



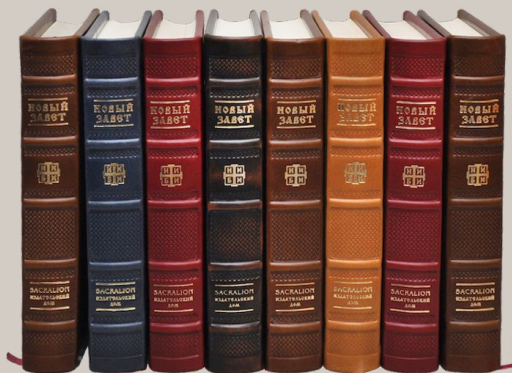
BookENDER

A book recommender

By Veena Kumar

Data

- ❖ Original dataset contains plot summaries for 16,559 books (taken from Wikipedia), author, and genre(s)
- ❖ Final data: **8,825 books**



Model

1. **Tokenize** words from book summaries



2. Run every word in each document through

`word2vec`'s model to get a **score** for the word

- 3,000,000 pre-trained words from Google News dataset
- Vectors are 300 dimensions (i.e 'features') x # of words



3. Take mean of every dimension

- Vectors become 300 x 1 for each document



4. Find **cosine similarity** between documents

5. Donezo! Almost...

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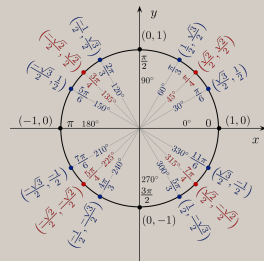
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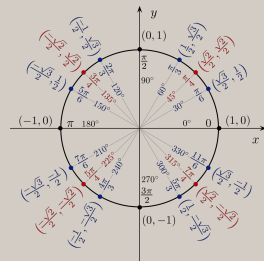
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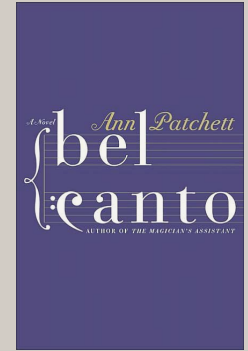
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Example



- ❖ **Book:** Bel Canto
- ❖ **Genres:** Thriller, Fiction, Romance
- ❖ **Summary of the summary:** A terrorist organization attempts to assassinate the President during a party thrown by the Vice President, however the President was not in attendance. The terrorists decide instead to take important guests hostage in order to leverage a ransom. The rest of the story is about the romances that develop during the hostage situation.

Example

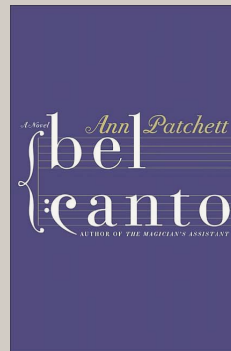
- ❖ Enter book title

`bookender('Bel Canto')`



- ❖ Function

```
def bookender(title):  
    title_index = books[books['Book Title']=='{}'.format(title)].index.tolist()[0]  
    top_5_values = dists[title_index][:].sort_values(ascending=False)[:6]  
    values_indexes = dists[title_index][:].sort_values(ascending=False).index.tolist()  
    top_5_indexes = values_indexes[:6]  
    top_5_books = books.iloc[top_5_indexes]  
    print top_5_values  
    return top_5_books
```





Example

Results



Scores	Book Title	Author	Genre
1.000000	Bel Canto	Ann Patchett	Thriller, Fiction, Romance
0.887298	We the Living	Ayn Rand	Autobiography, Historical
0.883713	Omerta	Mario Puzo	Thriller, Crime Fiction
0.883078	Farnham's Freehold	Robert A. Heinlein	Science Fiction
0.882075	Twilight at the Well of Souls	Jack L. Chalker	Science Fiction,
0.881482	Becoming Madame Mao	Anchee Min	Historical



Future Enhancements

- ❖ Add extra layer to match by genre
- ❖ Sentiment analysis
- ❖ [Website!](#)
 - Links to purchase book
 - Reviews
 - Scrape Wikipedia to obtain summary, if book is not in database



References

Data:

- ❖ David Bamman and Noah Smith (2013), "New Alignment Methods for Discriminative Book Summarization"

Google's word2vec:

- ❖ Tomas Mikolov, Kai Chen, Greg Corrado, and Jeffrey Dean. Efficient Estimation of Word Representations in Vector Space. In Proceedings of Workshop at ICLR, 2013.
- ❖ Tomas Mikolov, Ilya Sutskever, Kai Chen, Greg Corrado, and Jeffrey Dean. Distributed Representations of Words and Phrases and their Compositionality. In Proceedings of NIPS, 2013.
- ❖ Tomas Mikolov, Wen-tau Yih, and Geoffrey Zweig. Linguistic Regularities in Continuous Space Word Representations. In Proceedings of NAACL HLT, 2013.

Questions?

