Лабораторна робота 5. Варіант 1.

Моделювання тем

Мета роботи: Ознайомитись з вирішенням задач пошуку ключових слів та моделювання тем.

- Застосувати приховане семантичне індексування бібліотеки Gensim для моделювання тем. Вивести документи, що зробили найбільший вклад в теми. Обрати або створити три нових документи (які модель ще не бачила) та визначити їх теми.
- 2. Використати текст austen-persuasion.txt з корпусу gutenberg бібліотеки nltk та вивести ключові біграми.

```
import pandas as pd
data = pd.read csv('news.csv')
data
                                                     text
                                                           label
       Here are Thursday's biggest analyst calls: App...
1
       Buy Las Vegas Sands as travel to Singapore bui...
                                                                0
2
       Piper Sandler downgrades DocuSign to sell, cit...
                                                                0
3
       Analysts react to Tesla's latest earnings, bre...
                                                                0
4
       Netflix and its peers are set for a 'return to...
                                                                0
16985 KfW credit line for Uniper could be raised to ...
                                                                3
16986 KfW credit line for Uniper could be raised to ...
                                                                3
                                                                3
16987 Russian https://t.co/R0iPhyo5p7 sells 1 bln r...
                                                                3
16988 Global ESG bond issuance posts H1 dip as supra...
                                                                3
16989 Brazil's Petrobras says it signed a $1.25 bill...
[16990 \text{ rows } x \text{ 2 columns}]
texts = data['text'].tolist()
from gensim.parsing.preprocessing import preprocess string
from gensim import corpora
# Попередня обробка текстів
processed texts = [preprocess string(text) for text in texts]
# Створення словника та корпусу
dictionary = corpora.Dictionary(processed texts)
corpus = [dictionary.doc2bow(text) for text in processed texts]
```

```
from gensim.models import LsiModel

# Побудова LSI моделі
lsi_model = LsiModel(corpus, id2word=dictionary, num_topics=10)
```

Вивести документи, що зробили найбільший вклад в теми.

```
topics = lsi model.print topics(num topics=10, num words=5)
for i, topic in topics:
    print(f"Topic #{i}: {topic}")
# Визначення внеску кожного документа в теми
corpus lsi = lsi model[corpus]
# Визначення топ документів для кожної теми
from collections import defaultdict
topic contributions = defaultdict(list)
for doc id, doc in enumerate(corpus lsi):
    for topic_id, contribution in doc:
        topic contributions[topic id].append((contribution, doc id))
for topic id, contributions in topic contributions.items():
    top docs = sorted(contributions, reverse=True)[:3]
    print(f"\nTop documents for topic #{topic id}:")
    for contribution, doc id in top docs:
        print(f"\nDocument #{doc id} with contribution {contribution}:
{texts[doc id]}")
Topic #0: 0.966*"http" + 0.060*"announc" + 0.058*"market" +
0.056*"new" + 0.055*"stock"
Topic \#1: -0.508*"stock" + -0.469*"market" + -0.295*"trade" + -
0.264*"economi" + -0.250*"invest"
Topic #2: 0.377*"quarter" + 0.354*"second" + 0.319*"result" +
0.291*"earn" + 0.251*"announc"
Topic #3: -0.351*"rate" + -0.341*"inflat" + -0.304*"quarter" + -
0.283*"second" + -0.260*"year"
Topic #4: 0.629*"new" + 0.427*"est" + 0.290*"prev" + 0.263*"jun" + -
0.257*"market"
Topic \#5: -0.730*"market" + 0.438*"stock" + -0.236*"est" + -
0.160*"jun" + -0.159*"prev"
Topic #6: -0.680*"new" + 0.392*"est" + 0.229*"prev" + 0.216*"jun" + -
0.182*"market"
Topic \#7: -0.456*"price" + -0.419*"stock" + 0.396*"bank" +
0.246*"rate" + 0.178*"invest"
Topic #8: -0.473*"stock" + 0.294*"price" + 0.289*"trade" +
0.258*"economi" + -0.256*"market"
Topic #9: -0.410*"beat" + 0.378*"announc" + -0.338*"revenu" + -
0.300*"ep" + -0.294*"earn"
```

Top documents for topic #0:

Document #927 with contribution 3.8801262379041486: NovaBay Pharmaceuticals' DERMAdoctor Products Now Available at https://t.co/tQjGxou7L8 and https://t.co/9nezzpbAhfhttps://t.co/XmjxG1Dxfq https://t.co/6d6mdehcp4

Document #10118 with contribution 3.2152883108474914: South Korea Air Treatment Systems Markets, 2021-2022 & Document Markets, Air Treatment Systems Markets, & Elter Replacement Markets - https://t.co/guyiBzPH8C https://t.co/SNWojILr2B https://t.co/kYjlxlrWE0

