



Lab 05: Transferring large data volumes with Fabric Data Factory

Introduction

In this lab, you will ingest data from a Azure SQL Database to the Managed Lakehouse Sales table. The objective is to understand how to connect to this external database and how the Fabric Data Factory Pipeline can scale when having to transfer 10M, 50M and 100M of rows.

Objectives

After completing this lab, you will be better able to:

1. Connect to an Azure SQL Database and read the tables
2. Set up and run the Fabric Data Factory pipeline to transfer data from Azure SQL tables to a table in the managed Lakehouse.

Estimated time to complete this lab

60 minutes

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Lab Prerequisites

- Workspace: Fabric, Power Premium or Fabric trial
- Individual license: Power Pro or Premium Per User account

Information provided by your training provider

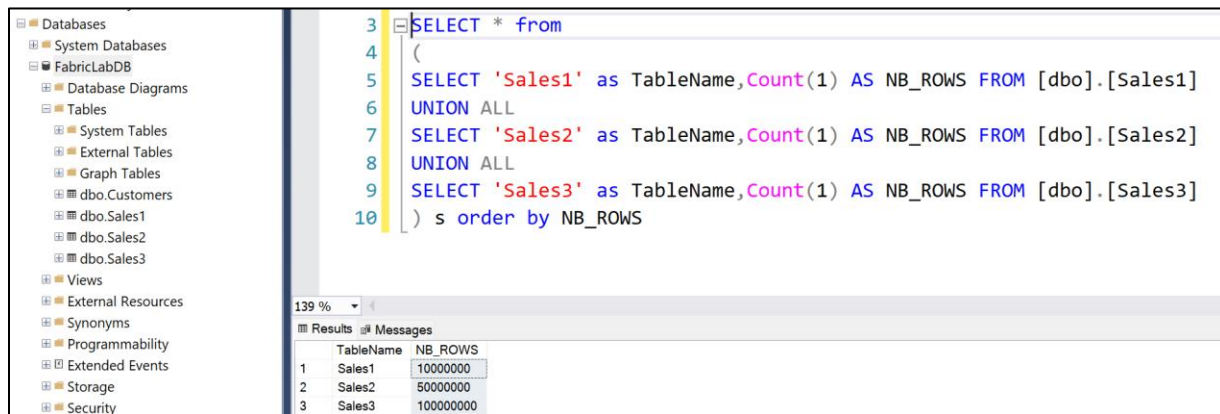
- Trial tenant (if applicable): login & password, workspace to use for the lab.
- Azure SQL Server Database: server name, login and password.

Task 1: Setup the connection to the Azure SQL Database

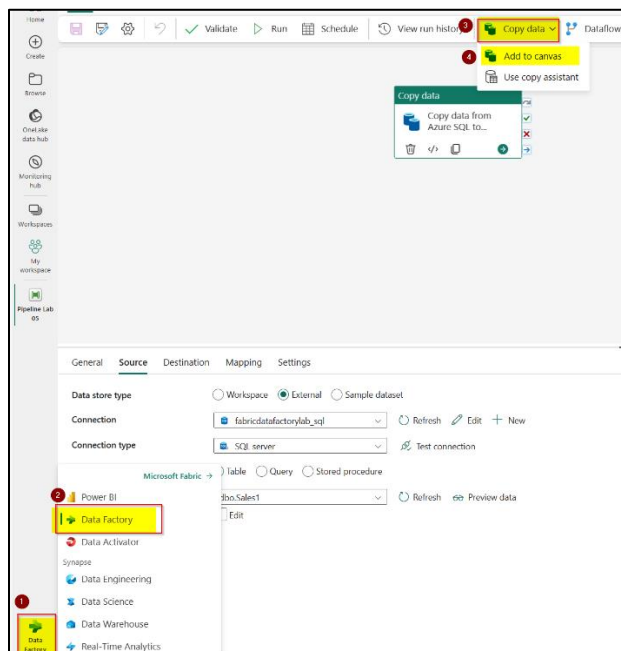
In this task, you will establish the Linked Service connection to an Azure SQL Database.

