Group-4

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Problem Statement

Analysis of News Data: Verifying Source Credibility and Truthfulness

This project will analyze a dataset of thousands of fact-checked news headlines to analyse the following points

Analysis to be performed:

- 1. Veracity count e.g. Number of false news, true news and other categories
- 2. Top 3 sources of most false and true statements e.g. News, social media etc

- 3. Top 3 originators of false and true statements which could be a person or post
- 4. Top 3 Month-Year with most false news by count and percentage
- 5. Top 5 keywords found in false news headlines

Dataset & Source Information

The dataset has been obtained from Kggle

https://www.kaggle.com/datasets/rmisra/politifact-fact-check-dataset

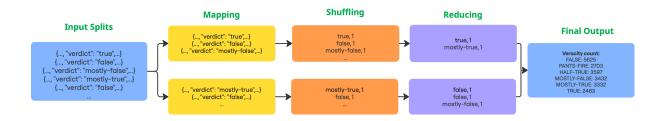
The data has been gathered from a website <u>PolitiFact</u> which fact checks the news. The news statements has been categorised into 6 categories: true, mostly true, half true, mostly false, false, and pants on fire

It has more than 21k news headlines fact checked

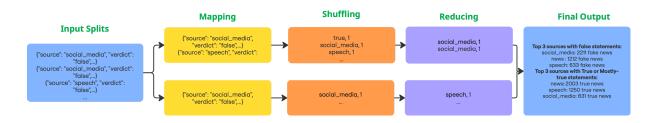
Source of dataset is <u>rishabhmisra.github.io/publications</u>

Map-Reduce Diagrams for each analysis task

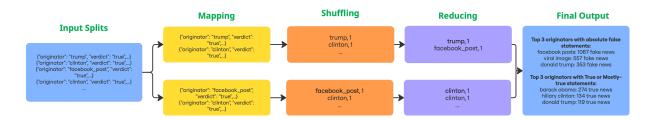
Veracity Count



Top Sources with false/true statements



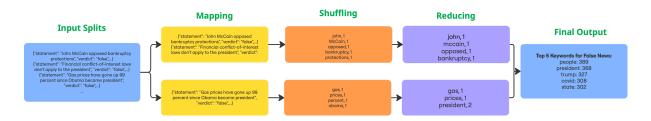
Top Originators with false/true statements



Top Months with false statements



Originators with false/true statements



Code

Pseudo Code

mapper.py

```
STOP_WORDS = {set of common English words}
def extract_month_year(date_string):
  # extract month-year from date string
def tokenize_and_filter_stop_words(text):
  # filter the line to remove the stop words and create a list of keywords, it
will remove words like a, an, the, is, and etc etc and keep keywords
for each line in standard input:
    row = json.loads(line) # parse the JSON
    # Extract the required vars from JSON
    headline = "<statement>"
    source = "<statement_source>"
    statement_originator = "<statement_originator>"
    posted_on = "<statement_date>"
    label = "<verdict>"
    # print source, label, originator, month-year
    if news is false:
         # print source, originator, month-year of news
         # Process the headline for keywords
      words = tokenize_and_filter_stop_words(headline)
      for word in words:
           print("false_news_keyword\t", word)
```

if news is true:

print source, originator

reducer.py

Declare dicts to keep count of label, source, originator, source with fake news and true news, originator with fake and true news, month-year, mont h-year of fake news, keywords in fake news

for line in sys.stdin:

split the input on "\t"

count each parameter from mapper and store in the dict declared above

print count of label

Sort and print top 3 sources of fake news

Sort and print top 3 sources of true news

Sort and print top 3 originators of fake news

Sort and print top 3 originators of true news

Sort and print top 3 month-year with most number of fake news overall

for month_year in month_year_fake_news_count:

Get total news for the month_year from month_year count

Get fake news count for the month_year from month_year_fake_news_count

Let's consider months-year with at least 10 news if total_news > 10:

calculate percentage of fake news

Sort and print top 3 month-year with highest percentage of fake news Sort and print top 5 keywords in fake news

