

LDA Topic Modeling

Steffen Pielström, Galway 06.12.2018



The Problem...



The Problem...

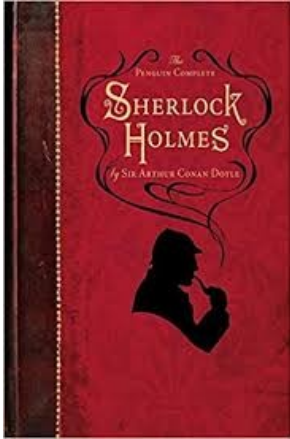


... what are all these documents about?

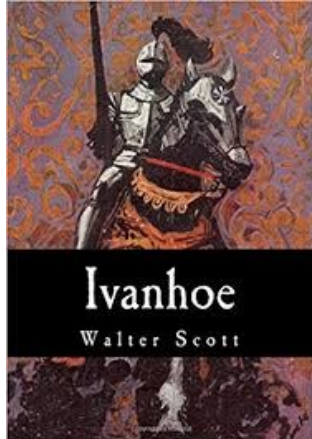
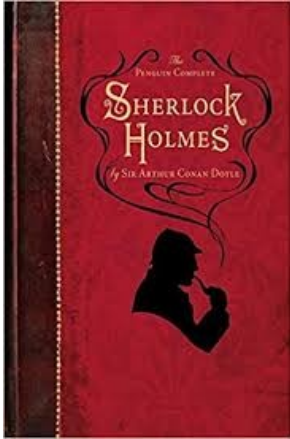
And which ones are interesting for me?

What are they about?

