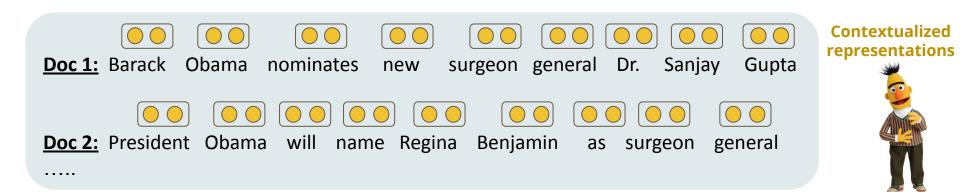
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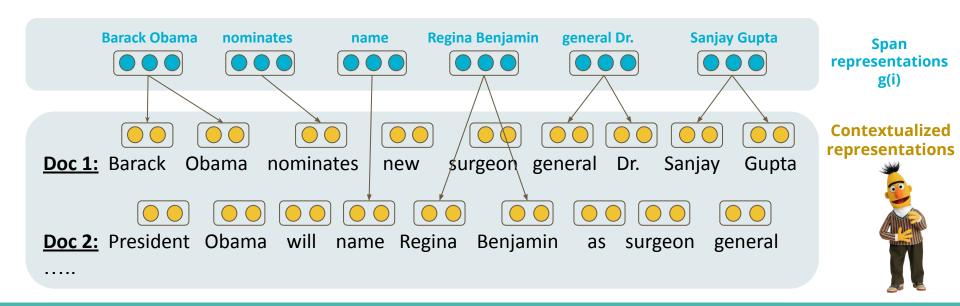
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 - <u>Caciularu et al. (2021)</u> apply CDLM on two full documents for every mention pair (!)

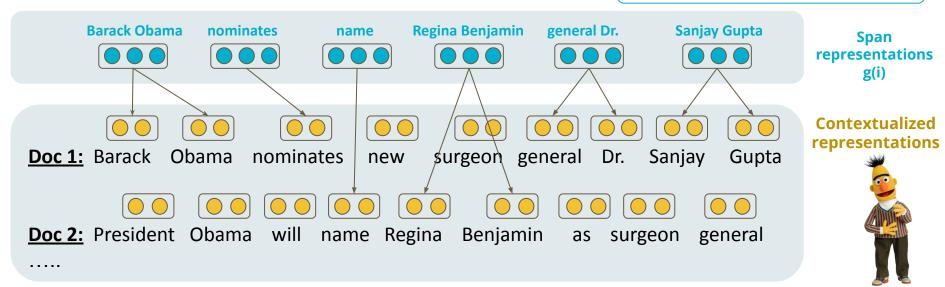
<u>Doc 1:</u> Barack Obama nominates new surgeon general Dr. Sanjay Gupta

