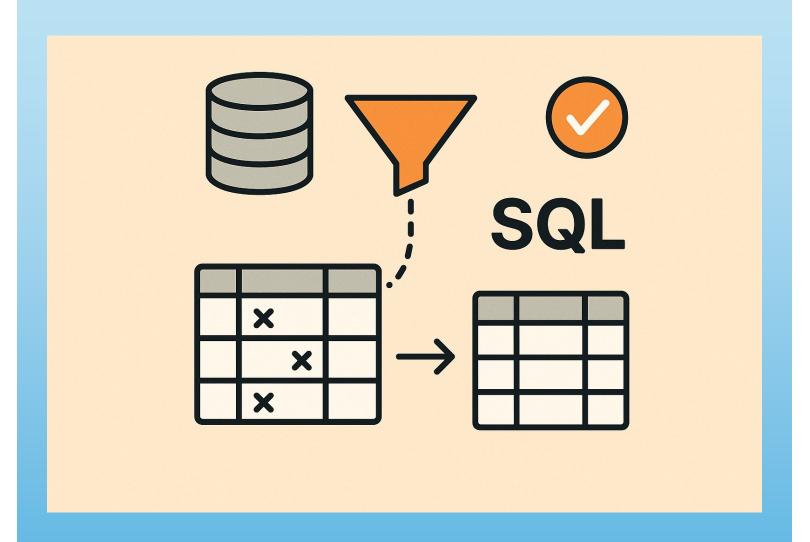
# How to Clean Data Using SQL



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# Introduction

Data cleaning is a critical step in any data analysis or data science project. Without proper data cleaning, your analysis may lead to inaccurate or misleading results. This enhanced guide covers essential SQL data cleaning techniques with practical examples, step-by-step strategies, and real-world input/output demonstrations.

# 1. Handling Missing Values

**Problem:** Missing values can lead to inaccurate analysis or cause errors during joins and aggregations.

Solution: Use COALESCE() or IFNULL() to replace missing values with defaults.

### **Example with Data:**

```
-- Input Data (users table)

/*

| user_id | email | | | |
|-----| | 1 | john@example.com | |
| 2 | NULL | |
| 3 | sarah@example.com | |
| 4 | NULL | |
| 5 | mike@example.com | |
*/
```

SELECT user\_id, COALESCE(email, 'unknown') AS cleaned\_email
FROM users;

### **Output:**

# 2. Removing Duplicates

**Problem:** Duplicates in data can distort results and lead to incorrect conclusions.

**Solution:** Use ROW\_NUMBER() to eliminate duplicate rows.

```
-- Input Data (orders table)
/*
order_id | user_id | created_at | amount |
|-----|----|-----|
                       | 2023-01-01 10:00:00 | 100
101

      | 102
      | 1
      | 2023-01-02 11:00:00 | 150

      | 103
      | 2
      | 2023-01-01 09:00:00 | 200

      | 104
      | 3
      | 2023-01-03 12:00:00 | 120

                       | 2023-01-03 13:00:00 | 130
105
*/
WITH RankedRows AS (
    SELECT *, ROW_NUMBER() OVER (PARTITION BY user_id ORDER BY
created at DESC) AS row num
    FROM orders
SELECT order_id, user_id, created_at, amount
FROM RankedRows
WHERE row num = 1;
```

/\*

	order_id			  -	created_at		amount	ļ
- 1		:			2023-01-02 11:00:00	<sup>-</sup>	150	1
		:		•		•		ł
		!		•	2023-01-01 09:00:00	•		ļ
	105	l	3		2023-01-03 13:00:00		130	ı

\*/

# 3. Standardizing Data Formats

**Problem:** Inconsistent data formats can cause issues in comparisons or analysis.

Solution: Use LOWER(), UPPER(), and TRIM() to standardize text.