Document #10143 with contribution 3.1315457616203695: Global Lending Market Report to 2031 - Featuring Citi Group, Bank of America and State Bank of India Among Others - https://t.co/guyiBzPH8C https://t.co/CaK9s3yvWN https://t.co/Q6l1W6mexs

Top documents for topic #1:

Document #1134 with contribution 0.5845055015481952: TotalEnergies SE UK Regulatory Announcement: Papua New Guinea: TotalEnergies Announces New Milestone towards Papua LNG Development https://t.co/IZaoBm04WIhttps://t.co/zsQ5lsjy3G

Document #943 with contribution 0.5332510955479566: TotalEnergies SE UK Regulatory Announcement: United States: TotalEnergies Announces the Start-up of New Ethane Cracker in Port Arthur https://t.co/FCZzlm1FWzhttps://t.co/6KfPlvGwzy

Document #12386 with contribution 0.5012628568440157: https://t.co/GGf0VJaxvZ Announces Appointment of Karen Drexler to Its Board of Directors and Jean-Olivier Racine to Its Advisory Board https://t.co/oGjocD0Mg9 https://t.co/5nkVakOc5Y

Top documents for topic #2:

Document #6890 with contribution 2.7706311737078253: Tesla reported second-quarter earnings above Wall Street projections, defying expectations. \$TSLA said it earned \$2.3 billion, or \$1.95 a share, in the second quarter, compared with \$1.1 billion, or \$1.02 a share, in the second quarter of 2021. https://t.co/3z9kegGUiG https://t.co/nBKoOf1J05

Document #4660 with contribution 2.355880077264461: USA Compression Partners Announces Second Quarter 2022 Distribution; Second Quarter 2022 Earnings Release and Conference Call Scheduled for August 2 https://t.co/kDuvPiRHCU https://t.co/ullVYUQGD5

Document #6904 with contribution 2.066809659256738: Tesla (\$TSLA) is

expected to report adjusted earnings of \$1.86 a share in the second quarter, which would compare with adjusted earnings of \$1.45 a share in the second quarter of 2021, according to analysts polled by FactSet. https://t.co/JQ7yfa5We9

Top documents for topic #3:

Document #10118 with contribution 0.6372971359487317: South Korea Air Treatment Systems Markets, 2021-2022 & Document Markets, Air Treatment Systems Markets, & Elter Replacement Markets - https://t.co/guyiBzPH8C https://t.co/SNWojILr2B https://t.co/kYjlxlrWE0

Document #9422 with contribution 0.5965214248772098: uncertainty https://t.co/rCq2sgw0ew WATCH: Elon Musk's effort to delay Company's trial against him flopped in court, after a Delaware judge ruled that Company's lawsuit seeking to hold Elon Musk to his \$44 billion takeover will go to trial in October https://t.co/rCq2sgw0ew https://t.co/3EDTIG0N83

Document #9595 with contribution 0.5488179314996751: Northleaf Capital Partners to Acquire 40% Interest in New Zealand Mobile Tower Infrastructure Business From Vodafone New Zealand Limited https://t.co/ruFWvKrjWT https://t.co/HFx3Ekz1gU

Top documents for topic #4:

Document #11112 with contribution 6.676393064391269: #0ATT | Global #WASDE Corn End Stocks New Jun: 313M (est 311M; prev 310M) - Soybean End Stocks New: 100M (est 99M; prev 100M) - Wheat End Stocks New: 268M (est 266M; prev 267M) - Cotton End Stocks New: 84M (est 82M; prev 83M)

Document #11113 with contribution 6.455319537916789: #0ATT | US #WASDE Corn End Stocks New Jul: 1470M (est 1450M; prev 1400M) - Soybean End Stocks New: 230M (est 203M; prev 280M) - Wheat End Stocks New: 639M (est 641M; prev 627M) - Cotton End Stocks New: 2.40M (est 2.79M; prev 2.90M) https://t.co/twS4Y9iqnH

Document #10762 with contribution 4.753817550619759: U.S CPI (MOM) (JUN) ACTUAL: 1.3% VS 1.0% PREVIOUS; EST 1.1% U.S CPI (YOY) (JUN) ACTUAL: 9.1% VS 8.6% PREVIOUS; EST 8.8% U.S CORE CPI (MOM) (JUN) ACTUAL: 0.7% VS 0.6% PREVIOUS; EST 0.5% U.S CORE CPI (YOY) (JUN) ACTUAL: 5.9% VS 6.0% PREVIOUS; EST 5.7%

Top documents for topic #5:

Document #14933 with contribution 1.327968131943603: Linear stocks > choppy stocks. Where a stock has come from > relative strength. Entries matter.

