

Introduction to Data Joins:

Matching Rows and Tables



- Combine info from multiple tables for a complete business picture



- Real-world data is scattered: customer, order, and product tables



- Correct joins prevent mismatches, missing data, and silent report errors



- Roadmap: keys, inner/left joins, unmatched values, common pitfalls, practice



How Tables Relate: Keys and Relationships



Primary key: uniquely identifies every row in a table, like a customer ID



Foreign key: references a primary key in another table to create links



One-to-many: one customer → many orders; foreign key lives on the "many" side



Visual: lines connect matching columns, arrows show relationship direction



Pitfall: using a non-unique key silently multiplies rows later

Primary Key

Customers	
	CustomerID
	CustomerName
	Email
	Phone
...	

Foreign Key

Orders	
	CustomerID
	OrderID
	OrderDate
	OrderTotal
...	

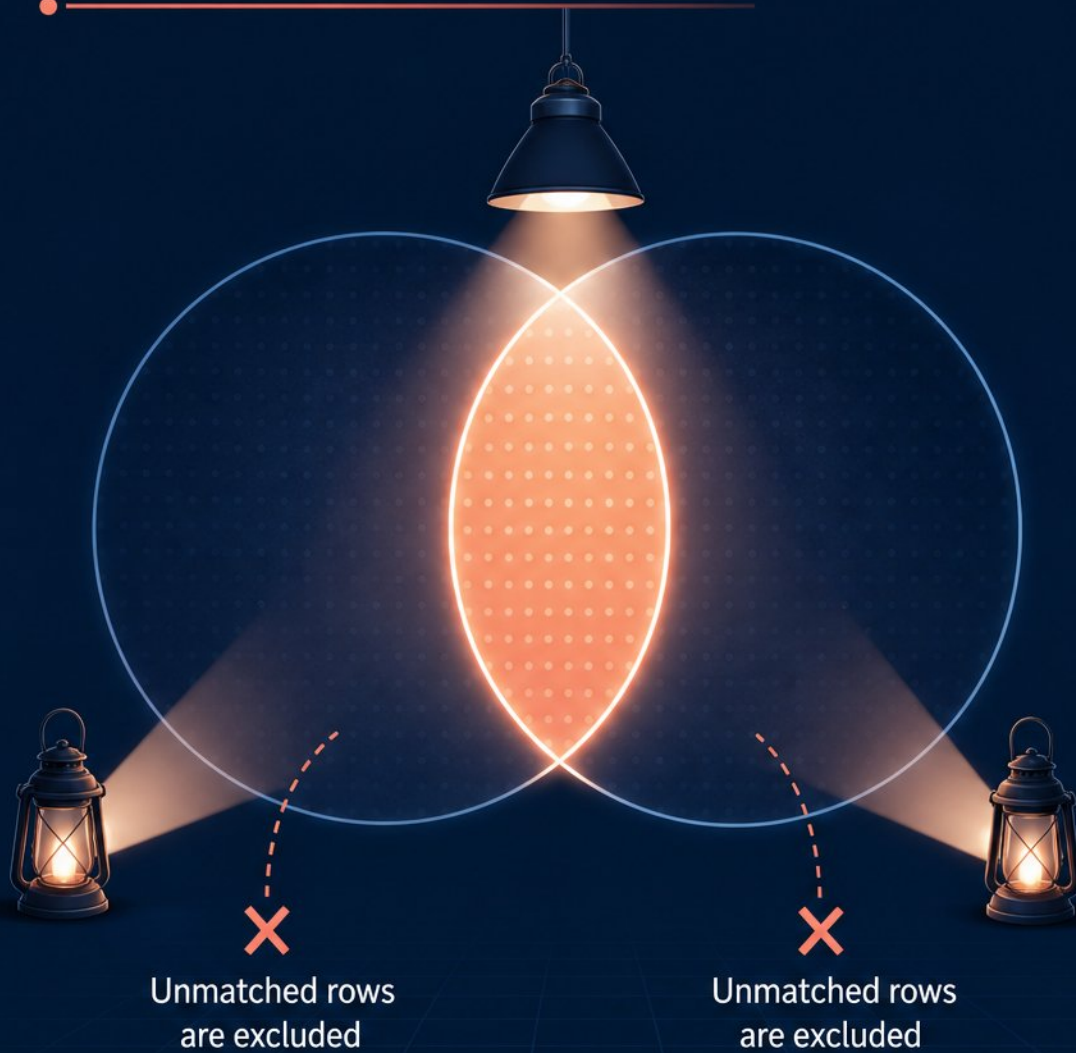


One
Customer



Many
Orders

Inner Join: Keeping Only Matched Rows



- An INNER JOIN returns rows only when the key matches in both tables
- Unmatched rows — such as customers with no orders — are excluded completely
- Missing or invalid key values in either table cause rows to disappear silently
- Use for clean reports: "show only orders that have a valid customer"

Left Join:

Keeping All Rows from the First Table

- Preserves every row from the left table, regardless of matches
- Right table columns fill with NULL when no match exists
- Compared to inner join: unmatched left rows appear instead of vanishing
- Business use: list all customers, including those with no orders



Inner vs. Left Join: Choosing the Right Tool



- Inner join: no match means no row in the result



- Left join: no match means row stays with NULLs



