# Global Layoffs SQL Project

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# Introduction

In this project dataset which is used is Global Layoffs, This project is designed to research global layoff trends in a more profound way by identifying patterns of layoffs across various industries, companies, and regions for a comprehensive overview of workforce reduction worldwide..

#### **Dataset Includes:**

- Company: The companies which layoffs (String)
- Location HQ: The Location of the company's headquarter (String)
- Industry: Company belong to which industry (String)
- Laid OFF: Number of people laid off (Int)
- Date: Date of layoffs (date), Data is till 5th of June 2024.
- Funds raised (In Millions or billions): Amount of funding company has raised (string)
- Stage: Stage of the company (string)
- Country: Company headquarter is in which country (string)
- Percentage : Layoff Percentage

# Start with Data Cleaning

• Step 1: Making copy of actual data and working on the copy data because if any mistake happen at least original data will be there. This is a good practice to do.

```
create table copy_global_layoff
like global_layoffs;
```

Use this query to create the same table as original one.

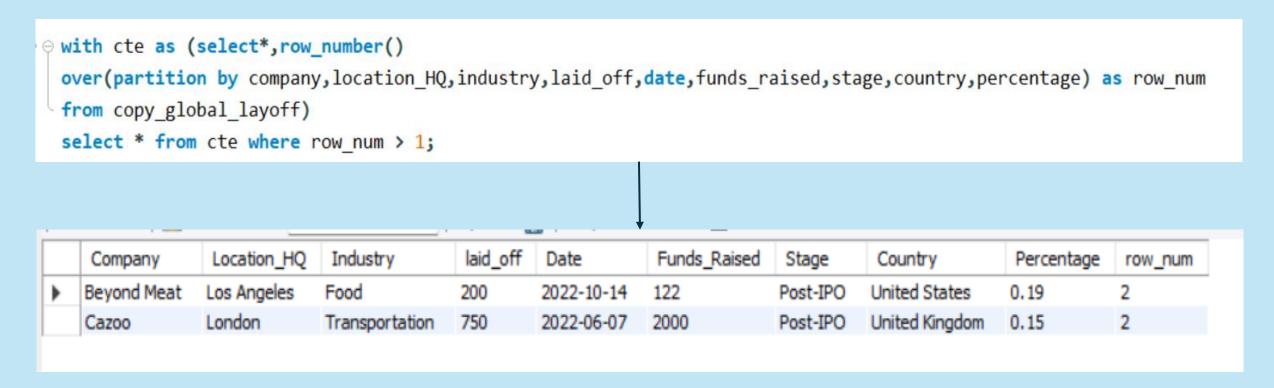
```
insert into copy_global_layoff
select * from global_layoffs;
```

Use this query to insert same data as the original table is having.

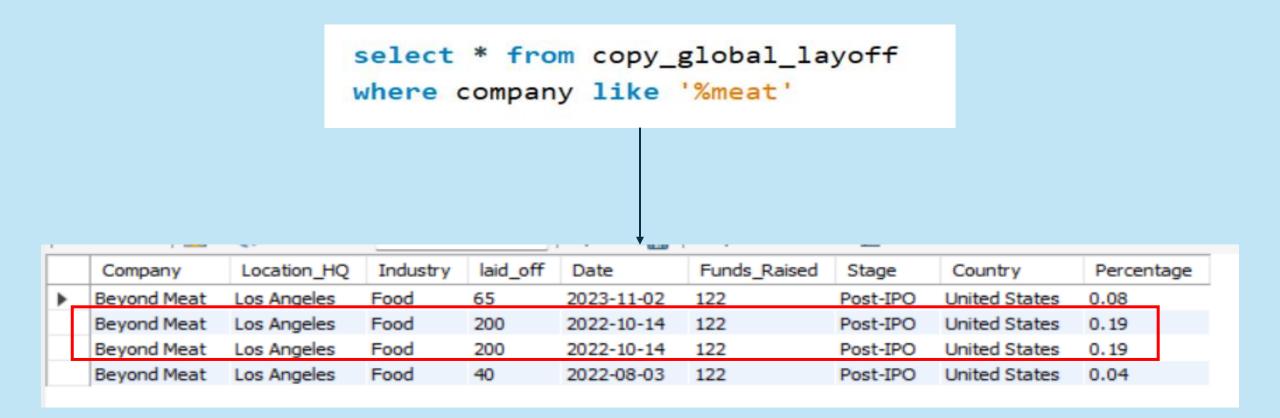
	Company	Location_HQ	Industry	laid_off	Date	Funds_Raised	Stage	Country	Percentage
▶	Oda	Oslo	Food	150	2024-06-05	691	Unknown	Norway	
	Pagaya	Tel Aviv	Finance	100	2024-06-05	2000	Post-IPO	Israel	0.2
	Aleph Farms	Tel Aviv	Food	30	2024-06-05	119	Unknown	Israel	0.3
	MoonPay	Dover	Crypto	30	2024-06-05	651	Unknown	United States	0.1
	Yext	New York City	Marketing		2024-06-05	117	Post-IPO	United States	0.12
	Microsoft	Seattle	Other	1000	2024-06-03	1	Post-IPO	United States	
	OrCam	Jerusalem	Healthcare	100	2024-06-03	86	Unknown	Israel	0.5
	Google	SF Bay Area	Consumer	100	2024-05-31	26	Post-IPO	United States	
	Tropic	New York City	Finance	40	2024-05-31	67	Series B	United States	
	Gro Intelligence	New York City	Food		2024-05-31	118	Series B	United States	0.1