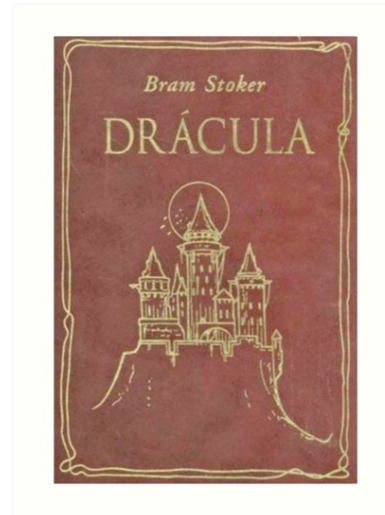
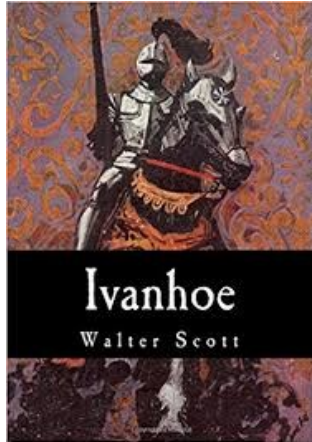
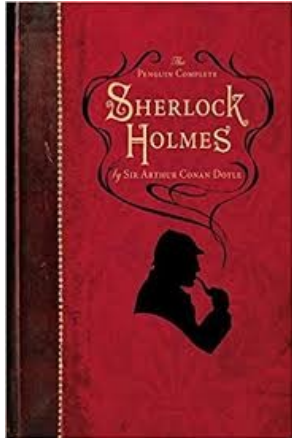
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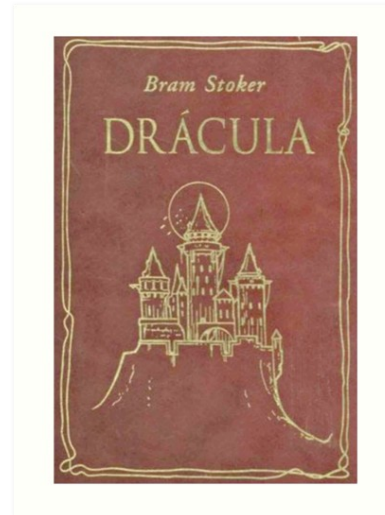
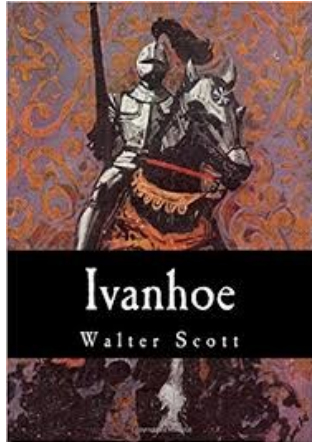
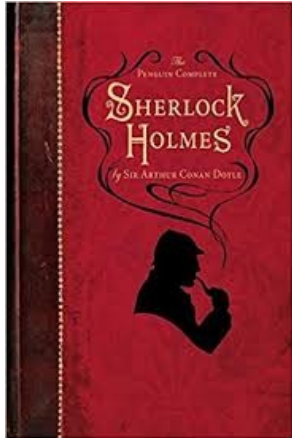
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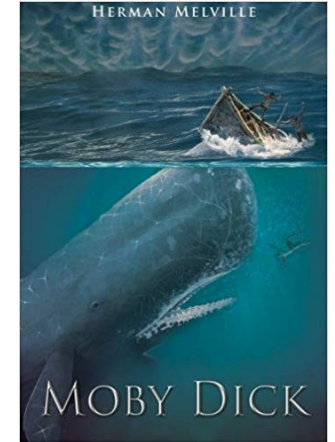
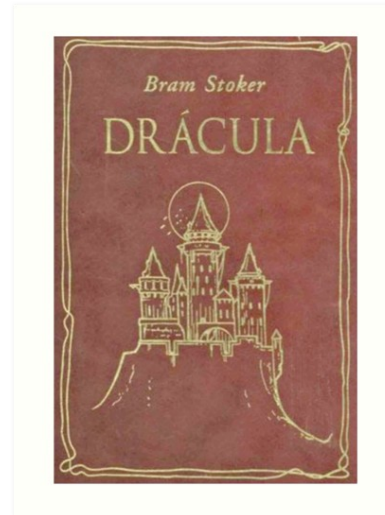
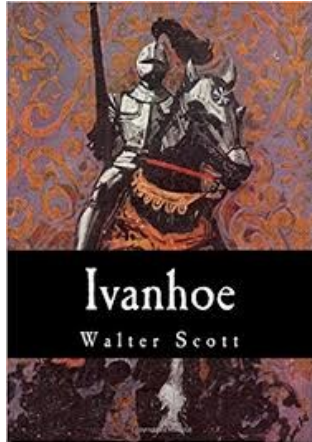
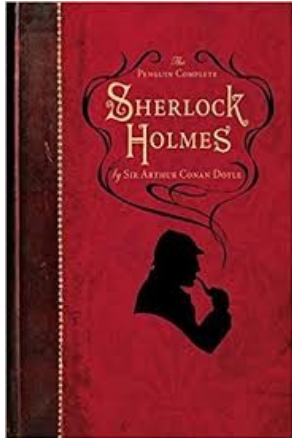
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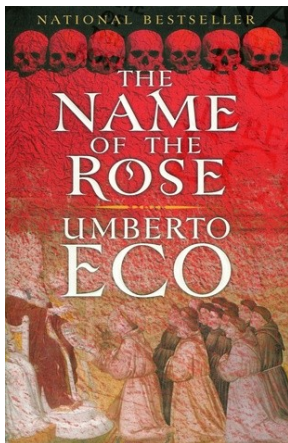
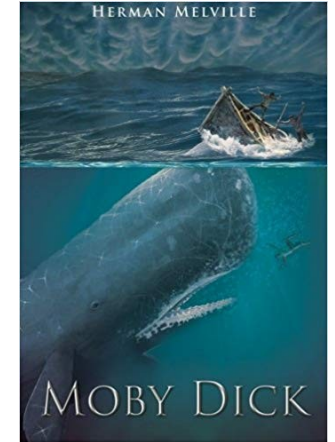