This database named FabricLabDB contains 3 tables with the same structure than the managed Sales table in the Lakehouse :

- Sales1 contains 10M rows
- Sales2 contains 50M rows
- Sales3 contains 100M rows



- On the MS Fabric menu, navigate to Data Factory and click on the Data pipeline. Provide the name Pipeline Lab 05 for this new Lab.
- On the newly created data pipeline, click on "Copy Data" then "Add to canvas"



- Configure the Data Copy Activity Task
 - Under the **General Tab**, provide the name for the Pipeline: ***Copy data from Azure SQL to Lakehouse***. The other settings remain unchanged.
 -
 - On the **Source Tab** choose the following options:
 - Data store type: External
 - Connection: Click on new and select Azure SQL Server, click on Continue.

Server: *<database server name>*.database.windows.net (information given by your trainer).

Database: fabriclabdb

Connection: Create new connection

Connection name: azuresqladb

Authentication kind: Basic

Username: LabUser (to be confirmed by your trainer)

Password: [MicrosoftFabric@2023](#) (to be confirmed by your trainer)

Thick Use **Encrypted Connection**

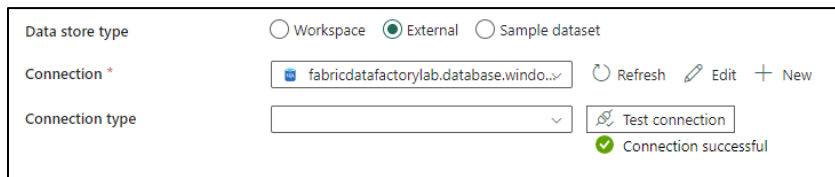
Make sure that the settings will look like this:

The screenshot shows the 'New connection' dialog in Microsoft Fabric Data Factory. The dialog is titled 'New connection' and has a 'SQL server' icon. The 'Connection settings' tab is active, showing the following fields:

- Server:** fabricdatafactorylab.database.windows.net
- Database:** fabriclabdb
- Connection:** Create new connection (dropdown menu)
- Connection name:** azuresqladb
- Authentication kind:** Basic (dropdown menu)
- Username:** LabUser
- Password:** (masked with dots)
- Use encrypted connection:** ☒

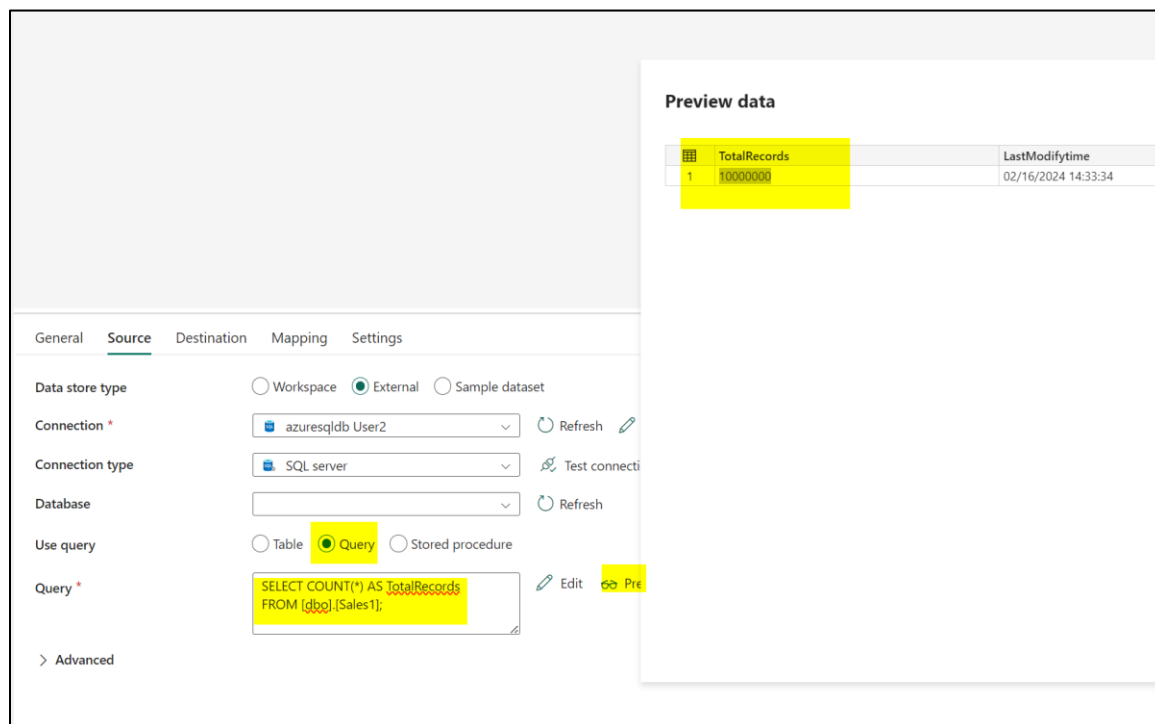
At the bottom of the dialog, there are 'Create' and 'Back' buttons.