# • **Step 2**: Removing duplicates values

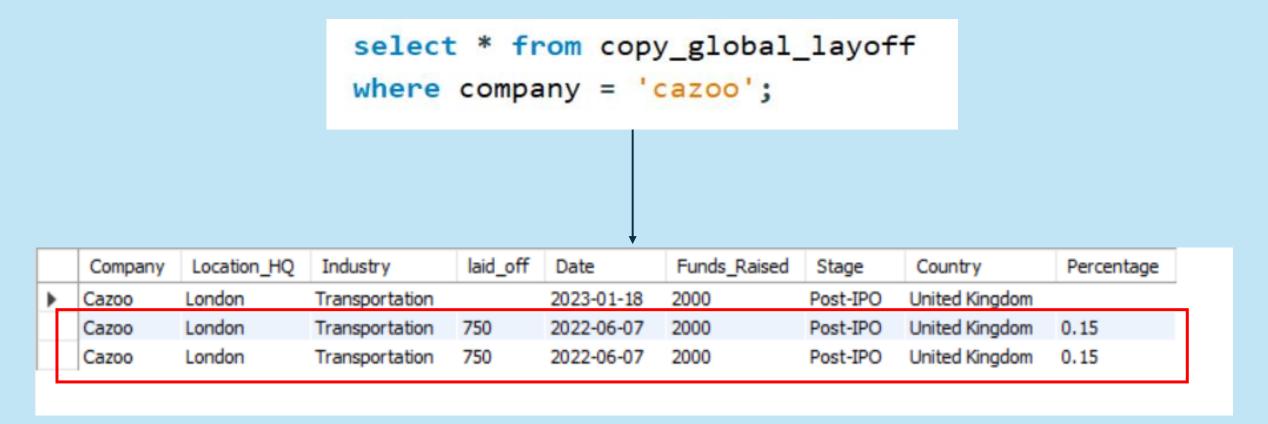
There was no key column in this dataset such as sr.no or something like uniquely identifier so I use window function (ROW\_Number) and cte to identify duplicates and remove it.



This query shows that there are 2 duplicate records for both company.



This record was duplicate so removing one of them.

