

Francesco De Chirico



# Il mio grosso, grasso Power BI model

 \*PASS  
SQLSATURDAY



# Francesco De Chirico

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## Consultant, trainer and speaker

Specializing in Power BI, SSAS, MDX, DAX and M.  
Using SSAS and MS BI Platform since 2001

## Main certifications

MCT since 2008, SSAS Maestro in 2012 and  
Microsoft Professional Program Data Science in 2017

## Projects

Ideator and co-developer of ASQA tool



<https://ssasqueryanalyzer.github.io/>



PASS



A large, teal-colored decorative graphic on the left side of the slide, consisting of several overlapping, curved, ribbon-like shapes that create a sense of movement and depth.

# Agenda

- Why To optimize – The problem
  - Demo
- Vertipaq engine overview
- How To optimize – Best practices
- How to verify optimization – DMV
  - Demo
- Tools available
- Connect to .pbix data model
  - SSAS sandbox
  - Problem: how to get port number?
  - Demo
- Build a .pbix template
  - Power Vertipaq Demo



# What we'll talk about


- What increase .pbix size
- What we can do to reduce .pbix size
- How to identify causes

# What we'll NOT talk about

- Performance
- DAX
- Power Query and M
- ...



DEMO



VertiPaq engine  
in ~~3~~ ... ~~4~~ ... ok, 5 slides!

# Vertipaq is an in-memory columnar database

**Row store**

| ID | Name       | Color | Price  |
|----|------------|-------|--------|
| 1  | Sneakers   | Red   | 139.99 |
| 2  | T-shirt    | Red   | 18.00  |
| 3  | Hat        | White | 24.75  |
| 4  | Shirt      | Black | 70.00  |
| 5  | Shoes      | Blue  | 185.50 |
| 6  | Polo-shirt | Red   | 49.99  |
| 7  | Scarf      | Blue  | 27.50  |
| 8  | Sweater    | Black | 95.00  |
| 9  | Jacket     | Black | 375.00 |
| 10 | Trousers   | Grey  | 175.99 |

**Column store**

| ID | Name       | Color | Price  |
|----|------------|-------|--------|
| 1  | Sneakers   | Red   | 139.99 |
| 2  | T-shirt    | Red   | 18.00  |
| 3  | Hat        | White | 24.75  |
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- Very fast on single-column access
- More columns require to reorganize the information
- The more columns you need the harder to obtain the result!

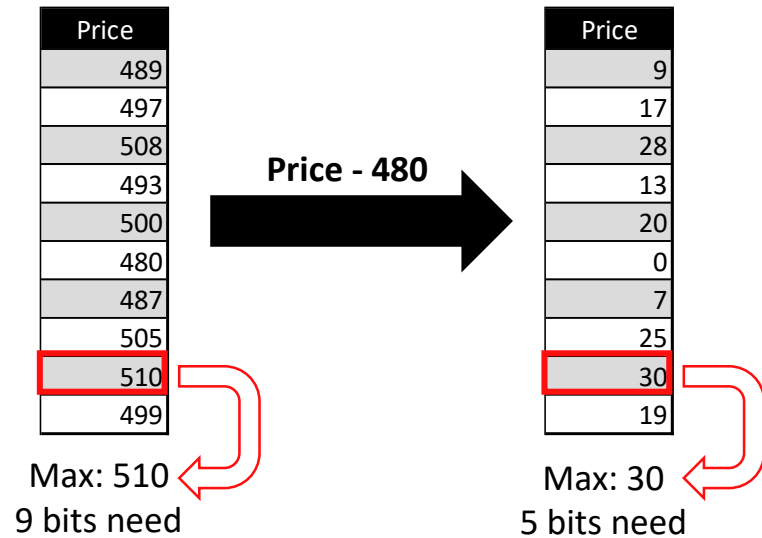
- Data is organized in rows
- Sum of Price:
  - Start reading the first row
  - Discard 75% of data (ID, Name, Color)
  - Retain the searched value (Price)
  - Move to the next row and repeat until the end of the table

