

# Understanding Topic Modeling with LDA and Visualizing Results with pyLDAvis

A Practical Guide to Text Analysis

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# LDA 토픽 모델링 분석을 통한 챗GPT 활용에 대한 탐색적 연구 -챗GPT와 교육을 중심으로-

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## An Exploratory Study on the Utilization of ChatGPT through LDA Topic Modeling Analysis -Focusing on ChatGPT and education-

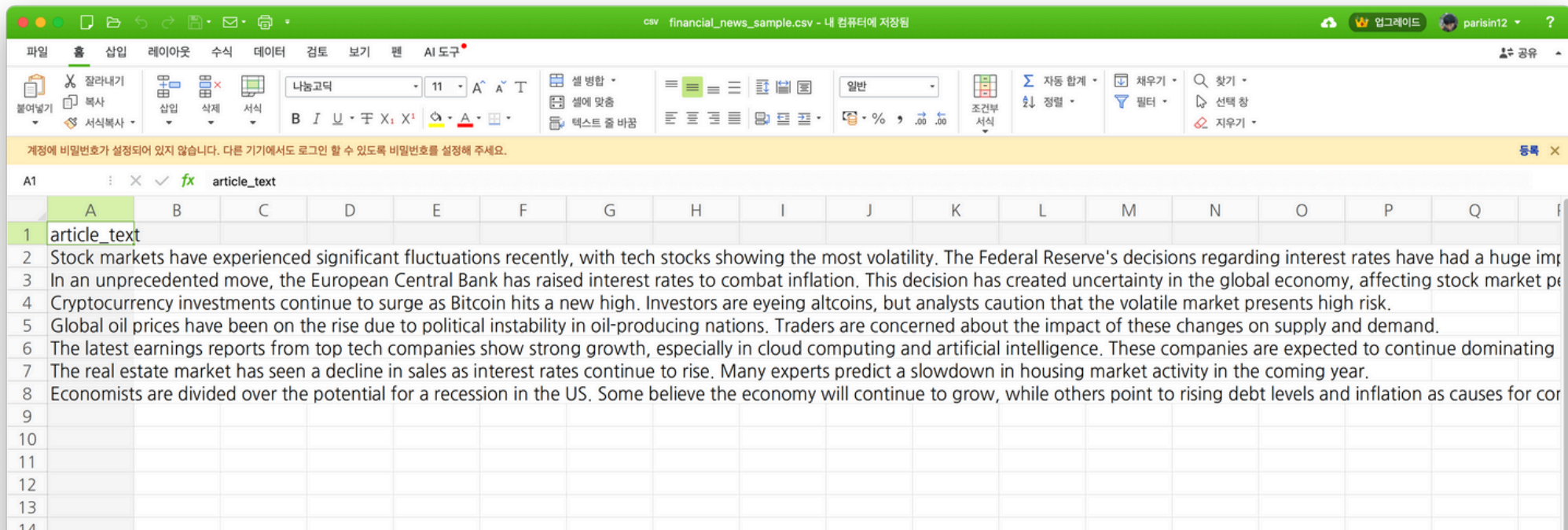
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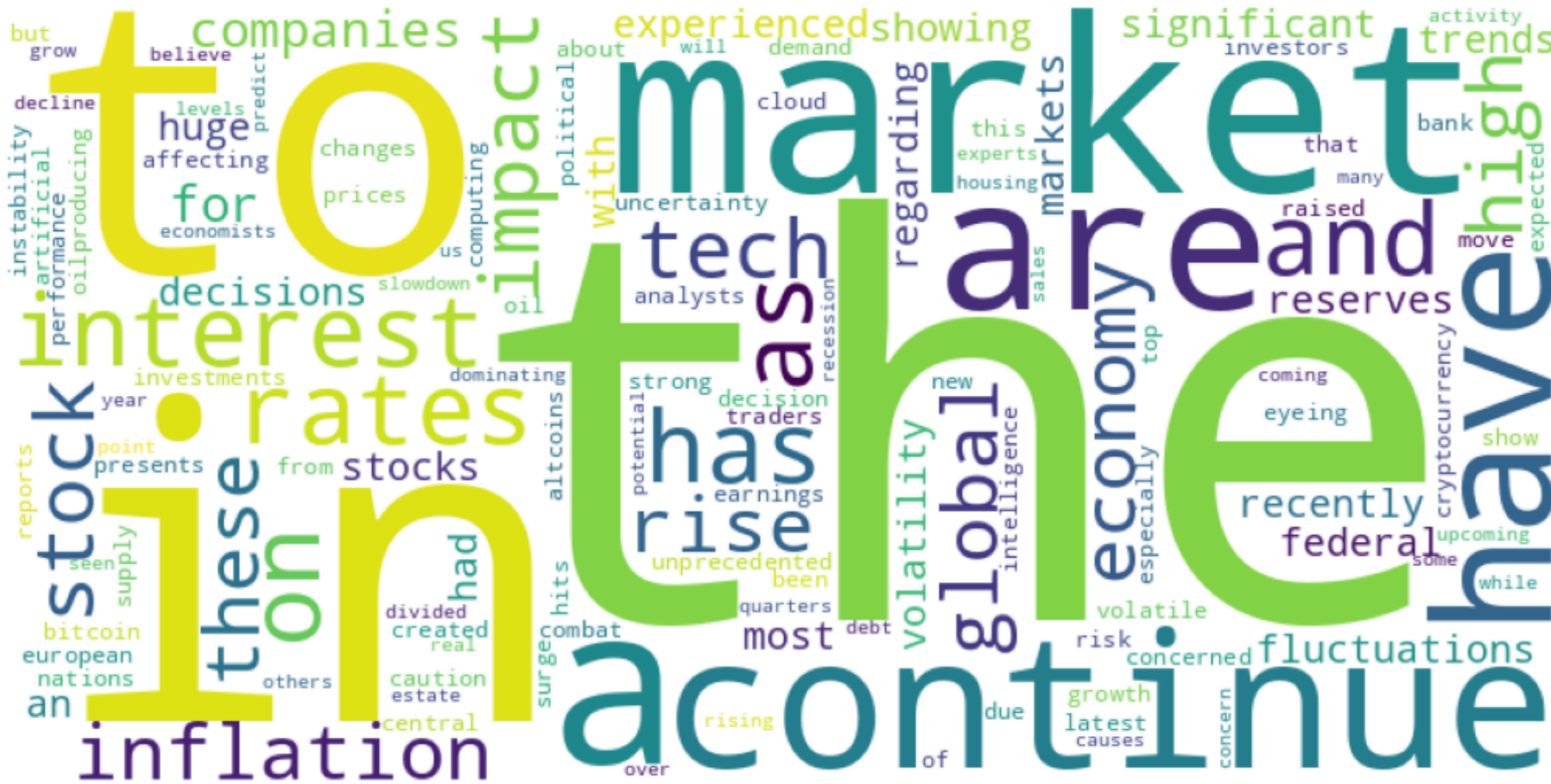
# Introduction to Topic Modeling