Functional code

The code is available in following repository aditya-bits-cc/BDS-assignment: Assignment 1 of BDS

stop_words.json

```
{
 "stop_words": [
  "call", "upon", "still", "nevertheless", "down", "every", "forty", "'re", "alwa
ys", "whole", "side",
  "n't", "now", "however", "an", "show", "least", "give", "below", "did", "so
metimes", "which", "'s",
  "nowhere", "per", "hereupon", "yours", "she", "moreover", "eight", "some
where", "within", "whereby",
  "few", "has", "so", "have", "for", "noone", "top", "were", "those", "thenc
e", "eleven", "after", "no",
  "'Il", "others", "ourselves", "themselves", "though", "that", "nor", "just",
"'s", "before", "had",
  "toward", "another", "should", "herself", "and", "these", "such", "elsewhe
re", "further", "next", "indeed",
  "bottom", "anyone", "his", "each", "then", "both", "became", "third", "who
m", "'ve", "mine", "take", "many",
  "anywhere", "to", "well", "thereafter", "besides", "almost", "front", "fiftee
n", "towards", "none", "be",
  "herein", "two", "using", "whatever", "please", "perhaps", "full", "ca", "w
e", "latterly", "here",
  "therefore", "us", "how", "was", "made", "the", "or", "may", "'re", "namel
y", "'ve", "anyway", "amongst",
  "used", "ever", "of", "there", "than", "why", "really", "whither", "in", "onl
v", "wherein", "last", "under",
  "own", "therein", "go", "seems", "'m", "wherever", "either", "someone", "u
p", "doing", "on", "rather",
  "ours", "again", "same", "over", "'s", "latter", "during", "done", "'re", "put",
"'m", "much", "neither",
  "among", "seemed", "into", "once", "my", "otherwise", "part", "everywher
e", "never", "myself", "must", "will",
```

```
"am", "can", "else", "although", "as", "beyond", "are", "too", "becomes",
"does", "a", "everyone", "but",
  "some", "regarding", "'II", "against", "throughout", "yourselves", "him",
"'d", "it", "himself", "whether",
  "move", "'m", "hereafter", "re", "while", "whoever", "your", "first", "amoun
t", "twelve", "serious", "other",
  "any", "off", "seeming", "four", "itself", "nothing", "beforehand", "make",
"out", "very", "already", "various",
  "until", "hers", "they", "not", "them", "where", "would", "since", "everythin
g", "at", "together", "yet", "more",
  "six", "back", "with", "thereupon", "becoming", "around", "due", "keep",
"somehow", "n't", "across", "all",
  "when", "i", "empty", "nine", "five", "get", "see", "been", "name", "betwee
n", "hence", "ten", "several", "from",
  "whereupon", "through", "hereby", "'ll", "alone", "something", "formerly",
"without", "above", "onto", "except",
  "enough", "become", "behind", "'d", "its", "most", "n't", "might", "wherea
s", "anything", "if", "her", "via",
  "fifty", "is", "thereby", "twenty", "often", "whereafter", "their", "also", "any
how", "cannot", "our", "could",
  "because", "who", "beside", "by", "whence", "being", "meanwhile", "this",
"afterwards", "whenever", "mostly",
  "what", "one", "nobody", "seem", "less", "do", "'d", "say", "thus", "unles
s", "along", "yourself", "former",
  "thru", "he", "hundred", "three", "sixty", "me", "sometime", "whose", "yo
u", "quite", "'ve", "about", "even",
  "says", "said"
]
}
```

