

# NLP + TDA

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# Have you seen this on Grammarly?

*Your text is likely to be understood by a reader who has at least a 9th-grade education (age 15). Aim for the score of at least 60-70 to ensure your text is easily readable by 80% of English speakers.*

The screenshot shows a Grammarly report with three main sections: Performance, Word Count, and Readability. The Performance section features a circular progress indicator with the score 86. The Word Count section lists 689 characters, 105 words, and 7 sentences, along with reading and speaking times. The Readability section includes horizontal bar charts for word and sentence length, both marked as 'Above average', and a readability score of 53. A footer contains a 'DOWNLOAD PDF REPORT' link and a 'Close' button.

## Performance

Text score: 86 out of 100. This score represents the quality of writing in this document. You can increase it by addressing Grammarly's suggestions.

**86**

## Word Count

Characters	<b>689</b>	Reading time	<b>25 sec</b>
Words	<b>105</b>	Speaking time	<b>48 sec</b>
Sentences	<b>7</b>		

## Readability

Metrics compared to other Grammarly users

Word length	<b>5.3</b>		Above average
Sentence length	<b>15</b>		Above average
Readability score	<b>53</b>		

Your text is likely to be understood by a reader who has at least a 9th-grade education (age 15). Aim for the score of at least 60-70 to ensure your text is easily readable by 80% of English speakers.

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# Semantic *tie-backs* in a text document

## Similarity Filtration (SIF).

1.  $D_{max} = \max D(x_i, x_j), \forall i, j = 1 \dots n$
2. **FOR**  $m = 0, 1, \dots, M$
3.     Add  $VR \left( \frac{m}{M} D_{max} \right)$  to the filtration
4. **END**
5. Compute persistent homology on the filtration

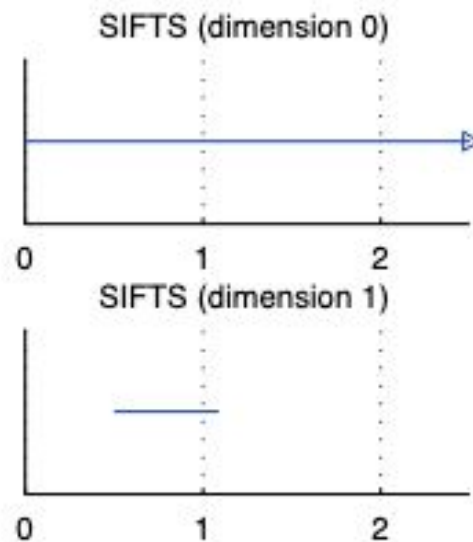
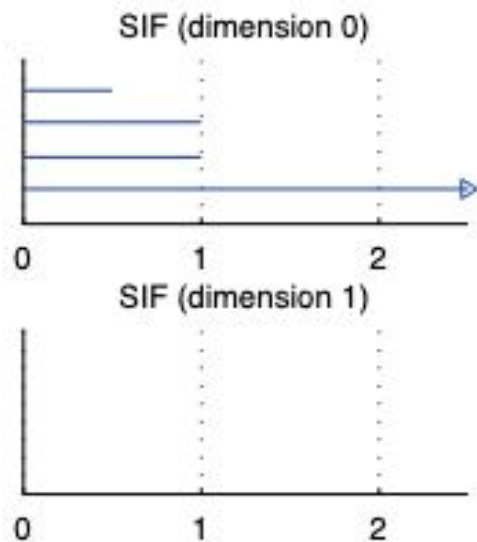


