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Software Requirements

- Power BI Desktop (Free) Power Query is built into Power BI for data transformation
- Excel (2016 and later, or Microsoft 365) Power Query is available in the "Get & Transform" section.
- Windows OS (Windows 10 or later recommended) Power Query in Power BI is optimized for Windows.

Optional:

 Power BI Service (Pro or Premium Per User License) – If publishing reports online, you'll need a Power BI account

Prereq	uisites	- Tec	hnical

Computer Capabilities & Performance Considerations

Power Query processes data transformations, and performance can be impacted by your system specs.

- RAM 8GB minimum; 16GB+ recommended for handling large datasets.
- **Processor** Intel i5/i7 or AMD Ryzen 5/7 or higher for better performance.
- Storage (SSD Recommended) Faster SSD drives improve data processing speed compared to HDD.
- Internet Speed If working with cloud data, a stable internet connection is necessary.

Prerequisites

Excel & Power Query Basics

Power Query is integrated into Excel and Power BI, so a solid understanding of **Excel functionalities** will be helpful:

Excel Tables & PivotTables & Power PivotTables

 If you've worked with Excel PivotTables, the transition to Power BI data modeling is easier.

Power Query (ETL Process)

- Extracting data from different sources.
- Transforming and shaping data using Power Query (e.g., removing duplicates, merging tables, unpivoting data)

Basic Data Modeling Concepts

Before diving into Data Modeling, a beginner needs a good grasp of:

Relational Databases & Tables

- Familiarity with concepts like tables, columns, rows, primary keys, and foreign keys
- Knowing how different tables relate to each other (one-to-many, many-to-one, many-to-many).

Data Types & Formatting

 Understanding text, numbers, dates, Boolean (True/False), and how they impact calculations.

Data Cleaning & Transformation Basics

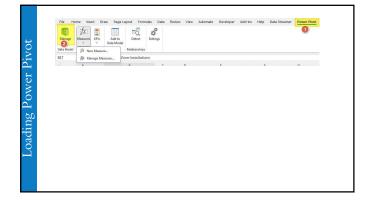
Basic knowledge of handling missing data, duplicates, and inconsistencies.

Who is Data Modeling For?

	User Group	How Data Modeling Benefits Them
1	Power BI Develops	Data modeling is essential for structuring Power BI datasets efficiently using star schema for optimal performance.
2	BI Analysts & Report Designers	Helps create models that support fast, accurate, and scalable reporting in tools like Power BI.
3	DAX Users	A strong data model minimizes the complexity of DAX formulas and improves report performance.
3	Financial Analysts and Accountants	A well-designed financial data model supports profitability analysis, forecasting, and budgeting.
4	Credit Managers & Sales Teams	Helps structure customer and transaction data for segmenting audiences and tracking key metrics.
5	Decision Makers & Executives	Understanding data models helps in financial reporting and operational efficiency.

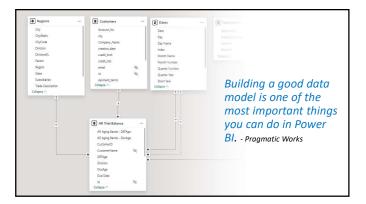
Benefits of Learning Data Modeling in Excel:

- 1. Excel is familiar territory for all of us
- 2. Increased opportunities for internal use leads to wider application
- 3. Increased usage cases leads to increased experience
- ${\bf 4.} \ \ {\bf Small \ quick \ wins \ will \ encourage \ motivation \ to \ learn \ more.}$
- 5. Logical transition to Power PI through PowerPivot



Why is a Data Model Important?

- Optimized Performance Reduces redundancy and improves efficiency.
- Scalability Allows easy expansion of data as your model grows.
- Accurate Insights Prevents errors and inconsistencies.
- Better Data Relationships Enables complex queries and drill-downs.
- Better DAX Performance & Simplicity Makes DAX Measures easier to write and troubleshoot.



	ad Data Models:	
	Poor Performance & Slow Reports	
	Incorrect Data & Calculation Errors	
	Difficulty Writing & Understanding DAX	
	Relationship & Cardinality Issues	
. 1	Data Model is Hard to Maintain	
23	• Security Risks	
	Unusable or Messy Reports	
5	Poor Data Refresh Performance	
	Inconsistent Reporting & Business Intelligence	
	Lack of Scalability & Flexibility	
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Importance of the Data Model

1. Performance Optimization

- A well-structured data model ensures faster query performance by reducing redundancy and improving compression.
- Proper relationships and indexing reduce the load on DAX calculations, improving responsiveness.
- Avoiding unnecessary calculated columns and instead using measure-based calculations ensures efficiency.