mapper.py

```
#!/usr/bin/python3
import sys
import json
import re
from datetime import datetime
```

```
FALSE_NEWS_LABELS = {"FALSE", "PANTS ON FIRE"}
TRUE_NEWS_NABELS = {"TRUE", "MOSTLY-TRUE"}
with open("stop_words.json") as f:
     STOP_WORDS = set((json.load(f)["stop_words"])
# Function to extract month and year
def extract_month_year(date_str):
  date = datetime.strptime(date_str.strip(), '%m/%d/%Y')
  return date.month, date.year
def tokenize_and_filter_stop_words(text):
  # Lowercase and split text into words
  words = re.findall(r'\b[a-zA-Z]+\b(?\'[a-zA-Z]+)?', text.lower())
  # Filter out stop words
  filtered_words = (word for word in words if word not in STOP_WORDS an
d len(word) > 1)
  return filtered words
# Reading input from stdin
for line in sys.stdin:
  line = line.strip()
  # line = line.encode('utf-8').decode('unicode_escape')
  # Parse the JSON object
  row = json.loads(line)
  # for row in reader:
  headline = row['statement']
  source = row['statement_source'].lower()
  statement_originator = row['statement_originator'].lower()
  posted_on = row['statement_date']
  label = str(row['verdict']).upper()
  # Count the labels
  print(f"label\t{label}")
  # Count the sources
  print(f"source\t{source}")
  # Count the person who gave the statement
```

```
print(f"statement_originator\t{statement_originator}")
  # Extract the month-year for the posted date
  month, year = extract_month_year(posted_on)
  if month and year:
    month_year = f"{month}-{year}"
    print(f"month_year\t{month}-{year}")
  if label in FALSE_NEWS_LABELS:
    print(f"fake_news_source\t{source}") # Print source if news is fake
    print(f"fake_news_statement_originator\t{statement_originator}") # Pri
nt statement_originator if news is fake
    print(f"fake_news_month_year\t{month_year}") # Print month-year Of
fake news
    # Process the headline for keywords
    words = tokenize_and_filter_stop_words(headline)
    for word in words:
       print(f"false_news_keyword\t{word}")
  if label in TRUE_NEWS_NABELS:
    print(f"true_news_source\t{source}") # Source name with true or mostl
y-true news
    print(f"true_news_statement_originator\t{statement_originator}") # Pri
nt statement_originator if news is true or mostly-true
```

