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Roll no:-ET1-37
Dataset:-WordNet

1. Display 1st five rows of the data set

```
[4] df.head()
```

	Word	Count	Term 1	Term 2	Term 3	Term 4	Term 5	Term 6	Term 7
0	's gravenhagen	13	The Hague	's Gravenhage	Den Haag	city (generic term)	metropolis (generic term)	urban center (generic term)	NaN
1	tween decks	12	between decks	NaN	NaN	NaN	NaN	NaN	NaN
2	0.22	4	twenty-two	firearms (generic term)	piece (generic term)	small-arm (generic term)	NaN	NaN	NaN
3	22-calibre	11	22 caliber	22-caliber	22 calibre	diameter	diam (related term)	NaN	NaN
4	22 caliber	11	22-caliber	22 calibre	22-calibre	diameter	diam (related term)	NaN	NaN

2. Display last 5 rows of the data set

```
df.tail()
```

	Word	Count	Term 1	Term 2	Term 3	Term 4	Term 5	Term 6	Term 7
204675	zymosis	7	zymolysis	fermentation	fermenting	ferment	chemical process (generic term)	chemical change (generic term)	chemical action (generic term)
204676	zymosis	7	infection (generic term)	NaN	NaN	NaN	NaN	NaN	NaN
204677	zymotic	7	zymolytic	chemical process	chemical change	chemical action (related term)	NaN	NaN	NaN
204678	zymotic	7	infection (related term)	NaN	NaN	NaN	NaN	NaN	NaN
204679	zymurgy	7	biochemistry (generic term)	NaN	NaN	NaN	NaN	NaN	NaN

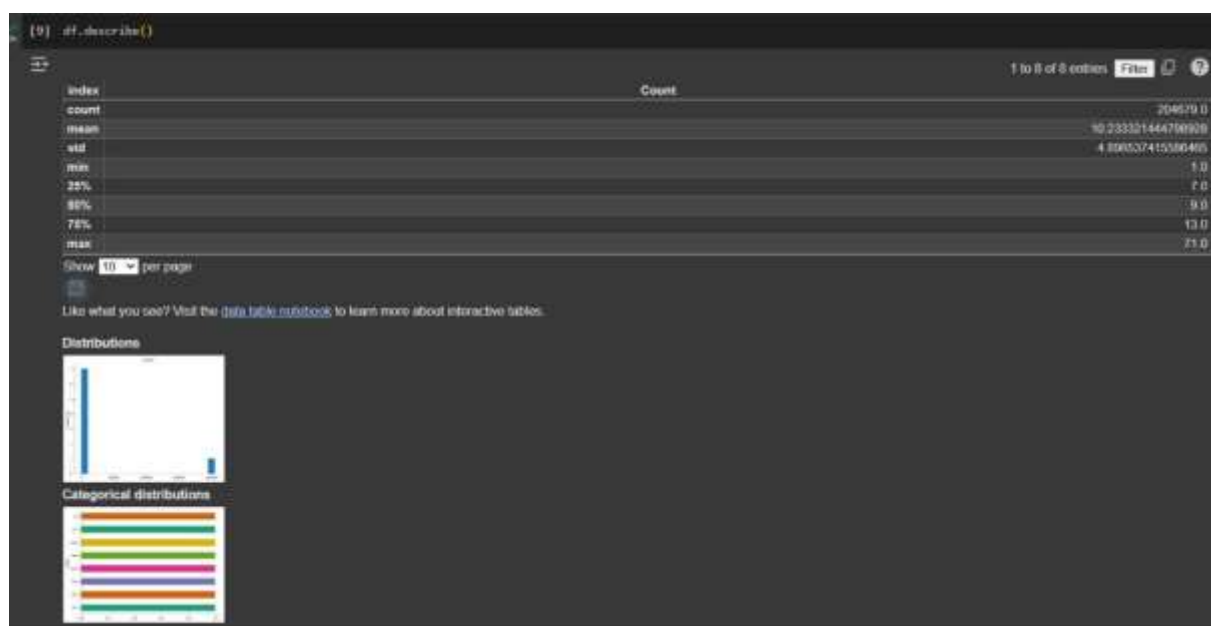
3. Easy to analyze, filter, and visualize large datasets.

[5] df

	Word	Count	Term 1	Term 2	Term 3	Term 4	Term 5	Term 6	Term 7
0	% greenhage	13	The Hague	% Greenhage	Den Haag	city (generic term)	metropolis (generic term)	urban center (generic term)	NaN
1	seven decks	12	between decks	NaN	NaN	NaN	NaN	NaN	NaN
2	0.22	4	heerly two	foam (generic term)	piece (generic term)	small arm (generic term)	NaN	NaN	NaN
3	22-calibre	11	22-caliber	22-caliber	22-calibre	diameter	dam (related term)	NaN	NaN
4	22-caliber	11	22-caliber	22-calibre	22-calibre	demoler	dam (related term)	NaN	NaN
...									
204675	zymosis	7	zymolysis	fermentation	fermenting	ferment	chemical process (generic term)	chemical change (generic term)	chemical action (generic term)
204676	zymosis	7	infection (generic term)	NaN	NaN	NaN	NaN	NaN	NaN
204677	zymotic	7	zymolytic	chemical process	chemical change	chemical action (related term)	NaN	NaN	NaN
204678	zymotic	7	infection (related term)	NaN	NaN	NaN	NaN	NaN	NaN
204679	zymurgy	7	biochemistry (generic term)	NaN	NaN	NaN	NaN	NaN	NaN

