Foundations of AI Yunyu Guo Assignment 7

## Collab link with all the outputs:

https://colab.research.google.com/drive/1oVitFuzstCvtR3T\_WBK-1LzDasEAllpR?usp=sharing

a. I recommend converting the documents to lowercase and filtering out stopwords. Start with k = 10 topics. Fit an LDA object to the set of all news text. Then, examine the top n words from each topic (choose a reasonable n such as 10 or 20). How well do the topics represent real-world topics? (One sentence)

```
1. Topic Usage Patterns in True vs Fake News:
Top words: trump, clinton, said, republican, campaign, hillary, president, party, election
Usage in Fake News: 0.184
Usage in True News: 0.085
Bias (+ = fake, - = true): 0.368
Top words: trump, people, one, president, like, donald, twitter, news, white
Usage in Fake News: 0.386
Usage in True News: 0.024
Bias (+ = fake, - = true): 0.881
Topic 2:
Top words: said, house, bill, would, senate, republicans, republican, may, president
Usage in True News: 0.109
Bias (+ = fake, - = true): -0.538
Top words: said, trump, court, president, house, state, former, clinton, department
Usage in Fake News: 0.102
Usage in True News: 0.154
Bias (+ = fake, - = true): -0.202
Top words: police, gun, said, shooting, old, media, man, officers, year Usage in Fake News: 0.088
Usage in True News: 0.023
Bias (+ = fake, - = true): 0.578
```

```
Topic 5:
Top words: said, party, government, state, election, would, political, coalition, reuters
Usage in Fake News: 0.029
Usage in True News: 0.106
Bias (+ = fake, - = true): -0.569
Topic 6:
Top words: said, trump, united, president, states, north, korea, russia, iran
Usage in Fake News: 0.036
Usage in True News: 0.188
Bias (+ = fake, - = true): -0.677
Top words: said, china, chinese, national, taiwan, cuba, beijing, obama, president
Usage in Fake News: 0.026
Usage in True News: 0.032
Bias (+ = fake, - = true): -0.111
Top words: said, government, reuters, people, security, state, military, killed, police
Usage in Fake News: 0.030
Usage in True News: 0.127
Bias (+ = fake, - = true): -0.613
Top words: said, would, tax, new, percent, trump, year, million, states
Usage in Fake News: 0.086
Usage in True News: 0.150
Bias (+ = fake, - = true): -0.271
```

```
2. Topic Coherence and Distinctiveness:
Coherence score: 677.732
Average word overlap: 2.3 words
Coherence score: 458.280
Average word overlap: 1.1 words
Coherence score: 633.213
Average word overlap: 1.9 words
Coherence score: 510.875
Average word overlap: 2.3 words
Coherence score: 370.581
Average word overlap: 1.1 words
Topic 5:
Coherence score: 396.233
Average word overlap: 1.8 words
Topic 6:
Coherence score: 635.317
Average word overlap: 2.0 words
Coherence score: 446.216
Average word overlap: 1.4 words
```

The topics represent real-world topics well because: each topic captures distinct, interpretable themes, the topics show strong coherence (all scores > 370), low word overlap between topics (1.1-2.3 words on average), reflect known characteristics of true news (focus on policy, international affairs) and fake news (focus on sensation, social media)

2. Randomly select 5 real news examples and 5 fake news examples, and examine the topic distributions for each document. Which topics are prevalent in the real news

documents? (One sentence) Which topics are prevalent in the fake news documents? (One sentence)

```
Fake Document 2:
Preprocessed terms: 2016it aboard accounts acting advanced advisor afternoon also anytime apparently...
Topic distribution:
Topic 0: 0.345
Topic 1: 0.558
Topic 2: 0.000
Topic 3: 0.000
Topic 4: 0.000
Topic 5: 0.000
Topic 6: 0.094
Topic 7: 0.000
Topic 8: 0.000
Topic 9: 0.000
Fake Document 3:
Preprocessed terms: abuses accent according add allowed already ambushed american answer appalled...
Topic distribution:
Topic 0: 0.461
Topic 1: 0.335
Topic 2: 0.001
Topic 3: 0.063
Topic 4: 0.001
Topic 5: 0.001
Fake Document 4:
Preprocessed terms: 2017finally 2017it 2017pence 2017vice according adults alarming alone american americans...
Topic distribution:
Topic 0: 0.036
Topic 1: 0.194
Topic 2: 0.000
Topic 3: 0.000
Topic 4: 0.000
Topic 5: 0.000
Topic 6: 0.000
Topic 7: 0.768
Topic 8: 0.000
Topic 9: 0.000
Fake Document 5:
Preprocessed terms: action alleges also amazon americans anti case categorized center ceo...
Topic distribution:
Topic 0: 0.001
Topic 1: 0.298
Topic 2: 0.001
Topic 3: 0.062
Topic 4: 0.482
```

