

## MySQL Queries

create database Financial\_Fraud

use Financial\_Fraud

select \* from Financial\_Fraud\_dataset

transaction_id	user_id	amount	transaction_type	merchant_category	country	hour	device_risk_score	ip_risk_score	is_fraud
9608	363	4922.5875423114285	ATM	Travel	TR	12	0.9923465237053082	0.9479076765522912	1
456	692	48.01830285876102	QR	Food	US	21	0.16857123943741714	0.22405740349119166	0
4747	587	136.88196000299405	Online	Travel	Travel	14	0.29612724879473956	0.12505769241015305	0
6934	445	80.53471896232148	POS	Clothing	TR	23	0.12480097836782779	0.1592425051585693	0
1646	729	120.04115818676428	Online	Grocery	FR	16	0.0981289537116425	0.02754196922389681	0
2183	944	97.10862458470542	POS	Clothing	DE	17	0.2353994323859285	0.1054540510901703	0
1919	829	166.2092618572779	Online	Travel	UK	12	0.11590590864268228	0.22371784341004333	0
3479	845	96.51263663096326	Online	Grocery	US	7	0.08225043328647746	0.034023010133471865	0
6796	129	83.33870063187989	QR	Food	DE	16	0.0217744413637976	0.27959750653342147	0
5129	249	89.69573134806983	QR	Grocery	UK	6	0.09535343717269791	0.13633641340833455	0
7402	701	101.40382672472745	QR	Travel	FR	7	0.16479393481940688	0.29193962463857537	0
6188	116	106.56344355857529	QR	Food	DE	23	0.13623596858798723	0.1358305529043107	0

DESCRIBE Financial\_Fraud\_dataset;

Field	Type	Null	Key	Default	Extra
transaction_id	int	YES		NULL	
user_id	int	YES		NULL	
amount	double	YES		NULL	
transaction_type	text	YES		NULL	
merchant_category	text	YES		NULL	
country	text	YES		NULL	
hour	int	YES		NULL	
device_risk_score	double	YES		NULL	
ip_risk_score	double	YES		NULL	
is_fraud	int	YES		NULL	

### Overall Fraud Ratio





SELECT

COUNT(\*) AS total\_transactions,

SUM(is\_fraud) AS fraud\_transactions,

ROUND(SUM(is\_fraud) / COUNT(\*), 4) AS fraud\_ratio

FROM financial\_fraud\_dataset;

Result Grid     Filter Rows: <input type="text"/>				Export: 	Wrap Cell Content: 
	total_transactions	fraud_transactions	fraud_ratio		
▶	10000	500	0.0500		

## Fraud vs Legit Comparison

SELECT

is\_fraud,

COUNT(\*) AS txn\_count,





ROUND(AVG(device\_risk\_score), 3) AS avg\_device\_risk,

ROUND(AVG(ip\_risk\_score), 3) AS avg\_ip\_risk,

ROUND(AVG(amount), 2) AS avg\_amount

FROM financial\_fraud\_dataset

GROUP BY is\_fraud;

Result Grid     Filter Rows: <input type="text"/>						Export: 	Wrap Cell Content: 
	is_fraud	txn_count	avg_device_risk	avg_ip_risk	avg_amount		
▶	1	500	0.858	0.852	1657.58		
	0	9500	0.148	0.15	100.28		

## Feature Influence

SELECT

'device\_risk\_score' AS feature,

ROUND(

```
AVG(CASE WHEN is_fraud = 1 THEN device_risk_score END) -  
AVG(CASE WHEN is_fraud = 0 THEN device_risk_score END),  
3  
) AS separation_score  
FROM financial_fraud_dataset
```

UNION ALL

```
SELECT  
'ip_risk_score',  
ROUND(  
AVG(CASE WHEN is_fraud = 1 THEN ip_risk_score END) -  
AVG(CASE WHEN is_fraud = 0 THEN ip_risk_score END),  
3  
)  
FROM financial_fraud_dataset
```

UNION ALL

```
SELECT  
'amount',  
ROUND(  
AVG(CASE WHEN is_fraud = 1 THEN amount END) -  
AVG(CASE WHEN is_fraud = 0 THEN amount END),  
2  
)  
FROM financial_fraud_dataset
```