reducer.py

```
#!/usr/bin/python3
import sys

# Dicts for keeping the count
label_counts = {}
source_counts = {}
statement_originator_counts = {}
source_fake_news_count = {}
```

```
source_true_news_count = {}
statement_originator_fake_news_count = {}
statement_originator_true_news_count = {}
month_year_counts = {}
month_year_fake_news_count = {}
month_year_percentage_fake_news = {}
false_news_keyword_counts = {}
# Reading the mapper output from stdin
for line in sys.stdin:
  line = line.strip()
  key, value = line.split('\t', 1)
  #1. Count labels
  if key == "label":
    if value in label_counts:
       label_counts[value] += 1
     else:
       label_counts[value] = 1
  # 2. Count sources
  elif key == "source":
     if value in source_counts:
       source_counts[value] += 1
     else:
       source_counts[value] = 1
  # 2. Count statement_originator
  elif key == "statement_originator":
     if value in statement_originator_counts:
       statement_originator_counts[value] += 1
     else:
       statement_originator_counts[value] = 1
  # 3. Count headlines posted per month-year
  elif key == "month_year":
     if value in month_year_counts:
```

```
month_year_counts[value] += 1
  else:
    month_year_counts[value] = 1
# 4. Sources with the most fake news (False, Pants on Fire)
elif key == "fake_news_source":
  if value in source_fake_news_count:
    source_fake_news_count[value] += 1
  else:
    source_fake_news_count[value] = 1
# 5. Sources with the most true news (True, Mostly-True)
elif key == "true_news_source":
  if value in source_true_news_count:
    source_true_news_count[value] += 1
  else:
    source_true_news_count[value] = 1
# statement_originator with most fake news
elif key == "fake_news_statement_originator":
  if value in statement_originator_fake_news_count:
    statement_originator_fake_news_count[value] += 1
  else:
    statement_originator_fake_news_count[value] = 1
# 5. statement_originator with the most true news (True, Mostly-True)
elif key == "true_news_statement_originator":
  if value in statement_originator_true_news_count:
    statement_originator_true_news_count[value] += 1
  else:
    statement_originator_true_news_count[value] = 1
# 6. Track fake news count per month-year
elif key == "fake_news_month_year":
  if value in month_year_fake_news_count:
    month_year_fake_news_count[value] += 1
  else:
    month_year_fake_news_count[value] = 1
```

```
#7. Fake news keyword count
  elif key == "false_news_keyword":
    if value in false_news_keyword_counts:
      false_news_keyword_counts[value] += 1
    else:
      false_news_keyword_counts[value] = 1
# Count of each label e.g. TRUE, FALSE, MOSTLY-TRUE etc
print("\nVeracity count:")
for label, count in label_counts.items():
  print(f"{label}) \rightarrow {count}")
# Top 3 sources with most absolute false news
source_fake_news_sorted = sorted(source_fake_news_count.items(), key=1
ambda x: x[1], reverse=True)
print("\nTop 3 sources with false statements:")
for i, (source, count) in enumerate(source_fake_news_sorted[:3]):
  print(f"{source}: {count} fake news")
# Top 3 Sources with most TRUE or MOSTLY-TRUE news
source_true_news_sorted = sorted(source_true_news_count.items(), key=la
mbda x: x[1], reverse=True)
print("\nTop 3 sources with True or Mostly-true statements:")
for i, (source, count) in enumerate(source_true_news_sorted[:3]):
  print(f"{source}: {count} true news")
# Top 3 originators with most absolute false news
source_fake_news_sorted = sorted(statement_originator_fake_news_count.i
tems(), key=lambda x: x[1], reverse=True)
print("\nTop 3 originators with absolute false statements:")
for i, (source, count) in enumerate(source_fake_news_sorted[:3]):
  print(f"{source}: {count} fake news")
# Top 3 originators with most TRUE or MOSTLY-TRUE news
source_true_news_sorted = sorted(statement_originator_true_news_count.it
ems(), key=lambda x: x[1], reverse=True)
print("\nTop 3 originators with True or Mostly-true statements:")
```

```
for i, (source, count) in enumerate(source_true_news_sorted[:3]):
  print(f"{source}: {count} true news")
# Month-year with most fake news
month_year_fake_news_sorted = sorted(month_year_fake_news_count.item
s(), key=lambda x: x[1], reverse=True)
print("\nTop 3 Month-Year with the highest overall false news count:")
for i, (month_year, count) in enumerate(month_year_fake_news_sorted[:3]):
  print(f"{month_year}: {count} fake news out of {month_year_counts.get
(month_year, 0)} news")
# Calculate the month-year with the most percentage of fake news
for month_year in month_year_fake_news_count:
  # Get the total headlines for this month-year
  total_headlines = month_year_counts.get(month_year, 0)
  fake_news_count = month_year_fake_news_count.get(month_year, 0)
  # Calculate the percentage of fake news for this month-year
  if total_headlines > 10:
    fake_news_percentage = (fake_news_count / total_headlines) * 100
    month_year_percentage_fake_news[month_year] = fake_news_percent
age
# Sort month-year by the percentage of fake news
sorted_month_year_percentage_fake_news = sorted(month_year_percentag
e_fake_news.items(), key=lambda x: x[1], reverse=True)
# month-year with the highest percentage of fake news
print("\nTop 3 Month-Year with the highest percentage of false news (consi
dering months with minimum of 10 total news):")
for month_year, percentage in sorted_month_year_percentage_fake_news[:
3]:
  print(f"{month_year}: {percentage:.2f}% fake news")
print("\nTop 5 Keywords for False News:")
sorted_keywords = sorted(false_news_keyword_counts.items(), key=lambd
a x: x[1], reverse=True)
```

for i, (keyword, count) in enumerate(sorted_keywords[:5]):
 print(f"{keyword}: {count}")

Commands executed to perform the analysis:

Download the dataset

curl -L -o politifact-fact-check-dataset.zip https://www.kaggle.com/api/v1/datasets/download/rmisra/politifact-fact-check-dataset

Unzip and rename the file unzip politifact-fact-check-dataset.zip mv politifact_factcheck_data.json newsdata.json

Put the dataset in hdfs hadoop fs -mkdir /newsdata hadoop fs -put newsdata.json /newsdata