- Click on Test Connection to make sure that the connection is successful.



- Use query: Select Table
 - **Table optional:** To find out how many records are on the dbo.Sales1, as a test you can select Query and paste this query:

```
SELECT COUNT(*) AS TotalRecords
FROM [dbo].[Sales1];
```



| | TotalRecords | LastModifytime |
|---|--------------|---------------------|
| 1 | 10000000 | 02/16/2024 14:33:34 |

- As you can see, the table dbo.Sales1 contains 10M of rows.
- **Table:** select **dbo.Sales1**. Click on Preview Data to make sure that you can read the data from the Database

General **Source** Destination Mapping Settings

Data store type ☐ Workspace ☒ External ☐ Sample dataset

Connection Refresh Edit + New

Connection type Test connection
Connection successful

Use query ☒ Table ☐ Query ☐ Stored procedure

Table Refresh Preview data
☐ Edit

> Advanced

Preview data

| | OnlineSalesKey | OrderDate | DeliveryDate | StoreKey | ProductKey | PromotionKey | CustomerKey | SalesOrderNumber | SalesOrd |
|----|----------------|---------------------|---------------------|----------|------------|--------------|-------------|------------------|----------|
| 1 | 19562610 | 2020-01-01T00:00:00 | 2020-01-04T00:00:00 | 307 | 293 | 5 | 19079 | 200701012CS425 | 2 |
| 2 | 19562623 | 2020-01-01T00:00:00 | 2020-01-06T00:00:00 | 307 | 672 | 5 | 19079 | 200701013CS425 | 2 |
| 3 | 19562636 | 2020-01-01T00:00:00 | 2020-01-03T00:00:00 | 307 | 1420 | 5 | 19079 | 200701015CS425 | 2 |
| 4 | 19562655 | 2020-01-01T00:00:00 | 2020-01-02T00:00:00 | 307 | 2167 | 5 | 19079 | 200701018CS425 | 2 |
| 5 | 19562863 | 2020-01-01T00:00:00 | 2020-01-10T00:00:00 | 307 | 47 | 5 | 19079 | 200701011CS425 | 2 |
| 6 | 19562869 | 2020-01-01T00:00:00 | 2020-01-06T00:00:00 | 307 | 1008 | 5 | 19079 | 200701014CS425 | 2 |
| 7 | 19563086 | 2020-01-01T00:00:00 | 2020-01-07T00:00:00 | 307 | 1688 | 5 | 18347 | 20070101729346 | 2 |
| 8 | 19563087 | 2020-01-01T00:00:00 | 2020-01-07T00:00:00 | 307 | 1688 | 5 | 2 | 20070101711001 | 2 |
| 9 | 19563088 | 2020-01-01T00:00:00 | 2020-01-07T00:00:00 | 307 | 1688 | 5 | 3 | 20070101711002 | 2 |
| 10 | 19563089 | 2020-01-01T00:00:00 | 2020-01-07T00:00:00 | 307 | 1688 | 5 | 4 | 20070101711003 | 2 |

