

# A Rank-Based Similarity Metric for Word Embeddings

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## Motivation

- **Distributional Semantic Models** build *high-dimensional* and *sparse* representations from co-occurrence statistics
- **Semantic similarity** is measured by vector cosine, which treats all features equally
- **Rank-based metrics** have been successfully applied to DSMs, but *not yet* on *low-dimensional* and *dense* Word Embeddings

## APSynP: Rank-Based Metric

### Hypotheses:

- Similarity = sharing a high number of relevant features
- Dissimilarity = either non-sharing relevant features or sharing non-relevant features
- Clustering = cluster members share their salient semantic dimensions to increase cluster cohesiveness

### APSyn for Sparse and High Dimensional Vectors:

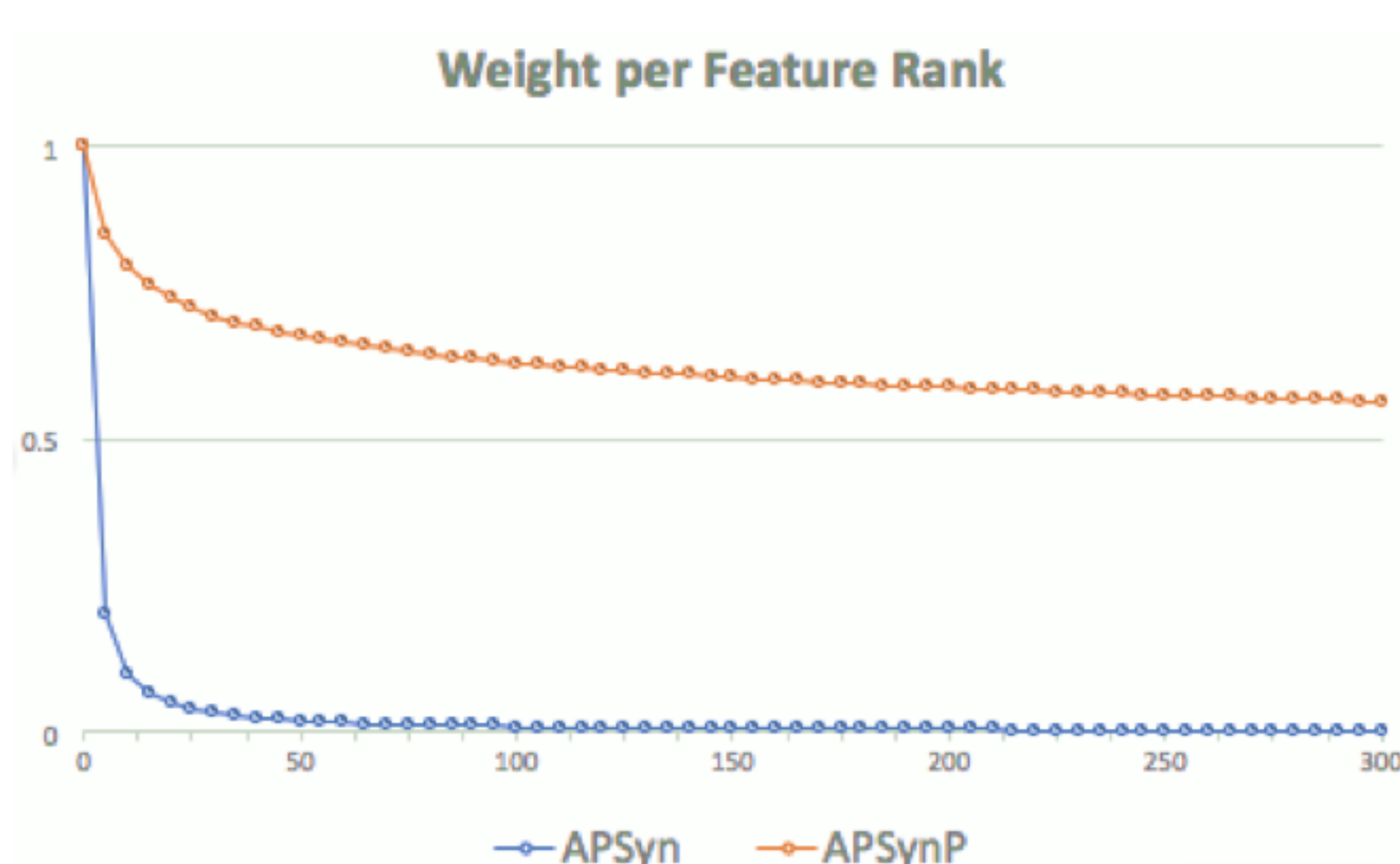
$$APSyn(w_x, w_y) = \sum_{i=0}^{i=N} \frac{1}{AVG(r_{s_x}(f_i), r_{s_y}(f_i))}$$

