SQL Data Cleaning & Exploratory Data Analysis (EDA)

1. Introduction

The objective of this project was to clean and analyze a dataset related to company layoffs. The dataset contained information about companies, industries, countries, and the number of employees laid off.

2. Data Cleaning Steps

The following steps were performed during data cleaning:

- Removed duplicate rows
- Standardized company names, industries, and country names
- Handled missing/null values appropriately
- Corrected date formats for consistency

3. Exploratory Data Analysis (EDA)

The following analyses were conducted:

- Maximum layoffs observed in dataset
- - Companies with 100% layoffs
- - Top companies with most layoffs
- Yearly layoff trends
- - Monthly layoff trends
- - Rolling total layoffs
- - Top companies by year

4. Key Insights & Findings

- Certain industries experienced significantly higher layoffs compared to others.
- Some companies had 100% of their workforce laid off.
- Layoffs showed clear yearly and monthly patterns, with spikes in specific periods.
- Rolling totals highlighted long-term layoff trends.

5. Conclusion

This project highlighted the importance of proper data cleaning before conducting analysis. The exploratory data analysis provided insights into how layoffs have impacted companies over time, revealing key trends and industry-specific effects.

6. Appendix - SQL Queries

The project included SQL queries for data cleaning and EDA. Example queries:

```
SELECT company, SUM(total_laid_off) FROM layoffs GROUP BY company;
```

SELECT YEAR(date), SUM(total_laid_off) FROM layoffs GROUP BY YEAR(date);

SQL Data Cleaning & Exploratory Data Analysis (EDA) - Layoffs Dataset:

1. Creating a Backup Table:

```
-- DATA CLEANING
1
2 •
       SELECT * FROM layoffs;
3
4 •
      CREATE TABLE layoffs1 LIKE layoffs;
5 •
                                                     SELECT * FROM layoffs1;
6
7 •
      INSERT INTO layoffs1
       SELECT *
9
       FROM layoffs;
10
```

2. Identifying and Removing Duplicates:

```
12
       ROW_NUMBER() OVER(PARTITION BY company, location, industry, total_laid_off, `date`, stage, country, funds_raised_millions) AS row_numbering
13
       FROM layoffs1;
15 • WITH duplicates AS
16 ⊝ (
17
       SELECT *,
18
       ROW_NUMBER() OVER(PARTITION BY company, location, industry, total_laid_off, `date`, stage, country, funds_raised_millions) AS row_numbering
19
       FROM layoffs1
20
21
      SELECT *
22
      FROM duplicates
23
       WHERE row_numbering > 1;
```

```
25 • SELECT *
      WHERE company LIKE 'casper';
31
       ROW_NUMBER() OVER(PARTITION BY company, location, industry, total_laid_off, `date`, stage, country, funds_raised_millions) AS row_numbering
32
33
       FROM layoffs1
35
       DELETE
       FROM duplicates
37
      WHERE row_numbering > 1;
39 • CREATE TABLE 'layoffs2' (
48
           'company' text,
41
            "location" text,
42
            'industry' text,
            'total_laid_off' int DEFAULT MULL,
43
44
          'percentage_laid_off' text,
45
            'date' text,
            'stage' text,
47
            'country' text.
           'funds_raised_millions' int DEFAULT NULL,
49
            "row numbering" INT
        ) ENGINE-InnoDB DEFAULT CHARSET-utf8mb4 COLLATE-utf8mb4_0908_ai_ci;
58
51
52 •
                                              SELECT * FROM layoffs2;
53 0
         INSERT INTO layoffs2
55
          ROW_NUMBER() OVER(PARTITION BY company, location, industry, total_laid_off, 'date', stage, country, funds_raised_millions) AS row_numbering
          FROM layoffs1;
57
58
        SELECT *
          FROM layoffs2
          WHERE row_numbering > 1;
68
61
62 0
        DELETE
           FROM layoffs2
          WHERE row_numbering > 1;
```

3. Standardizing Text Data:

```
-- STRANDANDIZING DATA
67 • SELECT company, TRIM(company)
      From layoffs2;
70 • UPDATE layoffs2
     SET company = TRIM(company);
71
72
73 • SELECT DISTINCT industry
74
     From layoffs2;
75
76 • SELECT *
77
     FROM layoffs2
    WHERE industry LIKE 'CRYPTO%'
78
79
80
81 • UPDATE layoffs2
      SET industry = 'Crypto'
82
    WHERE industry LIKE 'Crypto%';
83
84
85 • SELECT DISTINCT industry
86 FROM layoffs2;
```

```
85 • SELECT DISTINCT industry
86 FROM layoffs2;
87
88 • SELECT DISTINCT country
89 FROM layoffs2;
90
91 • UPDATE layoffs2
92 SET country = 'United States'
93 WHERE country LIKE 'United States%';
```