- Data is organized in columns
  - Optimize vertical scanning
- Sum of Price:
  - scan the Price column ONLY



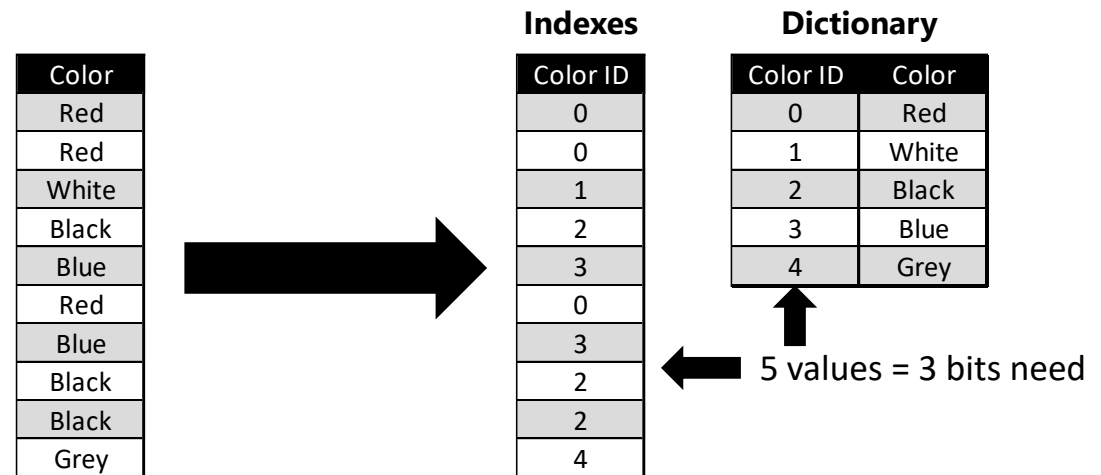
# Vertipaq compression in pills

## Value encoding



Only for integer columns

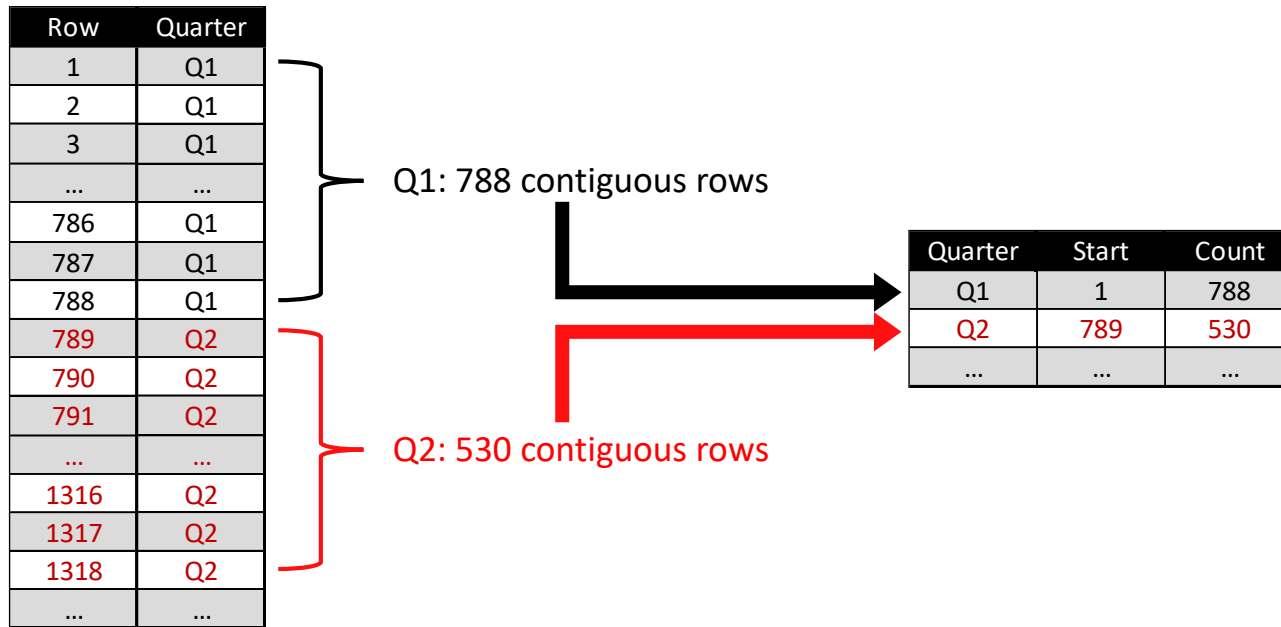
## Dictionary encoding



Bits to store a single value = minimum bits of index entry  
All columns contain ONLY integer values  