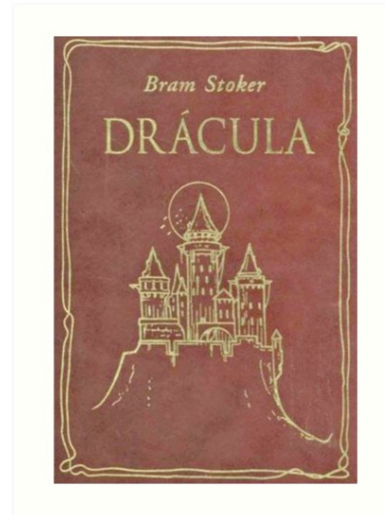
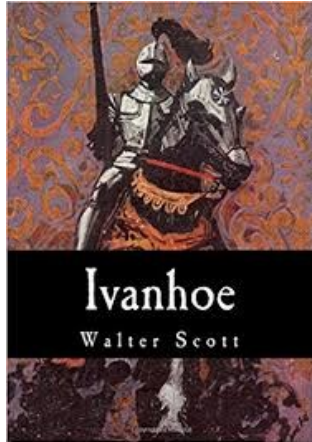
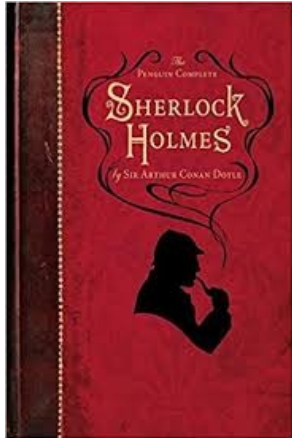
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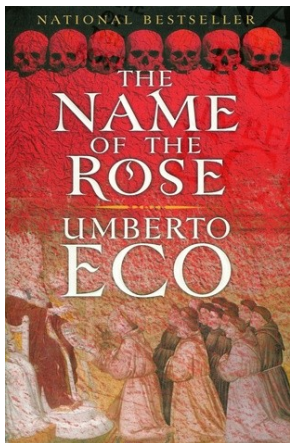
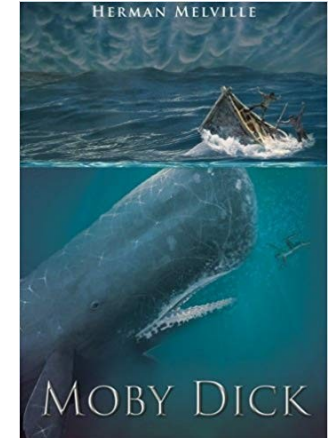
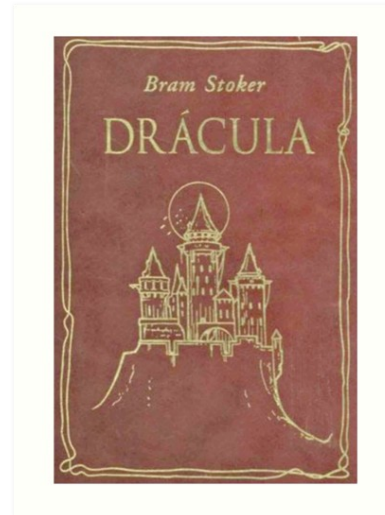
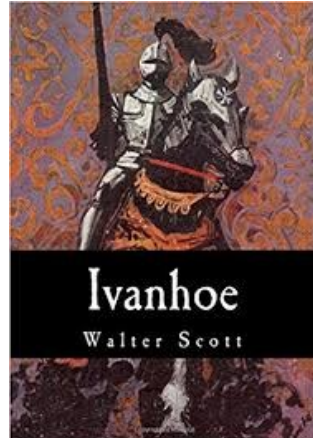
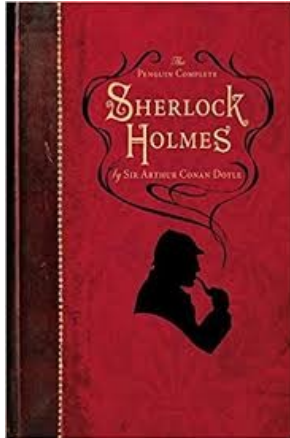
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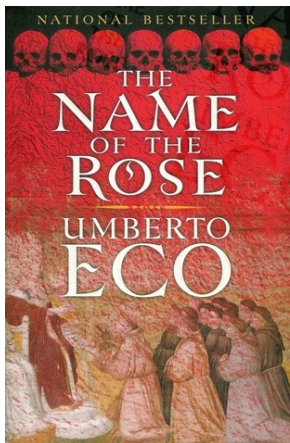
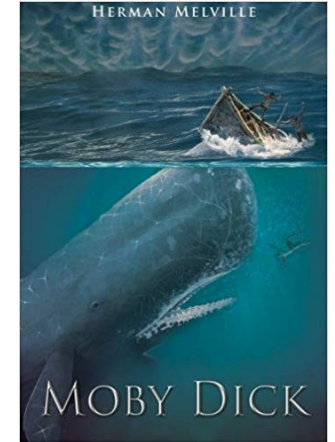
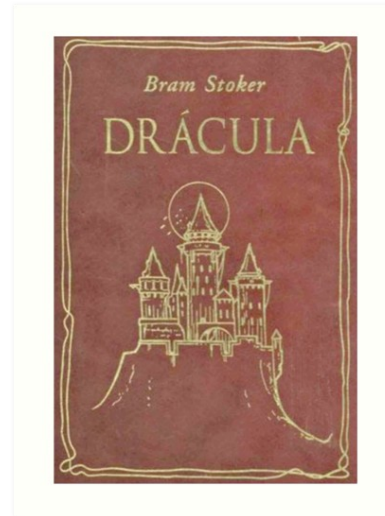
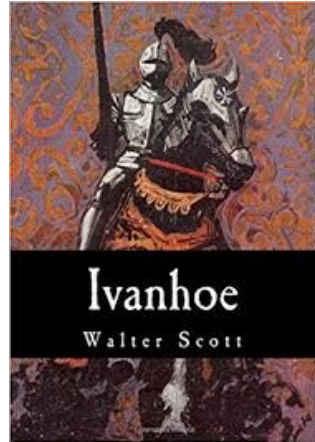
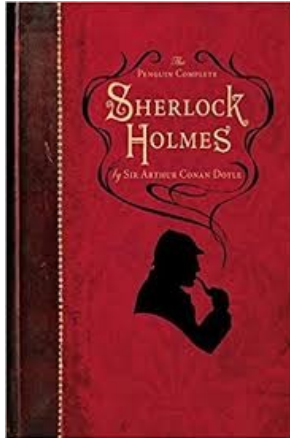
What are they about?