- Accidental inner joins silently drop legitimate rows



- Default to left join when matches are uncertain



- Row count shrinking signals wrong join choice



Other Join Types: Right, Full Outer, and Cross Joins



- **RIGHT JOIN:** mirror of LEFT JOIN — keeps all right-table rows; swap tables and use LEFT JOIN instead
- **FULL OUTER JOIN:** keeps every row from both tables, NULLs where unmatched — ideal for data reconciliation
- **CROSS JOIN:** Cartesian product — every row paired with every other; use for size×color grids or date scaffolds
- **Rule of thumb:** LEFT JOIN ~80% of time, INNER ~15%, others ~5% — know them, but reach for them sparingly



Understanding Unmatched Values: Where NULLs Come From



NULLs signal missing matches — not errors, but clues about absent data



Causes: key typos, format mismatches, incomplete data, genuine missing links



Spot unmatched rows: filter with WHERE right_table.key IS NULL after left join



Business value: find customers who never ordered or products never sold



Use a non-nullable right-table key (like order_id) for reliable NULL checks



The **WHERE** vs. **ON** Trap: Accidentally Breaking Left Joins

- Filtering right table in **WHERE** turns left join into inner join
- **WHERE** `o.status = 'shipped'` drops rows where status is **NULL**
- Fix: move right-table filters into the **ON** clause
- Rule: right-table filters go in **ON**; left-table filters go in **WHERE**





Avoiding Join Pitfalls:

Duplicate Keys and Row Multiplication



 Duplicate keys cause fan-out:
one left row matches many right rows,
repeating left data

 Summing order totals after joining
to line items inflates revenue by
the item count

 Compare `COUNT(*)` after the join to
the left table's row count — larger
means fan-out