## Similarity Filtration with Time Skeleton (SIFTS).

0.  $D(x_i, x_{i+1}) = 0$  for  $i = 1, \dots, n - 1$



# Semantic *tie-backs* in a text document



# Semantic *tie-backs* in a text document - *Nursery Rhymes*

Euclidean distance between sentence-level bag-of-words count vectors

Filtrations has  $M = 100$  steps

*The itsy bitsy spider climbed up the waterspout.*

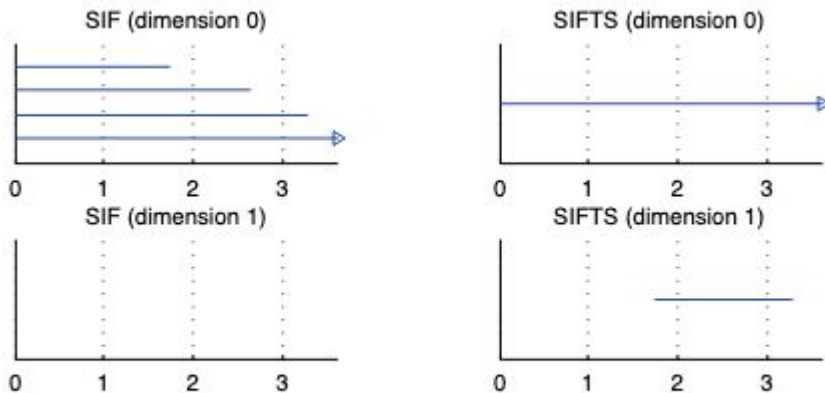
*Down came the rain*

*And washed the spider out.*

*Out came the sun*

*And dried up all the rain*

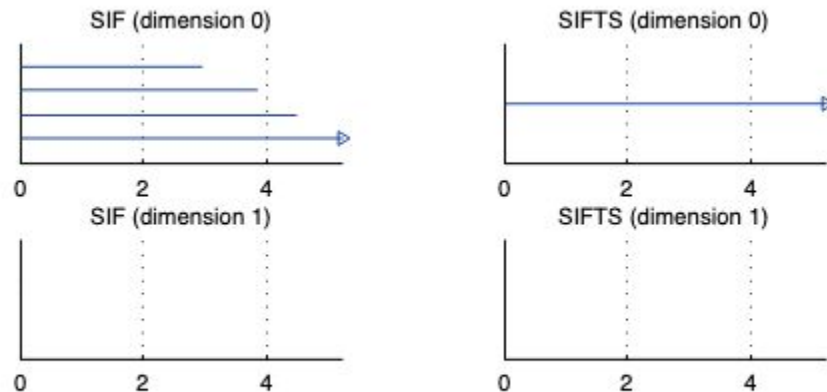
*And the itsy bitsy spider climbed up the spout again.*



(a) Itsy Bitsy Spider

# Semantic *tie-backs* in a text document - *Nursery Rhymes*

*Row, row, row your boat  
Gently down the stream  
Merrily merrily, merrily, merrily  
Life is but a dream*



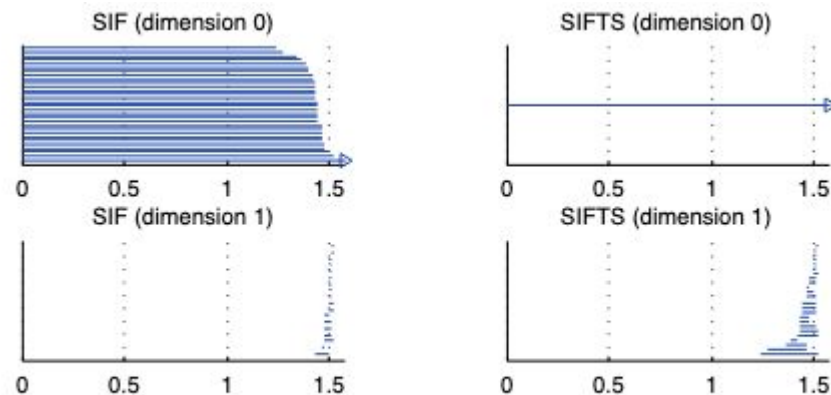
(b) Row Row Row Your Boat

# Semantic *tie-backs* in a text document - *Longer Documents*

Penn Treebank tokenization,  
case-folding, punctuation removal,  
and SMART stopwords removal

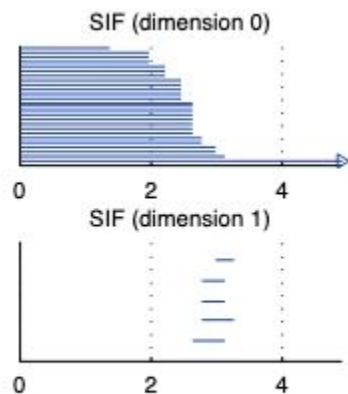
Each text unit is converted to a  
tf.idf vector

$$D(x_i, x_j) = \cos^{-1} \left( \frac{x_i^\top x_j}{\|x_i\| \cdot \|x_j\|} \right).$$