2. Data Accuracy & Consistency

- A properly normalized data model prevents duplicate data and inconsistencies.
- Correctly defining relationships (one-to-many, many-to-one) ensures that data aggregates accurately without errors.
- Using surrogate keys instead of composite keys avoids ambiguity in relationships.

Importance of the Data Model

3. Better DAX Performance & Simplicity

- A well-designed model simplifies DAX formulas, making them easier to write and troubleshoot.
- Flat or poorly structured models can lead to complex, slow DAX queries.
- Keeping a star schema instead of a "snowflake" or "spaghetti" model improves
 the readability and maintainability of DAX calculations.

4. Scalability & Flexibility

- As the business requirements and personnel change, modifying a solid model is easier to understand and restructure than a messy one.
- A structured model makes it easier to scale as data volume grows. It allows for the
 addition of new tables and data sources without breaking existing reports.

Importance	01	f the	Data	Mod	e

5. Improved Report Performance

- The right relationships and calculated fields reduce the need for excessive visuals and filters, leading to faster dashboards.
- A model optimized for import mode or DirectQuery enhances report performance by minimizing unnecessary calculations.

6. Security & Row-Level Security (RLS) Implementation

- A well-structured model supports role-based access and Row-Level Security (RLS).
- Poorly designed models can lead to security loopholes or data leakage.

7. Easier Maintenance & Collaboration

- A clean model helps other analysts understand and build on it without confusion.
- If multiple developers work on the model, a structured approach ensures team collaboration without causing disruptions.

8. Efficient Storage & Reduced File Size

- A normalized model with proper relationships results in better compression and reduced memory consumption.
- Flattened models with excessive columns increase file size and slow down queries.

Importance of the Data Model

9. Data Reusability

- A well-structured data model can serve multiple dashboards and reports. Cross Reporting Drill Through.
- Rather than reinventing data structures for every new report, a good model supports different business needs efficiently.

10. Business Logic Centralization

- Keeping business logic in the model (through measures and relationships) ensures consistency across all reports.
- If logic is embedded in visuals, it becomes harder to manage and leads to inconsistencies.



Data Modeling Done Right

OVERVIEW

Prerequisites

Basic Data Modeling Concepts

Basic Power BI Navigation

Data Modeling in Power BI

Data Modeling in Power Pivot Excel

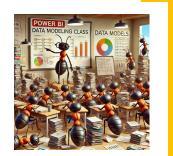
Basic Data Modeling Concepts Organizing Your Analytical Data

Fact vs. Dimension Tables

 Learning the Star Schema and Snowflake Schema structures.

Relationships in Power BI

- · Cardinality (one-to-many, manyto-one, many-to-many).
- Direction of relationships (single vs. both directions)
- Active vs. inactive relationships



What is a Data Model

A data model is an structured framework that defines how data is stored, organized, and

In the context of Power BI, a data model consists of tables, relationships, and calculations (measures, calculated columns) that help transform raw data into meaningful insights.

Different Data Model Types are:

- Flat Model A single table with all data (inefficient for large datasets).
- Relational Model Multiple tables with defined relationships. (most common in Power BI). Star Schema A central fact table surrounded by multiple dimension tables (ideal for
- reporting). Often called dimensional data modeling.