What are they about?



What are they about?



LILY JAMES SAM RILEY JACK HUSTON BELLA HEATHCOTE DOUGLAS BOOTH MATT SMITH CHARLES DANCE LENA HEADEY



Based on the Best-Selling Novel by Jane Austen & Seth Grahame-Smith

IN CINEMAS FEB 2016

LIONSGATE

what are all these documents about?

Some are about more than one thing!

Which ones are interesting?

Should I use

“castle” ...

as a search term?

Which ones are interesting?

Should I use

“castle” ... “fortress” ...

as a search term?

Which ones are interesting?

Should I use

“castle” ... “fortress” ... “stronghold”...

as a search term?

Which ones are interesting?

Should I use

“castle” ... “fortress” ... “stronghold”...

“fastness”

as a search term?

Which ones are interesting?

Should I use

“castle” ... “fortress” ... “stronghold”...

“fastness” or “burgh”

as a search term?

We want...

- Groups of semantically related words

We want...

- Groups of semantically related words
- To see the contribution of each group to each text

We use...

- Distributional semantics

We use...

- Distributional semantics
- Generative model

We use...

- Distributional semantics
- Generative model
- Probabilistic procedure

We can do...

We can do...

- Searching relevant texts

We can do...

- Searching relevant texts
- Text classification

We can do...

- Searching relevant texts
- Text classification
- Diachronic analysis of themes in literary history

We can do...

- Searching relevant texts
- Text classification
- Diachronic analysis of themes in literary history
- Topics as a research topic

MALLET

<http://mallet.cs.umass.edu/>

```
bin/mallet import-dir --input ~/MALLET_test/corpus --output topic-input.mallet --keep-sequence --remove-stopwords
```

```
bin/mallet train-topics --input topic-input.mallet --num-topics 10 --output-state topic-state.gz
```

Introduction - Chromium

Introduction

localhost:8888/notebooks/Code/Python/Projects/Topics/Introductic

jupyter Introduction Last Checkpoint: 04/04/2017 (autosaved)

File Edit View Insert Cell Kernel Help Python 3

Markdown Cell Toolbar: None

Topics - Easy Topic Modeling in Python

The text mining technique Topic Modeling has become a popular statistical method for clustering documents. This notebook introduces a user-friendly workflow, basically containing data preprocessing, an implementation of LDA (Latent Dirichlet Allocation) topic modeling which learns the relationships between words, topics, and documents, as well as a visualization to explore the trained LDA model.

Preparations

The following tutorial will explain how to perform LDA topic modeling with a programming library in Python. If you have not done so yet, please follow the instructions for installing jupyter and all necessary python libraries mentioned in [readme.txt/installation_instructions](#).

Before you start, make sure to have Git and the Topics repository ready to use on your computer:

1. Download and install [Git](#)
2. Open the [command-line interface](#), type `git clone https://github.com/DARIAH-DE/Topics.git` and press Enter
3. Access the new folder `topics` in your package explorer
4. To install the required packages, simply run `setup.py`
5. Install [Jupyter](#) and run it by typing `jupyter notebook` in the command-line
6. Access the folder `Topics` through Jupyter in your browser, open `IntroductionTopics.ipynb` and follow the instructions

1. Preprocessing

1.1. Loading modules

Loading modules from DARIAH-Topics library

First, we have to get access to the functionalities of the library by importing them. For using its functions we use the prefix of the toolbox's submodules (`pre`, `visual` and `mallet`).

```
In [1]: from dariah_topics import preprocessing as pre
from dariah_topics import visualization as visual
from dariah_topics import mallet as mal
```

Load required functions from Gensim

Furthermore, we will need some additional functions from external libraries.

```
In [2]: from gensim.models import LdaModel
from gensim.corpora import MmCorpus
```

Deactivating annoying deprecation warnings

```
In [3]: import warnings
warnings.filterwarnings('ignore', category = DeprecationWarning)
```

Activating inline output in Jupyter notebook

The following line will just tell the notebook to show graphics in the output frames.

```
In [4]: %matplotlib inline
```

1.2. Reading a corpus of documents

Defining the path to the corpus folder

In the present example code, we are using a folder of `txt` documents provided with the package. For using your own corpus, change the path accordingly.

```
In [5]: path = "corpus_txt"
```

List all documents in the folder

We begin by creating a list of all the documents in the folder specified above. That list will tell function `pre_read_from_txt()` (see below) which text

dariah-de.github.io/TopicsExplorer/



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Evelyn Gaier
Blerta Veseli
Thora Hagen
Stefan Krywinski
Thorsten Vitt