- – What is Topic Modeling?
- – A type of statistical model used to discover topics that appear in a collection of documents.
- – Purpose: Identify hidden thematic structures in large text corpora.
- – Applications:
- – Text summarization, content recommendation, document clustering, information retrieval.

# dataset



# total word crowd



Topic #1



# What is LDA (Latent Dirichlet Allocation)?

- – LDA is a generative probabilistic model used to discover topics in a collection of documents.
- – Key Ideas:
  - – Topics: distributions over words.
  - – Documents: mixtures of topics.
  - – Words: assigned to topics based on the model.

# How LDA Works – Simplified

- 1. Input: A collection of documents (text corpus).
- 2. Process:
  - – LDA assigns each word in each document to a topic.
  - – Adjusts topic assignments to maximize likelihood.
- 3. Output: Topics that best represent the themes in the documents.



## 1. Importing Libraries:

```
content_copy
import pandas as pd
import numpy as np
import nltk
from nltk.corpus import stopwords
from sklearn.feature_extraction.text
import TfidfVectorizer
from gensim import corpora
from gensim.models import LdaModel
import pyLDAvis.gensim_models
import matplotlib.pyplot as plt
from wordcloud import WordCloud
from sklearn.decomposition import
LatentDirichletAllocation
```

```
# TF-IDF Vectorization
vectorizer =
TfidfVectorizer(max_df=0.95,
min_df=2, stop_words='english')
X =
vectorizer.fit_transform(data['pr
ocessed_text'])
```

# Example of LDA Topics

- – Topic 1: Stock Market, Interest Rates, Inflation
- – Topic 2: Cryptocurrency, Bitcoin, Investment
- – Topic 3: Real Estate, Housing Market, Sales
- – Topic 4: Tech Earnings, Artificial Intelligence, Growth

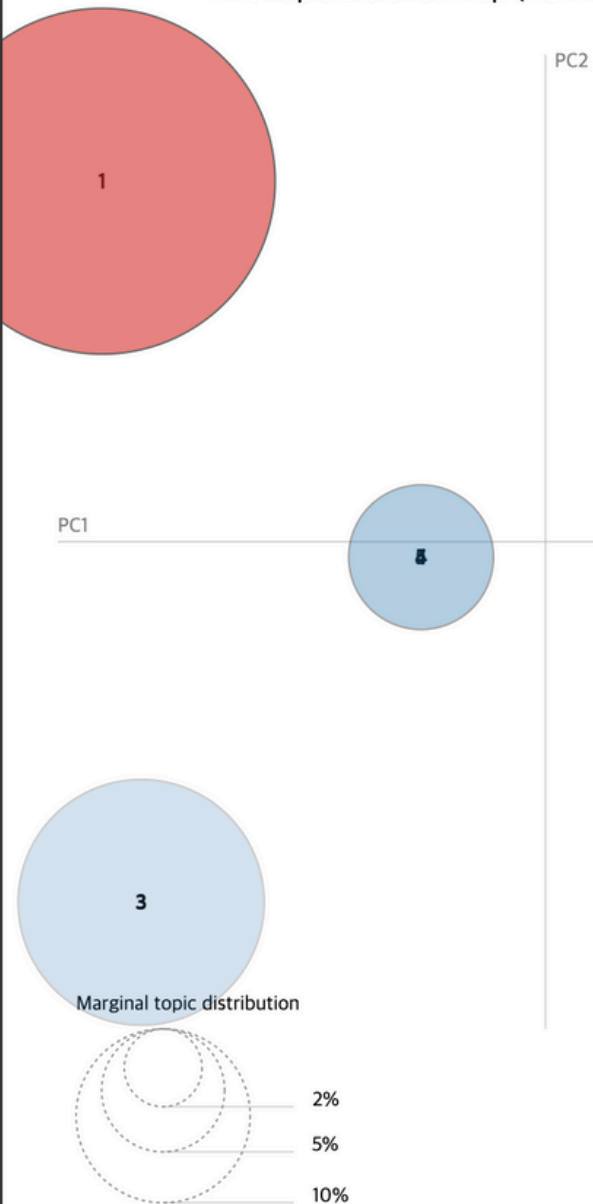
# Why LDA is Useful

- – Topic Discovery: Helps identify underlying themes in large datasets.
- – Text Categorization: Improves document classification.
- – Data Exploration: Identifies meaningful patterns in unstructured data.