 Snowflake Schema A more normalized version of the star schema with further breakdown

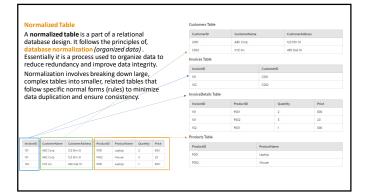
Key Components of Dimensional Data Modeling

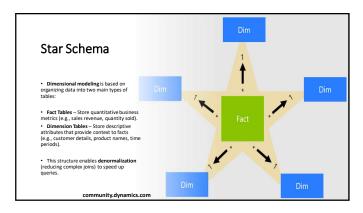
- Fact Tables Store measurable numerical data (e.g., sales, revenue, credit limits).
- Dimension Tables Store descriptive data (e.g., customers, products, regions)
- Relationships Define how dimensions relate to facts (typically one-to-many)
- Surrogate Keys Unique identifiers used in dimension tables instead of Natural keys.
- Hierarchies Logical drill-down structures (e.g., Year \rightarrow Quarter \rightarrow Month \rightarrow Day).

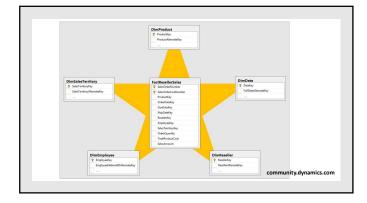
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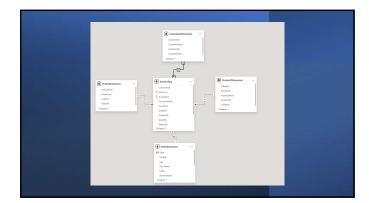
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ontrol Systems	Northeast Region	04	SC _h	requiring joins to other tables. Each row in the table
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ng Services	Northeast Region	04	SCR	attributes included as columns.
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urity	Northeast Region	04	SCR	 No Relationships or Joins Required – All data is stored in a
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Integation	Northeast Region	04	SCR	multiple tables. (No VLOOKUP/XLOOKUP, SUMIFS or
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ontrol Systems	Midwest Region	03	SPR	 Wide Table Structure – The table may have many columns
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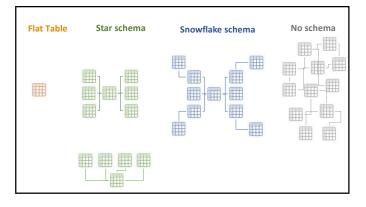
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20	9546985	Mertz LLC	1002844	4641 SWD-CH	N30	INV100284400	7/29/2024	8/28/2024	s	332.564.00	s	332,564.00	12/31/2024
41	1742275	Hoeger, Ouitzon an	1002893	4202 SWD-PEO	N60	INV100289300	8/3/2024	10/2/2024	s	272.625.00	s	272,625.00	12/31/2024
22	9879585	Raynor-Ebert	1002858	3581 MON-CHI	N90	INV100285800	8/3/2024	11/1/2024	\$	395,059.00	\$	395,059.00	12/31/2024
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24	1680902	Willms Group	1002866	1306 ACS-CHI	DR	INV100286600	8/6/2024	8/6/2024	\$	407,142.00	\$	407,142.00	12/31/2024
25	6803991	Smitham, Koch and	1002869	2676 OCTV-SCR	N30	INV100286900	8/7/2024	9/6/2024	\$	196,971.00	s	196,971.00	12/31/2024
28	1032609	VonRueden Group	1002877	1051 MON-PEO	DR	INV100287700	8/11/2024	9/30/2024	\$	57,543.00	\$	57,543.00	12/31/2024
33	1752145	Baumbach-Wilkins	1002884	4818 ACS-CHI	N90	INV100288400	8/18/2024	11/16/2024	\$	367,536.00	s	367,536.00	12/31/2024
36	3460752	Mayer-Kozey	1002888	1115 MON-CHI	N60	INV100288800	8/19/2024	10/18/2024	\$	322,423.00	\$	322,423.00	12/31/2024
42	2449050	Kohler-Nolan	1002894	4077 CYB-PEO	N60	INV100289400	8/21/2024	10/20/2024	\$	321,975.00	s	321,975.00	12/31/2024
49	3160691	Swaniawski-Weiss	1002901	4116 ACS-PEO	N60	INV100290100	8/23/2024	10/22/2024	\$	336,570.00	\$	336,570.00	12/31/2024
51	1195833	Daugherty Inc	1002903	4003 SWD-SPR	N90	INV100290300	8/27/2024	11/25/2024	\$	438,170.00	s	438,170.00	12/31/2024
54	1632070	Hansen-McGlynn	1002906	1254 ACS-PEO	DR	INV100290600	8/30/2024	8/30/2024	\$	327,898.00	\$	327,898.00	12/31/2024
55	6655263	Harber-Parisian	1002907	1290 SYSI-SCR	N90	INV100290700	9/2/2024	12/1/2024	\$	39,553.00	s	39,553.00	12/31/2024
58	6850075	Mertz, Kub and Bar	1002910	2496 HWD-PEO	N60	INV100291000	9/4/2024	11/3/2024	\$	412,417.00	\$	412,417.00	12/31/2024
109	4934876	Wunsch Inc	1002961	2680 SWD-SPR	N60	INV100296100	9/5/2024	11/4/2024	\$	112,190.00	s	112,190.00	12/31/2024
120	8054195	Abernathy, Kuhlma	1002972	4490 ACS-CHI	N60	INV100297200	9/5/2024	11/4/2024	\$	380,343.00	s	380,343.00	12/31/2024
121	1597766	Walsh-Batz	1002973	1433 MON-PEO	N60	INV100297300	9/5/2024	11/4/2024	\$	493,666.00	\$	493,666.00	12/31/2024
60	1043931	Purdy, Ortiz and Ro	1002912	4866 ACS-CHI	N90	INV100291200	9/5/2024	12/4/2024	\$	41,467.00	s	41,467.00	12/31/2024
85	2657814	Halvorson, Predovi	1002937	2354 HWD-SCR	N90	INV100293700	9/5/2024	12/4/2024	\$	105,388.00	\$	105,388.00	12/31/2024
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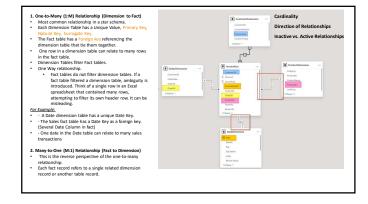


