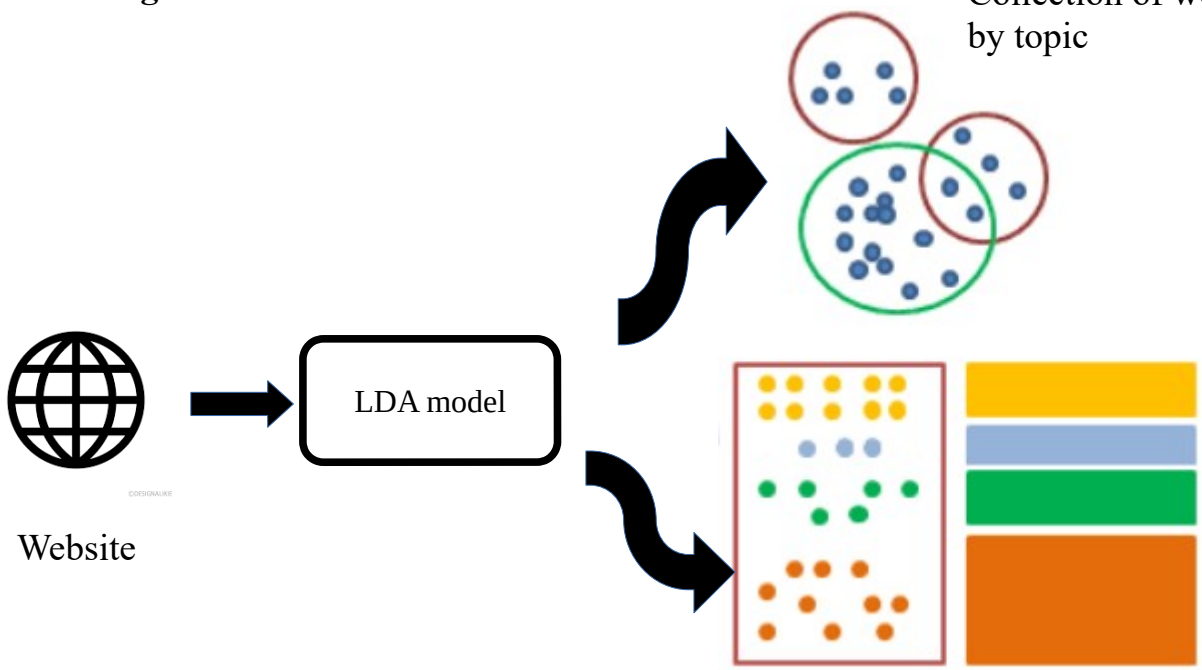
# Topic Modelling

**Introduction:**
Topic modelling is a technique used in natural language processing (NLP) to uncover hidden thematic structures in a collection of texts. This software report provides an overview and analysis of topic modelling performed on a web page using the given code snippet.

**Code Overview:**
The provided code uses the Selenium library to scrape text from a webpage (https://www.studytoday.net/acid-rain/). The extracted text is then preprocessed and subjected to topic modelling using the gensim library. The code generates a visualization of the topics using pyLDAvis.

**Context Diagram:**




**Software Used:**

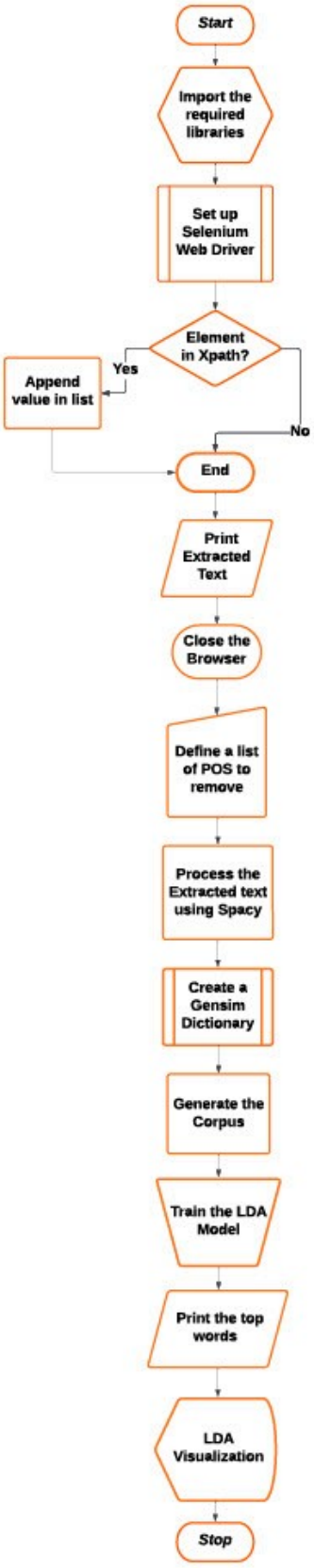
| Tools | Description |
|---|---|
| Python | A high-level, general-purpose programming language |
| Jupyter Notebook | A project to develop open-source software, open standards, and services for interactive computing across multiple programming languages |