- Under the **Destination** tab, select the following options:
 - Data store type: Workspace
 - Workspace data store type: Lakehouse
 - Lakehouse: Contoso (your managed Lakehouse created previously)
 - Root folder: Tables
 - Table name: Sales
 - Make sure that the settings will look like this:

General Source **Destination** Mapping Settings

Data store type ☒ Workspace ☐ External

Workspace data store type

Lakehouse Refresh Open + New

Root folder ☒ Tables ☐ Files

Table name * Refresh Preview data + New

> Advanced

- On the **Mapping** tab select Import schemas:

General Source Destination **Mapping** Settings

> Type conversion settings

Import schemas Preview source

+ New mapping Reset Delete

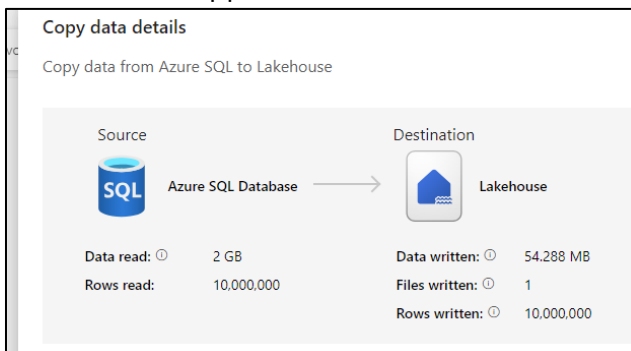
| Source | Type | Destination | Type | | |
|------------------------------------------------|---------------------------|------------------------|-----------|---|----|
| <input type="checkbox"/> SalesOrderNumber | nvarchar | → SalesOrderNumber | string | + | 🗑️ |
| <input type="checkbox"/> SalesOrderLineNumber | int | → SalesOrderLineNumber | integer | + | 🗑️ |
| <input type="checkbox"/> OnlineSalesKey | int | → OnlineSalesKey | integer | + | 🗑️ |
| <input type="checkbox"/> OrderDate | datetime | → OrderDate | timestamp | + | 🗑️ |
| <input type="checkbox"/> DeliveryDate | datetime | → DeliveryDate | timestamp | + | 🗑️ |
| <input type="checkbox"/> StoreKey | int | → StoreKey | integer | + | 🗑️ |
| <input checked="" type="checkbox"/> ProductKey | int | → ProductKey | integer | + | 🗑️ |
| <input type="checkbox"/> PromotionKey | int | → PromotionKey | integer | + | 🗑️ |
| <input type="checkbox"/> CustomerKey | int | → CustomerKey | integer | + | 🗑️ |
| <input type="checkbox"/> SalesQuantity | int | → SalesQuantity | integer | + | 🗑️ |
| <input type="checkbox"/> SalesAmount | money | → SalesAmount | string | + | 🗑️ |
| | Precision: 19 Scale: 4 | | | | |
| <input type="checkbox"/> ReturnQuantity | int | → ReturnQuantity | integer | + | 🗑️ |
| <input type="checkbox"/> ReturnAmount | money | → ReturnAmount | string | + | 🗑️ |
| | Precision: 19 Scale: 4 | | | | |
| <input type="checkbox"/> DiscountQuantity | int | → DiscountQuantity | integer | + | 🗑️ |
| <input type="checkbox"/> DiscountAmount | money | → DiscountAmount | string | + | 🗑️ |
| | Precision: 19 Scale: 4 | | | | |
| <input type="checkbox"/> TotalCost | money | → TotalCost | string | + | 🗑️ |
| | Precision: 19 Scale: 4 | | | | |

- On the **Settings** tab, no changes are required
- Validate and Save the Pipeline
- You can now **Run** your pipeline

- During the pipeline execution, you can monitor the progression by clicking on the Activity in the Output section

| Parameters | Variables | Settings | Output |
|----------------------------------------------------------------------|-----------|-----------------|--------|
| Pipeline run ID: 9f092682-6b46-4b36-9f44-a0f0f4b86187 [🔗] 🔄 ⓘ | | | |
| Showing 1 - 1 items | | | |
| Activity name | ↑↓ | Activity status | ↑↓ |
| Copy data from Azure SQL to Lakehouse | | 🔄 In progress | |