### **Example with Data:**

### SELECT

```
customer_id,
   TRIM(LOWER(first_name)) AS standardized_name
FROM customers;
```

### **Output:**

```
/*
```

customer_id   standard	1
1	       

\*/

# 4. Handling Outliers

**Problem:** Outliers can distort analysis results.

**Solution:** Identify and either remove or cap outliers.

```
-- Input Data (orders table)
/*
order id amount
|----|
| 101 | 100
102
       150
| 103 | 200 |
       | 1200 | -- Outlier
104
     | 130 |
105
*/
-- Identifying outliers
SELECT order id, amount
FROM orders
WHERE amount > (SELECT AVG(amount) + 3 * STDDEV(amount) FROM orders);
-- Capping outliers
UPDATE orders
SET amount = (SELECT AVG(amount) + 3 * STDDEV(amount) FROM orders)
```

```
WHERE amount > (SELECT AVG(amount) + 3 * STDDEV(amount) FROM orders);
SELECT * FROM orders;
```

### Output (after capping):

/\*

order_id 	amount	
101	100	
102	150	
103	200	
104	356	Capped value
105	130	

\*/

# 5. Date Format Standardization

**Problem:** Inconsistent date formats can cause issues in time-based analysis.

Solution: Use to date() or extract() functions.

### **Example with Data:**

-- Standardizing dates

# SELECT order\_id, TO\_DATE(order\_date, 'YYYY-MM-DD') AS standardized\_date FROM orders; -- Extracting components SELECT order\_id, EXTRACT(YEAR FROM TO\_DATE(order\_date, 'YYYY-MM-DD')) AS year, EXTRACT(MONTH FROM TO\_DATE(order\_date, 'YYYY-MM-DD')) AS month FROM orders;

### **Output:**

/\*

order_id	standardized_date	year	month
101	2023-01-01	2023	1
102	2023-02-15	2023	2
103	2023-03-03	2023	3
104	2023-04-04	2023	4
105	2023-05-05	2023	5

\*/

# 6. Correcting Data Entry Errors

**Problem:** Manual data entry often leads to formatting errors.

**Solution:** Use REGEXP to detect and correct errors.

```
-- Input Data (customers table)
/*
| customer_id | phone_number |
|-----|
| 1 | 1234567890 |
```

/\*

customer_id	phone_number
3 4	234-567-8901     34567890     (456)7890123     5678901234

\*/

# 7. Handling Null Values in Aggregations

**Problem:** Null values in aggregations can cause incorrect results.

**Solution:** Use COALESCE () to handle nulls.

```
104
         NULL
105
          150
*/
SELECT SUM(COALESCE(amount, 0)) AS total_amount FROM orders;
Output:
```

```
/*
| total_amount |
|-----|
450
*/
```

# 8. Removing Leading/Trailing Spaces

**Problem:** Extra spaces can cause comparison issues.

**Solution:** Use TRIM() to remove unnecessary whitespace.

### **Example with Data:**

```
-- Input Data (employees table)
/*
emp id | first name |
|----|
    | " John "
2
     | " Mary "
| 3 | "Peter "
     | " Alice"
    Bob "
| 5
*/
```

SELECT emp id, TRIM(first name) AS trimmed name FROM employees;

/\*

emp_id	trimmed_name
1	] John
2	Mary
3	Peter
4	Alice
5	Bob

\*/

### 9. Splitting Combined Columns into Multiple Columns

**Problem:** Data often comes combined in a single column (e.g., full names, addresses) and needs to be split for analysis.

**Solution:** Use Substring(), Split part(), or similar functions to separate values.

```
customer_id,
   SUBSTRING(full_name, 1, POSITION(' ' IN full_name) - 1) AS
first_name,
   SUBSTRING(full_name, POSITION(' ' IN full_name) + 1) AS last_name
FROM customers;
```

/\*

customer_id	first_name	last_name
1	John	Smith
2	Mary	Johnson
3	Peter	Parker
4	Alice	Williams
5	Bob	Brown

\*/

### 10. Handling Inconsistent Categorical Values

**Problem:** Categorical data (e.g., product categories) may have inconsistent labels (e.g., "Electronics" vs. "ELECTRONICS").

Solution: Standardize categories using CASE statements or UPDATE queries.

### **Example with Data:**

```
SELECT
```

```
product_id,
CASE
```

WHEN LOWER(category) LIKE '%electronic%' THEN 'Electronics'

```
WHEN LOWER(category) LIKE '%book%' THEN 'Books'
WHEN LOWER(category) LIKE '%stationery%' THEN 'Stationery'
ELSE category
END AS standardized_category
FROM products;
```

/\*

product_id	standardized_category
1	Electronics
2	Electronics
3	Books
4	Books
5	Stationery

\*/

# **Conclusion**

This guide provides practical, real-world examples of data cleaning techniques in SQL. Each concept is demonstrated with sample input data and the corresponding output after cleaning, making the techniques more tangible and easier to understand. By following these methods, you can ensure your data is clean, consistent, and ready for analysis.