

The image features a central logo for 'sqlbits 2025' enclosed in a blue neon rectangle. The text 'sqlbits' is in a large, orange-outlined font, and '2025' is in a smaller, pink-outlined font below it. The background is a dark brick wall. Surrounding the central logo are several neon icons: an ice cream cone (top left), two cherries (top left), a girl's face with a bow (top right), a four-pointed star (top right), a planet with a ring (middle right), a lightning bolt (middle right), a four-pointed star (middle right), a leaf (bottom right), a four-pointed star (bottom right), a face with a lightbulb (bottom left), a four-pointed star (bottom left), a swan (middle left), a four-pointed star (middle left), and a four-pointed star (top left).

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Optimizing Power BI: Enhancing Performance Through Data Modeling



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Geek by design

Because performance isn't optional



Why is the Data Model Important?

- The data model is the foundation of Power BI.
- A well-designed data model improves performance and speeds up queries.
- 80% of performance issues are related to the data model.
- A better data model makes DAX queries easier and more efficient.
- Reduces data redundancy and minimizes errors.

VertiPaq Compression

UnitCost	StoreName
€ 10,00	Contoso North America Online Store
€ 11,00	Contoso Europe Online Store
€ 14,00	Contoso Europe Online Store
€ 18,00	Contoso Europe Online Store
€ 15,00	Contoso North America Online Store
€ 14,00	Contoso North America Online Store
€ 19,00	Contoso Asia Online Store
€ 14,00	Contoso North America Online Store
€ 17,00	Contoso Europe Online Store
€ 19,00	Contoso Asia Online Store

Dictionary Encoding

- Replaces unique text values with numeric keys
- Saves memory: text → integer
- Values stored once in a dictionary, referenced via keys
- Especially effective for low-cardinality columns

StoreName	StoreName.ID
Contoso North America Online Store	1
Contoso Europe Online Store	2
Contoso Europe Online Store	2
Contoso Europe Online Store	2
Contoso North America Online Store	1
Contoso North America Online Store	1
Contoso Asia Online Store	3
Contoso North America Online Store	1
Contoso Europe Online Store	2
Contoso Asia Online Store	3

Run-Length Encoding (RLE)

- Compresses consecutive repetitions of the same value
- Stores as (value + count)
- Only works well when the column is sorted
- Applied only if it actually saves memory

StoreName.ID	StoreName	StoreName.ID	Count	Rows
1	1	1	0	4
2	1	2	4	4
2	1	3	7	2
2	1			
1	2			
1	2			
3	2			
1	2			
2	3			
3	3			

Value Encoding (Bit-Packing)

- VertiPaq chooses the smallest number of bits per column
- Smaller range of values = fewer bits needed
- Works after dictionary encoding or on numeric columns

Original	Optimized
UnitCost	UnitCost
€ 10,00	0 +10
€ 11,00	1 +10
€ 14,00	4 +10
€ 18,00	8 +10
€ 15,00	5 +10
€ 14,00	4 +10
€ 19,00	9 +10
€ 14,00	4 +10
€ 17,00	7 +10
€ 19,00	9 +10

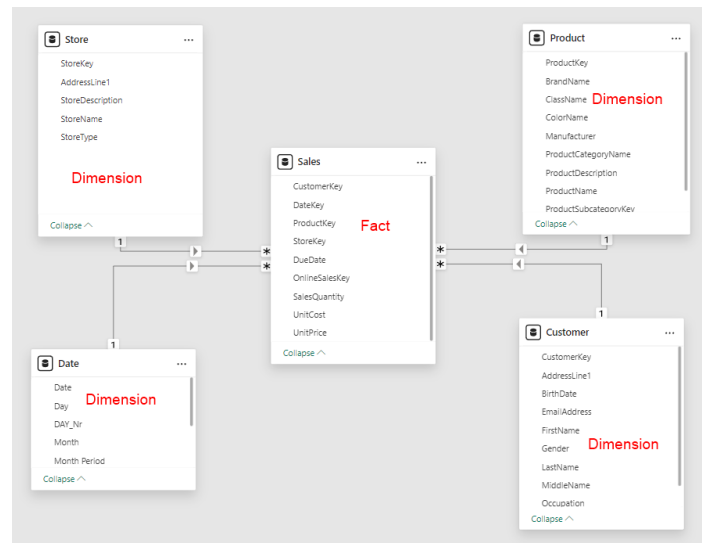
Star Schema

Fact Tables:

- Contain facts, representing an event with dimensions.
- A sale includes a product, a customer, and a date.
- Metrics that can be aggregated to gain insights.