| | |
|---|---|
| Selenium | Used for web scraping and interacting with web browsers |
| Pandas | Used for data manipulation and analysis |
| Matplotlib | Used for data visualization |
| Seaborn | Used for data visualization |
| Spacy | A library for natural language processing tasks |
| Gensim | A library for topic modelling and document similarity analysis. |
| pyLDAvis | A library for interactive topic model visualization |

**Algorithm:**

1. Import the required libraries: selenium, pandas, matplotlib.pyplot, seaborn, spacy, pyLDAvis.gensim_models, en_core_web_md, gensim, pickle.
2. Set up a Selenium WebDriver to control the web browser (Chrome) and open the desired website.
3. Find all elements on the page using XPath and store them in a list.
4. Iterate through the elements and extract the text from each element, excluding empty text.
5. Print the extracted text.
6. Close the browser.
7. Load the required libraries and models for natural language processing and topic modeling: en_core_web_md from spaCy, Dictionary and LdaMulticore from Gensim.
8. Define a list of POS tags to remove from the text.
9. Process the extracted text using spaCy to tokenize, lemmatize, and filter out unwanted POS tags.
10. Create a Gensim dictionary from the processed tokens and generate the document-term matrix (corpus).
11. Save the corpus and dictionary objects to disk.
12. Train an LDA (Latent Dirichlet Allocation) model using Gensim on the corpus, specifying the number of topics and passes.
13. Save the trained LDA model to disk.
14. Print the top words for each topic in the LDA model.
15. Load the saved dictionary, corpus, and LDA model from disk.
16. Prepare the LDA visualization using pyLDAvis and display it.

## Flowchart:


Start
Import the required libraries
Set up Selenium Web Driver
Element in Xpath?
Yes
Append value in list
No
End
Print Extracted Text
Close the Browser
Define a list of POS to remove
Process the Extracted text using Spacy
Create a Gensim Dictionary
Generate the Corpus
Train the LDA Model
Print the top words
LDA Visualization
Stop