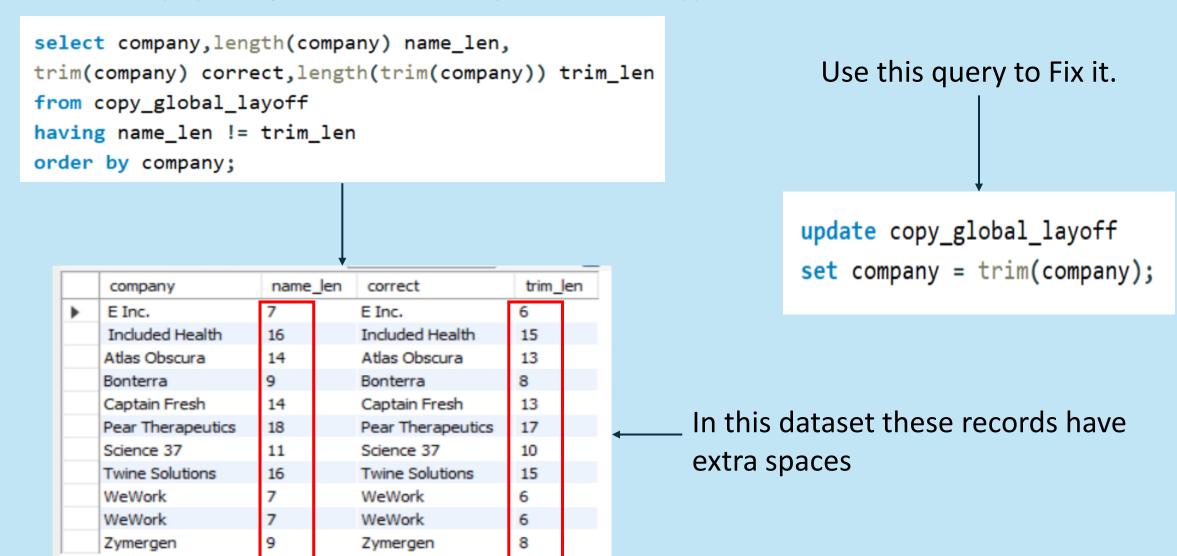


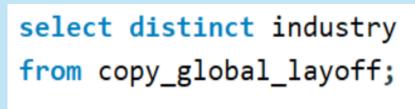
This record was duplicate so removing one of them.

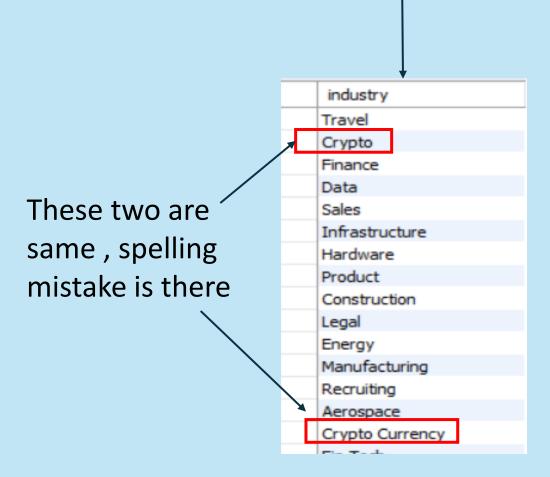
If there were lots of duplicate records, then creating a new table and inserting data from the window function (Row\_Number) which was used to identify duplicate records. And then delete the duplicate records from the new table.

# • **Step 3**: Standardizing Formats

Check for any spelling errors or extra spaces or datatype of columns.







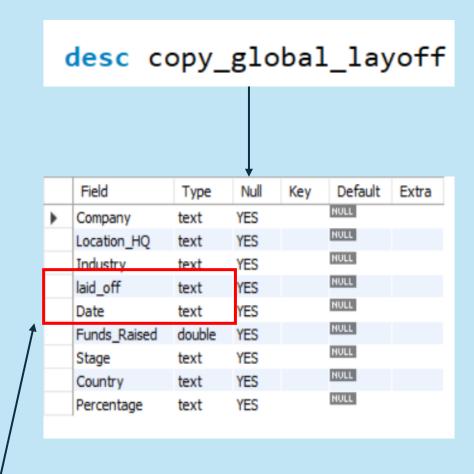
Use this query to Fix it.

update copy\_global\_layoff
set industry = 'Crypto'
where industry like 'Crypto%'