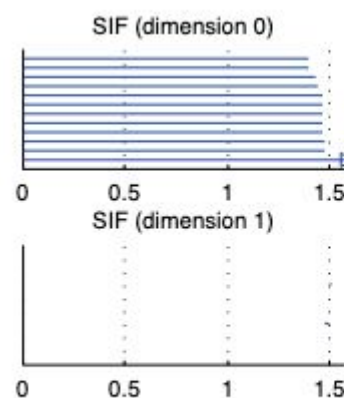
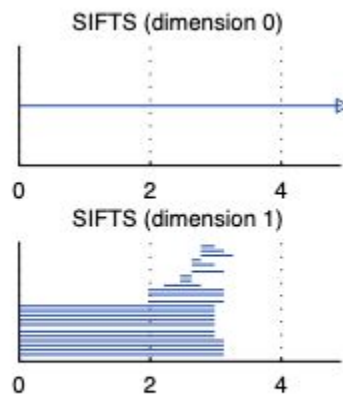


(d) The Emperor's New Clothes

# Semantic *tie-backs* in a text document - *Longer Documents*



(c) London Bridge



(f) Alice in Wonderland



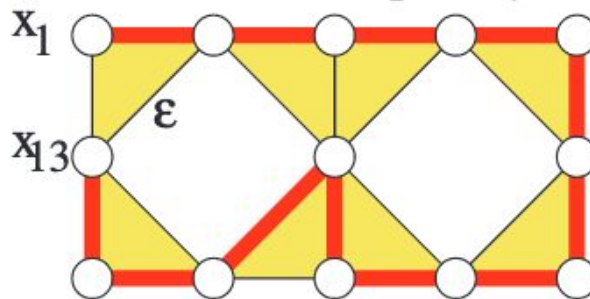
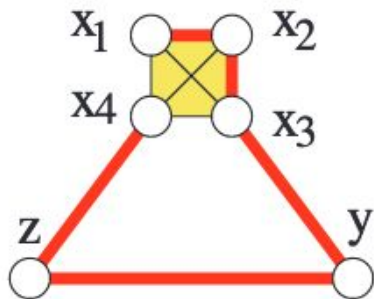
## Semantic *tie-backs* in a text document - *Observations*

- *Older writers have more 1-homology groups than younger writers*
- $|H_1|$ , the total number of 1st-order persistent homology classes (holes) over the whole range epsilon
  - counting the number of bars
- Epsilon-star - the smallest epsilon when the first hole in  $H_1$  forms

	child	adolescent	adol. trunc.
holes?	87%	100%*	98%*
$ H_1 $	3.0 ( $\pm 0.2$ )	17.6 ( $\pm 0.9$ )*	3.9 ( $\pm 0.2$ )*
$\epsilon^*$	1.35 ( $\pm .02$ )	1.27 ( $\pm .02$ )*	1.38 ( $\pm .01$ )

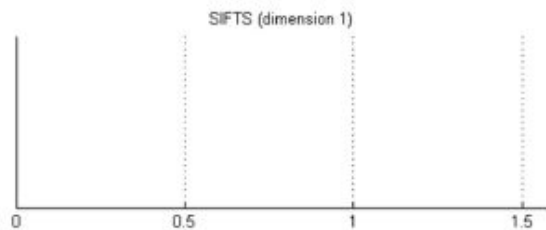
## Semantic *tie-backs* in a text document - *Observations*

- *Homology is not just counting repeated text units*

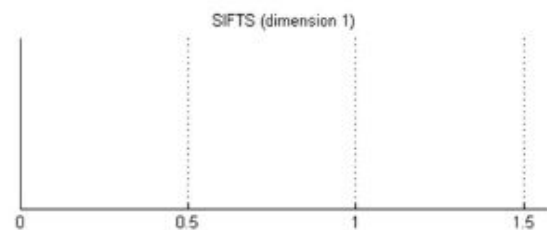


# Movie Genre Detection Using Topological Data Analysis

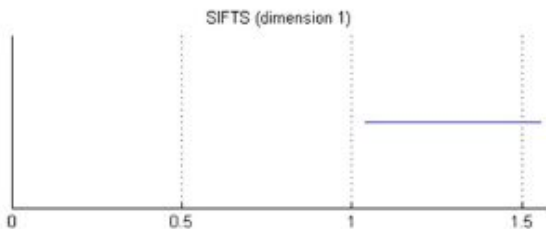
- Predicting movie genres based on plot descriptions