**Working Code:**

```
from selenium import webdriver

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# Create a new instance of the Chrome driver
browser = webdriver.Chrome()

# Open the website
browser.get("https://www.studytoday.net/acid-rain/")

# Get all elements on the page
elements = browser.find_elements("xpath", "//*")  # Using the correct method
find_elements()

# Iterate through the elements and extract text
all_text = []
for element in elements:
    text = element.text.strip()
    if text:
        all_text.append(text)

# Print the extracted text
for text in all_text:
    print(text)

# Close the browser
browser.quit()

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import spacy
import pyLDAvis.gensim_models
pyLDAvis.enable_notebook()# Visualise inside a notebook
import en_core_web_md
from gensim.corpora.dictionary import Dictionary
from gensim.models import LdaMulticore
from gensim.models import CoherenceModel

# Our spaCy model:
nlp = en_core_web_md.load()
# Tags I want to remove from the text
removal= ['ADV','PRON','CCONJ','PUNCT','PART','DET','ADP','SPACE', 'NUM',
'SYM']
```
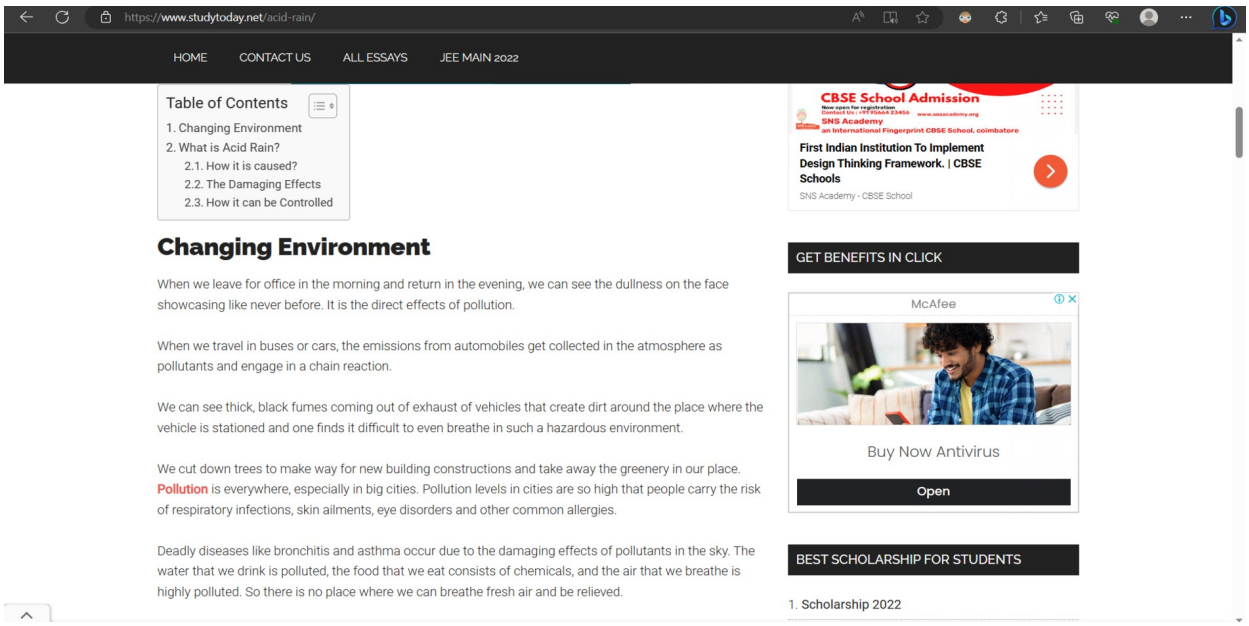
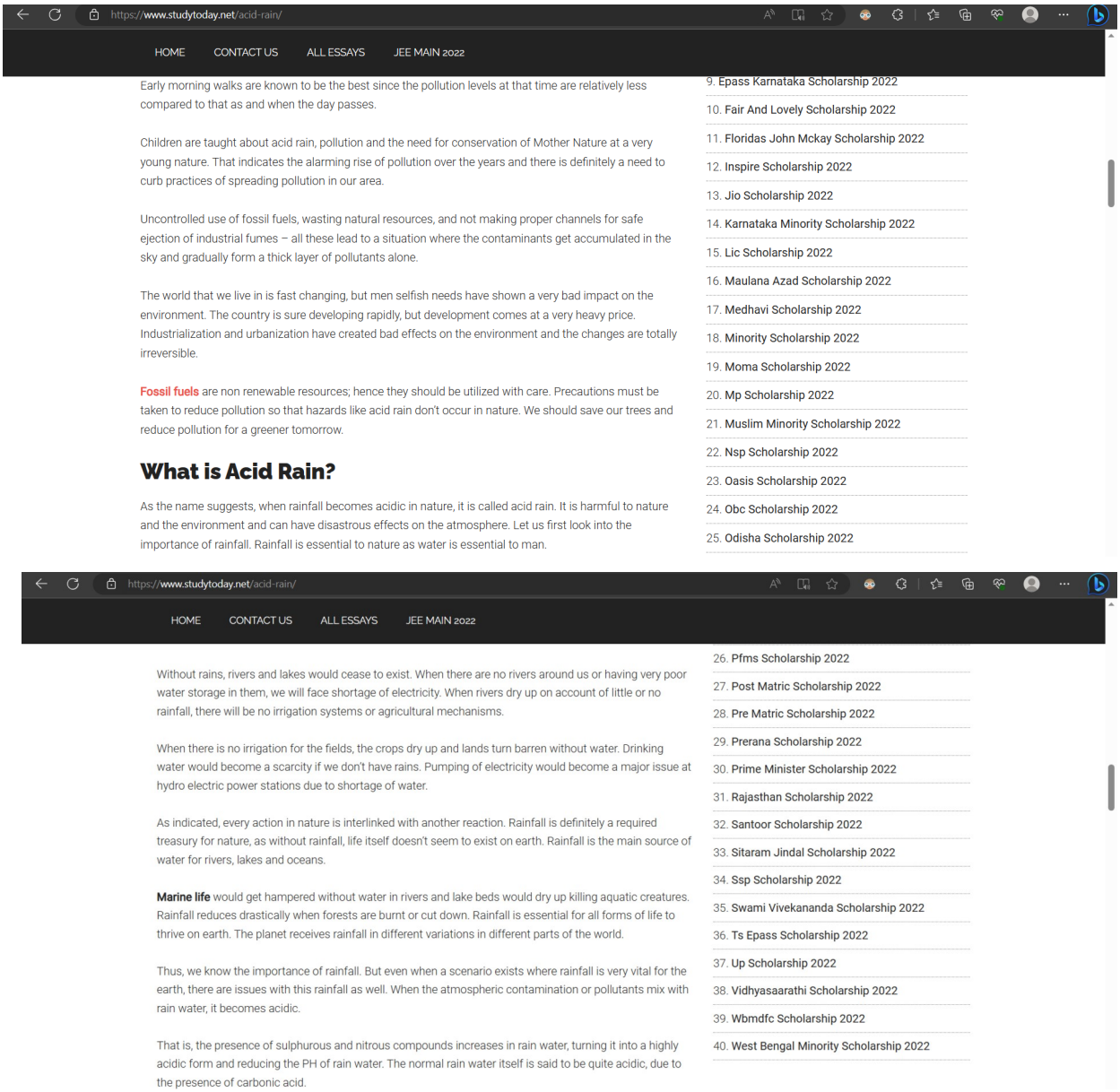

```
tokens = []
for summary in nlp.pipe(all_text[:]):
  proj_tok = [token.lemma_.lower() for token in summary if token.pos_ not in
removal and not token.is_stop and token.is_alpha]
  tokens.append(proj_tok)

from gensim import corpora
dictionary = corpora.Dictionary(tokens)
corpus = [dictionary.doc2bow(text) for text in tokens]
import pickle
pickle.dump(corpus, open('corpus.pkl', 'wb'))
dictionary.save('dictionary.gensim')

import gensim
NUM_TOPICS = 5
ldamodel = gensim.models.ldamodel.LdaModel(corpus, num_topics =
NUM_TOPICS, id2word=dictionary, passes=15)
ldamodel.save('model5.gensim')
topics = ldamodel.print_topics(num_words=4)
for topic in topics:
   print(topic)

dictionary = gensim.corpora.Dictionary.load('dictionary.gensim')
corpus = pickle.load(open('corpus.pkl', 'rb'))
lda = gensim.models.ldamodel.LdaModel.load('model5.gensim')
import pyLDAvis.gensim
lda_display = pyLDAvis.gensim.prepare(lda, corpus, dictionary, sort_topics=False)
pyLDAvis.display(lda_display)
```