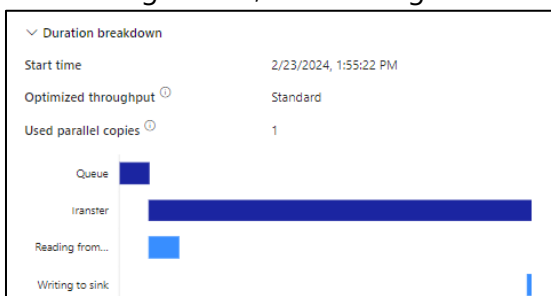
- Details should appear, and 10M rows should we be written.



- Note the time spent to execute the pipeline to transfer 10M of rows

| Activity runs | | | |
|-------------------------------------------------------|----|-----------------|----|
| Showing 1 - 1 items | | | |
| Activity name | ↑↓ | Activity status | ↑↓ |
| Copy data from Azure SQL to Lakehouse | | ✅ Succeeded | |
| Run start | ↑↓ | Duration | ↑↓ |
| 2/23/2024, 1:55:21 PM | | 1m 41s | |

- As demonstrated in the lab 3, you can use the Duration Breakdown to understand where the execution time is mostly consumed – depending on the status of the Azure SQL DB used during the lab, the Reading from source (Light Blue) might be bigger.



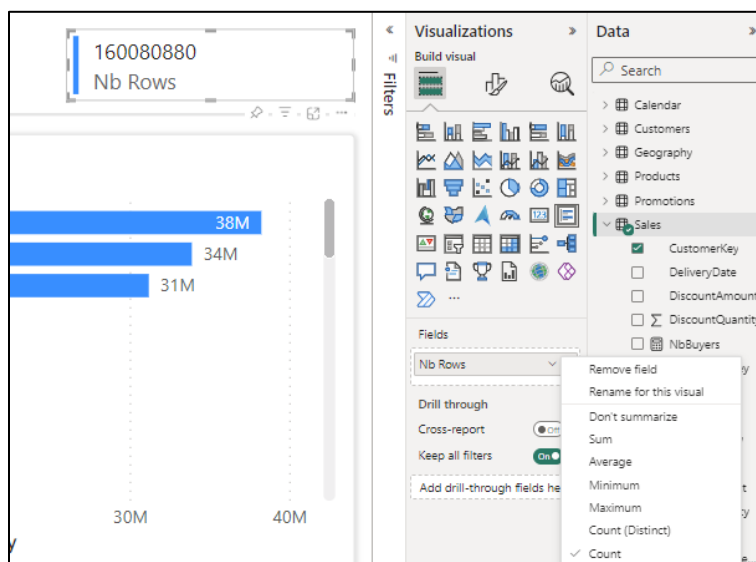
- Using the Lakehouse explorer, you can also see the last Parquet file written during the last pipeline execution. This parquet file weights 50MB and contains 10M of records.

| Contoso | Name | Date modified | Type | Size |
|----------------------|----------------------------------------------|----------------------|---------|----------|
| Tables | | | | |
| Calendar | _delta_log | 2/22/2024 7:31:05 PM | Folder | 15 Items |
| CustomerGroupMapping | 8de707fc-ee9a-428a-bb5d-b9925c478396.parquet | 2/23/2024 1:57:00 PM | PARQUET | 51 MB |
| CustomerGroups | 8d1cce7b-7c16-4209-b2b7-f04571301a84.parquet | 2/22/2024 9:49:44 PM | PARQUET | 10 MB |
| Customers | 5fd29b93-d807-4503-911a-4c2042009e77.parquet | 2/22/2024 9:49:43 PM | PARQUET | 11 MB |
| Geography | 74c835d9-6eb2-473f-b403-b21cd37bc30d.parquet | 2/22/2024 9:49:39 PM | PARQUET | 11 MB |
| ProductCategories | 18b71ae7-4e6f-48e1-87e8-2ffcd21fcd77.parquet | 2/22/2024 9:49:37 PM | PARQUET | 11 MB |
| Products | d87a190b-2100-48ba-9f34-dd629de6eef5.parquet | 2/22/2024 9:49:27 PM | PARQUET | 7 MB |
| ProductSubCategories | 1af472d1-2f0a-4bcb-bf39-8bd34cee3e2c.parquet | 2/22/2024 9:49:24 PM | PARQUET | 10 MB |
| Promotions | 59f64907-986e-4077-a9e7-a8722af0b762.parquet | 2/22/2024 9:49:23 PM | PARQUET | 11 MB |
| Sales | | | | |
| Stores | | | | |

