



Connecting the Dots Between News Articles



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Motivation

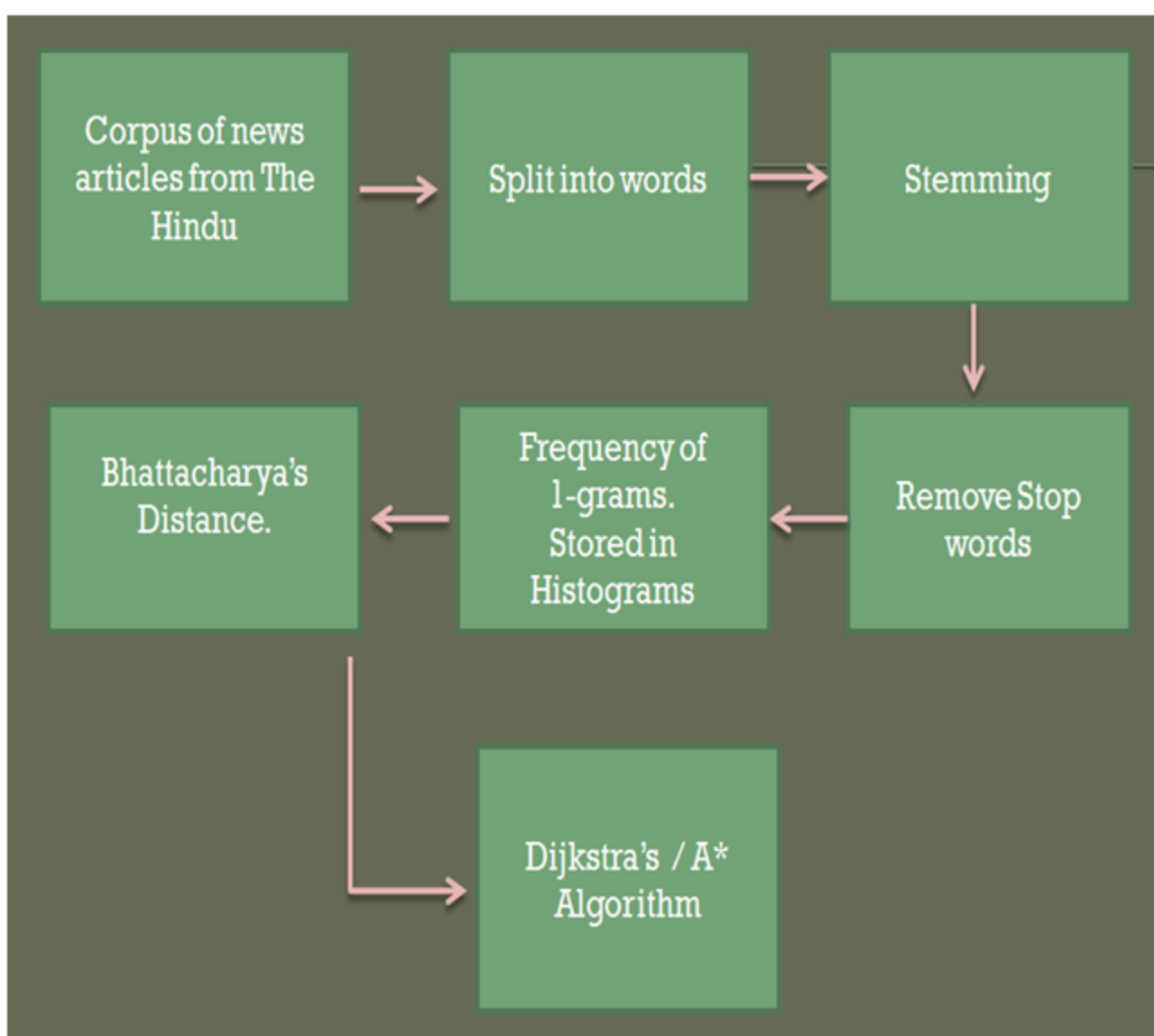
- Tackling Information Overload
- Navigate Between Topics
- Seeing the bigger picture
- News Browsing : Primary use of Internet