 Always check key uniqueness
(`GROUP BY key, COUNT(*) > 1`)
before joining

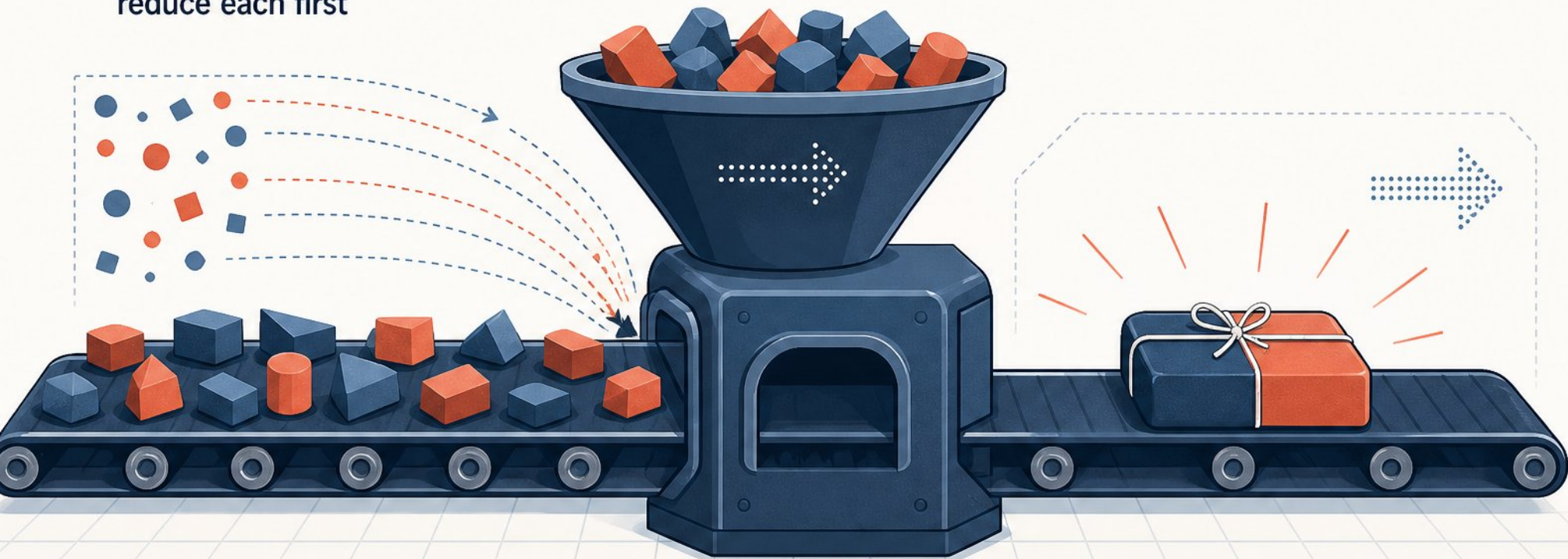
 Count rows before and after
every join as a quick sanity check