4. Fixing Dates:

```
95 • SELECT 'date',
96 STR_TO_DATE('date', '%m/%d/%Y')
97 FROM layoffs2;
98
99 • UPDATE layoffs2
100 SET 'date' = STR_TO_DATE('date', '%m/%d/%Y');
101
102 • ALTER TABLE layoffs2
103 MODIFY COLUMN 'date' DATE;
```

5. Handling Null & Missing Values:

```
111 •
        SELECT *
112
         FROM layoffs2
          WHERE industry IS NULL
113
114
         OR industry = '';
116 • SELECT t1.industry, t2.industry
117
        FROM layoffs2 t1
       JOIN layoffs2 t2
118
119
           ON t1.company = t2.company
            WHERE t1.industry IS NULL
120
            OR t1.industry = ''
121
            AND t2.industry IS NOT NULL;
122
123
124 • UPDATE layoffs2
       SET industry = NULL
125
       WHERE industry = '';
126
128 • UPDATE layoffs2 t1
129
    JOIN layoffs2 t2
       ON t1.company = t2.company
130
131 SET t1.industry = t2.industry
132 WHERE t1.industry IS NULL
        AND t2.industry IS NOT NULL;
```

Exploratory Data Analysis (EDA):

5 **Maximum layoffs**:

```
155 -- EXPLATORY DATA ANALYSIS
156 • SELECT *
157 FROM layoffs2;
158
159 • SELECT MAX(total_laid_off), MAX(percentage_laid_off)
160 FROM layoffs2;
```

6 Companies with 100% layoffs:

```
162 • SELECT *

163 FROM layoffs2

164 WHERE percentage_laid_off = 1

165 ORDER BY total_laid_off DESC;
```

7 Companies with most layoffs:

```
172 • SELECT company, SUM(total_laid_off)
173 FROM layoffs2
174 GROUP BY company
175 ORDER BY 2 DESC;
```

8 Layoffs by Year:

```
177 • SELECT YEAR(`date`), SUM(total_laid_off)
178 FROM layoffs2
179 GROUP BY YEAR(`date`)
180 ORDER BY YEAR(`date`) DESC
181 ;
```

9 **Monthly Trends**:

```
183 • SELECT SUBSTRING(`date`, 1,7) AS `MONTH`, SUM(total_laid_off) AS total_off
184 FROM layoffs2
185 WHERE SUBSTRING(`date`, 1,7) IS NOT NULL
186 GROUP BY `MONTH`
187 ORDER BY 1 ASC
188
;
```

10 Rolling Total of Layoffs:

```
190 • WITH ROLLING_TOTAL AS
192
       SELECT SUBSTRING('date', 1,7) AS 'MONTH', SUM(total_laid_off) AS total_off
     FROM layoffs2
193
      WHERE SUBSTRING('date', 1,7) IS NOT NULL
195
       GROUP BY 'MONTH'
196
       ORDER BY 1 ASC
197
     SELECT 'MONTH', SUM(total_off) OVER(ORDER BY 'MONTH') AS rolling_total
198
       FROM ROLLING_TOTAL;
200
201 • SELECT company, YEAR('date'), SUM(total_laid_off)
       FROM layoffs2
202
203
       GROUP BY company, YEAR('date')
     ORDER BY company ASC
205
206
```

11 Top Companies by Year:

```
WITH COMPANY_YEAR (company, years, total_laid_off)AS
209
       SELECT company, YEAR(`date`), SUM(total_laid_off)
210
       FROM layoffs2
211
       GROUP BY company, YEAR('date')
212 ORD ),
       ORDER BY company ASC
214
      COMPANY_YEAR_RANK AS
215 \ominus (SELECT *, DENSE_RANK() OVER(PARTITION BY years ORDER BY total_laid_off DESC) AS DENSE_RANKING
216
217
       WHERE YEARS IS NOT NULL
218
219
      SELECT *
220
       FROM COMPANY_YEAR_RANK
221
       WHERE DENSE_RANKING <= 5;
222
```

Conclusion:

This project demonstrated the importance of structured data cleaning and exploratory analysis in deriving meaningful insights from raw datasets. By addressing duplicates, inconsistencies, and missing values, the data was transformed into a reliable foundation for analysis. The EDA highlighted key layoff patterns, industry impacts, and temporal trends, offering valuable perspectives for strategic decision-making. Overall, the study reinforces that clean, well-analyzed data is critical to driving informed business outcomes.