Document #4366 with contribution 1.3091793308907116: \$C bucks the banking trending and reports surprisingly big upside \$JPM \$XLF \$MS \$WFC https://t.co/z0XuABKmsY #earnings #economy #banks #financials #inflation #Investment #banking #bank #stocks

Document #8653 with contribution 1.3043316607763182: So levered single-stock ETFs are okay but volatility products using most liquid option contracts (SPY) in the world are not? @HesterPeirce Investors in U.S. have a new way to supersize bets on high-profile stocks, with the launch of single-stock ETFs https://t.co/zgB5b3DsqD

Top documents for topic #6:

Document #11026 with contribution 4.157251779070796: Eurozone CPI (Y/Y) Jun F: 8.6% (est 8.6%; prev 8.6%) - Eurozone CPI (M/M) Jun F: 0.8% (est 0.8%; prev 0.8%) - Eurozone CPI Core (M/M) Jun F: 0.2% (est 0.2%; prev 0.2%) - Eurozone CPI Core (Y/Y) Jun F: 3.7% (est 3.7%; prev 3.7%)

Document #10762 with contribution 4.140361563349039: U.S CPI (MOM) (JUN) ACTUAL: 1.3% VS 1.0% PREVIOUS; EST 1.1% U.S CPI (YOY) (JUN) ACTUAL: 9.1% VS 8.6% PREVIOUS; EST 8.8% U.S CORE CPI (MOM) (JUN) ACTUAL: 0.7% VS 0.6% PREVIOUS; EST 0.5% U.S CORE CPI (YOY) (JUN) ACTUAL: 5.9% VS 6.0% PREVIOUS; EST 5.7%

Document #11013 with contribution 4.139541012334429: South African CPI (M/M) Jun: 1.1% (est 0.9%; prev 0.7%) - South African CPI (Y/Y) Jun: 7.4% (est 7.3%; prev 6.5%) - South African CPI Core (M/M) Jun: 0.6% (est 0.5%; prev 0.2%) - South African CPI Core (Y/Y) Jun: 4.4% (est 4.3%; prev 4.1%)

Top documents for topic #7:

Document #5308 with contribution 1.6930081714065317: \$BAC | Bank Of America Q2 22 Earnings: - EPS: 0.73\$ (est \$0.76) - Revenue: \$22.69B (est \$22.86B) - Wealth & (est \$10.76) - FICC Sales & (est \$5.43B) - Trading Revenue EX DVA: \$4B (est \$4.01B) - FICC Sales & (est \$2.29B)

Document #4366 with contribution 1.6015048204900764: \$C bucks the banking trending and reports surprisingly big upside \$JPM \$XLF \$MS \$WFC https://t.co/z0XuABKmsY #earnings #economy #banks #financials #inflation #Investment #banking #bank #stocks

Document #5969 with contribution 1.4221833967948219: UK Foreign Secretary Liz Truss's citing of the Bank of Japan's inflation mandate as a potential model for the Bank of England isn't convincing central bank watchers https://t.co/bQMHhN5YaU