Dimension Tables:

- Descriptive attributes of entities such as a product, customer, employee, or patient.
- Dimensions have attributes like color, category, manufacturer, or price.



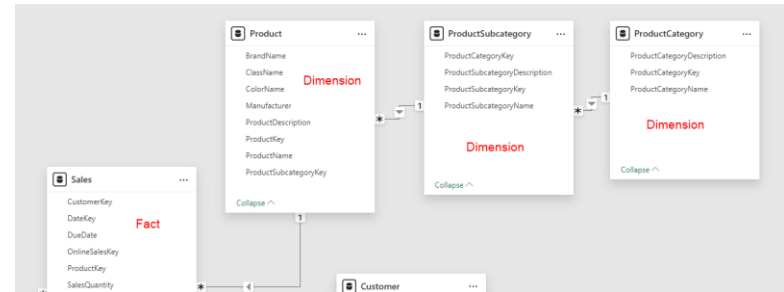
Snowflake-model

The Snowflake Schema is a Variant of the Star Schema

The difference is that dimension tables that are related to each other are connected.

For example:

- Product
- Product Subcategory
- Product Category





What does an Optimized Data Model consist of?

The model consists of facts and dimensions.

- A fact table contains values that you can calculate, such as:
 - Revenue, purchase date, sold products, etc.
- A dimension table contains values you want to filter, such as:
 - Year, month, manufacturer, customer, etc.



Relationships and Keys

To create relationships between tables, keys are used.

Primary Key:

- A unique value that appears only once in a dimension table.

Foreign Key:

- Used in a fact table to indicate how often, for example, a product has been sold.
- The foreign key can appear multiple times in a fact table, as a product is typically sold more than once.

Relationships and Filtering

A relationship must be created between facts and dimensions. Possible relationships include:

- One-to-many
- One-to-one
- Many-to-many

You can also choose the filter direction:

- Single
- Both

Multiple Fact Tables

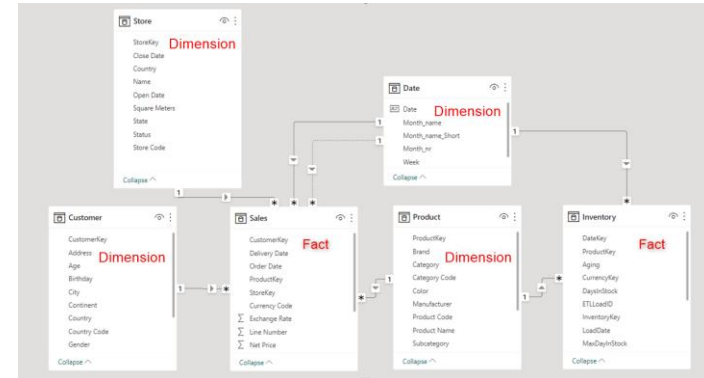
It is possible to use multiple fact tables in your model.

These fact tables may not have much in common with each other

- Inventory and sales.

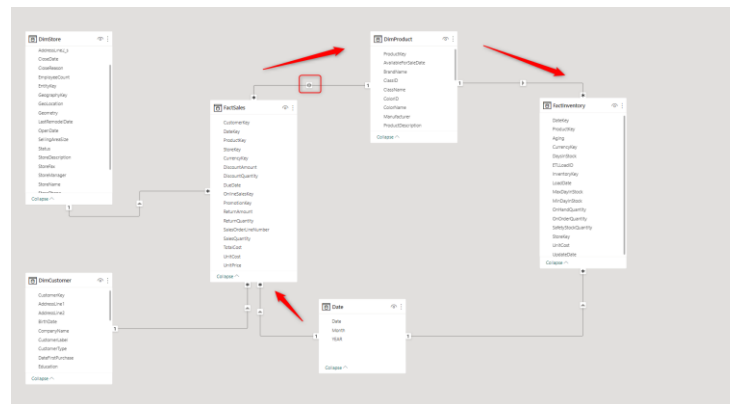
However, it is necessary to have some dimension tables that are connected to the fact tables

- date and product.