<u>Doc 2:</u> President Obama will name Regina Benjamin as surgeon general

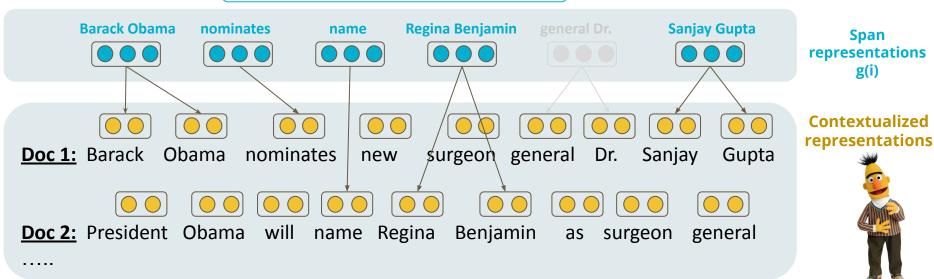


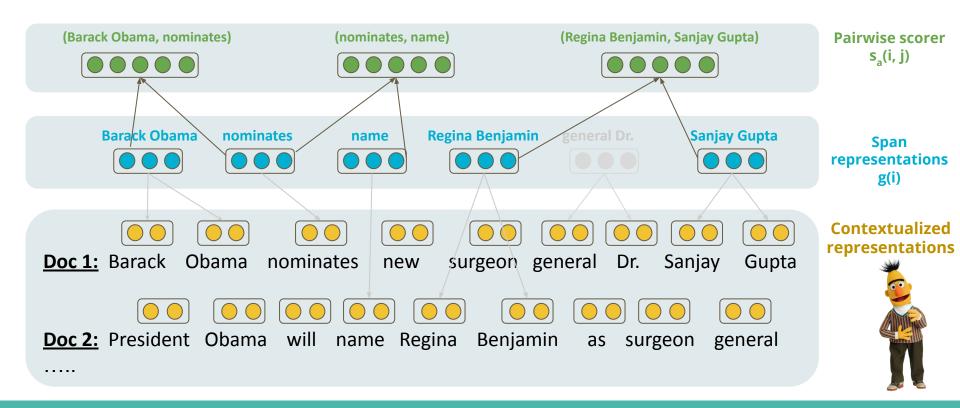


$$g_i = [x_{ ext{FIRST(i)}}, x_{ ext{LAST(i)}}, \hat{x}_i, \phi(i)]$$

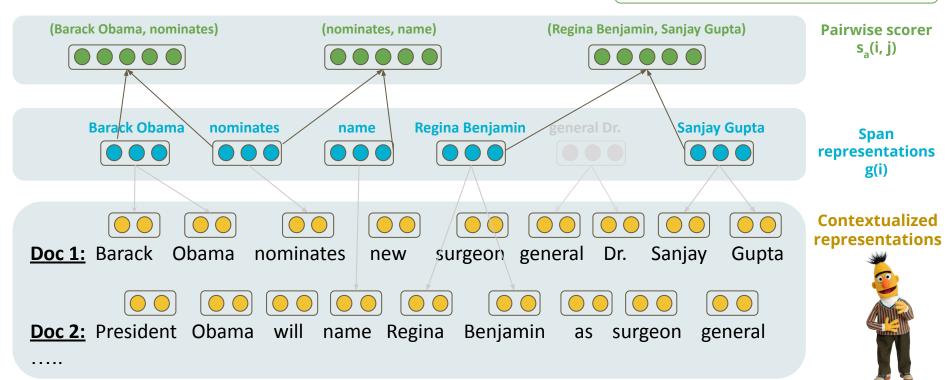


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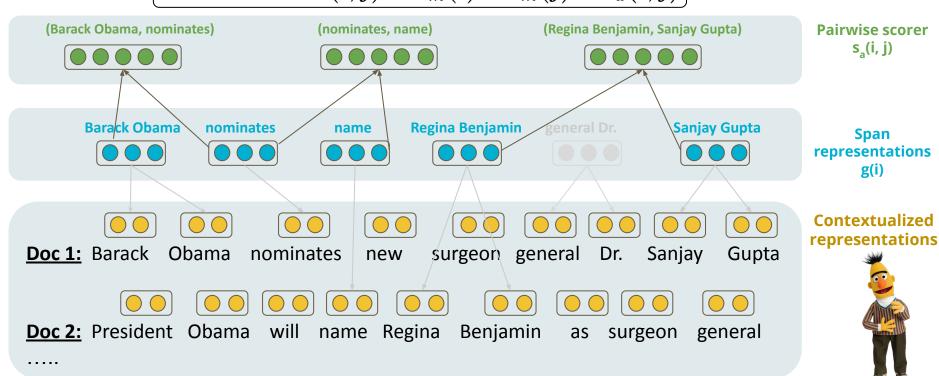




$$s_a(i,j) = extsf{FFNN}_a([g_i,g_j,g_i\circ g_j])$$



Overall score: $s(i,j) = s_m(i) + s_m(j) + s_a(i,j)$



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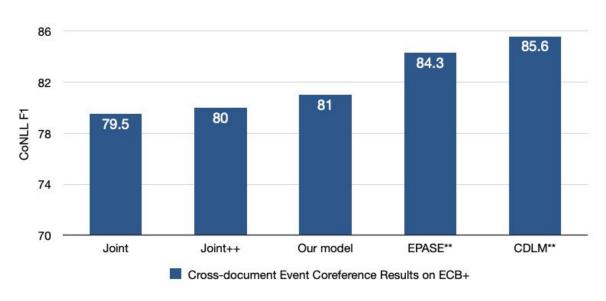
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Advantages:

- Single pass of RoBERTa for each <u>document</u>
- Pairwise scores are calculated only once
- Fast inference on a single 12GB GPU

Results — Gold mentions on ECB+



** Apply BERT for all mention pairs

Joint: Barhom, S., Shwartz, V., Eirew, A., Bugert, M., Reimers, N., & Dagan, I. (2019). Revisiting Joint Modeling of Cross-document Entity and Event Coreference Resolution. *ACL*. Joint++: Meged, Y., Caciularu, A., Shwartz, V., & Dagan, I. (2020). Paraphrasing vs Coreferring: Two Sides of the Same Coin. *FINDINGS*. Epase: Zeng, Y., Jin, X., Guan, S., Guo, J., & Cheng, X. (2020). Event Coreference Resolution with their Paraphrases and Argument-aware Embeddings. COLING.. CDLM: Caciularu, A., Cohan, A., Beltagy, I., Peters, M.E., Cattan, A., & Dagan, I. (2021). Cross-Document Language Modeling. ArXiv, abs/2101.00406.

Results

- Our model set first results over predicted mentions (54.4 F1)
 - Large room for improvement under realistic conditions

Refer to the paper for the results on entity coreference

Analysis and Ablations

1. Cross-document coreference is **harder** than within-document coreference

	Gold		Predicted	
	WD	CD	WD	CD
Event	86.6	81.0	59.6	54.4
Entity	81.2	73.1	39.7	35.7
ALL	83.9	76.7	46.3	43.4

Table 2: Results (CoNLL F1) of our model, on within-document (WD) vs. cross-document (CD), using gold and predicted mentions.

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2. Document clustering significantly **simplifies** the task, especially for event

	Gold	Δ	Predicted	Δ
Event	76.0	-5.0	48.2	-6.2
Entity	70.9	-2.2	34.4	-1.3
ALL	74.1	-2.6	41.4	-2.0

Table 3: CoNLL F1 results of our model without document clustering, using gold and predicted mentions. Δ shows the relative drop in performance.

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3. Ablation studies show the **effectiveness** of each component

	F1	Δ
Our model	58.1	
- pre-train of mention scorer	54.9	-3.2
- dynamic pruning	54.1	-4.0
 negative sampling 	56.7	-1.4

Table 4: Ablation results (CoNLL F1) of our model on the development set of ECB+ event coreference.

Thanks!

Questions?

Arie Cattan

github.com/ariecattan/coref

arie.cattan.github.io