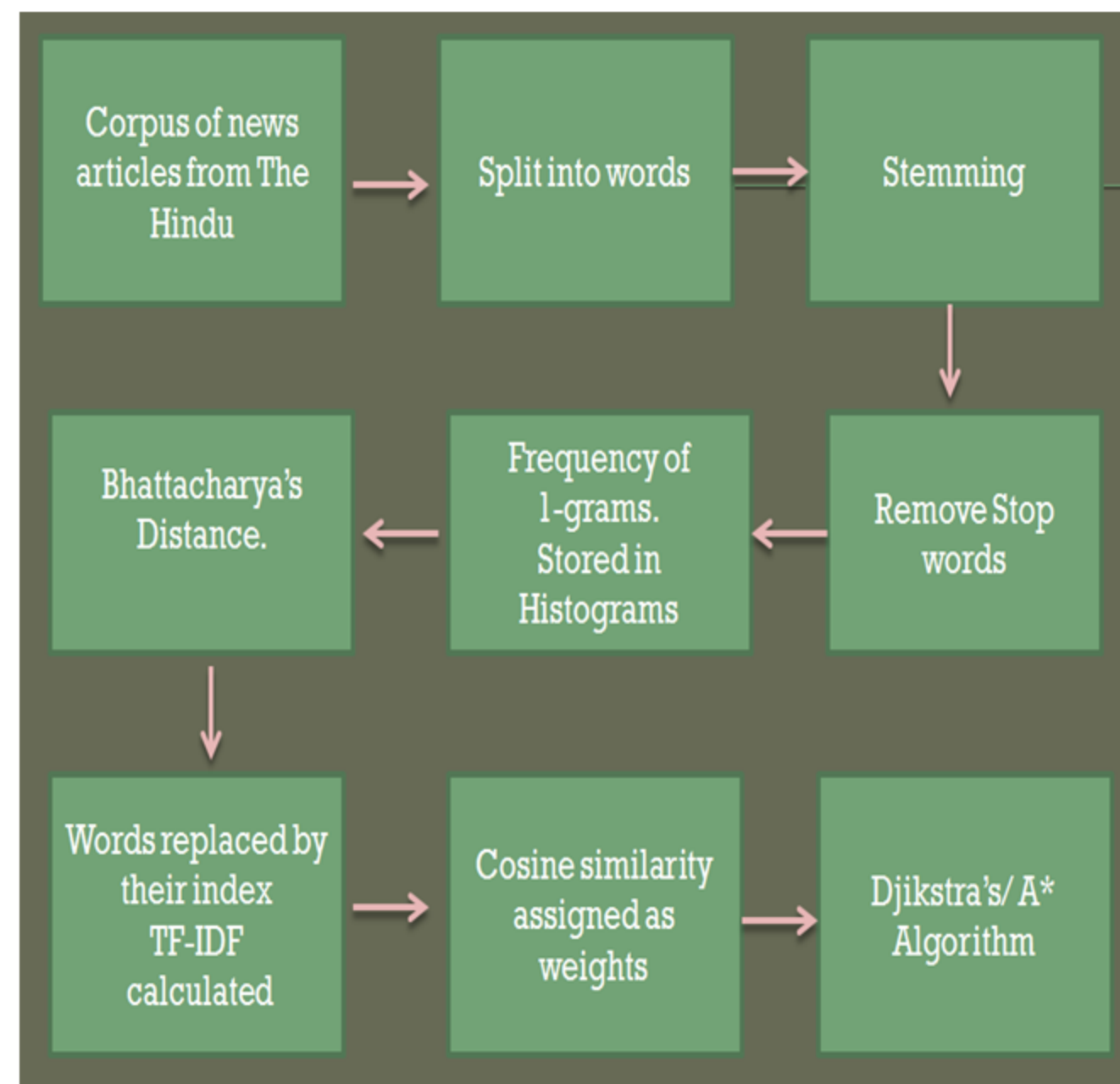
Related work

- Dafna Shahaf and Prof. Carlos Guestrin used Bipartite Graph and notion of influence (2010)
- Hossain, Gresock, Edmond, helm, potts and ramakrishnan used *Soergel Distance* and *A* algorithm* (2012)
- Shahaf, Guestrin and Horvitz used *m-coherence* to compare similarity between articles (2012)
- R. Nallapati, A. Feng, F. Peng, and J. Allan. Event threading within news topics. In *CIKM '04, 2004*
- Q. Mei and C. Zhai. Discovering evolutionary theme patterns from text: an exploration of temporal text mining. In *KDD '05, 2005*

Approach 1



Approach2



Coherence

We evaluate our results using *coherence* as a parameter. We used two different methods to calculate *coherence*.

$$\bullet \text{Coherence1}(d_1, \dots, d_n) = \min_{i=1, \dots, n-1} \sum_w 1(w \in d_i \cap d_{i+1})$$

This is the minimal transition score.

Ref: Dafna Shahaf and Prof. Carlos Guestrin : Connecting the dots between news articles. *ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD) 2010*.