Fixing Duplicate Key Problems: Aggregate Before You Join



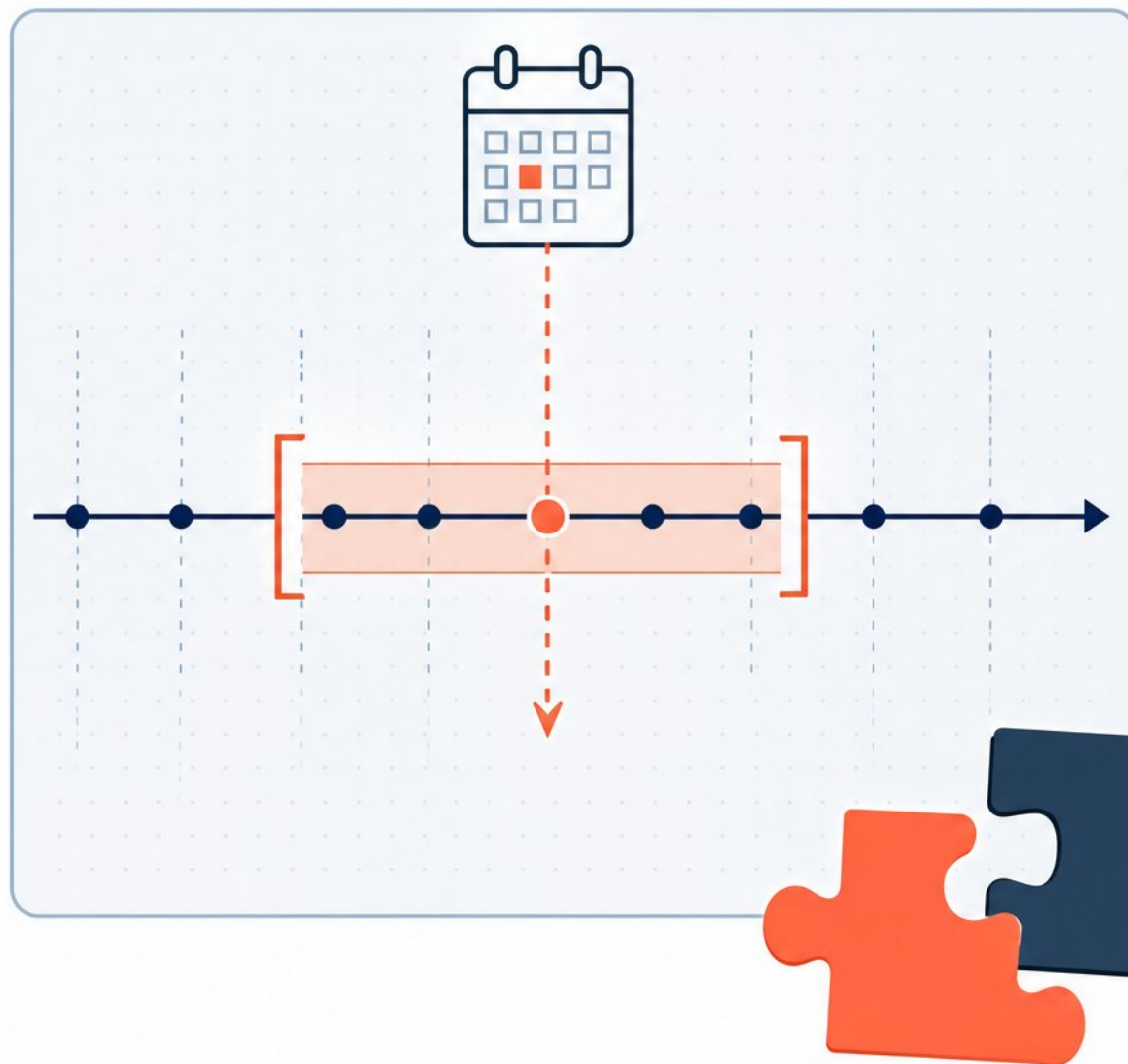
- **Match the grain:** sum columns at the join's level, not the parent's level
- **Fix 1:** sum `line_total` (item grain) instead of `order_total` (order grain)
- **Fix 2:** pre-aggregate the many-side to one row per key before joining
- **Many-to-many danger:** joining two child tables creates a cross product — reduce each first





Join Conditions Beyond Equality: Range and Non-Equi Joins

- Non-equi joins match on inequalities: $>$, $<$, $>=$, $<=$, BETWEEN, $\neq$
- Example: Assign each sale to a price tied to its active date window
`sale_date BETWEEN valid_from
AND valid_to`
- Another example: Assign employees to salary grades based on salary ranges
- Performance tip: Pair range conditions with an equality check to avoid slow scans



Dealing with NULLs After a Join:

COALESCE and Defaults

- NULLs from unmatched joins break arithmetic and comparisons — use COALESCE to supply defaults
- COALESCE(order_total, 0) for aggregates; COALESCE(category, 'Unknown') for display labels
- The WHERE-vs-ON trap: filtering right-table columns in WHERE silently drops unmatched rows
- Distinguish "missing match" NULLs from "genuinely unknown" NULLs to avoid hiding data quality issues