- It is also recommended to use the available analytics tools in Fabric to see the new ingested in the Sales Tables
 - The SQL Analytics Endpoint:**

| | |
|-------------------------------------------------------|-----------------------------|
| Run | Save as view |
| 1 SELECT COUNT(*) AS TotalRecords FROM [dbo].[Sales]; | |
| Messages | Results |
| Open in Excel | Explore this data (preview) |
| 123 TotalRecords | |
| 1 | 160080880 |

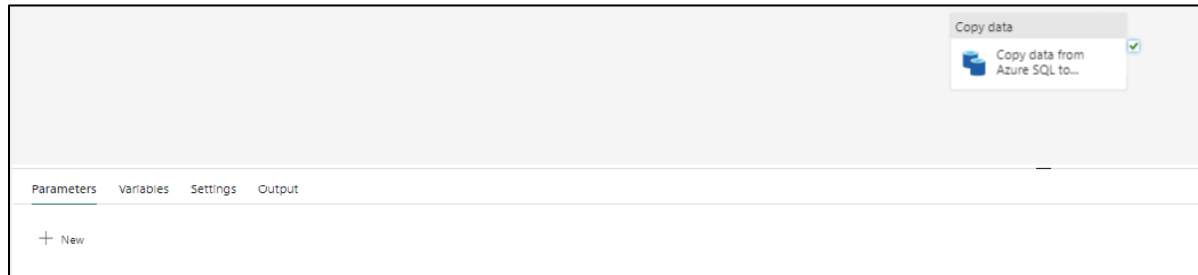
- The Power BI report**
 - You can add a simple visual card to count the number of rows in the Sales tables, and you don't have to refresh the report or the semantic model, thanks to the **Direct Lakehouse Mode**



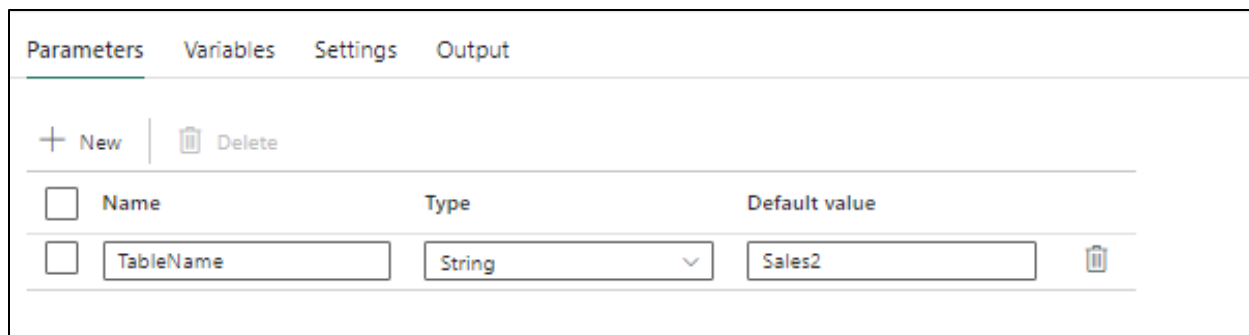
Task 2: Use Data Pipeline Parameters to change the source table

In this task, you will declare a Parameter to be used in the Pipeline. Then you will load data from the second table (50M of records)

- Open the Fabric Data Factory Pipeline, and go to the **Parameters** section



- Define the following parameter:
 - Name : **TableName**
 - Type : **String**
 - DefaultValue : **Sales2**