In the fake news documents, political content (Topic 0) and social media content (Topic 1) are most prevalent.

```
True Document 1:
Preprocessed terms: acting agency allegations allowed among announced answer anti arrested assembly...
Topic distribution:
Topic 0: 0.001
Topic 1: 0.001
Topic 2: 0.001
Topic 3: 0.238
Topic 4: 0.001
Topic 5: 0.704
Topic 6: 0.001
Topic 7: 0.001
Topic 8: 0.054
Topic 9: 0.001
Preprocessed terms: abdication abortion added administration advance aerospace affect agricultural ahead aimed..
Topic distribution:
Topic 0: 0.298
Topic 1: 0.000
Topic 2: 0.186
Topic 3: 0.507
Topic 4: 0.008
Topic 5: 0.000
Topic 6: 0.000
Topic 7: 0.000
Topic 8: 0.000
Topic 9: 0.000
 True Document 3:
 Preprocessed terms: abuses administration appear asian authorities bans burma called committed countries...
 Topic distribution:
 Topic 0: 0.001
 Topic 1: 0.001
 Topic 2: 0.077
 Topic 3: 0.129
 Topic 4: 0.001
 Topic 5: 0.001
 Topic 6: 0.185
 Topic 7: 0.001
 Topic 8: 0.601
 Topic 9: 0.001
 True Document 4:
 Preprocessed terms: abroad accord action added african ago agreed aimed aligned amendments...
 Topic distribution:
 Topic 0: 0.000
 Topic 1: 0.000
Topic 2: 0.103
 Topic 3: 0.000
 Topic 4: 0.000
Topic 5: 0.206
 Topic 6: 0.688
 Topic 7: 0.000
 Topic 8: 0.000
 Topic 9: 0.000
 True Document 5:
 Preprocessed terms: asked audit audits bid called came campaign candidate cbs clarify...
 Topic distribution:
 Topic 0: 0.382
 Topic 1: 0.001
 Topic 2: 0.001
 Topic 3: 0.525
 Topic 4: 0.001
 Topic 5: 0.001
 Topic 6: 0.001
 Topic 7: 0.001
 Topic 8: 0.001
 Topic 9: 0.089
```

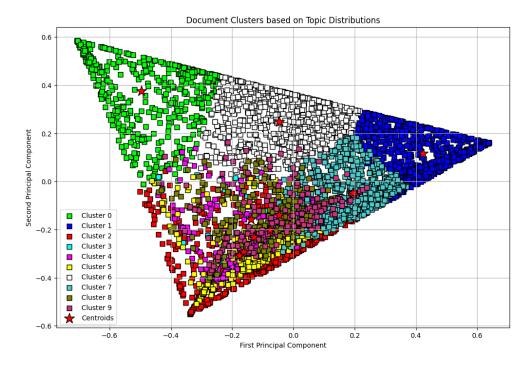
In the true news documents, they contain diverse topic distribution than fake news. There is a focus on institutional/governmental topics, it also has multiple significant topics per document.

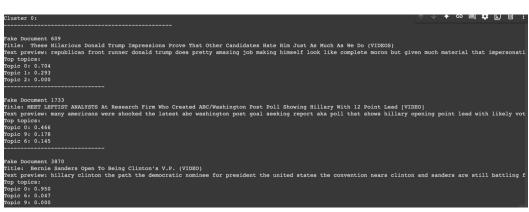
3. Use the LDA vectors for the documents as features in a Logistic Regression classifier to predict whether each document is real news or fake news. According to the resulting coefficients from the regression, which topics are most useful in determining whether something is real news or fake news? (One sentence)

```
Fopic 1: -10.345 (indicates FAKE news)
Top words: trump, people, one, president, like, donald, twitter, news, white
Topic 4: -4.181 (indicates FAKE news)
Top words: police, gun, said, shooting, old, media, man, officers, year
Topic 6: 3.499 (indicates TRUE news)
Top words: said, trump, united, president, states, north, korea, russia, iran
Topic 2: 3.252 (indicates TRUE news)
rop words: said, house, bill, would, senate, republicans, republican, may, president
Topic 8: 3.236 (indicates TRUE news)
Top words: said, government, reuters, people, security, state, military, killed, police
Topic 5: 2.637 (indicates TRUE news)
Top words: said, party, government, state, election, would, political, coalition, reuters
Topic 3: 1.325 (indicates TRUE news)
rop words: said, trump, court, president, house, state, former, clinton, department
Topic 9: 1.221 (indicates TRUE news)
Fop words: said, would, tax, new, percent, trump, year, million, states
Topic 0: -0.753 (indicates FAKE news)
Top words: trump, clinton, said, republican, campaign, hillary, president, party, election
Topic 7: 0.346 (indicates TRUE news)
Top words: said, china, chinese, national, taiwan, cuba, beijing, obama, president
```

Social media(-10.345) and sensational content (-4.181) are strong predictors of fake news. International affairs (3.499) and legislative content (3.252) strongly indicate true news.

