Automotive

Banking

Biotech

EDS Theory Assignment

Blog Authorship Corpus Data Analysis Name: Atharva Gajbhiye Divison:CS3 Roll_no: CS3-42 PRN: 202401040259 Loading the dataset. #mount the google datset from google.colab import drive drive.mount('/content/drive') # load the dataset !pip install numpy pandas import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt path='/content/drive/MyDrive/blogtext.csv' df=pd.read_csv(path) df.info() 🕁 Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True). Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (2.0.2) Requirement already satisfied: pandas in /usr/local/lib/python3.11/dist-packages (2.2.2) Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas) (2.8.2) Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas) (2025.2) Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas) (2025.2) Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->pandas) (1.17.0) <class 'pandas.core.frame.DataFrame'> RangeIndex: 681284 entries, 0 to 681283 Data columns (total 7 columns): # Column Non-Null Count Dtype --------0 id 681284 non-null int64 1 gender 681284 non-null object 681284 non-null int64 2 age 3 topic 681284 non-null object 681284 non-null object sign 681284 non-null object date 6 text 681284 non-null object dtypes: int64(2), object(5) memory usage: 36.4+ MB Problem statements 1. Display First 15 entries and and show variance of age distribution across industries. #1. Display first 15 entries and show variance of age distribution across industries. print("Problem 1: Display first 15 entries and show variance of age distribution across industries.") df.head(15) variance=df.groupby('topic')['age'].var().round(2) print(variance) Problem 1: Display first 15 entries and show variance of age distribution across industries. →▼ topic Accounting 108.83 Advertising 55.32 Agriculture 41.81 Architecture 36.43 Arts 47.94

59.21

17.43 44.71

```
BusinessServices
                            64.36
Chemicals
                            21.13
Communications-Media
                             45.34
Construction
                            65.73
Consulting
                            90.75
Education
                            64.55
Engineering
                            20.30
Environment
                            27.98
Fashion
                            63.66
Government
                            55.00
HumanResources
                            68.15
Internet
                            65.04
{\tt InvestmentBanking}
                            37.83
                            24.04
Law
LawEnforcement-Security
                            72.85
Manufacturing
                            35.69
Maritime
                            60.52
Marketing
                            30.80
                            83.15
Military
Museums-Libraries
                            77.65
Non-Profit
                            54.95
Publishing
                            51.46
RealEstate
                            56.76
Religion
                            71.72
                            22.40
Science
Sports-Recreation
                            45.12
Student
Technology
                            45.92
Telecommunications
                            41.67
Tourism
                            48.90
Transportation
                            47.31
indUnk
                            63.86
Name: age, dtype: float64
```

2. Display data type of each column.

```
#2. Display data type of each column.
print("Problem 2: Display data type of each column")
df.dtypes
```

Problem 2: Display data type of each column

```
id int64
gender object
age int64
topic object
sign object
date object
text object
```

3. Checking for missing values in all columns.

```
#Checking for missing values in all columns.
print("Problem 3: Checking for missing values in all columns.")
df.isnull().sum()
```

```
→ Problem 3: Checking for missing values in all columns.
             0
             n
       id
     gender 0
             0
      age
      topic
             0
             0
      sian
             0
      date
      text
             0
     dtuna inte/
```

4. Find the unique bloggers by gender and and list them all.

Double-click (or enter) to edit

5. Find the minimun and maximum bloggers age.

```
#Find the minimun and maximum bloggers age.
print("Problem 5: Find the minimun and maximum bloggers age.")
print(df['age'].head(5))
print("Minimum age:",df['age'].min())
print("Maximum age:",df['age'].max())
     Problem 5: Find the minimun and maximum bloggers age.
         15
     1
          15
     2
          15
     3
          15
          33
     Name: age, dtype: int64
     Minimum age: 13
     Maximum age: 48
6. Calculate the average, median, and mode of the blooger's age.
#Calculate the average , median, and mode of the blooger's age.
print("Problem 6: Calculate the average , median, and mode of the blooger's age.")
print("Average age:",np.mean(df['age']).round(2))
print("Median age:",np.median(df['age']))
print("Mode age:",df['age'].mode()[0])
    Problem 6: Calculate the average , median, and mode of the blooger's age.
     Average age: 23.93
```