- Select the **Copy Data Activity** in the Pipeline, go to the **Source**, and enable the **Enter manually** check box
- Click on the **Add dynamic content** link to access to the Pipeline expression builder

The screenshot shows the 'Source' tab of a Copy Data Activity configuration. The 'Data store type' is set to 'External'. The 'Connection' is 'fabricdatafactorylab.database.windows'. The 'Connection type' is 'Azure SQL Database'. The 'Database' is 'fabriclabdb'. The 'Use query' is set to 'Table'. The 'Table' is 'dbo'. The 'Enter manually' checkbox is checked. The 'table name' field is highlighted with a green box, and the 'Add dynamic content [Alt+Shift+D]' link is visible next to it. The 'Preview data' link is also visible.

- From the Pipeline Expression Builder, drag and drop the TableName parameter, and the following expression should appear. Click on OK.

The screenshot shows the 'Pipeline expression builder' dialog. The text area contains the expression '@pipeline().parameters.TableName'. The 'Parameters' tab is selected. The search bar shows 'Search' and the 'TableName' parameter is listed below it.

- The expression should be now visible in the table name section.

The screenshot shows the 'Source' tab of the Copy Data Activity configuration, similar to the first screenshot. The 'table name' field now contains the expression '@pipeline().parameters.TableName' and is highlighted with a green box. The 'Enter manually' checkbox is still checked. The 'Preview data' link is visible.

- The expression should be now visible in the table name section.

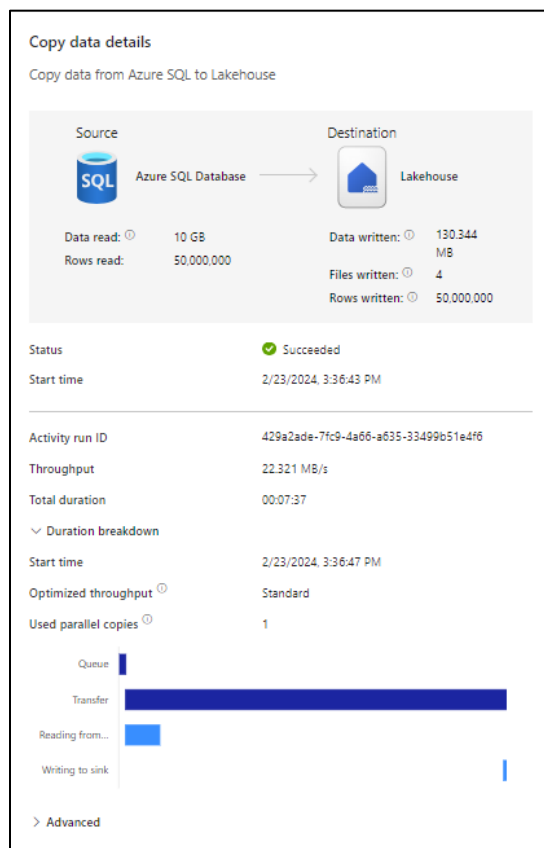
- Save the pipeline and click on Run.
- As the Pipeline now contains a parameter, you will have the choice to keep the default parameter value, or change it before the execution. Keep the default parameter value for this execution to get data from the table Sales2.

Pipeline run

Parameters

| Name | Type | Value |
|-----------|--------|-------------------------------------|
| TableName | string | <input type="text" value="Sales2"/> |

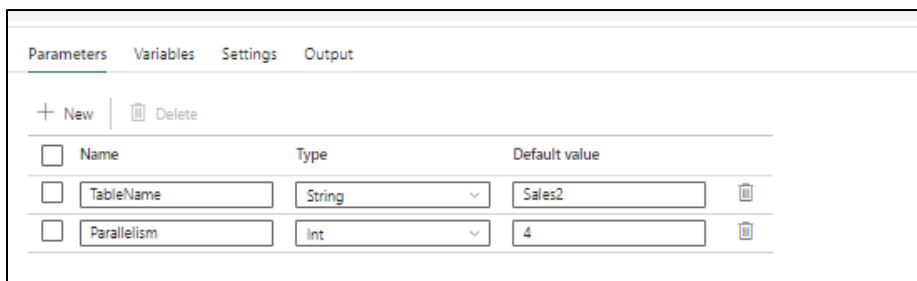
- At the end of the pipeline execution, you can observe that :
 - 50M of records were loaded
 - 4 files were written
 - The parallelism was 1 (in the next lab you will configure the parallelism using a parameter).