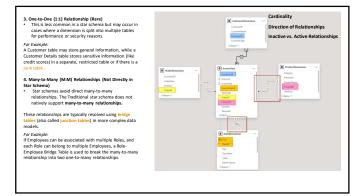


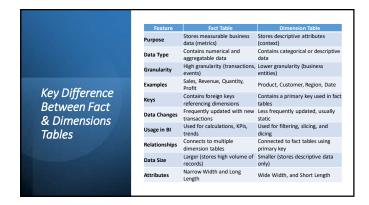


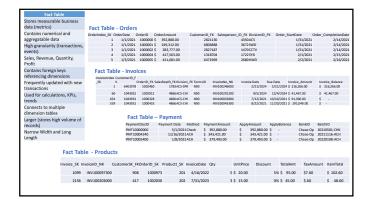




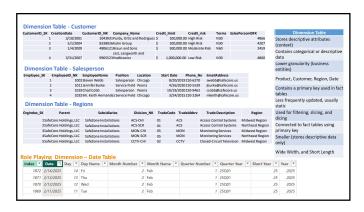








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Date Table - DAX Method

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CALENDAR (DATE(2015,1,1), DATE(2030,12,31)),
"Year", YEAR[[Date]),
"Month", FORMAT[[Date], "MMMM"),
"Month Number", MONTH([Date]),
"Quarter", "Q" & FORMAT[[Date], "Q"),
"Day of Week", FORMAT[[Date], "DDD"),
"Week Number", WEEKNUM([Date])

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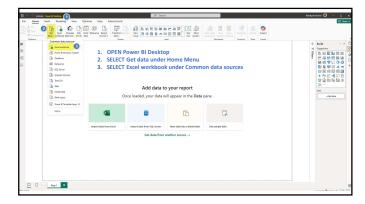
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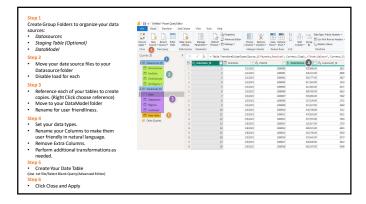
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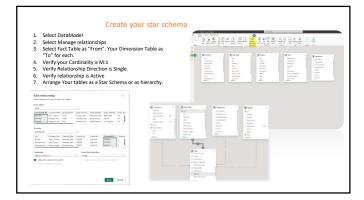
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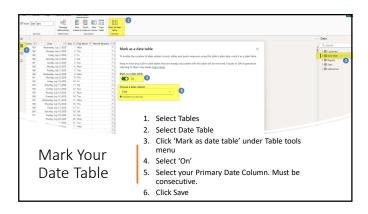
Basic Power BI Navigation Importing Data • Connecting Power BI to dataset. Building Simple Data Model • Connect our Data Model • Create a Date Table • Mark our Date Table Building Simple Reports • Dragging fields into visuals. • Formatting charts and tables.