$$\bullet \text{Coherence2}(d_1, \dots, d_n) = \min_{i=1, \dots, n-1} \{ \text{cosine-similarity}(d_i, d_{i+1}) \}$$

This measure guarantees a value between 0 and 1, and hence, gives the percentage similarity between documents.

Tools And Resources used

- Gensim: www.radmehurek.com/gensim
- NetworkX: <http://networkx.github.io/documentation/latest/index.html>
- matplotlib: <http://matplotlib.org/>

Data Set

We have created our own corpus of 1000 articles downloaded from The Hindu spread over around 40 topics. Some of the topics are Nitish Katara murder case, Jessica Lal murder case, US presidential elections, India's world cup triumph etc. We expect it to be very useful for future projects related to Data-Mining. Following is a snapshot of our data set

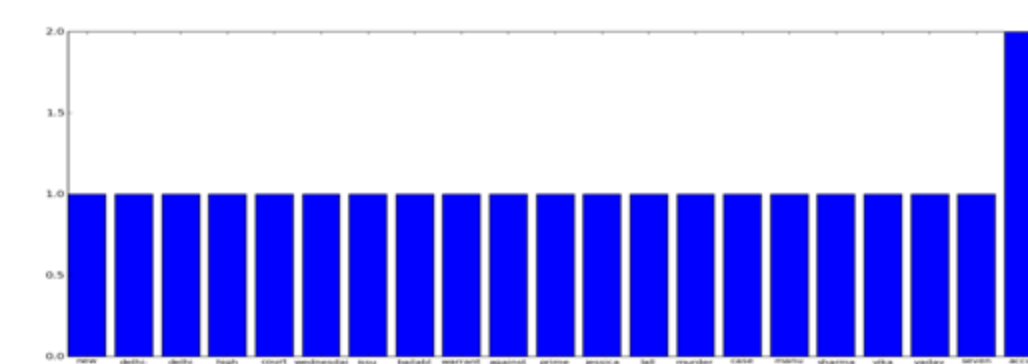
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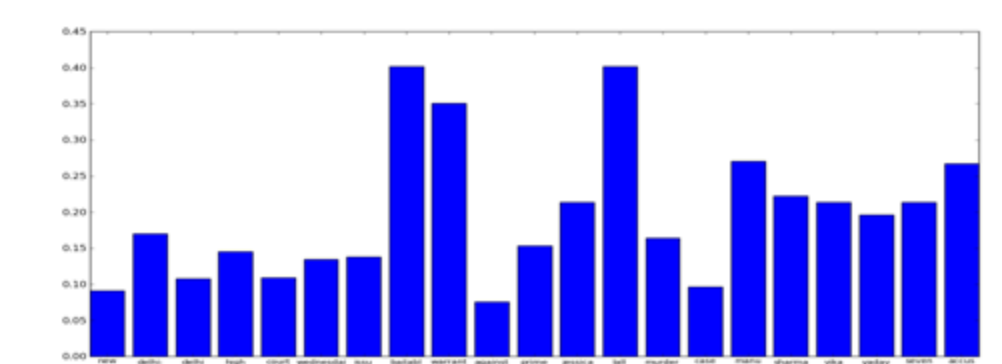
Results