Ambiguity

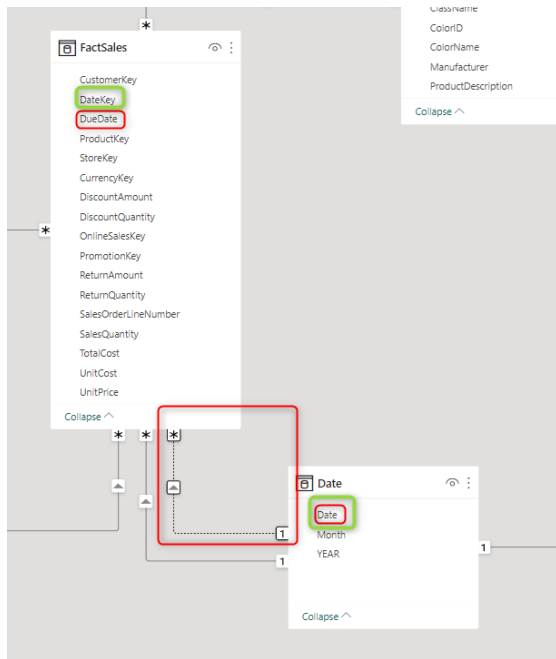
- When a model includes multiple fact tables and uses bidirectional relationships, there is a risk of ambiguity.
- This means the model may not know which path to follow when filtering.
- If it can filter through multiple tables, it may display incorrect values.



Multiple date tables

- When comparisons need to be made between dates, multiple date tables are sometimes used.
- The reason for this is that you can only have one active relationship between tables.
- This adds extra data to your model, which can be inefficient.
- It is better to use the USERELATIONSHIP function in DAX.
- This allows for easier comparisons, and the relationship only becomes active when the measure is used.

Example of multiple date tables



```
€ Total Sales DueDate =  
CALCULATE(  
    SUMX(  
        FactSales,  
        FactSales[SalesQuantity] * FactSales[UnitPrice]  
    ), USERELATIONSHIP('Date'[Date], FactSales[DueDate])  
)
```

YEAR	Month	€ Total Sales	€ Total Sales DueDate
2020	1	33.009.186,00	37.071.480,00
2020	2	35.875.667,00	32.881.523,00
2020	3	37.972.129,00	39.884.837,00
2020	4	23.246.283,00	24.885.072,00
2020	5	25.203.379,00	25.347.549,00
2020	6	22.731.355,00	22.762.326,00
2020	7	31.128.116,00	29.408.149,00
2020	8	26.639.125,00	28.419.751,00
2020	9	27.177.507,00	25.827.772,00
2020	10	29.276.942,00	30.009.064,00
2020	11	24.816.322,00	25.134.687,00
2020	12	28.318.035,00	27.652.516,00
Total		345.394.046,00	349.284.726,00

Level of detail in the tables

The level of detail at the lowest level in your table.

Consider what is needed at the lowest level of the table, for example:

- What is the lowest level of each individual purchase?
- Is it sufficient to store the revenue per day?

DateKey	TotalCost	Date	€ Total Sales
7-4-2019 0:00:00	52,00	7-4-2019 0:00:00	1.529.817,00
7-4-2019 0:00:00	56,00	8-4-2019 0:00:00	1.140.754,00
7-4-2019 0:00:00	66,00	9-4-2019 0:00:00	1.671.361,00
7-4-2019 0:00:00	115,00	10-4-2019 0:00:00	1.988.986,00
7-4-2019 0:00:00	131,00	11-4-2019 0:00:00	1.707.435,00
7-4-2019 0:00:00	171,00	12-4-2019 0:00:00	1.127.218,00
7-4-2019 0:00:00	242,00	13-4-2019 0:00:00	1.564.933,00
7-4-2019 0:00:00	254,00	14-4-2019 0:00:00	1.837.926,00
7-4-2019 0:00:00	285,00	15-4-2019 0:00:00	1.436.617,00
7-4-2019 0:00:00	356,00	16-4-2019 0:00:00	1.808.614,00
7-4-2019 0:00:00	408,00	17-4-2019 0:00:00	2.025.175,00
7-4-2019 0:00:00	413,00	18-4-2019 0:00:00	1.873.287,00
7-4-2019 0:00:00	436,00	19-4-2019 0:00:00	1.633.725,00
7-4-2019 0:00:00	484,00	20-4-2019 0:00:00	1.544.020,00
7-4-2019 0:00:00	509,00	21-4-2019 0:00:00	1.459.055,00
7-4-2019 0:00:00	758,00	22-4-2019 0:00:00	1.883.524,00
7-4-2019 0:00:00	827,00	23-4-2019 0:00:00	1.520.738,00
7-4-2019 0:00:00	831,00	24-4-2019 0:00:00	1.570.734,00
7-4-2019 0:00:00	1.274,00	25-4-2019 0:00:00	1.704.843,00