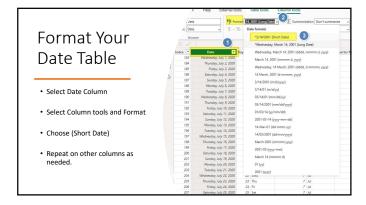


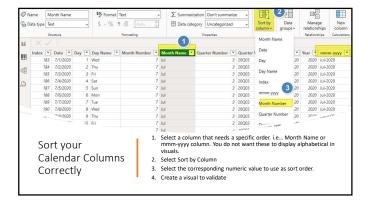


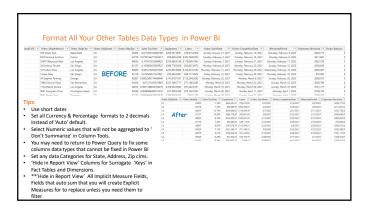












Date Table - DAX Method

DateTable =
ADDCOLUMNS {
CALENDAR (DATE(2015,1,1), DATE(2030,12,31)),
"Year", YEAR([Date]),
"Month, "FORMAT([Date], "MMMM"),
"Month Number", MONTH([Date]),
"Quarter", "Q" & FORMAT([Date], "Q"),
"Day of Week", FORMAT([Date], "DDD"),
"Week Number", WEEKNUM([Date])
"Week Number", WEEKNUM([Date])

PowerPivot Excel Comparison

Importing Data

Connecting Power BI to dataset.

Building Simple Data Model

- Connect our Data Model
- Create a Date Table

Mark our Date Table

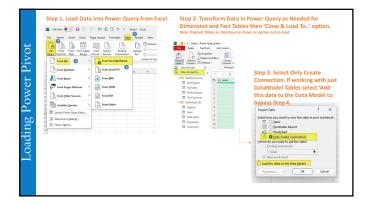
Building Simple Reports

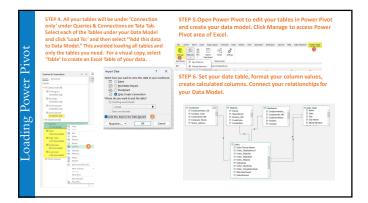
- Dragging fields into visuals.
- Formatting charts and tables.

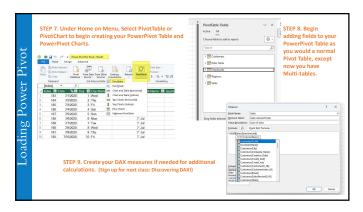


Benefits of Learning Power Pivot in Excel:

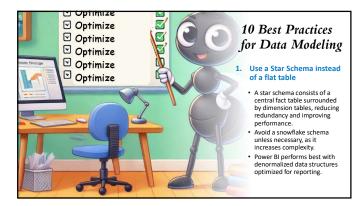
- 1. Excel is familiar territory for all of us
- 2. Increased opportunities for internal use leads to wider application
- 3. Increased usage cases leads to increased experience
- 4. Small quick wins will encourage motivation to learn more.
- 5. Logical transition to Power PI through PowerPivot