UNION ALL

SELECT

'hour',

ROUND(

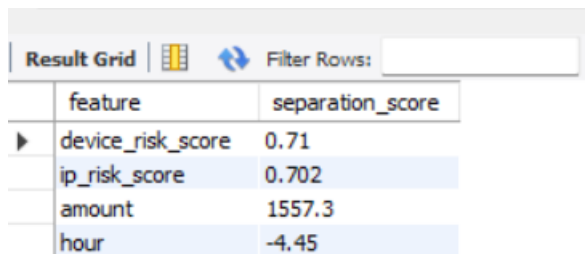
AVG(CASE WHEN is\_fraud = 1 THEN hour END) -

AVG(CASE WHEN is\_fraud = 0 THEN hour END),

2

)

FROM financial\_fraud\_dataset;



	feature	separation_score
▶	device_risk_score	0.71
	ip_risk_score	0.702
	amount	1557.3
	hour	-4.45

### Hourly Fraud Pattern

SELECT

hour,



SUM(is\_fraud = 0) AS legit\_txns,

SUM(is\_fraud = 1) AS fraud\_txns

FROM financial\_fraud\_dataset



GROUP BY hour

ORDER BY hour;

Result Grid |   Filter Rows:

	hour	legit_txns	fraud_txns
0	0	16	
1	0	11	
2	0	16	
3	0	18	
4	0	117	
5	0	15	
6	538	22	
7	530	15	
8	511	19	
9	526	15	
10	543	17	
11	515	19	

Result 6 x

Result Grid |   Filter Rows:

	hour	legit_txns	fraud_txns
12	519	22	
13	561	23	
14	512	10	
15	538	15	
16	540	15	
17	552	12	
18	514	22	
19	525	14	
20	530	18	
21	538	22	
22	485	11	
23	523	16	

### Transaction Amount Distribution

```

SELECT
    is_fraud,
    MIN(amount) AS min_amount,
    MAX(amount) AS max_amount,
    ROUND(AVG(amount), 2) AS avg_amount
FROM financial_fraud_dataset
GROUP BY is_fraud;

```

Result Grid	Filter Rows:	Export:	Wrap Ce
is_fraud	min_amount	max_amount	avg_amount
1	10	11628.213880743357	1657.58
0	1	277.20325054798906	100.28

### High-Value Fraud Outliers



```

SELECT
    transaction_id,
    amount
FROM financial_fraud_dataset
WHERE is_fraud = 1

```

ORDER BY amount DESC

LIMIT 10;

Result Grid   Filter Rows: <input type="text"/>		
	transaction_id	amount
▶	9747	11628.213880743357
	9956	11085.081507626275
	9657	10430.158834513746
	9593	9647.70321201071
	9538	8893.770032426824
	9638	8065.537607919381
	9896	7801.156913071603
	9918	7653.453836441954
	9518	7504.225429074358
	9845	7029.217542179787

### Device Risk Score

SELECT

FLOOR(device\_risk\_score \* 10) / 10 AS risk\_bucket,



COUNT(\*) AS total\_txns,

SUM(is\_fraud) AS fraud\_txns

FROM financial\_fraud\_dataset

GROUP BY risk\_bucket

ORDER BY risk\_bucket;

Result Grid   Filter Rows: <input type="text"/>			
	risk_bucket	total_txns	fraud_txns
▶	0	3257	0
	0.1	3125	0
	0.2	3118	0
	0.7	154	154
	0.8	163	163
	0.9	183	183

### IP Risk Score

SELECT

FLOOR(ip\_risk\_score \* 10) / 10 AS risk\_bucket,

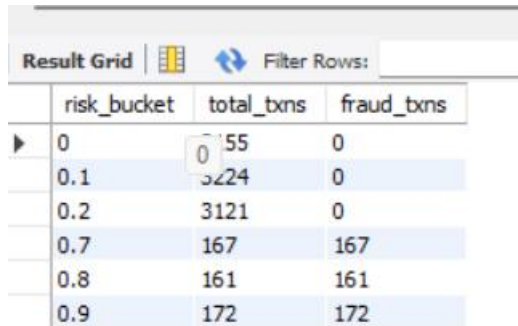
COUNT(\*) AS total\_txns,

SUM(is\_fraud) AS fraud\_txns

FROM financial\_fraud\_dataset

GROUP BY risk\_bucket

ORDER BY risk\_bucket;