Weight	Algorithm	Chain	Coherence1	Coherence2
Bhatt. Dist	Dijkstra	123-117-105-113-111	11 (105-113)	16.05% (105-113)
Cosine-Similarity	Dijkstra	123-103-102-110-111	16 (110-111)	22.88% (110-111)
Bhatt. Dist	A*	123-118-103-117-111	18 (117-111)	19.24% (118-103)
Cosine-similarity	A*	123-103-117-110-111	16 (110-111)	20.45% (117-110)

Comparison between Word-count and TF-IDF

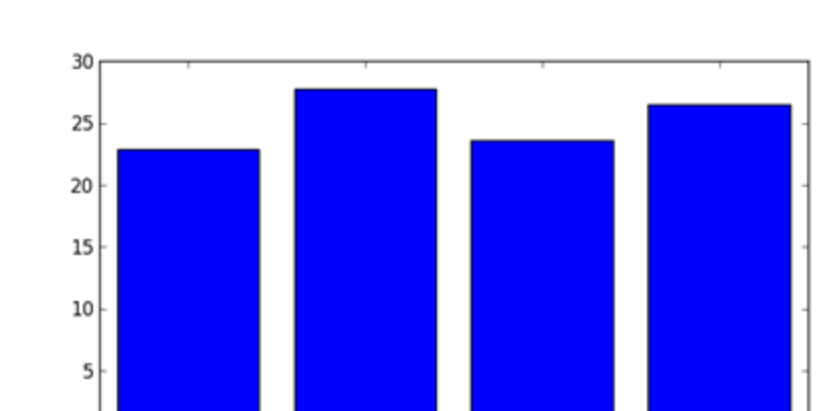
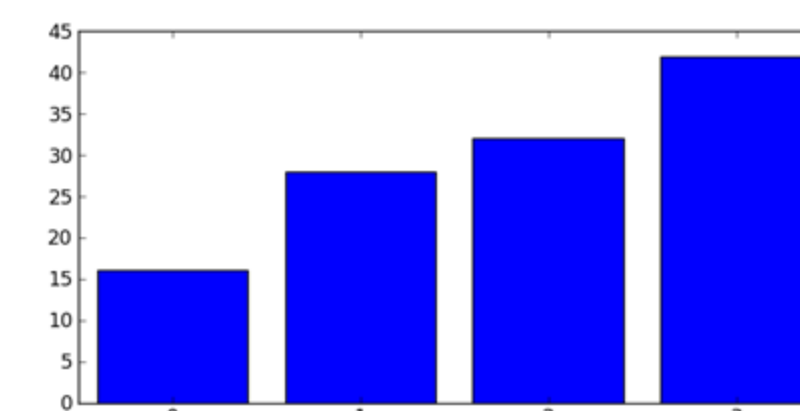
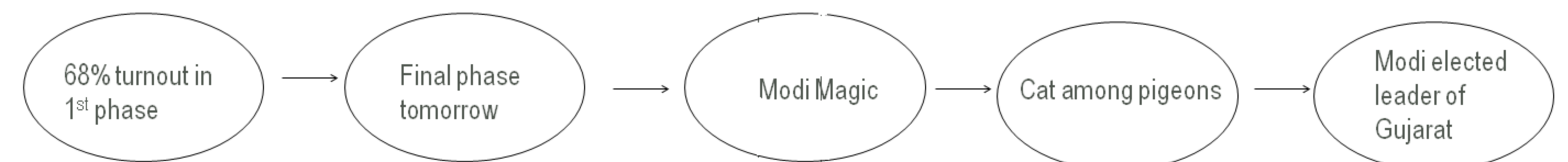