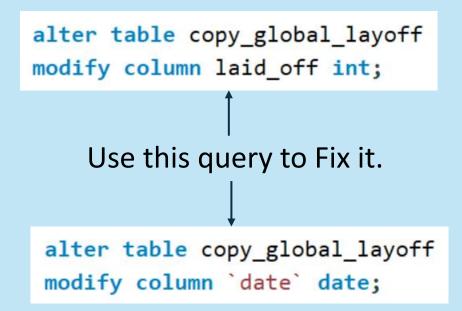


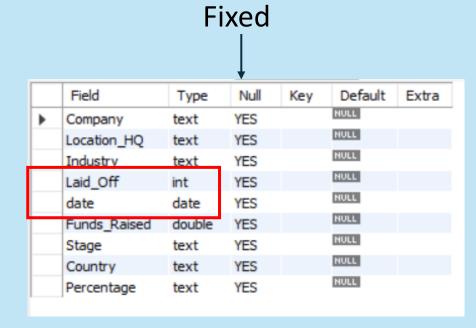
```
Use this query to Fix it.

update copy_global_layoff
set country = 'United States'
where country = 'United States.';
```



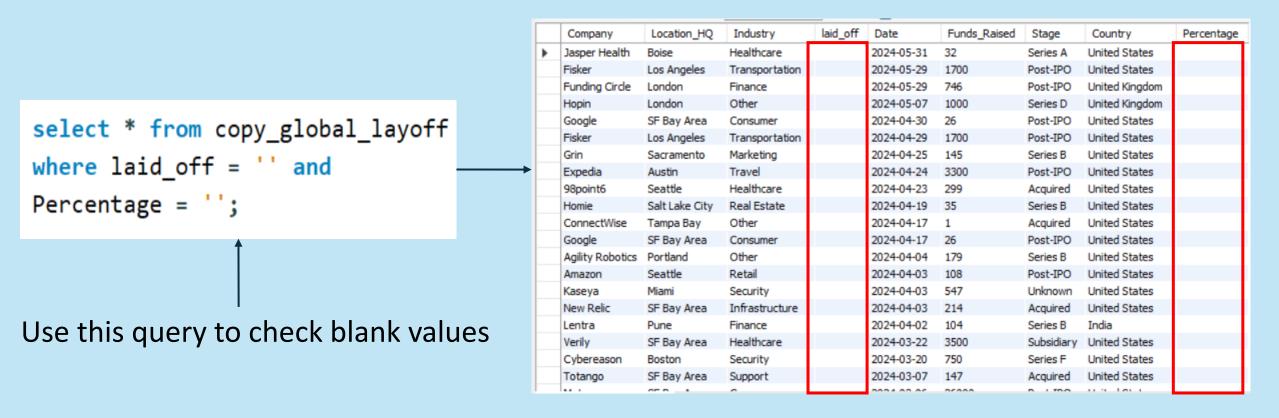
Here these two-column datatype is wrong Laid\_off should be integer and Date column datatype should be date.





# • Step 4: Removing Null or Blank values

In this dataset there were no (Null) values but there are (Blank) values in Laid\_off and percentage columns. But the thing was if there is a blank value in Laid\_off column then there was no blank value in percentage column and vice versa. So, I decided to check whether both columns have blank value in the same row. If yes, then those records are of no use because there are no values then how can any own gather information from those records. So, I remove those records



```
delete from copy_global_layoff
where laid_off = '' and
Percentage = '';
```

This delete all blank values from dataset. These records were of no use because how can we show information with blank values. So, removing blank values are the best solution I can say.

	Company	Location_HQ	Industry	Laid_Off	date	Funds_Raised	Stage	Country	Percentage
<b>&gt;</b>	&Open	Dublin	Marketing	9	2022-11-17	35	Series A	Ireland	0.09
	#Paid	Toronto	Marketing	19	2023-01-27	21	Series B	Canada	0.17
	10X Genomics	SF Bay Area	Healthcare	100	2022-08-04	242	Post-IPO	United States	0.08
	1stdibs	New York City	Retail	70	2020-04-02	253	Series D	United States	0.17
	23andMe	SF Bay Area	Healthcare	71	2023-08-08	1100	Post-IPO	United States	0.11
	23andMe	SF Bay Area	Healthcare	75	2023-06-09	1100	Post-IPO	United States	0.09
	2TM	Sao Paulo	Crypto	100	2022-09-01	250	Unknown	Brazil	0.15
	2TM	Sao Paulo	Crypto	90	2022-06-01	250	Unknown	Brazil	0.12
	54gene	Washington D.C.	Healthcare	95	2022-08-29	44	Series B	United States	0.3
	Acence	CE Ray Area	Sales	150	2022-10-12	476	Series F	United States	0.1

During the cleaning process, 2 duplicate records were removed; 11 records were having extra spaces, so they were trimmed; spelling errors were found in 2 records and then corrected. Corrected datatypes for Laid\_off and date columns and removed the records with blank values. Now this dataset is ready for further in-depth analysis.