Source: [wordnet](#)

4. To describe the dataset

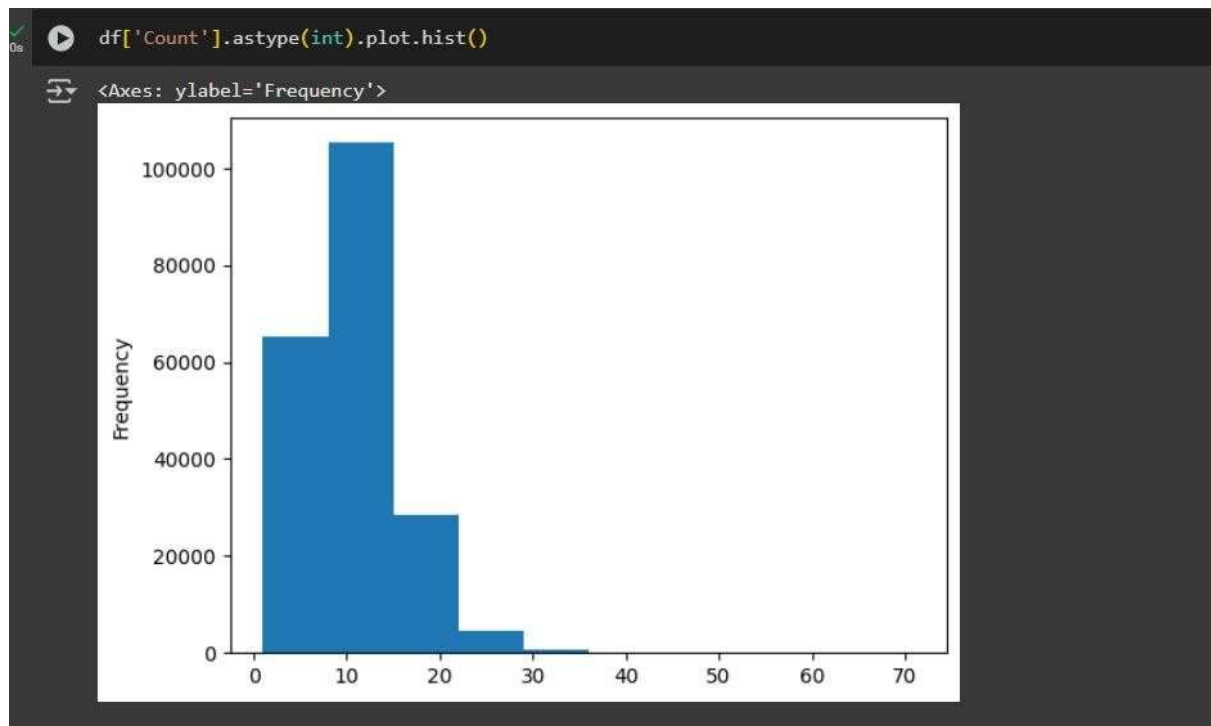


5. To display information about dataset

```
[11] df.info()

<class 'pandas.core.frame.DataFrame'>
Index: 204679 entries, 0 to 204679
Data columns (total 9 columns):
#   Column      Non-Null Count  Dtype
---  ---
0    Word        204674 non-null  object
1    Count       204679 non-null  int64
2    Term 1      204679 non-null  object
3    Term 2      178804 non-null  object
4    Term 3      148133 non-null  object
5    Term 4      93700 non-null   object
6    Term 5      60820 non-null   object
7    Term 6      38155 non-null   object
8    Term 7      24151 non-null   object
dtypes: int64(1), object(8)
memory usage: 15.6+ MB
```

