

SQL Window Functions

LESSON

- ✦ Generate a quiz about this unit
- ❖ Summarize this unit
- ★ Have any doubt about this content? Ask!

Learning Objectives

By the end of this lesson, you will be able to:

- Explain how window functions work and describe their utility.
- Apply various window functions in MySQL to perform calculations across sets of table rows related to the current row.
- Utilize different window frame specifications to refine calculations.
- Combine window functions with other SQL clauses for advanced data analysis.

Introduction to Window Functions

Window functions, sometimes referred to as analytical or OLAP (Online Analytical Processing) functions, allow you to perform calculations across a set of rows related to the current row, without collapsing groups of rows into a single output row like an aggregate function does. They operate within a "window" of rows defined by the OVER clause.

Basic Syntax

```
SELECT column1, column2,

WINDOW_FUNCTION(column3) OVER (

PARTITION BY column4

ORDER BY column5

[window_frame_clause]

) AS alias_name

FROM tablename;

Explain this code
```

- 1. **PARTITION BY**: Divides the result set into partitions (groups) to which the window function is applied. Think of it as a way to apply the window function to "subgroups" of your result set.
- 2. **ORDER BY**: Defines an order within each partition.
- 3. **Frame Specification**: Defines which rows are included in the window to be processed by the window function relative to the current row.

Common Window Functions

1. ROW_NUMBER()

Assigns a unique sequential integer to each row within its partition.

```
1 SELECT column1, column2,
2 ROW_NUMBER() OVER (ORDER BY column3) AS row_num
3 FROM tablename;

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Explain this code
```

Example:

Here is a table sales with columns date, salesperson_id, and sales_amount.

date	salesperson_id	sales_amount
2023-09-01	1	100
2023-09-02	1	150
2023-09-01	2	200

Here is how to assign a unique sequential number to each sale made by each salesperson ordered by date:

```
SELECT date, salesperson_id, sales_amount,

ROW_NUMBER() OVER (PARTITION BY salesperson_id ORDER BY date) as r

FROM sales;

*Explain this code
```

This is the result:

date	salesperson_id	sales_amount	row_num
2023-09-01	1	100	1
2023-09-02	1	150	2
2023-09-01	2	200	1

► Detailed explanation

2. RANK() and DENSE_RANK()

Assigns a unique rank to each distinct row, with the same rank assigned to rows with the same values. If two rows have the same rank, the next rank is skipped.

DENSE_RANK() is similar to RANK(), but without skipping any ranks.

```
SELECT column1, column2,

RANK() OVER (ORDER BY column3) AS rank_num,

DENSE_RANK() OVER (ORDER BY column3) AS dense_rank_num

FROM tablename;

Explain this code
```

Example:

Take a look at this table:

salesperson_id	total_sales
1	250
2	200

salesperson_id	total_sales	
3	250	

Let's calculate the total sales for each salesperson and then rank them based on the total sales in descending order.

```
SELECT salesperson_id, SUM(sales_amount) as total_sales,

RANK() OVER (ORDER BY SUM(sales_amount) DESC) as rank,

DENSE_RANK() OVER (ORDER BY SUM(sales_amount) DESC) as dense_rank

FROM sales

GROUP BY salesperson_id;

Explain this code
```

This is the result:

salesperson_id	total_sales	rank	dense_rank
1	250	1	1
3	250	1	1
2	200	3	2

► Detailed explanation

3. NTILE(n)

Divides the result set into 'n' number of parts with approximately equal number of rows.

```
1 SELECT column1, column2,
2 NTILE(4) OVER (ORDER BY column3) AS quartile
3 FROM tablename;

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Explain this code
```

Example:

Take a look at this table:

salesperson_id	total_sales
1	300
2	200
3	400
4	100

Let's categorize salespeople into two groups based on their total sales, dividing them into lower and upper halves. This can be useful for a quick bifurcation (split into two groups) to understand which salespeople fall below the median in terms of sales and which ones are above the median.

```
SELECT salesperson_id, SUM(sales_amount) as total_sales,

NTILE(2) OVER (ORDER BY SUM(sales_amount)) as half

FROM sales

GROUP BY salesperson_id;

Explain this code
```

The result would be:

salesperson_id	total_sales	quartile
4	100	1
2	200	1
1	300	2
3	400	2

Detailed explanation

4. LEAD() and LAG()

LEAD() and LAG() functions allow you to access data from subsequent or preceding rows, respectively.

```
SELECT column1, column2,

LEAD(column3, 1) OVER (ORDER BY column4) AS next_value,

LAG(column3, 1) OVER (ORDER BY column4) AS prev_value

FROM tablename;
```

Example:

Imagine you have a sales team and you want to analyze the monthly performance of each salesperson. Specifically, you would like to understand how the sales of each individual has changed from one month to the next.

Here's a sample dataset from the monthly_sales table:

month	salesperson_id	monthly_sales
January	1	100
February	1	120
March	1	110
January	2	200
February	2	210
March	2	195

You want to determine the difference in monthly_sales for each salesperson_id from one month to the next.

Using LAG()

To achieve this, we can use the LAG() function to compare the monthly_sales of the current month with the monthly_sales of the previous month:

```
SELECT month, salesperson_id, monthly_sales,
monthly_sales - LAG(monthly_sales) OVER (PARTITION BY salesperson_
FROM monthly_sales
ORDER BY salesperson_id, month;

Explain this code
```

This is the result:

month	salesperson_id	monthly_sales	sales_difference
January	1	100	NULL

month	salesperson_id	monthly_sales	sales_difference
February	1	120	20
March	1	110	-10
January	2	200	NULL
February	2	210	10
March	2	195	-15

This result table illustrates:

- For salesperson_id 1, sales increased by 20 from January to February and then decreased by 10 from February to March.
- For salesperson_id 2, sales increased by 10 from January to February and then decreased by 15 from February to March.

The LAG() function provides insights into the month-over-month sales difference, helping to quickly identify trends and performance changes for each salesperson.

Window Frame Specification

Window frame specifications allow you to define which rows are included in the frame for each row's calculation. Common specifications include:

- ROWS BETWEEN n PRECEDING AND m FOLLOWING: Includes the 'n' rows before the current row and the 'm' rows after the current row.
- **ROWS UNBOUNDED PRECEDING/FOLLOWING**: Includes all rows from the start to the current row or from the current row to the end, respectively.

Example:

Here is how to calculate a moving average:

```
1 SELECT date, sales,
2 AVG(sales) OVER (
3 ORDER BY date
4 ROWS BETWEEN 1 PRECEDING AND 1 FOLLOWING
5 ) AS moving_avg
6 FROM sales_data;
```

Summary

Window functions are powerful tools for analytics and complex calculations in SQL without collapsing multiple rows into a single output row. They let you perform calculations across related data points, making them indispensable for time series analyzes, rankings, and other analytical challenges. Effectively using window functions in MySQL enables you to derive deeper insights from your data and perform advanced data processing tasks with ease.

→ Any doubt? Ask our Al Chatbot!

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How would you rate the content of this unit?











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SQL Window Functions Hands On

