DIFFERENT WAYS OF REPRESENTING HISTORY

An overview of techniques to track historical data in databases

DEFINING THE ISSUE

- Maintaining historical data is critical for accurate reporting, compliance, and business analysis.
- Different data modeling techniques offer various ways to represent history, each with trade-offs in complexity, storage, and performance.
- Organizations need to **choose the right method** to track data changes while ensuring data integrity and efficient querying.

Key Questions:

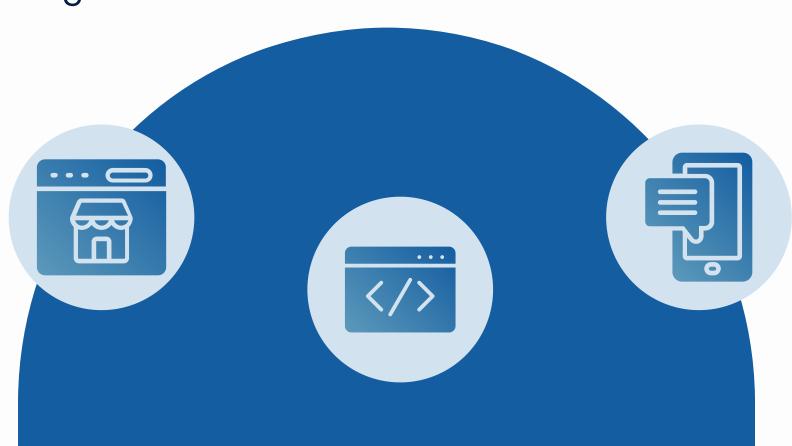
- How can we track changes in data over time effectively?
- What are the best techniques?
- How do we ensure historical accuracy without compromising performance?

What Does Representing History Mean?

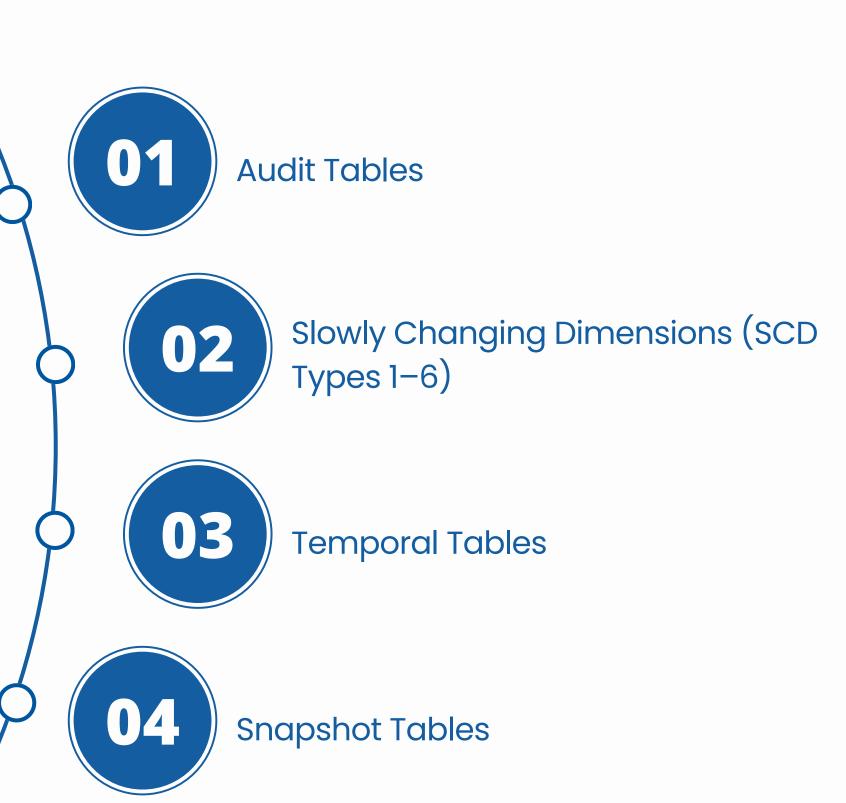
- Tracking Changes Over Time: Capturing how data evolves or changes over time, not just storing current data.
- **Preserving Past Versions:** Storing past versions of data to retain a historical record of all updates or modifications.
- Supporting Business Decisions: Enabling analysis of data trends and behavior over time to make informed business decisions.
- Ensuring Accountability: Keeping records of who made changes to data, when, and why, often for auditing purposes.