**Sample Input:**



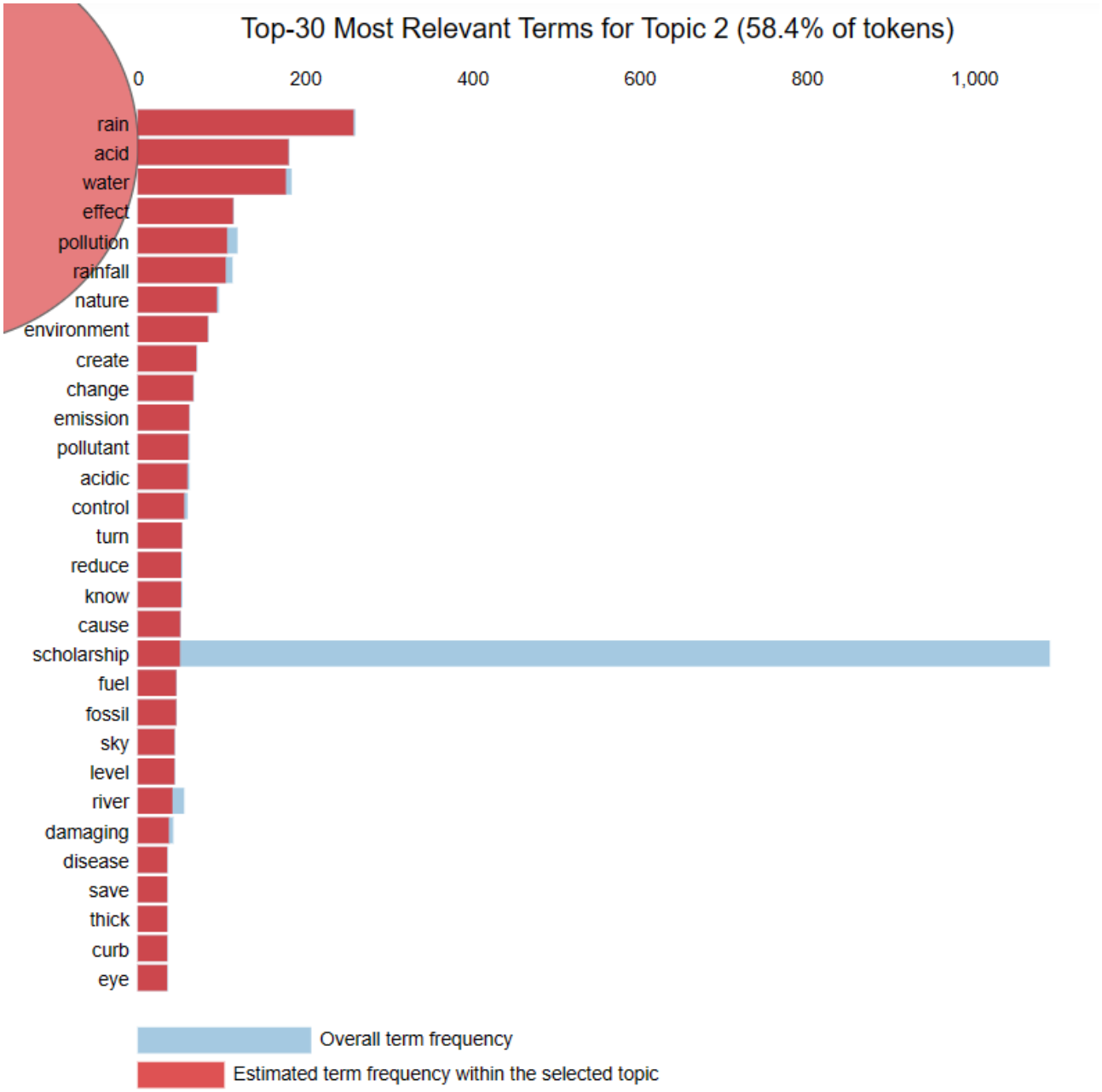






# Output:

```
(0, '0.359*"scholarship" + 0.036*"minority" + 0.018*"karnataka" + 0.018*"matric"')
(1, '0.039*"rain" + 0.028*"acid" + 0.027*"water" + 0.017*"effect"')
(2, '0.249*"class" + 0.229*"scholarship" + 0.200*"students" + 0.063*"student"')
(3, '0.146*"essay" + 0.062*"favourite" + 0.036*"speech" + 0.028*"policy"')
(4, '0.047*"rainfall" + 0.037*"water" + 0.021*"control" + 0.021*"life"')
```




**Code Execution Flow:**

1. Web Scraping:

- The Selenium library is imported.
- An instance of the Chrome driver is created.
- The web page "https://www.studytoday.net/acid-rain/" is opened.
- All elements on the page are retrieved using the XPath expression.
- Text is extracted from the elements and stored in the all_text list.

2. Text Preprocessing:

- The spaCy library is imported, and the en_core_web_md model is loaded.
- Parts of speech (POS) tags to be removed from the text are defined.
- Tokenization, lemmatization, and removal of stop words and specified POS tags are performed using spaCy.
- The preprocessed tokens are stored in the tokens list.

3. Topic Modelling:

- The gensim library is imported.
- The tokens list is converted into a gensim dictionary and corpus.
- The dictionary and corpus are saved as files for future use.
- An LDA (Latent Dirichlet Allocation) model is trained using the corpus and dictionary.
- The trained model is saved as a file.
- The top words for each topic are printed.

4. Topic Visualization:

- The saved dictionary, corpus, and LDA model are loaded.
- The pyLDAvis library is imported and enabled for notebook visualization.
- The pyLDAvis visualization is generated using the loaded LDA model, corpus, and dictionary.
- The visualization is displayed.

**Summary and Analysis:**

The provided code demonstrates a complete workflow for topic modelling. It starts with web scraping to gather text data from a webpage. Then, the text is preprocessed by removing stop words, applying lemmatization, and filtering based on specific parts of speech. After preprocessing, the code uses the gensim library to train an LDA model on the preprocessed text.

The trained LDA model is then used to generate the top words for each topic. The number of topics is set to 5 in this example. The code prints the top words for each

topic, allowing users to gain insights into the main themes present in the extracted text.

Additionally, the code employs the pyLDAvis library to create an interactive visualization of the topic model. The visualization provides an intuitive representation of the topics and their relationships, making it easier to interpret and explore the results.

The generated visualization can help users understand the underlying themes in the extracted text and explore the relationships between topics. It can be a useful tool for researchers, content analysts, and anyone dealing with large amounts of text data.

Overall, the provided code demonstrates an effective implementation of topic modelling using web scraping, text preprocessing, and the gensim library. The combination of these techniques enables the identification and exploration of latent topics in textual data.