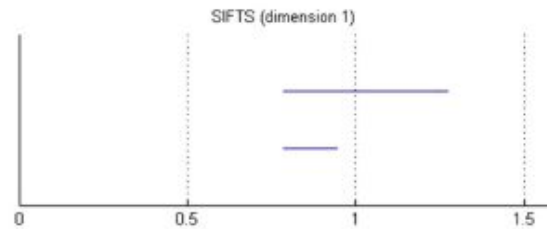
(a) Barcode using Action words



(b) Barcode using Horror words



(c) Barcode using Comedy words



(d) Barcode using Romance words

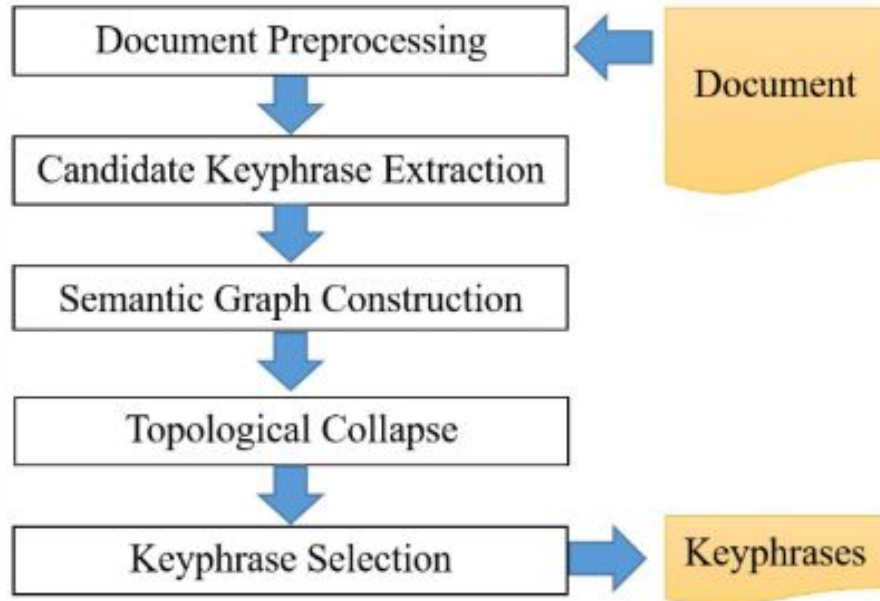
# Movie Genre Detection Using Topological Data Analysis

- Identify the top words using the TF-IDF measure
- Generate Term Frequency matrix for both the movie plots
- Find the 1-dimension holes across the sentences
  - Using the barcode representation of the 1-dimension homology complexes, the program is able to correctly identify the genres of 208 movies with overlapping genres, giving a hit rate of 0.8333%

# A Topological Collapse for Document Summarization

- DoCollapse: vertex dominance criterion
  - vertex  $v$  is dominated by vertex  $w$  if all vertices that share an edge with  $v$  also share an edge with vertex  $w$
- Key Idea: *In a document semantic graph, if one candidate keyphrase dominates another one, then the dominating candidate should convey more important information and thus, is more likely to be a keyphrase*

# A Topological Collapse for Document Summarization



# A Topological Collapse for Document Summarization

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**Algorithm 1** Topological Collapse Algorithm