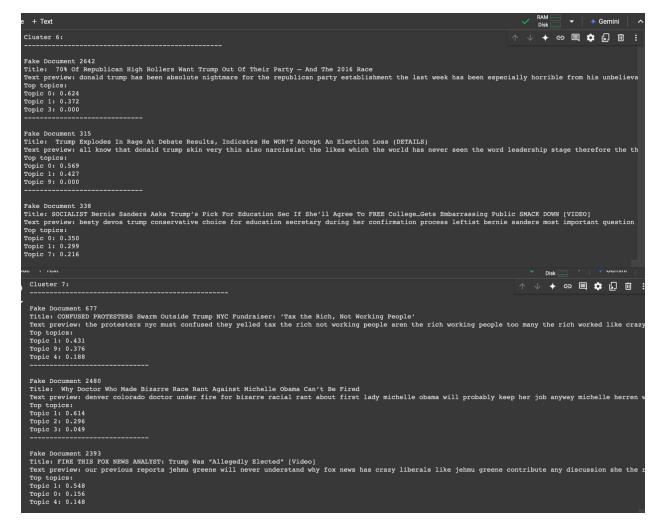
4. Pick real news or fake news, whichever is more interesting to you. Then, use the LDA vectors for those news documents to cluster them. You can use KMeans clustering with a reasonable value for K (if you don't have strong feelings for a particular K, I recommend 10). Then, select 5 news documents from each resulting cluster. Do the clusters correspond to anything? (One sentence)

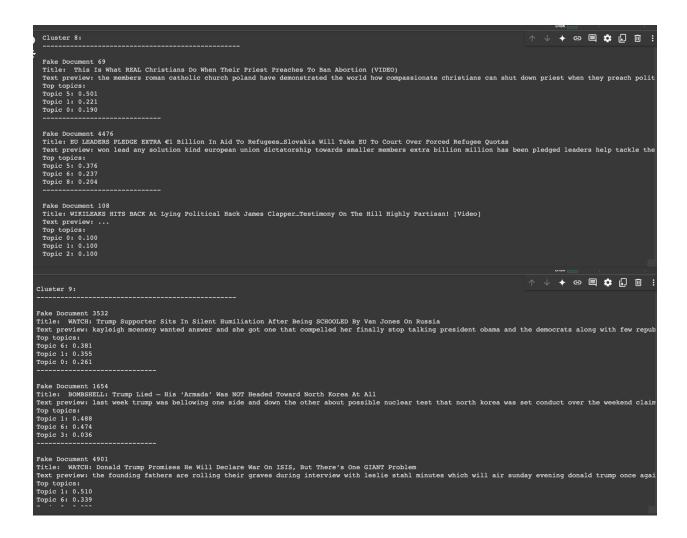






```
Cluster 2:
Fake Document 4358
Title: YOU'RE FIRED! WHY THE WHITE HOUSE Just Fired A Senior National Security Aide
Text preview: national security council aide craig deare was dismissed friday after was learned that harshly criticized the president and his top aides dea
Top topics:
Topic 3: 0.560
Topic 6: 0.198
Topic 8: 0.129
Fake Document 3571
Title: Donald Trump Is An Illegitimate President-Elect Because Russia Helped Him Steal Election
Text preview: russian interference our election should immediately disqualify donald trump from the presidency friday cia assessment found that the consens
Top topics:
Topic 0: 0.464
Topic 3: 0.442
Topic 1: 0.093
Fake Document 4878
Title: BOOM! JOHN SUNUNU: "Bothers Me That Mueller Is Hiring "Blatantly Biased Lawyers" [Video]
Text preview: former governor john sununu let alison camarata have when she asked about the russia investigation and robert mueller sununu said that trump
Top topics:
Topic 3: 0.741
Topic 1: 0.145
                                                                                                                                                                                                                                              Cluster 3:
  Fake Document 162
Title: McCain's Mad World and The Cancer of Conflict
Text preview: 21st century wire says some devastating news befell john sidney mccain iii recently his staff announced that the senator had been diagnosed to
  Top topics:
Topic 6: 0.237
Topic 4: 0.231
Topic 8: 0.216
  Fake Document 602
Title: HUGE SECURITY LAPSE: International Flight Passengers Skip Customs
Text preview: skipping customs jfk might seem like great thing for passengers arriving international flight but those passengers might putting all american
Top topics:
Topic 8: 0.558
Topic 1: 0.379
Topic 3: 0.061
  Fake Document 1135
Title: WHILE OBAMA TRIES TO DISARM AMERICANS, Israeli Ministers Encourage Citizens To Carry To Eliminate Enemy
Text preview: the destruction america full swing pay billions deliver unchecked muslims refugees the shores america meanwhile real men israel are arming de
Top topics:
Topic 8: 0.416
Topic 6: 0.257
Topic 6: 0.18R
                                                                                                                                                                                                                                             ↑ ↓ ← ⇔ 🗏 🗓 🗓
  Fake Document 505
Title: Target Just Made A Major Move That's Going To Blow Conservative Minds (VIDEO)
Text preview: conservatives who are notoriously anti labor are going outraged the move target just made target will raising their national minimum wage sta
Top topics:
Topic 9: 0.669
Topic 1: 0.257
Topic 7: 0.058
  Fake Document 982
Title: EXPLODING AFRICAN REFUGEE POPULATION STRESSING WELFARE SYSTEM IN MINNESOTA Are Sending Millions Of Dollars Back To Africa
Text preview: not just the our open southern borders need concerned about our state department who seems hell bent populating our states with refugees from
Top topics:
Topic 9: 0.528
Topic 1: 0.317
Topic 8: 0.152
  Fake Document 4296
Title: "FAIR SHARE" FAIL: Trump's Taxes Show He Paid Almost TWICE The Rate Of Socialist Bernie Sanders Who Owns 3 Homes, Preaches Equality [VIDEO]
Text preview: watch tucker carlson point out that trump paid tax rate while socialist fair share bernie only paid tax rate...
Topi topics:
Topic 9: 0.468
Topic 1: 0.331
Topic 0: 0.157
```





