

# Semantic Similarity Using Word2Vec

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# **Abstract**

Finding semantic similarity touches wide range of problems in real life scenarios. It may be used to summarize the news articles, to find similarities in the reviews of some product so as to get pros and cons of a product, etc. For doing semantic similarity, the problem with Indian language is that there is no proper parser available available for the Indian languages. Hence we proceed through WordVector implementation.

#### Motivation

- No proper parser available for Indian language.
- Local language user makes up to large user base on the internet.

	Internet		user	Penetration		
ALL INDIA	269Mn	127Mn	47%			
URBAN INDIA	188Mn	81Mn	51%	No Yes 57% 43%		
RURAL INDIA	81Mn	46Mn	41%	No Yes 43% 57%		

# Dataset

Hindi news dataset of about 300 news articles

# Project Goals/Objectives

- Collect and create dataset in Hindi language.
- Word level similarity using WordVector.
- Sentence level similarity using word level similarity.
- Semantic Vector for sentences.
- Being able to merge news articles using the above sentence similarity

# Methods/Process

The approach is similar to Yuhua Li et al[2] for "Sentence Similarity Based on Semantic Nets and Corpus Statistics".

Get hierarchical knowledge base by using Word2Vec and then getting a raw word similarity based on common synsets. We take the joint word set of the two sentences that we have to compare then we make an n-dimensional vector each for sentence T1 and T2 where n = #words in T1 + #words in T2 followed by calculating weights of each dimension as:-

$$\begin{cases} S1[i] = \frac{1}{similarity\ score} & \forall w[i]\ in\ the\ joint\ word\ set \in T1\\ \forall w[i]\ in\ the\ joint\ word\ set \notin T2\\ \end{cases} \\ \begin{cases} S2[i] = \frac{1}{similarity\ score} & \forall w[i]\ in\ the\ joint\ word\ set \in T2\\ \forall w[i]\ in\ the\ joint\ word\ set \notin T2\\ \end{cases} \\ S_s = \frac{\mathbf{s_1} \cdot \mathbf{s_2}}{\|\mathbf{s_i}\| \cdot \|\mathbf{s_2}\|}. \\ \begin{cases} R1[w[i]] = \frac{index[w[i]]}{index[similar\ word]} & \forall w[i]\ in\ the\ joint\ word\ set \notin T1\\ \forall w[i]\ in\ the\ joint\ word\ set \notin T2\\ \end{cases} \\ \begin{cases} R2[w[i]] = \frac{index[w[i]]}{index[similar\ word]} & \forall w[i]\ in\ the\ joint\ word\ set \notin T2\\ \forall w[i]\ in\ the\ joint\ word\ set \notin T2\\ \end{cases} \\ S_r = 1 - \frac{\|\mathbf{r_1} - \mathbf{r_2}\|}{\|\mathbf{r_1} + \mathbf{r_2}\|}. \\ \end{cases}$$

Once we have the results, recognize the threshold to classify similar and dissimilar by ROC curve.

The binary classifier can then be used to merge the news articles based on what are similar and dissimilar sentences in the articles.

	RAM	keeps	things	being	worked	with	The	CPU	uses	as	а	short-term	memory	store
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š	1	1	4	1	1	1	0	0.2802	0.4433	0	0	0	0.8147	0.8147
Weight	/(RAM)		I(things) I(things)			/(with) /(with)	I STATE OF THE STA	/(CPU) /(things)	/(uses) /(things)	l(as) l(as)	I(a) I(a)	/(short-term) /(short-term)	/(memory) /(RAM)	/(store) /(RAM)

#### Results Similarity S.No. Similarity बीजेपी की बिहार में करारी हार महागठबंधन ने बीजेपी को बिहार में 100 सीटों से हराया नरेंद्र मोदी अपने मंत्रियों के बड़बोलेपन से परेशान 0.558449 नरेंद्र मोदी का मंत्रियों के साथ विदेश दौरा टला बीफ की अफवाह पर भीड़ ने एक इंसान को जान से मार उमा जी कहती है नरेंद्र मोदी विनाश पुरुष नरेंद्र मोदी के गुजरात में विकास नहीं सिर्फ दंगे हुए हैं 0.319521 दादरी में भीड़ ने एक मासूम को जान से मारा छपरा के मढ़ौरा में प्रधानमंत्री नरेंद्र मोदी की रैली बिहार चुनाव के मद्देनज़र पार्टियों की रैली जोर-शोर पर 0.436039 यह एक जूता है मैं कल जूता लेने गया नीतीश कुमार सन्निपात रोगी की तरह परस्पर विरोधी बातें कर रहे हैं आज NLP की कक्षा है नीतीश कुमार प्रधानमंत्री नरेंद्र मोदी की यात्रा से भी परेशान 0.4790112 यह एक जूता है

# Conclusions/Recommendations

An approach was given to calculate sentence similarity score with the help of Word2Vec. Sentences having similar words tend to give more score irrespective of the overall sentence meaning, hence considered the impact of word order on sentence meaning. However, still without proper word disambiguation, similarity score was much higher then human score in some cases like in example 5 and vice-versa as in example 6. However in real life scenarios, sentences having many similar words tend to convey same meaning most of the time.

# References

- Yuhua Li et al "Sentence Similarity Based on Semantic Nets and Corpus Statistics" (AUGUST 2006) (IEEE TRANSACTIONS ON KNOWLEDGE AND DATA ENGINEERING, VOL. 18, NO. 8) <a href="http://www.aicit.org/AISS/ppl/AISS1666PPL.pdf">http://www.aicit.org/AISS/ppl/AISS1666PPL.pdf</a>
- Poster Template:- <a href="https://goo.gl/dd5j74">https://goo.gl/dd5j74</a>
- Userbase information: <a href="http://goo.gl/vqnhNU">http://goo.gl/vqnhNU</a>
- Userbase information: <a href="http://goo.gl/6nTxEP">http://goo.gl/6nTxEP</a>
- List of hindi stop words:- <a href="https://goo.gl/a0gRzd">https://goo.gl/a0gRzd</a>