	Feature/Aspect	Power Pivot in Excel	Power BI	
Excel Power Pivot vs Power Bl Considerations	Purpose	Enhances Excel's data modeling and analysis capabilities Good for small scale, one time, ad-hoc analysis reports.	A full-fledged BI and reporting tool for analytics Power BI is better when the analysis needs to be repeated, refreshed, or shared with others dynamically.	
	Data Modeling	Supports data modeling with DAX and relationships	More advanced modeling, including DAX, M	
	Data Sources	Limited connectivity (Excel, SQL, Access, OData, etc.)	Extensive support (SQL, APIs, cloud sources, etc.)	
	User Interface	Excel-based, uses Power PivotTables & Power Query integration	Dedicated UI with visuals, slicers, dashboards	
	Visualization	Basic (PivotTables, PivotCharts)	Advanced (interactive dashboards, custom visuals) Creating interactive dashboards	
	Data Refresh	Manual or scheduled via Power Query and Excel features	Automated refresh via Power BI Service	
	Data Volume	Limited to Excel's memory constraints (~1M rows)	Can handle billions of rows with efficient compression	
	Sharing & Collaboration	Via Excel file sharing or OneDrive	Cloud-based sharing via Power BI Service	
	Security	Limited security controls (password, OneDrive settings)	Row-level security (RLS), role-based access	
	Advanced Features	Some support for DAX, Measures, KPIs	Advanced AI, predictive analytics, Python, R	
	Deployment	Local desktop use only	Can be deployed to cloud, Power BI Service, Power BI Report Server	
	Automation	Limited (Excel Macros, VBA, Power Automate)	Supports Power Automate, APIs, Al-driven insights	
	Updates & Support	Slow updates (dependent on Excel updates)	Frequent updates with new features monthly	
	Licensing	Included in Excel (with Power Pivot enabled)	lequires Power BI Free, Pro, or Premium	



10 Best Practices for Data Modeling

- 2. Reduce Relationships & Cardinality Issues
 - Use one-to-many relationships rather than many-to-many to prevent performance issues.
 - Reduce high-cardinality columns in relationships (e.g., avoid using unique transaction IDs unless necessary).
 - Avoid bidirectional filtering unless essential—it can introduce ambiguity and slow down calculations.
- 3. Keep the Model as Simple as Possible
 - Remove unnecessary tables and columns that don't add value.
 - Create a single source of truth by defining clear measures and relationships.
 - Consolidate multiple fact tables if they serve the same purpose.

<i>10</i>	Best	Practices	for	Data	Mod	lelin	Q

4. Optimize Table Relationships

- Ensure fact tables are on the "many" side and dimension tables are on the "one" side.
- · Use integer keys instead of text keys for relationships (e.g., use CustomerID instead of CustomerName).
- Use surrogate keys where possible to improve relationship efficiency.

5. Use Aggregated Columns & PreCalcs in Data Transformation

- Perform aggregations (e.g., totals, averages) before loading data into Power BI to reduce DAX calculations. (Do in SQL or PowerQuery)
- Avoid row-by-row transformations in Power Query when table-wide calculations are more efficient.

 Remove unnecessary decimal precision in numerical columns.

10 Best Practices for Data Modeling

6. Use Measures Instead of Calculated Columns

- Measures are dynamic and computed at query time, whereas calculated columns increase memory usage.
- Use DAX measures for calculations that involve aggregations (e.g., SUM. AVERAGE. COUNT).
- Only create calculated columns when they are required for relationships or filters.

7. Leverage Power BI's Performance Optimization Features

- Enable query reduction options to minimize unnecessary queries.
- Use aggregation tables for large datasets.
- Utilize composite models when working with large datasets in DirectQuery mode.

10 Best Practices for Data Modeling

8. Use Role-Based Security Properly

- Implement Row-Level Security (RLS) to restrict data access by users or to make reports more region or user specific.
- Avoid hardcoding filters in RLS—use role-based tables instead.
- Test RLS configurations thoroughly to ensure security and performance balance.

9. Index & Partition Large Datasets

- Partition large fact tables in the data source to improve query performance.
 Use indexes on foreign keys in your SQL source system for faster indexes.
- Leverage incremental refresh for large datasets to optimize load times.

10 Best Practices for Data Modeling

10. Document Your Data Model

- Use clear naming conventions for tables, columns, and measures
- Create a data dictionary explaining key business logic.
- Add descriptions in Power BI for datasets, relationships, and measures

Remember: Good Data Models Provide: - Performance Optimization - Data Accuracy & Consistency - Better DAX Performance & Simplicity - Scalability & Flexibility - Improved Report Performance - Security & Row-Level Security (RLS) Implementation - Easier Maintenance & Collaboration - Efficient Storage & Reduced Data Size - Data Reusability - Business Logic & Centralization