7. Gender distribution in percentage.

Median age: 24.0 Mode age: 17

```
#Top 10 industries
print("Problem 8: Top 10 industries by blogger count\n")
print(df['topic'].value_counts().head(10))
```

→ Problem 8: Top 10 industries by blogger count

```
topic
indUnk
                         251015
Student
                         153903
                          42055
Technology
                          32449
Arts
Education
                          29633
Communications-Media
                          20140
                          16006
Internet
Non-Profit
                          14700
                          11653
Engineering
                           9040
Law
Name: count, dtype: int64
```

9. Drop duplicate blogger IDs

```
#Drop duplicate blogger IDs
print("Problem 9: Drop duplicate IDs\n")
df = df.drop_duplicates(subset='id')
print("Remaining records after duplicate removal:",len(df))
print(df[['id','gender','age','topic','sign','date']].head(15))
```

→ Problem 9: Drop duplicate IDs

```
Remaining records after duplicate removal: 19320
          id gender age
                                                      sign
                                                                       date
                                       topic
                                                               14,May,2004
a
     2059027
                male
                       15
                                      Student
                                                       Leo
4
     3581210
                male
                       33
                           InvestmentBanking
                                                  Aquarius
                                                               11, June, 2004
    3539003
              female
                                       indUnk
                                                     Aries
                                                               07, June, 2004
                       14
95
     4172416
              female
                       25
                                       indUnk
                                                 Capricorn 08, August, 2004
97
     3668238
              female
                       17
                                      Student
                                                    Gemini
                                                               30, June, 2004
107 4030905
                                      Student
                                                               31,July,2004
              female
                       17
                                                     Aries
132 4198080
                                       indUnk
                                                  Aquarius 10, August, 2004
              female
                       23
133 3705830
                male
                       25
                                   Non-Profit
                                                    Cancer
                                                               29, June, 2004
179 3649763
                                      Banking
                                                               20,June,2004
              female
                                                  Aquarius
                                                                23,May,2004
     3389918
                       37
                                       indUnk
192
              female
                                                  Aquarius
                                       indUnk
                                               Sagittarius 23, August, 2004
211
    4304458
              female
                       25
214 3429420
                male
                       15
                                      Student
                                                  Aquarius
                                                                30, May, 2004
     3389671
                male
                       26
                                       indUnk
                                                       Leo
                                                                28, May, 2004
283
     649790
             female
                       24
                                       indUnk
                                                   Scorpio
                                                                31.May.2002
474 3022585
             female
                       27
                                    Education
                                                  Aquarius 01, August, 2004
```

10. Create 'decade' column based on age

```
#Create 'decade' column based on age
print("Problem 10: Creating 'decade' column\n")
df.loc[:,'decade'] = (df['age'] // 10) * 10
print(df[['age', 'decade']].head(15))
```

```
→ Problem 10: Creating 'decade' column
               decade
     0
           15
                   10
     4
           33
                   30
     74
                   10
           14
     95
           25
                   20
     97
           17
     107
           17
                   10
     132
           23
                   20
     133
           25
                   20
     179
           33
                   30
     192
           37
                   30
     211
           25
                   20
     214
           15
                   10
     248
           26
                   20
     283
           24
                   20
     474
           27
                   20
  11. Number of bloggers with posts exceeding 5000 words
#Number of bloggers with posts exceeding 5000 words
print("Problem 11: Bloggers with posts >5000 words\n")
df.loc[:,'word_count'] = df['text'].apply(lambda x: len(str(x).split()))
print(f"Bloggers\ with\ \gt{5000}\ words:\ \{(df['word\_count']\ \gt{5000}).sum()\}")
print(df['word_count'])
→ Problem 11: Bloggers with posts >5000 words
     Bloggers with >5000 words: 20
     0
                28
     4
                65
     74
               185
     95
                16
     97
                41
     681004
               257
     681025
               874
     681141
                 7
     681156
               415
     681161
               13
     Name: word_count, Length: 19320, dtype: int64
12.Create 'text_length' column
#Create 'text length' column
print("Problem 12: Create 'text_length' (characters)\n")
df.loc[:,'text_length'] = df['text'].apply(lambda x: len(str(x)))
display(df[['id', 'text_length']].head())
Problem 12: Create 'text_length' (characters)
               id text_length
                                  ▦
      0 2059027
                           157
         3581210
                           402
      74 3539003
                           932
                           103
      95 4172416
      97 3668238
                           476
  13. Calculate average number of sentences per blog post
```

```
#Calculate average number of sentences per blog post
print("Problem 13: Average sentences per post\n")
df.loc[:,'sentence_count'] = df['text'].apply(lambda x: str(x).count('.')+str(x).count('!')+str(x).count('?'))
print("Average sentences per blog post:" ,df['sentence_count'].mean().round(2))
→ Problem 13: Average sentences per post
     Average sentences per blog post: 33.63
```