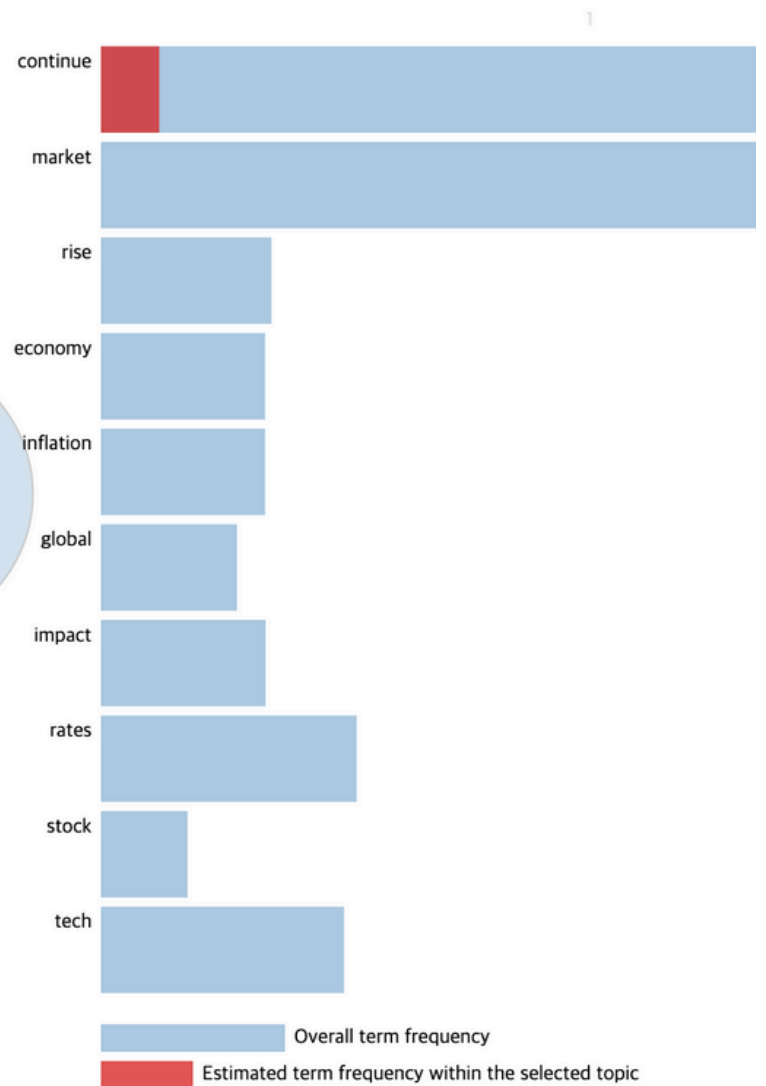
# pyLDAvis: Visualization of LDA Results

- – pyLDAvis provides interactive ways to visualize topic modeling results.
- – Key Features:
- – Interactive visualization of topic distribution and word frequency.
- – Topic–topic distance using dimensionality reduction (t-SNE, MDS).
- – Global and local views of topics and word associations.

Intertopic Distance Map (via multidimensional scaling)



Top-10 Most Relevant Terms for Topic 1 (3)



1.  $\text{saliency}(\text{term } w) = \text{frequency}(w) * [\sum_t p(t | w) * \log(p(t | w)/p(t))]$  for top
2.  $\text{relevance}(\text{term } w | \text{topic } t) = \lambda * p(w | t) + (1 - \lambda) * p(w | t)/p(w)$ ; see Sievert

# Key Components of pyLDAvis

- 1. Topic–Word Distribution: Most frequent words in each topic.
- 2. Topic–Topic Distance: Relationship between topics using 2D map.
- 3. Interactivity: Hover over points to view top words for each topic.

# pyLDAvis

```
# Replace the import statement
import pyLDAvis
import pyLDAvis.lda_model

# Replace pyLDAvis.sklearn.prepare with pyLDAvis.lda_model.prepare
# Use 'pcoa' instead of 'tsne' for MDS
vis = pyLDAvis.lda_model.prepare(lda, X, vectorizer, mds='pcoa') # Updated visualization
                        function call

# Display the visualization
pyLDAvis.display(vis) # Assuming you want to display the visualization in a notebook
                        environment
```



# How to Use pyLDAvis with LDA

- – Train an LDA model on your text data.
- – Prepare the output from the LDA model and term-document matrix.
- – Use pyLDAvis to visualize the topics and their distributions.

# Visualizing Financial Data Topics

- – Use pyLDAvis to explore how topics are distributed across financial articles.
- – Identify key words for each topic (e.g., “stocks,” “tech,” “cryptocurrency”).
- – Explore topic relationships and gain insights into market trends.

# Applications of LDA and pyLDAvis in Finance

- – Financial Text Analysis: Discover trends and themes in financial news.
- – Sentiment Analysis: Combine LDA with sentiment analysis to understand market sentiment.
- – Market Prediction: Predict trends by analyzing the emergence of key topics.

# Conclusion

- – LDA and pyLDAvis help uncover topics and visualize them in large datasets.
- – Applications in finance include analyzing market trends, understanding sentiments, and predicting market behavior.
- – pyLDAvis enhances topic model interpretability through interactive visualization.

# Q&A

- Questions?