unigram

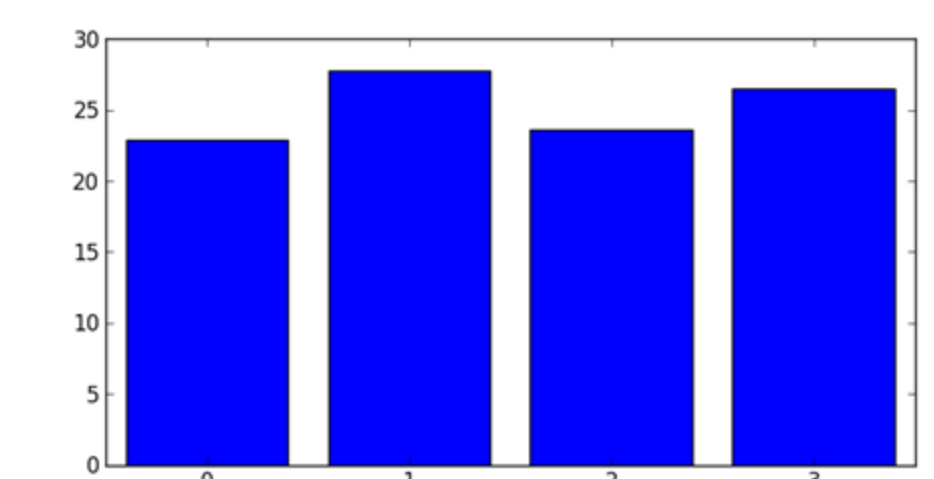
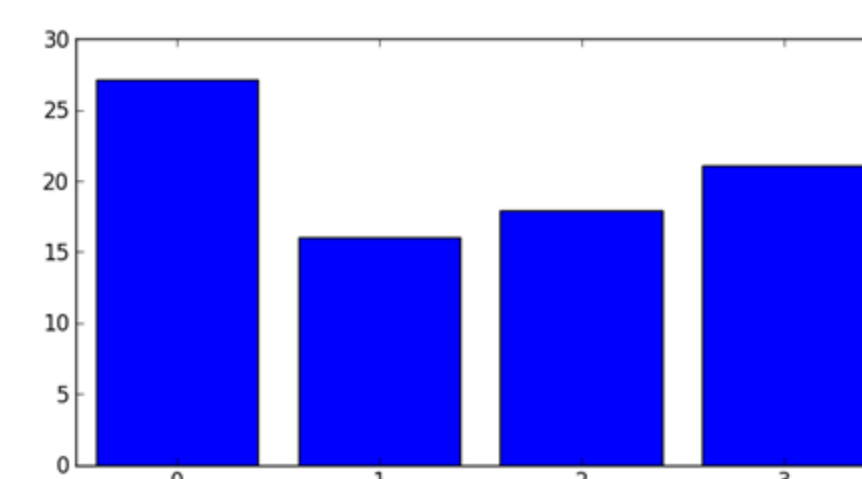


TF-IDF

Comparison between Coherence1 and Coherence2



Comparison using coherence2 between Bhattacharya's Distance and Cosine Similarity



Conclusion

- TF-IDF is better than unigram
- *Coherence2* is a better evaluating parameter than *coherence1*
- Cosine similarity using TF-IDF gives more coherent chain than Bhattacharya's distance.

Future Work

- Notion of *m-coherence* can be used instead of *coherence* for better evaluation of results.
- Using different corpora (other than News articles) may help in important scientific discoveries.