14. Average text length by gender

```
#Average text length by gender
print("Problem 9: Average text length by gender")
print(df.groupby('gender')['text_length'].mean().round(2))

Problem 9: Average text length by gender
gender
female 1596.30
male 1612.39
Name: text_length, dtype: float64

15. Most common industry by decade
```

#Most common industry by decade
print("Problem 15: Most common industry per age decade\n")
common = df.groupby('decade')['topic'].agg(lambda x: x.mode()[0])

display(common)

→ Problem 15: Most common industry per age decade

topic

decade	
10	Student
20	indUnk
30	indUnk
40	indUnk
ما بمصد ما	siaat

16. Create a pivot table: average age and average word count per gender and industry

```
#Create a pivot table: average age and average word count per gender and industry
print("Problem 16: Pivot table of age and word count\n")
pivot = pd.pivot_table(df.head(30), index=['gender', 'topic'], values=['age', 'word_count'], aggfunc='mean')
print(pivot.round(2))
```

→ Problem 16: Pivot table of age and word count

		age	word_count
gender	topic		
female	Banking	33.00	196.00
	Education	27.00	550.00
	Student	16.33	248.67
	indUnk	26.00	472.92
male	BusinessServices	24.00	386.00
	Communications-Media	41.00	723.00
	Engineering	24.00	30.00
	InvestmentBanking	33.00	65.00
	Non-Profit	25.00	270.00
	Science	15.00	47.00
	Sports-Recreation	17.00	95.00
	Student	15.25	242.75
	indUnk	24.50	246.00

17. Blogger with the maximum number of posts

```
#Blogger with the maximum number of posts
print("Problem 17: Blogger with most posts\n")
topposts = df['id'].value_counts().idxmax()
print("Blogger with most posts:",topposts)
```

 \rightarrow Problem 17: Blogger with most posts

Blogger with most posts: 1713845

18. Percentage of short posts greater than 200 words

```
# Percentage of short posts greater than 200 words
print("Problem 18: Percentage of short posts (<200 words)\n")</pre>
short_posts = (df['word_count'] < 200).mean() * 100</pre>
print("Percentage:", short_posts.round(2))
→ Problem 18: Percentage of short posts (<200 words)
     Percentage: 54.47
  19. How many blog posts mention 'life'.
#How many blog posts mention 'life'
print("Problem 19: Posts containing 'life'\n")
life_posts = df['text'].str.lower().str.contains('life')
print("Number of posts mentioning 'life':", life_posts.sum())
→ Problem 19: Posts containing 'life'
     Number of posts mentioning 'life': 4457
 20. Create a new column of name Word density (words per character).
#Create a new column of name Word density (words per character).
print("Problem 20: Word density (words per character)")
df.loc[:,'word_density'] = df['word_count'] / df['text_length']
print(df[['word_density']].head(15))
    Problem 20: Word density (words per character)
          word_density
     0
              0.178344
     4
              0.161692
     74
              0.198498
     95
              0.155340
     97
              0.086134
     107
              0.164539
              0.193997
     132
     133
              0.194665
     179
              0.177215
     192
              0.187382
     211
              0.173625
     214
              0.181200
     248
              0.184008
              0.147826
     283
     474
              0.187521
```

Thank You