# Moving To Analysis

Building on the cleaned dataset prepared, this section further analyzes the layoff trend across the globe by industry, country, and time periods. The objective here is to extract out useful insights that will inform the trends and patterns for workforce reduction.

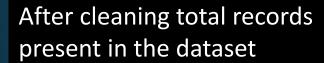
#### select \* from layoffs\_analysis;

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	Company	Location_HQ	Industry	Laid_Off	date	Funds_Raised	Stage	Country	Percentage
<b>•</b>	&Open	Dublin	Marketing	9	2022-11-17	35	Series A	Ireland	0.09
	#Paid	Toronto	Marketing	19	2023-01-27	21	Series B	Canada	0.17
	10X Genomics	SF Bay Area	Healthcare	100	2022-08-04	242	Post-IPO	United States	0.08
	1stdibs	New York City	Retail	70	2020-04-02	253	Series D	United States	0.17
	23andMe	SF Bay Area	Healthcare	71	2023-08-08	1100	Post-IPO	United States	0.11
	23andMe	SF Bay Area	Healthcare	75	2023-06-09	1100	Post-IPO	United States	0.09
	2TM	Sao Paulo	Crypto	100	2022-09-01	250	Unknown	Brazil	0.15
	2TM	Sao Paulo	Crypto	90	2022-06-01	250	Unknown	Brazil	0.12
	54gene	Washington D.C.	Healthcare	95	2022-08-29	44	Series B	United States	0.3
	6sense	SF Bay Area	Sales	150	2022-10-12	426	Series E	United States	0.1
	7shifts	Saskatoon	Food	30	2023-09-15	131	Series C	Canada	0.07
	7Shifts	Saskatoon	Food	68	2024-01-11	131	Series C	Canada	0.19
	8x8	SF Bay Area	Support	155	2023-01-18	253	Post-IPO	United States	0.07
	8x8	SF Bay Area	Support	200	2022-10-04	253	Post-IPO	United States	0.09
	99	Sao Paulo	Transport	75	2022-09-20	244	Acquired	Brazil	0.02
	Abra	SF Bay Area	Crypto	12	2022-06-30	106	Series C	United States	0.05
	Absci	Portland	Healthcare	30	2023-09-05	238	Post-IPO	United States	0.15
	Acast	Stockholm	Media	70	2022-09-15	126	Post-IPO	Sweden	0.15
	Acko	Mumbai	Finance	45	2020-04-01	143	Unknown	India	0.09
	Ada	Toronto	Support	78	2022-09-20	190	Series C	Canada	0.16
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"With the data cleaned and standardized, we're ready to begin our analysis."

# Start with basic EDA

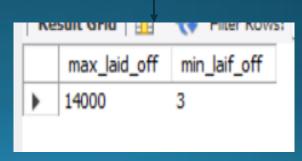
select count(\*) from layoffs\_analysis;



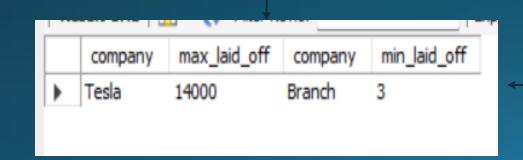


select max(laid\_off) max\_laid\_off,min(laid\_off) min\_laif\_off
from layoffs\_analysis;

Maximum & Minimum Layoff in on go



```
select max_company.company as company,max_company.max_laid_off,
min_company.company as company,min_company.min_laid_off
from (select company, max(laid_off) max_laid_off
from layoffs_analysis
group by company
order by max laid off desc
limit 1) as max company
join
(select company,min(laid_off) min_laid_off
from layoffs_analysis
group by company
order by min laid off
limit 1) as min company
on 1 = 1;
```



Name of the companies who laid off maximum & minimum in one go.

```
select company,sum(laid_off) total_laid_off
from layoffs_analysis
group by company
order by total_laid_off desc;
```