**It does not matter what the original data type is!**

# Vertipaq compression in pills

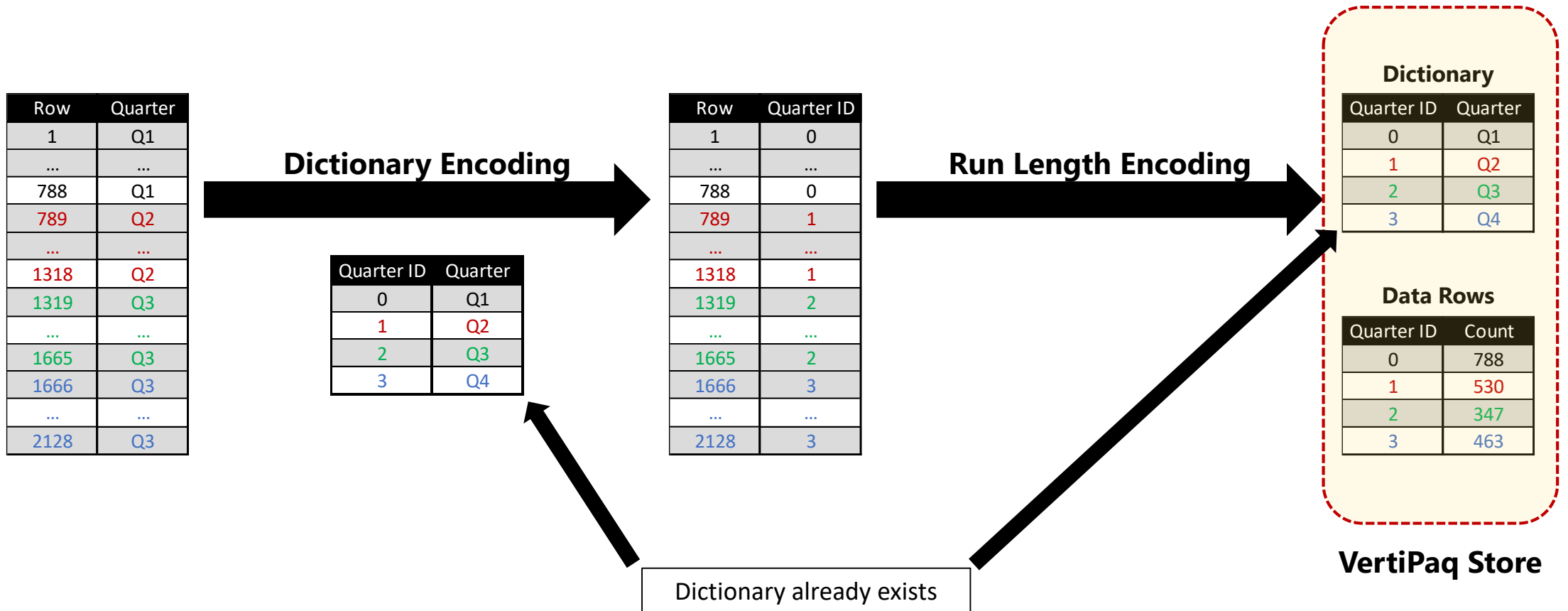
## Run Length Encoding (RLE)



- RLE's efficiency strongly depends on the repetition pattern of the column
- Two rows with the same cardinality can have very different compression ratio
- Sorting of data is extremely important to improve the compression ratio
- In columns with very high cardinality (i.e. primary keys) RLE is larger than the column itself.