Top documents for topic #8:

```
Document #5579 with contribution 1.353332097960103: "I think you're
going to see supply upended at the end of the year, if not before,"
Truist's Neal Dingmann says, adding: "I do think... that if trade
flows and price caps happen... there's a very good chance for some
price spikes in oil." https://t.co/Jb7k9tf4PH
https://t.co/zlGyhqqXLf
Document #15702 with contribution 1.2075195651548056: $TPG - TPG:
Short-Term Price Risk But Attractive Long-Term Business Fundamentals.
https://t.co/6JlcLtEySr #stockmarket #economy #investing
Document #11188 with contribution 1.1623802065821698: U.S.-China Trade
Rebounds As China Eases Its Zero-COVID Policy Lockdowns.
https://t.co/ZqQEdwk5IQ #trading #business #economy
Top documents for topic #9:
Document #10666 with contribution 1.7498802371380688: CANADA CPI (MOM)
(JUN) ACTUAL: 0.7% VS 1.4% PREVIOUS; EST 0.9% CANADA CPI (YOY) (JUN)
ACTUAL: 8.1% VS 7.7% PREVIOUS; EST 8.4% CANADA CORE CPI (MOM) (JUN)
ACTUAL: 0.3% VS 0.8% PREVIOUS CANADA CORE CPI (YOY) (JUN) ACTUAL:
6.2% VS 6.1% PREVIOUS; EST 5.9% @MtlExchange
Document #10762 with contribution 1.7327823286831745: U.S CPI (MOM)
(JUN) ACTUAL: 1.3% VS 1.0% PREVIOUS; EST 1.1% U.S CPI (YOY) (JUN)
ACTUAL: 9.1% VS 8.6% PREVIOUS; EST 8.8% U.S CORE CPI (MOM) (JUN)
ACTUAL: 0.7% VS 0.6% PREVIOUS; EST 0.5% U.S CORE CPI (YOY) (JUN)
ACTUAL: 5.9% VS 6.0% PREVIOUS; EST 5.7%
Document #11026 with contribution 1.6504801282488137: Eurozone CPI
(Y/Y) Jun F: 8.6% (est 8.6%; prev 8.6%) - Eurozone CPI (M/M) Jun F:
0.8% (est 0.8%; prev 0.8%) - Eurozone CPI Core (M/M) Jun F: 0.2% (est
0.2%; prev 0.2%) - Eurozone CPI Core (Y/Y) Jun F: 3.7% (est 3.7%;
prev 3.7%)
new texts = [
    "Analysts say Apple will continue to grow in the next quarter.",
    "Tesla's new model has impressed the market with its advanced
features.",
    "Amazon's stock prices are predicted to rise due to increased
sales."
new processed texts = [preprocess string(text) for text in new texts]
new corpus = [dictionary.doc2bow(text) for text in
new processed texts]
new corpus lsi = lsi model[new corpus]
for i, doc in enumerate(new corpus lsi):
```

```
print(f"New document #{i + 1}: {new_texts[i]}")
    for topic id, contribution in doc:
        print(f" - Topic #{topic id + 1} with contribution
{contribution}")
New document #1: Analysts say Apple will continue to grow in the next
quarter.
 - Topic #1 with contribution 0.0701707757892538
 - Topic #2 with contribution 0.02605186813960929
 - Topic #3 with contribution 0.33907283961691104
 - Topic #4 with contribution -0.3150724896901351
 - Topic #5 with contribution 0.009183449897806381
 - Topic #6 with contribution -0.08875207098864706
 - Topic #7 with contribution -0.11517570500442723
 - Topic #8 with contribution -0.13454558586140025
 - Topic #9 with contribution 0.03997909510178691
 - Topic #10 with contribution 0.008109876492719797
New document #2: Tesla's new model has impressed the market with its
advanced features.
 - Topic #1 with contribution 0.12970955198419243
 - Topic #2 with contribution -0.4276164854458416
 - Topic #3 with contribution -0.23161052124978343
 - Topic #4 with contribution 0.16704065259199743
 - Topic #5 with contribution 0.38628609795852714
 - Topic #6 with contribution -0.7702722717163895
 - Topic #7 with contribution -0.8674165923519286
 - Topic #8 with contribution -0.00011779822446590519
 - Topic #9 with contribution -0.2648436688181909
 - Topic #10 with contribution -0.2804336953230337
New document #3: Amazon's stock prices are predicted to rise due to
increased sales.
 - Topic #1 with contribution 0.1374158945219514
 - Topic #2 with contribution -0.565506307733751
 - Topic #3 with contribution -0.34045088648857896
 - Topic #4 with contribution -0.3248558149483241
 - Topic #5 with contribution 0.2076756959083238
 - Topic #6 with contribution 0.3597957302689959
 - Topic #7 with contribution 0.17682690435718748
 - Topic #8 with contribution -0.9868809722971388
 - Topic #9 with contribution -0.11910469732916602
 - Topic #10 with contribution 0.12563470079476136
```

1. Використати текст austen-persuasion.txt з корпусу gutenberg бібліотеки nltk та вивести ключові біграми.

```
import nltk
from nltk.corpus import gutenberg, stopwords
from nltk.collocations import BigramCollocationFinder
from nltk.metrics import BigramAssocMeasures
import string
```

```
# Завантажимо текст
nltk.download('gutenberg')
nltk.download('punkt')
persuasion text = gutenberg.raw('austen-persuasion.txt')
stop words = set(stopwords.words('english'))
# Токенізація
tokens = nltk.word tokenize(persuasion text)
cleaned tokens = [token for token in tokens if token not in stop words
and token not in string.punctuation and token not in ['``', "''"]]
# Знаходження біграм
bigram finder = BigramCollocationFinder.from words(cleaned tokens)
bigrams = bigram finder.nbest(BigramAssocMeasures.likelihood ratio,
10)
print("Key Bigrams:")
for bigram in bigrams:
    print(bigram)
[nltk data] Downloading package gutenberg to
[nltk data]
                 C:\Users\murat\AppData\Roaming\nltk data...
[nltk data]
               Package gutenberg is already up-to-date!
[nltk data] Downloading package punkt to
                 C:\Users\murat\AppData\Roaming\nltk data...
[nltk data]
               Package punkt is already up-to-date!
[nltk data]
Key Bigrams:
('Captain', 'Wentworth')
('Lady', 'Russell')
('Sir', 'Walter')
('Mr', 'Elliot')
('Mrs', 'Clay')
('Mrs', 'Smith')
('Captain', 'Benwick')
('Mrs', 'Musgrove')
('Camden', 'Place')
('great', 'deal')
```