- Maps the average feature ranks to a non-linear function, emphasizing the contribution of top-ranked feature
- Performs well on synonymy detection and similarity estimation, and SOTA results in thematic fit estimation
- Contribution of lower ranks are negligible

### APSynP for Dense and Low dimensional Vectors:

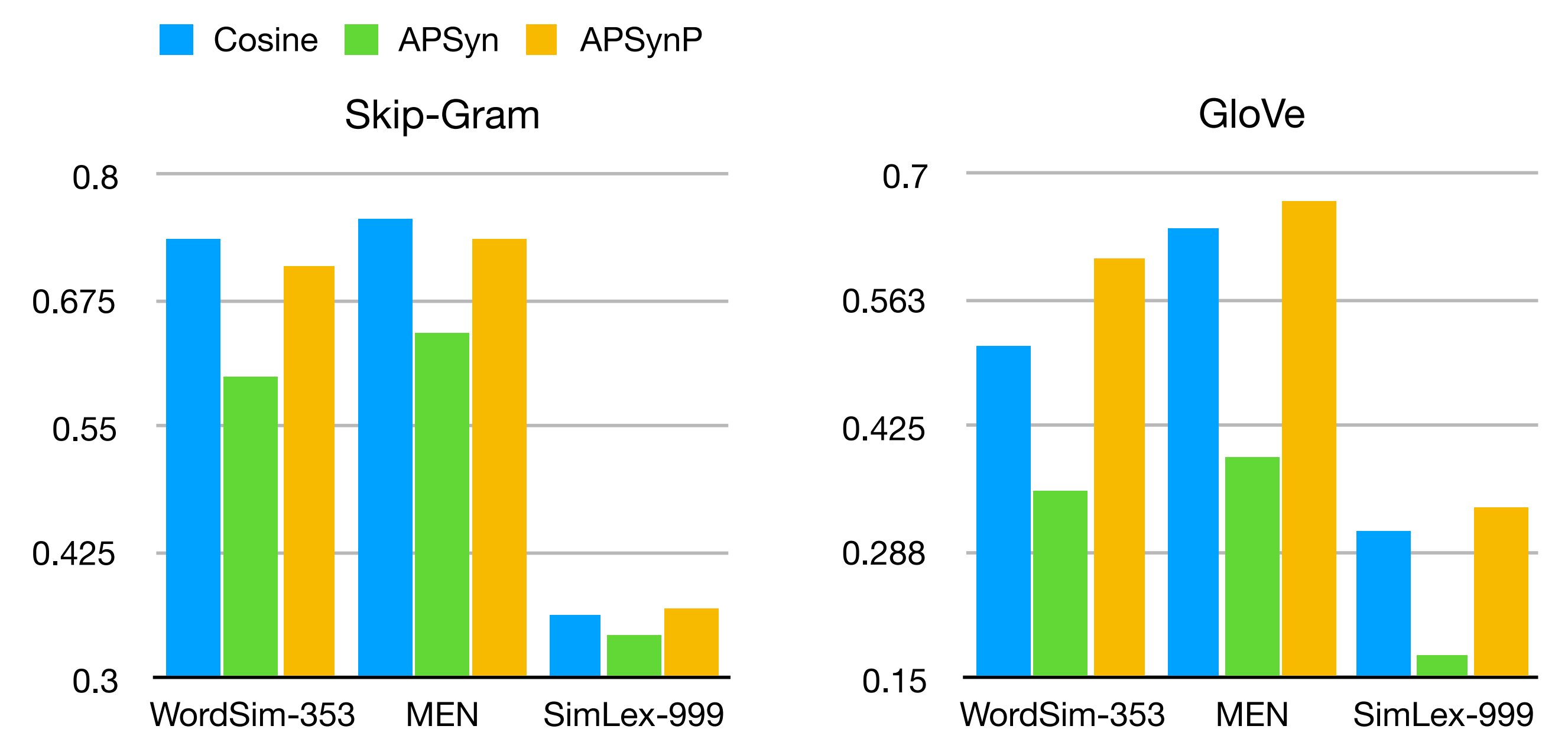
$$APSynP(w_x, w_y) = \sum_{i=0}^{i=|f|} \frac{1}{AVG(r_{s_x}(f_i)^p, r_{s_y}(f_i)^p)}$$