# select industry,sum(laid\_off) total\_laid\_off from layoffs\_analysis group by industry order by total\_laid\_off desc;

#### Company wise total laid off

#### total\_laid\_off company 18000 Amazon Tesla 14000 Google 12000 Meta 11000 SAP 11000 Salesforce 10700 Microsoft 10000 Ericsson 8500 Cisco 8350 8100 Flink 7525 Uber 7200 Micron 4500 PayPal 4484 Peloton Wayfair 4270 Carvana 4000 Better.com 3900 3900 Twitter 3300 Groupon Shopify 3300

Till now Amazon has layoffs the most

Transportation industry appears to be heavily impacted by layoffs

# Industry wise total laid off

		_
	industry	total_laid_off
D <sub>2</sub>	Transportation	49834
	Retail	49454
	Other	47341
	Consumer	44626
	Finance	31545
	Food	30922
	Real Estate	14831
	Healthcare	14677
	Sales	14346
	Travel	13313
	Infrastructure	11165
	Hardware	10920
	Crypto	10581
	Education	9126
	Fitness	8728
	Marketing	7806
	Security	7392
	HR	6674
	Media	6571
	Data	4518
	l	****

```
select country,sum(laid_off) total_laid_off
from layoffs_analysis
group by country
order by total_laid_off desc;
```

# Country wise total laid off

	country	total_laid_off		
Þ	United States	299385		
	India	27537		
	Germany	22793		
	Sweden	12442		
	United Kingdom	11888		
	Canada	8312 6929		
	Brazil			
	Singapore	5090		
	Israel	4515		
	Indonesia	2721		
	Australia	2614		
	France	1040		
	New Zealand	1025		
	United Arab E	995 982 970 755		
	Kenya			
	Nigeria			
	China			
	Hong Kong	730		
	Ireland	416		
	Russia	400		
		222		

Huge workforce reduction in USA till now

```
select stage,sum(laid_off) total_laid_off
from layoffs_analysis
group by stage
order by total_laid_off desc;
```

# Company stage wise total laid off

Post-IPO are experiencing higher layoffs

	stage	total_laid_off		
1	Post-IPO	244064		
	Unknown	29901		
	Acquired	26599		
	Series B	23767		
	Series D	19940		
	Series C	16984		
	Series E	15342		
	Series F	9070		
	Private Equity	7982		
	Series H	6781		
	Series A	4403		
	Series G	4352		
	Series J	2350		
	Series I	2055		
	Seed	1253		
	Subsidiary	714		

```
select company,sum(Funds_Raised) total_funds_raised,sum(laid_off) total_laid_off
from layoffs_analysis
group by company,Funds_Raised
order by total_Funds_Raised desc;
```

Netflix raised the highest amount of funds, but it had a relatively low number of layoffs compared to others.

	company	total_funds_raised	total_laid_off
+	Netflix	243800	450
	Uber	98800	7525
	Xerox	27200	3000
	Meta	26000	11000
	Twitter	25800	3900
	Tesla	20200	14000
	Lyft	19600	2814
	WeWork	19500	100
	Robinhood	16800	1203
	Lucid Motors	16600	1700
	Grab	16500	1000
	Cruise	15000	900
	Flipkart	12900	1100
	Byju's	11000	3000
	Swiggy	10800	1030
	Rivian	10700	840
	Gopuff	10200	2050
	Delivery Hero	9900	156
	Grab	9900	360
	Snap	9800	1780
	n 1 :	7000	****

Comparing company's funds raised and their total layoffs

```
select year(date) Years, sum(Laid_Off) total_laid_off
from layoffs_analysis
group by years
order by years;
```