Reading and Writing Join Logic: From Problem to Plan



- Identify the driving table, lookup table, and matching keys



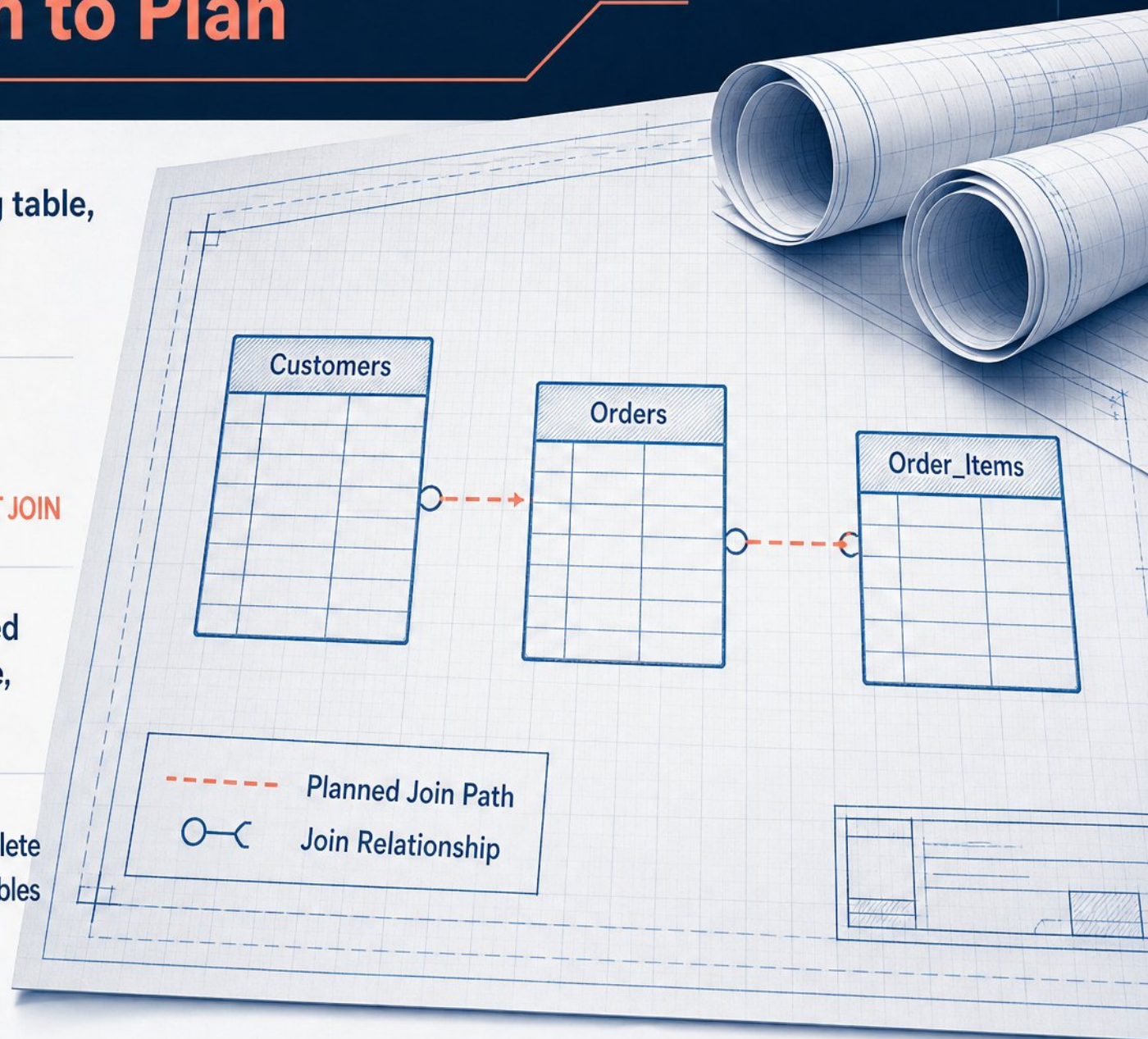
- Translate requests: “all customers and their last order” → **LEFT JOIN**



- Verify plan: expected row count, NULL side, unique keys?



- Practice: build a complete order report across 3+ tables



Key Takeaways and Next Steps



Inner joins keep matched rows; left joins preserve all left rows with NULLs for unmatched

Verify key uniqueness and check row counts before and after joining

Use anti-joins (LEFT JOIN + IS NULL) to find orphans; right-table filters belong in ON

NULLs are signals of missing matches, not errors — COALESCE fills defaults

Practice with real datasets, explore full outer joins for reconciliation, and study performance

Next Steps

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