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```
1:  $\forall v \in V, \text{label}(v) \leftarrow p_v$ 
2:  $V_C \leftarrow V, E_C \leftarrow E$ 
3: while True do
4:    $\text{del} \leftarrow \emptyset$ 
5:   for  $v \in V_c$  do
6:     for  $u \in \mathcal{N}(v)$  do
7:       if  $\mathcal{N}(u) \subseteq \mathcal{N}(v)$  then
8:          $\text{del} \leftarrow u$ 
9:          $\text{label}(v) \leftarrow \text{label}(u)$ 
10:  if  $\text{del}$  is  $\emptyset$  then
11:    Break
12:  else
13:     $E_C \leftarrow \{(u, v) \mid (u, v) \in E_C, u, v \notin \text{del}\}$ 
```

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# A Topological Collapse for Document Summarization



$w$ : {“web service”}

$u$ : {“web service community”}

$v$ : {“scallable web service”}

$w$ : {“web service”, “scallable web service” }

$u$ : {“web service community”}



# A Topological Collapse for Document Summarization

DATA STATISTICS

Dataset	Documents	Tokens	Keys	Candidates	Matches
SemEval-2010	100	9398.6	14.4	841.4	9.59
NUS Corpus	151	8295.1	13.4	809.9	8.87

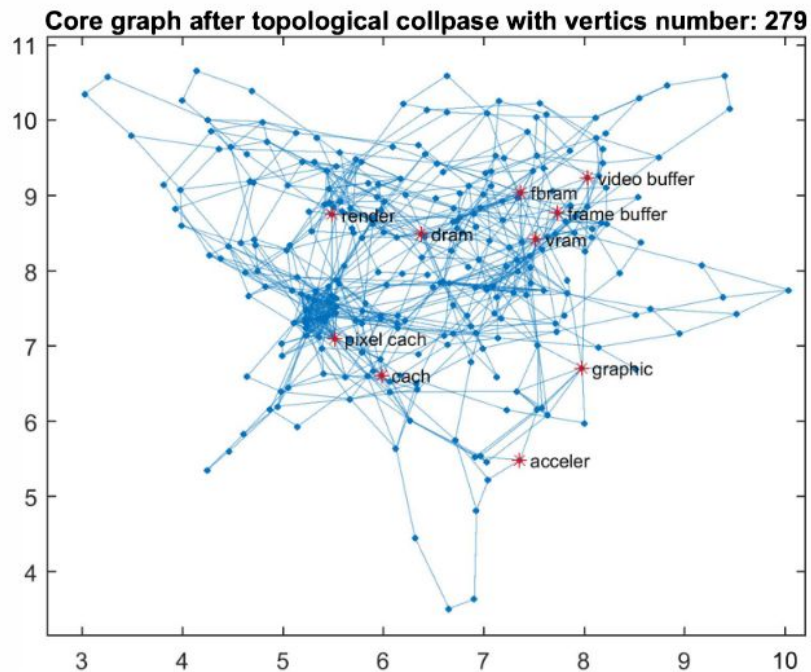
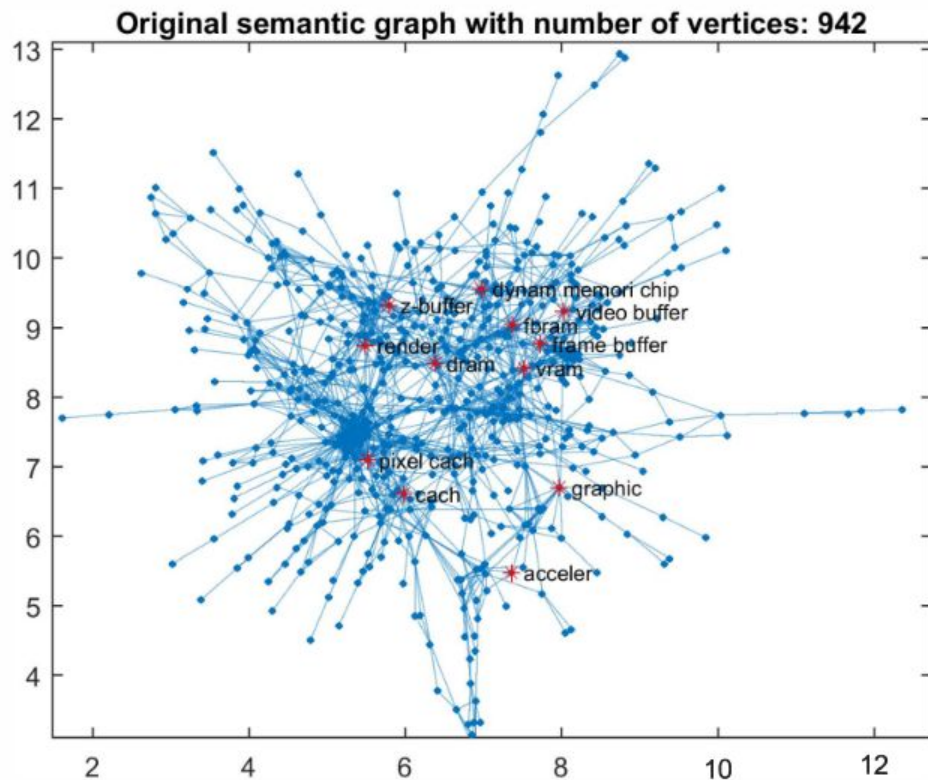
EVALUATION RESULTS ON SEMEVAL-2010

Methods	$M$	$P$	$R$	$F$
TF-IDF	2.32	15.47	16.57	15.85
TextRank	1.51	10.07	10.49	10.17
TopicRank	1.87	12.47	13.54	12.87
DoCollapse	<b>2.52</b>	<b>16.8</b>	<b>18</b>	<b>17.22</b>

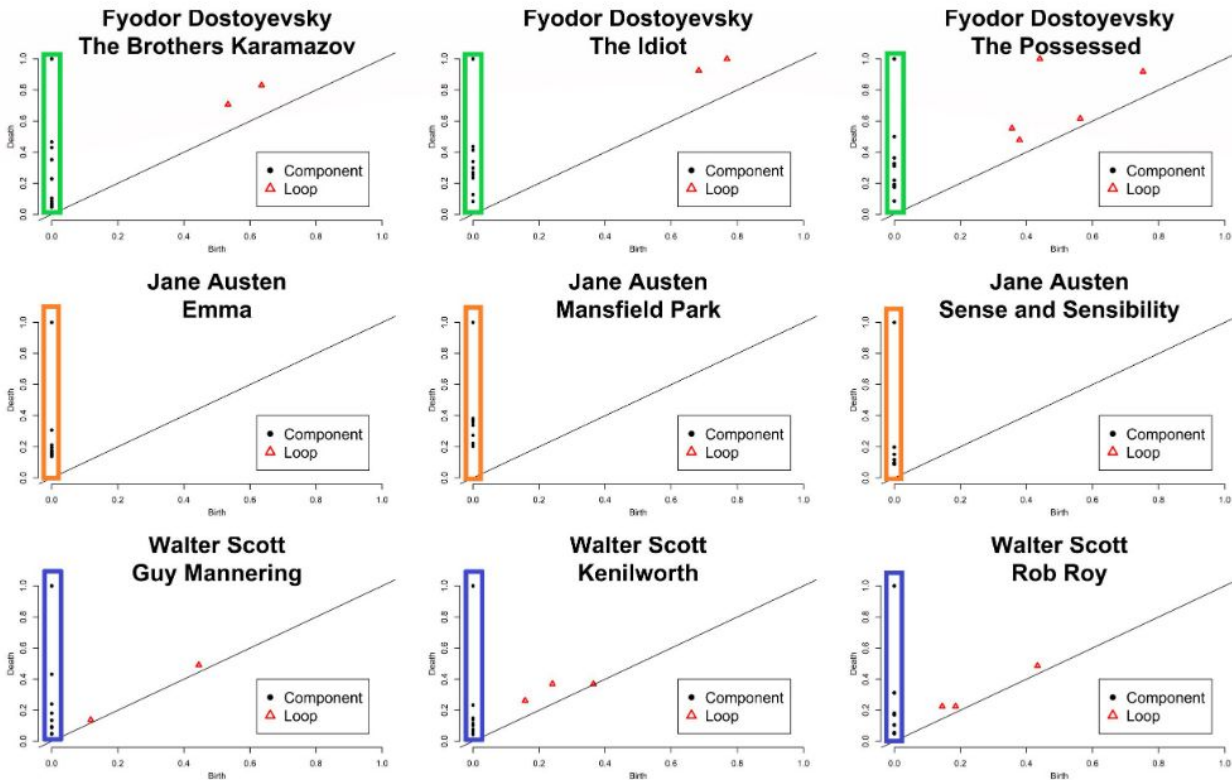
EVALUATION RESULTS ON NUS CORPUS

Methods	$M$	$P$	$R$	$F$
TF-IDF	2.62	17.44	21.61	18.57
TextRank	1.7	11.35	14.25	12.09
TopicRank	1.92	12.8	16.07	13.66
DoCollapse	<b>3.23</b>	<b>21.51</b>	<b>26.13</b>	<b>22.64</b>