# Vertipaq compression in pills

Run Length Encoding (RLE) applied to the dictionary-encoded version of a column



# Data models in pills of pills!

## Denormalized

| ID  | Product Name | Product Code | Price  |
|-----|--------------|--------------|--------|
| 1   | Sneakers     | 00-3475      | 139.99 |
| 2   | T-shirt      | 17-8553      | 18.00  |
| ... | ...          | ...          | ...    |

VS

## Normalized

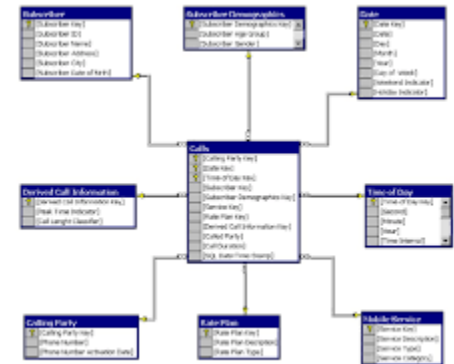
| ID  | Product Code | Price  |
|-----|--------------|--------|
| 1   | 00-3475      | 139.99 |
| 2   | 17-8553      | 18.00  |
| ... | ...          | ...    |

| Product Code | Product Name |
|--------------|--------------|
| 00-3475      | Sneakers     |
| 17-8553      | T-shirt      |
| ...          | ...          |



## Star schema




- Every relationship has an additional memory cost
  - Normalized Memory cost:  
 $ColumnCost(Price[Product\ Code]) + ColumnCost(Product[Product\ Code]) + ColumnCost(Product[Product\ Name]) + RelationshipCost(Price[Product\ Code])$
  - Denormalized Memory cost:  
 $ColumnCost(Price[Product\ Code]) + ColumnCost(Price[Product\ Name])$
- Theoretically the “optimal” model:
  - One single table
- In reality it is less usable and would force to have a single granularity for all the measures

# VertiPaq compression ratio - summary

Factor to consider (in order of importance):

- The cardinality of the column
  - Determines the number of bits used to store a value
- The distribution of data in the column
  - Many repeated values = high compression ratio
  - Frequently changing values = low compression ratio
- The number of rows in the table
- The datatype of the column
  - It affects only the dictionary size



Well, now we know how  
VertiPaq engine works  
(more or less 😊), so ...

# Best practices: at least 6 tips

- Import only useful columns
- Remove unused fields
  - Especially those with high cardinality
  - Best candidate: ID's
- Set correct data type
  - Typical issue: strings containing only numbers
- No datetime type!
  - Use ONLY date ← If you can 😊
  - If you need also time → next tips

# Best practices

- Split fields in two (or more) fields
  - Do you remember the previous tip? 😊
    - Split a datetime field in two: Date AND Time
- Shrink fields to reduce cardinality
  - Changing the Precision of Numeric Columns
- Disable auto date/time functionality
  - Particularly when you have many dates
  - **Cons: you need to manually build hierarchies!**



# Best practices – how to split

- Datetime

- date → (cardinality = 365 \* number of years)

- time → (cardinality = 24h \* 60min \* 60sec = 86.400)

- Ex: `CAST( MyDateTime AS DATE ) AS MyDate` → 2019/02/23 ~~12:37:45~~

- Ex: `CAST( MyDateTime AS TIME( 0 ) ) AS MyTime` → ~~2019/02/23~~ 12:37:45

- Big integer

- High value (INT)

- Low value (INT)

- ex: `CAST( MyBigValue / 1000000 AS INT ) AS MyValue_High` 1368405 ~~885980~~

- ex: `CAST( MyBigValue % 1000000 AS INT ) AS MyValue_Low` ~~1368405~~ 885980

- To obtain the original value:


- `MyValue_High * 1000000 + MyValue_Low`

# Best practices – how to split

- Long strings
  - LEFT substrings (leftmost "n" chars)
  - RIGHT substring (rightmost "n" chars)
    - ex: `LEFT(MyLongString, "n") AS MyString_Left`
    - ex: `SUBSTRING(MyLongString, "n" + 1, LEN(MyLongString) - "n") AS MyString_Right`
    - To obtain the original value:
      - `CONCAT(MyString_Left, MyString_Right)`
- Decimal numbers
  - Integer part
  - Decimal part
    - ex: `FLOOR(MyDecimalValue) AS MyValue_IntegerPart`
    - ex: `MyDecimalValue - FLOOR(MyDecimalValue) AS MyValue_DecimalPart`

# Best practices – shrink fields

- long strings used as ID's
  - replace with integer ←If you can 😊
  - If used for a distinct count measure, replace with RANK()
- Decimal numbers
  - Reduce decimal precision ←If you can 😊
  - i.e. a temperature value with more than 2 decimal
- Calculations
  - Do not import columns containing result of calculation
  - Calculate the value in the model:
    - TotalSale = `SUMX( Sales, Sales[Price] * Sales[Quantity])`




Now the problem is ...  
"how to identify critical  
columns?"

# Retrieve metadata info of Tabular model

- The “rough” way:
  - Connect to SSAS Tabular instance with SSMS
  - Execute a set of DMV using MDX/DMX queries
  - Store the results in a SQL database
  - Write some views to analyze data
- The “easy” way:
  - Connect to SSAS Tabular instance with Vertipaq Analyzer
  - Refresh the Power Pivot model
- Cons:
  - DMV’s results are not easy to analyze
  - Cannot join DMV’s
  - Many DMV’s to execute to retrieve useful data
- Pros:
  - Very simple to use
  - All calculations are provided in the Power Pivot model

A large, stylized teal graphic on the left side of the slide, consisting of several overlapping, curved, ribbon-like shapes that form a partial frame or abstract shape.

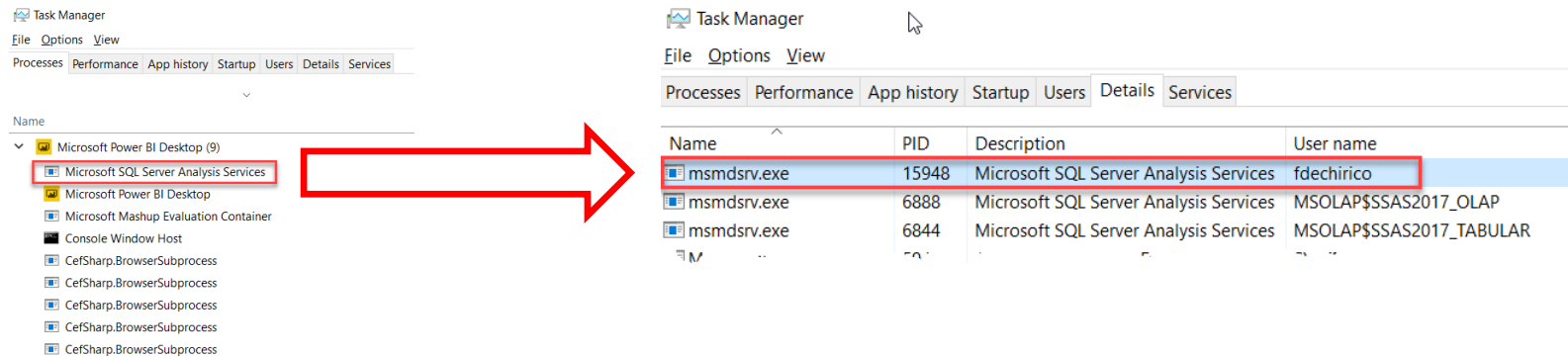
# DMV and VertiPaq Analyzer demo



Ok, but what about  
Power BI model?  
Is it possible to connect  
to a .pbix data model?