- $N = |f|$  (removing a parameter)
- Adding a smoothing parameter, which can be tuned but tends to be **constant** ( $p=0.1$ ) across all experiments
- Preserving the non-linear weight allocation across the average feature ranks during the summation
- Ranks of all features contributing to the final score



## Similarity Estimation

- **Spearman Correlation** between system-generated scores and human judgments
- **Benchmark:** WordSim-353, MEN, SimLex-999



## Outlier Detection

### Benchmark: 8-8-8 dataset

- 64 sets of 8 words + 1 outlier for the evaluation.

$$OPP = \frac{\sum_{W \in D} \frac{OP(W)}{|W|-1}}{D} \times 100$$

$$Accuracy = \frac{\sum_{W \in D} OD(W)}{D} \times 100$$

- **Pairwise Comparisons:** outlier has the lowest average similarity score with the other words in the cluster
- **Cluster Prototype:** outlier has the lowest similarity score with the average vector of the other N-1 words

	Skip-Gram				GloVe			
	UMBC		Wiki		UMBC		Wiki	
	OPP	Acc.	OPP	Acc.	OPP	Acc.	OPP	Acc.
<i>CC - Cos</i>	92.6	64.1	93.8	70.3	81.6	40.6	91.8	56.3
<b>Pairwise</b>								
<i>APSyn</i>	93.0	67.2	94.0	68.8	78.7	40.6	89.3	53.1
<i>APSynP</i>	<b>94.0</b>	<b>68.8</b>	<b>94.5</b>	<b>73.4</b>	<b>81.8</b>	<b>42.2</b>	<b>92.8</b>	<b>61.0</b>
<b>Prototype</b>								
<i>PT - Cos</i>	93.4	65.6	93.8	68.8	80.3	40.6	90.6	54.7
<i>APSyn</i>	92.6	70.3	91.0	62.5	81.6	40.6	88.7	54.7
<i>APSynP</i>	<b>94.0</b>	<b>70.3</b>	<b>94.9</b>	<b>73.4</b>	<b>82.2</b>	<b>43.8</b>	<b>92.0</b>	<b>60.9</b>

## Contributions

- **APSynP**, a rank-based similarity measure adapted with a smoothing parameter for word embeddings
- Setting  $N=|f|$  and using a constant parameter makes APSynP unsupervised
- Comparable or better performances than cosine and APSyn on similarity estimation, clustering and outlier detection
- Pilot studies suggest that other rank-based metrics can outperform vector cosine in multiple settings

## Reference

- Jose Camacho-Collados and Roberto Navigli. 2016. Find the Word that Does not Belong: A Framework for an Intrinsic Evaluation of Word Vector Representations. In Proceedings of ACL Workshop on Evaluating Vector Space Representations for NLP
- Enrico Santus, Emmanuele Chersoni, Alessandro Lenci, Chu-Ren Huang, and Philippe Blache. 2016a. Testing Apsyn against Vector Cosine on Similarity Estimation. In Proceedings of PACLIC.