Many-to-many relationships

Many-to-many (M:M) relationships

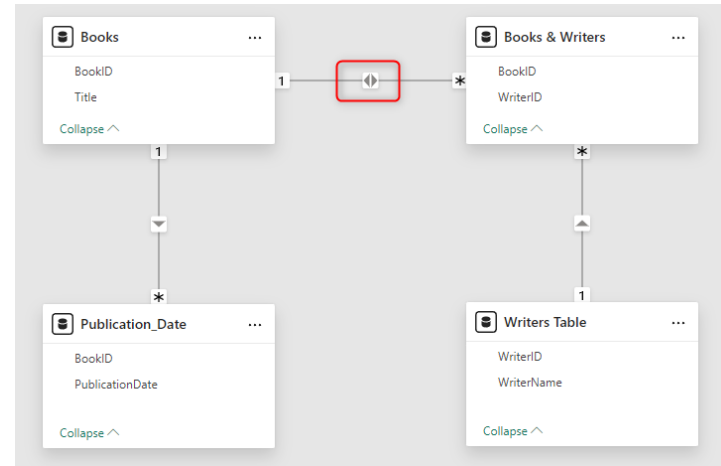
Example: Books and writers

How to solve this?

- Create a bridge table
- Make a table containing all keys

Pay attention to the direction of the cross-filter:

- Filtering only works from the 1 to * side






Importance of the Star Schema for VertiPaq

- Star schema boosts performance:
 - Faster queries by separating dimensions and facts.
 - Simplifies data for efficient VertiPaq processing.
- Proper data preparation is key:
 - Consistent data format ensures better compression and lower RAM usage.
- Small tweaks, big gains:
 - Minor adjustments can greatly improve performance.
 - Next: Key factors for optimizing compression.



Importance of Choosing the Right Data Types

- Select the correct data types to optimize model performance
 - Understand data requirements and report needs
 - Power Query defaults to "Decimal" for numeric columns:
 - May lead to unnecessary digits after the decimal point
 - Evaluate if fewer decimal places or whole numbers are sufficient
 - Correct data type choice is crucial for VertiPaq compression
- 

Example Decimal number

Metadata Functions DMV

Test_decimal_test

Model

Search

Demo_distinct

id

1.2 Numbers

Numbers
Demo_distinct[Numbers]

Type Column

Data Type 1.2 Double

Distinct Values 41,264,868

Min Value 0,00228174037

Max Value 99999,9964857808

Log Results History VertiPaq Analyzer

Tables

Columns

Relationships

Partitions

Summary

Test_decimal_test.pbix

Total Size in Memory 2,06 GB

Last Data Refresh 1-10-2024 18:19:56 +02:00

Analysis Date 2-10-2024 18:44:44 +02:00

Compatibility 1550

Tables 1

Columns 2

Server localhost:57692

File Home Insert Modeling View Optimize Help

Name Numbers

Format General

Data type Decimal number

Formatting

Whole number

Decimal number

Fixed decimal number

Date/time

Date

Time

Text

True/false

Binary

Example Fixed Decimal number

Metadata Functions DMV

Test_decimal_test

Model

Search

Demo_distinct

123 id

1.2 Numbers

Numbers
Demo_distinct[Numbers]

Type	Column
Data Type	1.2 Double
Distinct Values	40,786,499
Min Value	0.0023
Max Value	99999.9965

Log Results History **VertiPaq Analyzer**

Tables

Columns

Relationships

Partitions

Summary

Test_decimal_test.pbix

Total Size in Memory: 579,46 MB

Last Data Refresh: 2-10-2024 18:47:24 +02:00

Compatibility	Tables	Columns
1567	1	2

Name: Numbers

Format: General

123 Data type: Fixed decimal number

Format: \$ % , .00 →%

Whole number

Decimal number

Fixed decimal number

Date/time

Date

Time

Text

True/false

Binary



Optimizing Data Loading for Compression

- Load only necessary data to achieve optimal compression
- Reduce the number of columns per table:
 - Fewer columns lead to more effective compression
 - VertiPaq sort order technique: Stores columns with the lowest cardinality first for better compression
 - More columns, especially with higher cardinality, reduce compression efficiency
- Limit to around 15 columns per table to maintain an efficient model



Calculated Columns: Pros and Cons

- Calculated columns are useful during development:
 - Allow for quick testing and validation without modifying data sources
 - Provide flexibility and speed up development
- Limitations of calculated columns in production:
 - Added after model compression, not compressed efficiently
 - Increase memory usage and reduce performance
 - Can undo optimization efforts by increasing model size

Best practices

- Always use a star schema, or a snowflake schema if necessary.
- A fact table contains values for calculations.
- A dimension table contains values for filtering.
- Avoid using bidirectional relationships.
- Include only the data you actually use.
- Determine the level of detail for your tables in advance.
- Choose the right data type for each column

If you have questions or insights, please contact me !



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