Examples

- 1. Sales Data
- 2. Customer Data
- 3. Financial Records



Ways To Represent
History In Data
Modelling



AUDIT TABLES

- Separate tables
- Track and log every change (insert, update, delete) in the main table.
- Metadata about the changes (who, what, when).

STRUCTURE OF AN AUDIT TABLE:

- Audit ID: Unique identifier for each log entry.
- Table Name: The name of the table where the change occurred.
- Operation Type: The type of operation (INSERT, UPDATE, DELETE).
- Old Values: The previous values of the affected fields before the change.
- New Values: The new values of the affected fields after the change.
- Timestamp: The date and time when the change was made.
- User ID: The ID of the user who made the change.

Main Table (orders)

order_id	customer_id	status	updated_at
101	1001	Shipped	2025-01-05 10:00
102	1002	Delivered	2025-01-06 12:30

Audit Table (orders_audit)

audit_id	order_id	status	operation	modified_at	modified_by
1	101	Pending	INSERT	2025-01-01 09:00	admin
2	101	Processing	UPDATE	2025-01-02 14:30	admin
3	101	Shipped	UPDATE	2025-01-05 10:00	system
4	102	Pending	INSERT	2025-01-03 11:00	admin
5	102	Delivered	UPDATE	2025-01-06 12:30	system

SLOWLY CHANGING DIMENSIONS (SCD)

- Tracking History in Dimensions
- Capturing and storing changes to attributes in dimensional data (like customer, product, or location data) over time.

TYPES OF SCD:

SCD Type 1: Overwrite Data, No History

- Updates existing data with new values, overwriting the old record.
- No historical data is kept.

SCD Type 2: Full History with valid_from and valid_to

- Records each change as a new row, tracking the validity period of each record.
- Full history of changes is maintained.

SCD Type 3: Limited History with Extra Columns

- Keeps only the current and previous values in the same record.
- Typically uses extra columns (e.g., current_city, previous_city) to store the historical change.

SCD TYPE 1

Example: Customer Address Change

customer_id	name	city	updated_at
1	John Doe	New York	2025-01-01

Update Scenario:

- John moves from New York to Los Angeles on 2025-02-01.
- The row is updated, and the old city (New York) is **lost**.

customer_id	name	city	updated_at
1	John Doe	Los Angeles	2025-02-01

SCD TYPE 2

Example: Customer Address History

customer_id	name	city	valid_from	valid_to
1	John Doe	New York	2024-01-01	2025-01-31
1	John Doe	Los Angeles	2025-02-01	9999-12-31

SCD TYPE 3

Example: Customer Address Change

customer_id	name	current_city	previous_city	updated_at
1	John Doe	Los Angeles	New York	2025-02-01

- current_city reflects John's current address (Los Angeles).
- previous_city stores his last known address (New York).
- No further historical changes are retained beyond one previous value.

TEMPORAL TABLES

- Temporal tables use valid_from and valid_to timestamps to track the validity period of each row.
- They allow you to query data as it existed at a specific point in time, preserving historical states.
- System-versioned tables that track data changes with timestamps in the same table

employee_id	name	role	salary	valid_from	valid_to
1	John Doe	Junior Developer	50000	2022-01-01	2023-12-31
1	John Doe	Developer	60000	2024-01-01	2024-12-31
1	John Doe	Senior Developer	75000	2025-01-01	9999-12-31

- What was John's salary on January 1, 2024?
- When did John become a Senior Developer?

SNAPSHOT TABLES

• Periodic copies of the data at specific points in time

January Snapshot (inventory_snapshot_2025_01)

product_id	product_name	quantity	snapshot_date
1	Laptop	100	2025-01-31
2	Phone	200	2025-01-31

February Snapshot (inventory_snapshot_2025_02)

product_id	product_name	quantity	snapshot_date
1	Laptop	80	2025-02-28
2	Phone	250	2025-02-28

- Useful for Reporting: Great for point-in-time reporting, financial summaries, or compliance requirements.
- Storage-Heavy: Requires a lot of storage since each snapshot is a full copy of the table.
- Limited Granularity: Cannot track changes between snapshots (you only see the state at snapshot times).
- Complex Queries: Comparing snapshots can get complicated for detailed change tracking.

THANK YOU!