6. Distribution of count values



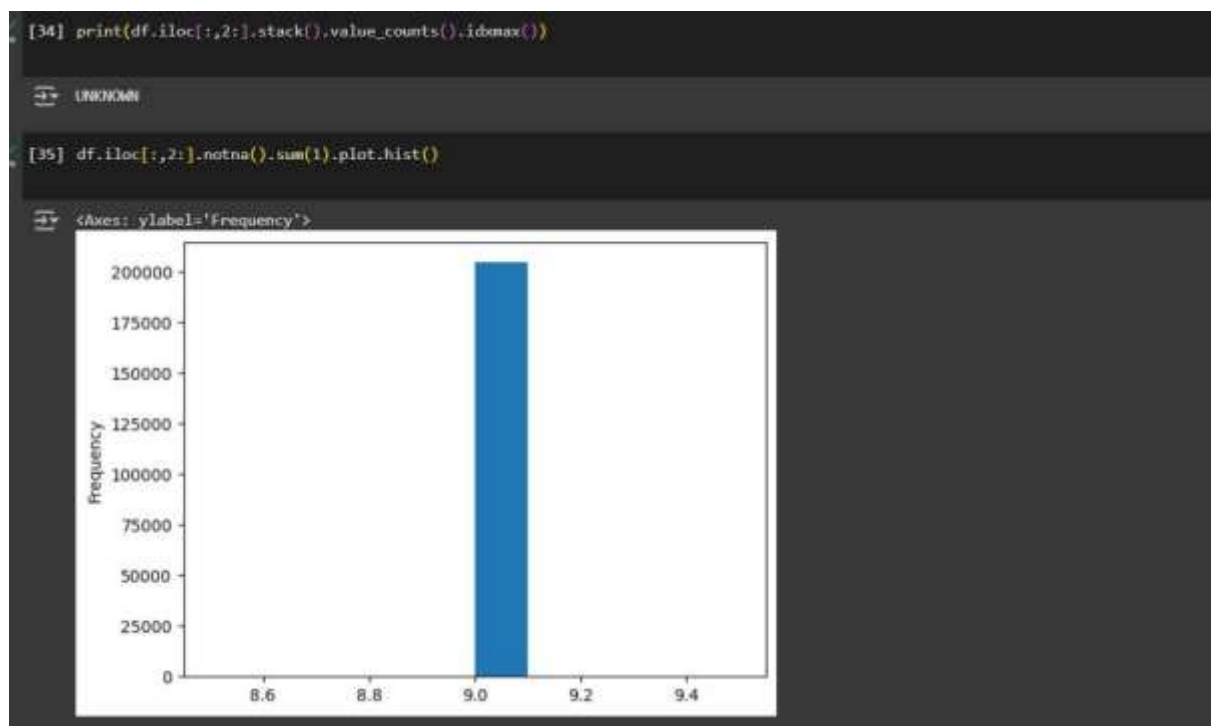
7. To clean the data by removing missing translation

```
[6] df.dropna(subset=['Term 1'], inplace=True)
    print(df[['Word', 'Term 1']].head())
```

	Word	Term 1
0	's gravenhage	The Hague
1	'tween decks	between decks
2	0.22	twenty-two
3	.22-calibre	.22 caliber
4	.22 caliber	.22-caliber

8. Find the most common value in the columns

9. To plot a histogram showing how many non missings entries each row



10. To find the cell that has shortest text length
11. To find rows where there is only 0-1 non null value

```
[30] print(df.iloc[:,2:].stack().dropna().str.len().idxmin())  
→ (np.int64(16), 'Term 2')  
  
[33] print(df[df.iloc[:,2:].notna().sum(1)<=1]['Word'])  
→ Series([], Name: Word, dtype: object)
```

12. Display first 5 rows of the columns
13. counts how many times the 'Word' column has duplicate entries.

```
[24] print(df[['Word', 'Term 2']].head())  
→  
      Word      Term 2  
0  's gravenhage  's Gravenhage  
1  'tween decks      UNKNOWN  
2    0.22 firearm (generic term)  
3  .22-calibre  .22-caliber  
4  .22 caliber   .22 calibre  
  
[25] print(df['Word'].duplicated().sum())  
→ 58888
```

14. counts the number of missing (NaN) values in each column
15. counts how many rows have *all* values missing across the selected columns

```
[36] print(df.iloc[:,2:].isna().sum().sort_values(ascending=False))
```

Term 1	0
Term 2	0
Term 3	0
Term 4	0
Term 5	0
Term 6	0
Term 7	0
rel_count	0
related_count	0

dtype: int64

```
print((df.iloc[:,2:].isna().all(1)).sum())
```

0

16. randomly inspecting 5 words and their related terms

```
[17] df.sample(5)
```

	Word	Count	Term 1	Term 2	Term 3	Term 4	Term 5	Term 6	Term 7
188801	brain	5	propose	swim (generic term)	study (generic term)	read (generic term)	bake (generic term)	NaN	NaN
91343	hypophyseal stalk	17	infundibulum (generic term)	NaN	NaN	NaN	NaN	NaN	NaN
67693	linball	0	bolide	meteor (generic term)	shooting star (generic term)	NaN	NaN	NaN	NaN
204803	/zoopsis	7	visual hallucination (generic term)	NaN	NaN	NaN	NaN	NaN	NaN
74395	generally	0	broeddy	loosely	broadly speaking	narrowly (antonym)	NaN	NaN	NaN

- 17. returns the **shape** of the DataFrame
- 18. checks for **missing values**

```
[12] df.shape
```

```
(204679, 9)
```

```
df.isnull().sum()
```

	0
Word	5
Count	0
Term 1	0
Term 2	25875
Term 3	64546
Term 4	110979
Term 5	143859
Term 6	166524
Term 7	180528

dtype: int64

- 19. calculates the **length** of each string in the **Word** column.

```
[29] print(df.loc[df['Word'].str.len().idxmax(), 'Word'])
```

```
blood-oxygenation level dependent functional magnetic resonance imaging
```

- 20. selects all **columns** starting from the 3rd **column** onward

```
[28] print((df.iloc[:,2:].isna().all(1)).sum())
```

```
0
```