# Connect to a .pbix data model


- YES, it is possible!!! 😊
- Power BI Desktop runs a local instance of SSAS Tabular engine in the background and it assigns a random port number to that local instance:



- It is possible to connect to that instance and to the Power BI model **BUT** you need to know that port number 😞



# How to get the port number?

- Command Line (CMD)
  - DAX Studio
    - Connect to your .pbix
    - Local SSAS instance address on the right bottom of the DAX studio window
- 
- The screenshot shows the status bar of DAX Studio. It contains the text 'Ln 1, Col 1', 'localhost:57142', '15.0.2.159', and '268'. The 'localhost:57142' text is highlighted with a red rectangular box.
- Tabular Editor
  - Power BI Desktop Temporary Location
    - Download Edition: *%LocalAppData%\Microsoft\Power BI Desktop\AnalysisServicesWorkspaces*
    - Store Edition: *%username%\Microsoft\Power BI Desktop Store App\AnalysisServicesWorkspaces*
    - Open "**msmdsrv.port.txt**"
  - PowerShell (using the PowerShell module "[PowerBIPS.Tools](#)")

# Retrieve metadata info of Power BI desktop model

- The “rough” way:
  - Launch the Power BI desktop report you want to analyze
  - Found the port used by the local SSAS Tabular instance
  - Same steps used for Tabular model
- The “easy” way:
  - Launch the Power BI desktop report you want to analyze
  - Found the port used by the local SSAS Tabular instance
  - Same steps used for Tabular model

# Retrieve metadata info of Power BI desktop model

- The “alternative” way:
  - Connect to the SQLite db used by the Power BI desktop model
    - Install SQLite ODBC driver (<http://www.ch-werner.de/sqliteodbc>) on your local machine
    - Open the .pbix model you want to analyze with Power Bi desktop
    - Go to “%LocalAppData%\Microsoft\Power BI Desktop\AnalysisServicesWorkspaces”
    - Identify the Workspace of your model → **can be tricky**
    - Copy the complete address of the **metadata.sqlitedb** (“.\Data\**<GUID>**.**<VERSION>**.db”)
    - Get Data → ODBC → SQLite 3 Datasource
    - In the “Advanced Options” connection string insert:  
“database=<your metadata.sqlitedb file address>”

From ODBC

Data source name (DSN)  
SQLite3 Datasource

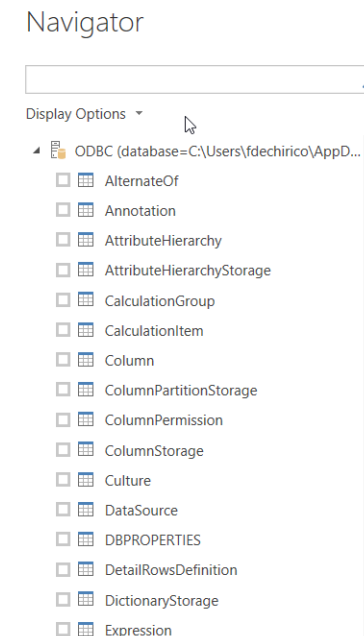
Advanced options

Connection string (non-credential properties) (optional) ⓘ

SQL statement (optional)

Supported row reduction clauses (optional)  
(None) Detect

OK Cancel



# Retrieve metadata info of Power BI desktop model

- The “ideal” (well, not exactly ... my “desiderata” 😊) way:
  - Launch the Power BI desktop report you want to analyze
  - Launch a second Power BI desktop report specifically designed to retrieve model info
  - Digit the name of the first Power BI desktop report (the one you want to analyze)
- The question is ...

**IS IT POSSIBLE?**

The answer is ...

~~NO!~~

The answer is ... Power Vertipaq!!!!



# Power VertiPaq demo