Run the job

hadoop jar /opt/hadoop-3.2.4/share/hadoop/tools/lib/hadoop-streaming-3. 2.4.jar -file stop_words.json -file mapper.py -file reducer.py -mapper "pyth on3 mapper.py" -reducer "python3 reducer.py" -input /newsdata/newsdat a.json -output /newsdata/analysis

Print the analysis output hadoop fs -cat /newsdata/analysis/*

Output screenshot

```
2025-03-17 18:23:22,484 INFO streaming.StreamJob: Output directory: /newsdata/analysis
[centos@master ~15 hadoop fs ~cat /newsdata/analysis/part*

Veracity count:
FALSE: 5025

PANTS-FIRE: 2703

HALF-TRUE: 3597

MOSTLY-FALSE: 3432

MOSTLY-FALSE: 3432

MOSTLY-FALSE: 3432

Top 3 sources with false statements:
social_media: 2211 fake news
news: 1212 fake news

speech: 633 fake news

Top 3 sources with True or Mostly-true statements:
news: 2003 true news
speech: 1250 true news
speech: 1250 true news
speech: 1250 true news
social_media: 631 true news

Top 3 originators with absolute false statements:
facebook posts: 1807 fake news

donald trump: 355 fake news

Top 3 originators with True or Mostly-true statements:
barack obama: 274 true news

folal media: 631 true news

social_media: 631 true news

Top 3 originators with True or Mostly-true statements:
facebook posts: 1807 fake news

donald trump: 1955 fake news

donald trump: 1959 fake news out of 217 news

Top 3 Month-Year with the highest false news count:
10-2020: 104 fake news out of 217 news

Top 3 Month-Year with the highest percentage of false news (considering months with minimum of 10 total news):
7-2022: 63.64% fake news
7-2021: 61.59% fake news
7-2021: 61.59% fake news
7-2021: 61.59% fake news
7-2021: 62.50% fake news
7-2021: 61.59% fake news
7-2021: 61.50% fake news
7
```

Output screenshot

Execution Statistics

- Number of Map tasks: 2
- Number of Reduce tasks: 1
- Memory consumption per task:
 - Peak Map Physical memory (bytes)=737529856
 - Peak Map Virtual memory (bytes)=3020111872
 - Peak Reduce Physical memory (bytes)=218513408
 - Peak Reduce Virtual memory (bytes)=4726763520
- Bytes Transferred (Reduce shuffle bytes)
 - Map Output Materialized Bytes: 4,444,908 bytes.
 - Reduce Shuffle Bytes: 4,444,908 bytes.

Execution statistics screenshot

```
| International Content | Standoop | Jar /opt/hadoop-2.4 //share/hadoop/tots/lb/hadoop-streaming-3.2 //shar-file mapper.py -file reducer.py -mapper "python3 mapper.py" -reducer "python3 reducer.py" -input /memodata/annoyis.ps. jon-output /memodata/annoyis.ps. jon-
```

Execution statistics 1

```
Map input records=164721
Map output records=164721
Map output bytes=4115454
Map output bytes=4115454
Map output materialized bytes=4444908
Input split bytes=186
Combine input records=0
Reduce input groups=10
Reduce shuffle bytes=4444908
Reduce input records=164721
Reduce output records=154721
Reduce output records=45
Spilled Records=329442
Shuffled Maps =2
Failed Shuffles=0
Merged Map outputs=2
GC time elapsed (ms)=182
CPU time spent (ms)=5450
Physical memory (bytes) snapshot=1693425664
Virtual memory (bytes) snapshot=10764891104
Total committed heap usage (bytes)=1646264320
Peak Map Physical memory (bytes)=737529856
Peak Map Virtual memory (bytes)=3020111872
Peak Reduce Physical memory (bytes)=3020111872
Peak Reduce Physical memory (bytes)=4726763520
Shuffle Errors
BAD_ID=0
CONNECTION=0
IO_ERROR=0
WRONG_REDUCE=0
File Input Format Counters
Bytes Read=9842215
File Output Format Counters
Bytes Read=9842215
File Output Format Counters
Bytes Written=1116
2025=03-17 18:23:22, 484 INFO streaming.StreamJob: Output directory: /newsdata/analysis
[centos@master ~]$ |
```

Execution statistics 2