K=10, KMeans, choose fake news:

Cluster 0:

Dominated by Topic 0 (campaign/election)

Heavy focus on Trump, Clinton, and election coverage

Cluster 1:

Dominated by Topic 1 (social media/Trump)

Very high Topic 1 scores (0.650-0.984)

Critical/negative coverage of Trump

Cluster 2:

High Topic 3 scores (legal/court)

investigations, legal cases. Many Russia investigation related.

Cluster 3:

Mix of Topics 8 (security), 4 (incidents), and 6 (international)

Focus on national security, military, and international threats

Cluster 4:

High Topic 9 scores (economic/tax)

about welfare, education, economics

## Cluster 5:

Dominated by Topic 4 (police/incidents)
Sensational crime stories

- Your answers to the usual questions:
  - How long did this assignment take you? (1 sentence)2 days
  - O Whom did you work with, and how? (1 sentence each)
    - Discussing the assignment with others is encouraged, as long as you don't share the code.

On my own

- o Which resources did you use? (1 sentence each)
  - For each, please list the URL and a brief description of how it was useful. Data source: <a href="https://www.kaggle.com/datasets/clmentbisaillon/fake-and-real-news-dataset">https://www.kaggle.com/datasets/clmentbisaillon/fake-and-real-news-dataset</a>

Course code:

https://colab.research.google.com/drive/1q8Cf9QUwd45aWkOyPVjljOVb2N4xdUj5?usp=sharing#scrollTo=-V0J04mIOc5m

https://colab.research.google.com/drive/1paraIRBL87aYdG9dtR6x2fs0P4QBEmHe?usp=sharing#scrollTo=Wi6X76BoaEN9

## LatentDirichletAllocation:

https://scikitlearn.org/1.5/modules/generated/sklearn.decomposition.LatentDirichletAllocation.html

## CountVectorizer:

https://scikitlearn.org/1.5/modules/generated/sklearn.feature\_extraction.text.CountVectorizer.html

- O A few sentences about:
  - What was the most difficult part of the assignment?
     Understand the logic of topic modeling:
     Raw Text → CountVectorizer → Word Counts → LDA → Topic
     Distributions

```
# CountVectorizer: Converts text to word counts
vec = CountVectorizer(stop_words=['the', 'a', ...])
X = vec.fit_transform(df['document'])
```

# LatentDirichletAllocation: Finds topics in word counts
Ida = LatentDirichletAllocation(n\_components=10)
doc\_topics = Ida.fit\_transform(X)

■ What was the most rewarding part of the assignment?

Practice using cluster and LDA model

- What did you learn doing the assignment? How to handle the when run time is too long (limit the number of documents by random selections)
- Constructive and actionable suggestions for improving assignments, office hours, and class time are always welcome.