# Layoffs over the years

Years total\_laid\_off

2020 60960

2021 6490

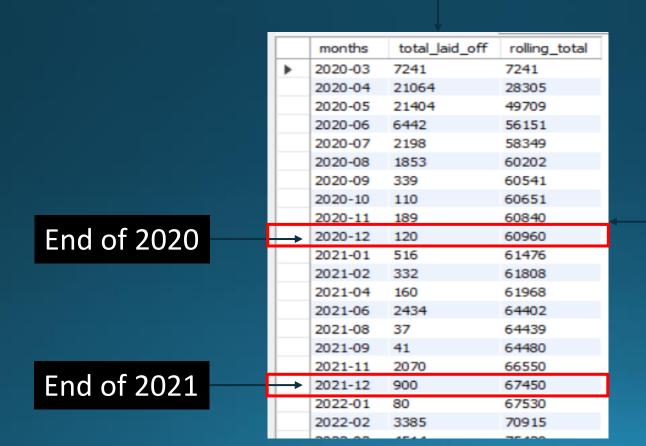
2022 126502

2023 158363

2024 63242

In 2023, layoffs were at their highest

```
with cte as (select substring(date,1,7) Months, sum(laid_off) Total_laid_off from layoffs_analysis group by months order by months) select months,total_laid_off,sum(total_laid_off) over(order by months) rolling_total from cte;
```



Rolling total layoffs by month

```
select company,substring(date,1,7) Months, sum(Laid_Off) total_laid_off
from layoffs_analysis
group by company,months
having months >= '2024-01'
order by total_laid_off desc;
```

Tesla is having maximum layoffs in 2024

ш.	company	Months	total_laid_off
	Tesla	2024-04	14000
	SAP	2024-01	8000
	Cisco	2024-02	4250
	Xerox	2024-01	3000
	PayPal	2024-01	2500
	Farfetch	2024-02	2000
	Unity	2024-01	1800
	Wayfair	2024-01	1650
	Expedia	2024-02	1500
	Flipkart	2024-01	1100
	Block	2024-01	1000
	Citrix	2024-01	1000
	eBay	2024-01	1000
	Indeed	2024-05	1000
	SolarEdge	2024-01	900
	Vacasa	2024-05	800
	Vroom	2024-01	800
	Salesforce	2024-01	700
	Electronic Arts	2024-02	670
	Take-Two	2024-04	579
	- ·	2024.00	

Company wise Total layoffs in 2024(Current year)

```
select

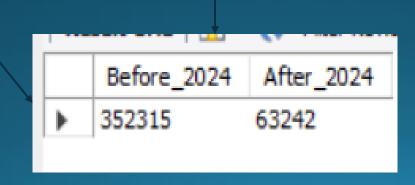
    sum(case)

     when date <'2024-01-01' then laid_off
     else 0
 end) as Before_2024,

    sum(case)

     when date >= '2024-01-01' then laid_off
     else 0
 end) as After_2024
 from layoffs_analysis;
```

Total layoffs before & after 1<sup>st</sup> Jan 2024. And the layoffs after 1<sup>st</sup> Jan 2024 is till 5<sup>th</sup> of June 2024 as by dataset.



```
with company_year as (select company, year(date) years, sum(laid_off) total_laid_off
from layoffs_analysis
group by company,years),
company_year_rank as

(select *,dense_rank () over(partition by years order by total_laid_off desc) as ranking
from company_year)
select * from company_year_rank
where ranking <=5;</pre>
```

	company	years	total_laid_off	ranking	
Þ	Uber	2020	7525	1	
	Groupon	2020	2800	2	
	Airbnb	2020	1900	3	1
	PaisaBazaar	2020	1500	4	
	Swiggy	2020	1450	5	
	Katerra	2021	2434	1	
	Zillow	2021	2000	2	
	Better.com	2021	900	3	4
	Dropbox	2021	315	4	
	Bounce	2021	200	5	
	Meta	2022	11000	1	
	Amazon	2022	10000	2	
	Cisco	2022	4100	3	
	Peloton	2022	4084	4	
	Carvana	2022	4000	5	
	Google	2023	12000	1	
	Microsoft	2023	10000	2	
	Ericsson	2023	8500	3	
	Flink	2023	8100	4	
	Amazon	2023	8000	5	
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In every year Top 5 companies who layoff maximum

# Findings

- This analysis meets our objective of identifying layoff trends by showing that the U.S. and Post-IPO companies are the most affected.
- Our analysis indicated that even with high fundraising, companies like Meta, Tesla,
   Uber, Twitter, and others still had major layoffs.
- Layoffs were highest in 2023, with the transportation industry highly affected.
- Also found that Tesla had the maximum number of layoffs in a single day in 2024.

That's all for this Project

Thank You