The screenshot shows a 'Result Grid' interface with a table containing three columns: risk\_bucket, total\_txns, and fraud\_txns. The data is sorted by risk\_bucket in ascending order. The values for total\_txns are 55, 224, 3121, 167, 161, and 172. The values for fraud\_txns are 0, 0, 0, 167, 161, and 172.

	risk_bucket	total_txns	fraud_txns
▶	0	55	0
	0.1	224	0
	0.2	3121	0
	0.7	167	167
	0.8	161	161
	0.9	172	172

### Geographic Fraud Distribution

SELECT

country,

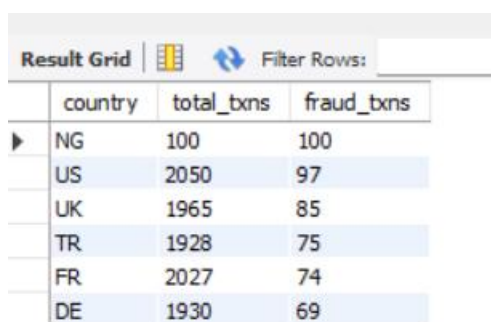
COUNT(\*) AS total\_txns,

SUM(is\_fraud) AS fraud\_txns

FROM financial\_fraud\_dataset

GROUP BY country

ORDER BY fraud\_txns DESC;



The screenshot shows a 'Result Grid' interface with a table containing three columns: country, total\_txns, and fraud\_txns. The data is sorted by fraud\_txns in descending order. The countries listed are NG, US, UK, TR, FR, and DE. The values for total\_txns are 100, 2050, 1965, 1928, 2027, and 1930. The values for fraud\_txns are 100, 97, 85, 75, 74, and 69.

	country	total_txns	fraud_txns
▶	NG	100	100
	US	2050	97
	UK	1965	85
	TR	1928	75
	FR	2027	74
	DE	1930	69

### Time Range

ALTER TABLE financial\_fraud\_dataset

ADD COLUMN time\_range VARCHAR(15);

```
UPDATE financial_fraud_dataset
```

```
SET time_range =
```

```
CASE
```

```
    WHEN hour BETWEEN 5 AND 11 THEN 'Morning'
```

```
    WHEN hour BETWEEN 12 AND 17 THEN 'Afternoon'
```

```
    WHEN hour BETWEEN 18 AND 22 THEN 'Evening'
```

```
    ELSE 'Night'
```

```
END;
```

### Rule-Based Fraud Flag

```
SELECT
```

```
    transaction_id,
```

```
    amount,
```

```
    device_risk_score,
```

```
    ip_risk_score,
```

```
CASE
```

```
    WHEN device_risk_score > 0.8
```

```
    AND ip_risk_score > 0.8
```

```
    AND amount > 1000
```

```
    THEN 'High Fraud Risk'
```

```
    ELSE 'Low Risk'
```

```
END AS risk_flag
```

```
FROM financial_fraud_dataset
```

```
LIMIT 6;
```

Result Grid					
		Filter Rows:		Export:	
		Wrap Cell Content:		Fetch rows:	
	transaction_id	amount	device_risk_score	ip_risk_score	risk_flag
▶	9608	4922.5875423114285	0.9923465237053082	0.9479076765522912	High Fraud Risk
	456	48.01830285876102	0.16857123943741714	0.22405740349119166	Low Risk
	4747	136.88196000299405	0.29612724879473956	0.12505769241015305	Low Risk
	6934	80.53471896232148	0.12480097836782779	0.1592425051585693	Low Risk
	1646	120.04115818676428	0.0981289537116425	0.02754196922389681	Low Risk
	2183	97.10862458470542	0.2353994323859285	0.1054540510901703	Low Risk