

## Task 2: Becoming a Data Detective

For the SQL exploration I will use duckdb as a SQL engine because of its easy of use and setup.

Importing duckdb and creating views for the given CSV. This makes writing SQL a bit more easier.

```
In [19]: # import the duckdb
import duckdb

# Connect to an in-memory DuckDB database
con = duckdb.connect(database=':memory:')

# Create views for the CSV files
con.execute("CREATE VIEW maintenance AS SELECT * FROM '../data/raw/maintenance_events.csv'")
con.execute("CREATE VIEW manufacturing AS SELECT * FROM '../data/raw/manufacturing_factory_dataset.csv'")
con.execute("CREATE VIEW operator AS SELECT * FROM '../data/raw/operators_roster.csv'")
```

```
Out[19]: <_duckdb.DuckDBPyConnection at 0x7eff3f6a42f0>
```

Before answering the question I performed basic SQL check on the dataset to ensure that the dataset was reliable:

### Uniqueness Check

Ensures that event\_id is unique and there are no duplicate records for the same event.

```
SELECT
    event_id,
    COUNT(*)
FROM maintenance
GROUP BY event_id
HAVING COUNT(*) > 1;
```

### NULL Value Checks

Identifies missing data in critical fields that could break your calculations (like downtime\_min or cost\_eur)

```
SELECT
    COUNT(*) - COUNT(event_id) AS missing_ids,
    COUNT(*) - COUNT(start_time) AS missing_start_times,
    COUNT(*) - COUNT(cost_eur) AS missing_costs,
    COUNT(*) - COUNT(technician_id) AS missing_technicians
FROM maintenance_table;
```

### Chronological Checks

Checks for time errors where the end time occurs before the start time, or the next maintenance is scheduled in the past.

```
SELECT * FROM maintenance
WHERE end_time < start_time
    OR next_due_date < end_time::date;
```

### Calculation Consistency

Verify that the downtime\_min column actually matches the difference between start\_time and end\_time.

```
SELECT
    event_id,
    start_time,
    end_time,
    downtime_min,
    date_diff('minute', start_time, end_time) AS calculated_min
FROM maintenance
WHERE downtime_min != date_diff('minute', start_time, end_time)
```

Once satisfied with the reliability of the data I went forward and queried the data to answer the questions.

```
In [20]: query = """
SELECT
    *
FROM maintenance
"""

df = con.execute(query).df()
print(df.head())
```

	event_id	factory_id	line_id	maintenance_type	\
0	MEV-202511031011-1684	FRA-PLANT-01	Line-B	Inspection	
1	MEV-202511031035-9997	FRA-PLANT-01	Line-A	Preventive	
2	MEV-202511031207-5834	FRA-PLANT-01	Line-C	Preventive	
3	MEV-202511031312-3195	FRA-PLANT-01	Line-B	Preventive	
4	MEV-202511031603-1385	FRA-PLANT-01	Line-A	Corrective	

  

	reason	start_time	end_time	downtime_min	\
0	Planned Maintenance	2025-11-03 10:11:00	2025-11-03 10:41:00	30	
1	Planned Maintenance	2025-11-03 10:35:00	2025-11-03 10:50:00	15	
2	Planned Maintenance	2025-11-03 12:07:00	2025-11-03 13:37:00	90	
3	Planned Maintenance	2025-11-03 13:12:00	2025-11-03 14:42:00	90	
4	Unplanned Breakdown	2025-11-03 16:03:00	2025-11-03 16:48:00	45	

  

	technician_id	parts_used	cost_eur	outcome	\
0	TECH-010	PART-022,PART-040	2201.88	Root Cause Identified	
1	TECH-012	PART-010,PART-038	1719.94	Restored	
2	TECH-002	None	683.64	Restored	
3	TECH-004	PART-034,PART-006	1439.96	Temporarily Fixed	
4	TECH-024	PART-021,PART-040	2615.84	Restored	

  

	next_due_date
0	2025-11-06
1	2025-11-04
2	2025-11-04
3	2025-11-17
4	2025-11-06

## 1. What's the total maintenance cost?

The total cost comes out at **total\_maintenance\_cost**: 169389.62

All the events in the table are unique, so a simple sum of the *cost\_eur* column give me the answer.

```
In [21]: query = """
SELECT
    SUM(cost_eur) AS total_maintenance_cost
FROM maintenance
"""

df = con.execute(query).df()
print(df)
```

	total_maintenance_cost
0	169389.62

## 2. How many minutes of downtime have there been?

Total downtime is **6180 minutes**.

The query in itself is simple, but I first checked the min, max values to make sure there were no negative values, checked if there were NULL rows:

```
SELECT
    COUNT(*) AS total_rows,
    COUNT(downtime_min) AS non_null_rows,
    MIN(downtime_min) AS min_val,
    MAX(downtime_min) AS max_val,
    SUM(downtime_min) AS total_downtime
FROM maintenance
```

I also doubled checked my results by comparing difference b/w the start and end time of maintenance by using the following query:

```
SELECT
    date_diff('minute', start_time, end_time) AS calculated_downtime,
    downtime_min
FROM maintenance
```

Once satisfied with the initial query results I went ahead and calculated the downtime by summing the *downtime\_min* column.

```
In [22]: query = """
SELECT
    SUM(downtime_min) AS total_downtime
FROM maintenance
"""

df = con.execute(query).df()
print(df)
```

	total_downtime
0	6180.0

### 3.How many maintenance events occurred?

**Total maintenance event occurred: 94**

```
In [23]: query = """
SELECT
    COUNT(*) AS maintenance_event_count
FROM maintenance
"""

df = con.execute(query).df()
print(df)
```

	maintenance_event_count
0	94

### 4. How many breakdowns (unplanned) happened?

**Total Unplanned Events: 23**

I first checked the values in the *reason* column by running **distinct** on the column. This revealed that there were only two values in the column (Planned Maintenance, Unplanned Breakdown). A simple filter on Unplanned Breakdown and count(\*) gave me the unplanned breakdowns.

```
In [24]: query = """
SELECT
    count(*) AS unplanned_maintenance_events
FROM maintenance
WHERE reason = 'Unplanned Breakdown'
"""

df = con.execute(query).df()
print(df)
```

	unplanned_maintenance_events
0	23

### 5. What's the average downtime per event?

**Average Downtime (min): 65.7 minutes**

Considering that there were no NULL events in the dataset the AVG downtime was calculated via AVG func.

```
In [25]: query = """
SELECT
    count(*) AS planned_maintenance_events,
    AVG(downtime_min) AS average_downtime
FROM maintenance
where downtime_min > 0
"""

df = con.execute(query).df()
print(df)
```

	planned_maintenance_events	average_downtime
0	94	65.744681