Task 3: Optimize Data Pipeline performance with parallelism and partitions

In this task, you will declare a second Parameter to be used as the level of parallelism in order to accelerate the data transfer and optimize performance by reducing the execution duration.

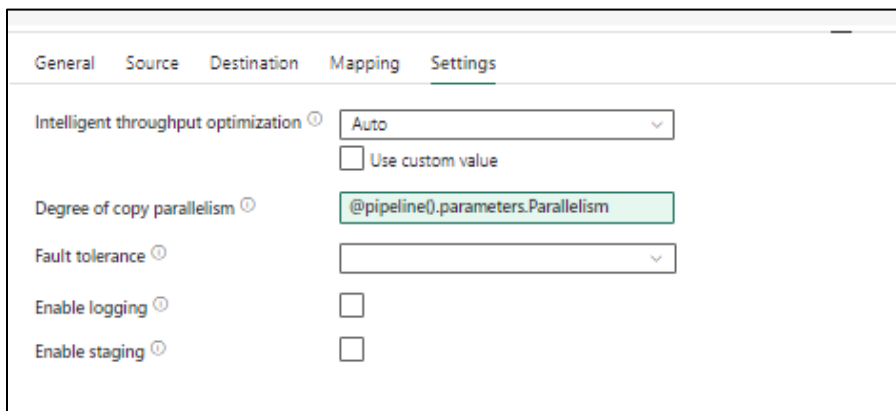
- Create a second parameter with the following settings:
 - Name: Parallelism
 - Type: Int
 - Default Value: 4



The screenshot shows the 'Parameters' tab in the Azure Data Factory interface. It contains a table with two parameters:

| Name | Type | Default value |
|-------------|--------|---------------|
| TableName | String | Sales2 |
| Parallelism | Int | 4 |

- Use this parameter to dynamically configure the Degree of copy parallelism (in the settings section of the Copy activity)



The screenshot shows the 'Settings' tab in the Copy activity configuration window. The 'Degree of copy parallelism' is set to '@pipeline().parameters.Parallelism'.

| Setting | Value |
|-------------------------------------|------------------------------------|
| Intelligent throughput optimization | Auto |
| Use custom value | <input type="checkbox"/> |
| Degree of copy parallelism | @pipeline().parameters.Parallelism |
| Fault tolerance | |
| Enable logging | <input type="checkbox"/> |
| Enable staging | <input type="checkbox"/> |

- On the Source section, enable the partitioning by selecting **Physical partitions** of table in the **Partition option** item.

The screenshot shows the 'Source' tab in the Azure Data Factory configuration interface. The 'Data store type' is set to 'External'. The 'Connection' is 'fabricdatafactorylab.database.windows.net'. The 'Connection type' is 'Azure SQL Database'. The 'Database' is 'fabriclabdb'. The 'Use query' is set to 'Table'. The 'Table' is 'dbo'. The 'Partition option' is highlighted in yellow and set to 'Physical partitions of table'. The 'Query timeout (minutes)' is 120. The 'Isolation level' is 'Select...'. The 'Additional columns' section has a '+ New' button.

- Save the pipeline and run it with the following configuration:
 - TableName : Sales3 (100M of rows)
 - Parallelism : 4

| Pipeline run | | |
|--------------|--------|--------|
| Parameters | | |
| Name | Type | Value |
| TableName | string | Sales3 |
| Parallelism | int | 4 |

- At the end of the pipeline execution you can observe :
 - A better throughput when transferring data
 - Good performance (only 2mn more when moving from 50M to 100M of rows)