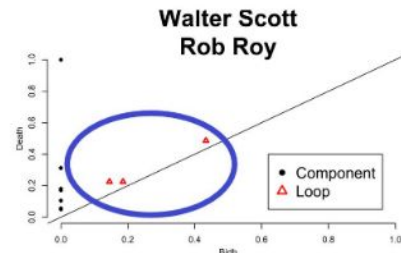
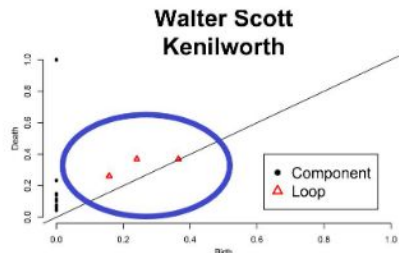
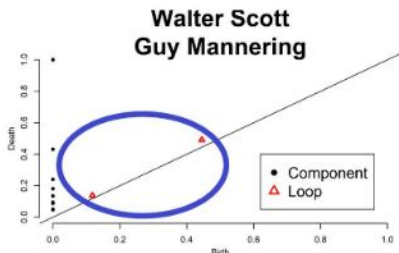
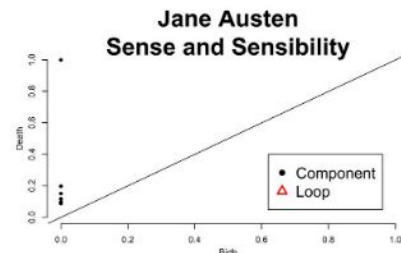
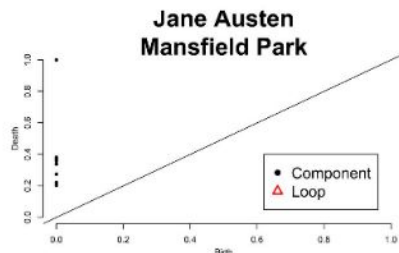
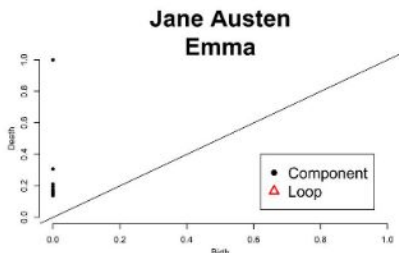
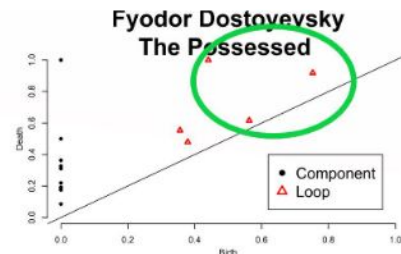
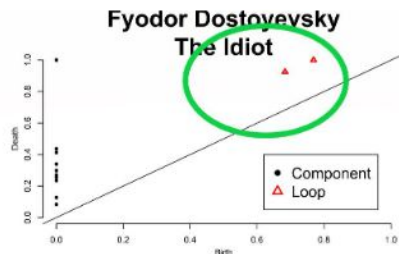
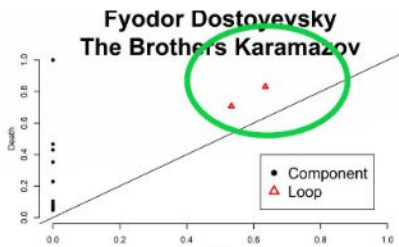
# A Topological Collapse for Document Summarization



# Topological Signature of 19th Century Novelists - a skim



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# Topological Signature of 19th Century Novelists - a skim

- Predicting the author
- Binary Classification (balanced sub-samples)
- 250 times 10-fold cross validation
- 60'000 total predictions
- Using a 5-*NN* algorithm
- Using Wasserstein distance of persistence diagrams

End