# **Data Engineering Batch 1**

# PROJECT 1

**Hybrid Cloud Data Movement** 

# Akilesh K

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## **Project Statement**

Implement a solution that involves moving data between on-premises data sources and Azure cloud using Azure Data Factory, and perform data processing tasks in Azure Databricks.

## **Project Overview**

The project entails orchestrating a data pipeline to facilitate seamless data movement between on-premises data sources and the Azure cloud environment. Initially, a SQL server will be set up, along with the creation of a database to house the source data. Azure Data Factory will play a pivotal role in this process, acting as the conduit for transferring data from the SQL database to Blob storage. This transfer will be facilitated by utilizing a self-hosted integration runtime, ensuring secure and efficient data transmission. Subsequently, Azure Databricks will be leveraged to execute essential data transformation tasks using PySpark, enabling the manipulation and refinement of the ingested data. Overall, the solution aims to streamline data integration and processing workflows, enhancing data accessibility and usability within the Azure ecosystem.

## **About the Project**

### **Database:**

The Northwind database is a sample database that was originally created by Microsoft and used as the basis for their tutorials in a variety of database products for decades. The Northwind database contains the sales data for a fictitious company called "Northwind Traders," which imports and exports specialty foods from around the world. The Northwind database is an excellent tutorial schema for a small-business ERP, with customers, orders, inventory, purchasing, suppliers, shipping, employees, and single-entry accounting.

The Northwind dataset includes sample data for the following.

Suppliers: Suppliers and vendors of Northwind

Customers: Customers who buy products from Northwind

Employees: Employee details of Northwind traders

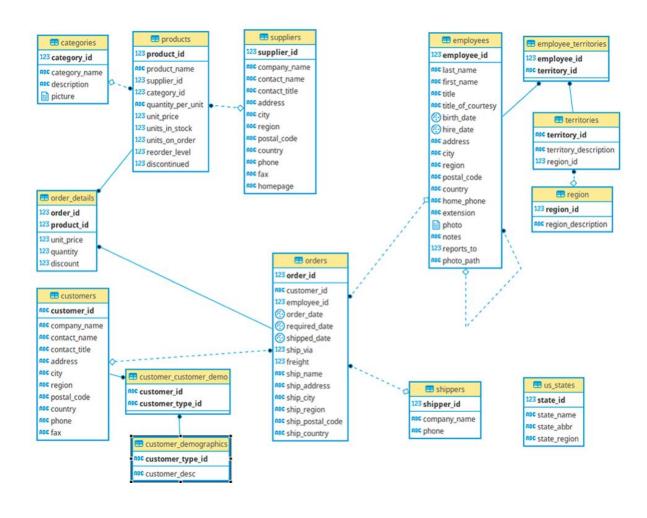
**Products: Product information** 

Shippers: The details of the shippers who ship the products from the traders to the end-

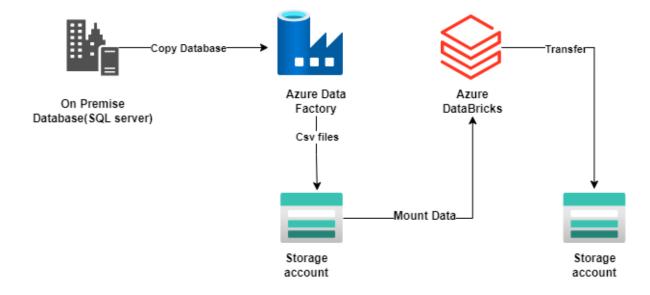
customers

Orders and Order\_Details: Sales Order transactions taking place between the customers & the company

### **Schema**



## **Architecture diagram**



The data flow in this architecture starts with SQL Server as the source, where data is extracted using ADF and loaded into Blob storage. Then, Azure Databricks performs the necessary data transformations using PySpark, leveraging the data stored in Blob storage. Overall, this architecture enables efficient data movement and processing between onpremises and cloud environments, facilitating analytics, reporting, and other data-driven tasks.

## **Azure Resources Used for this Project**

### • Azure Data Factory (ADF):

An ADF instance will be created to orchestrate and automate the data movement process.

Linked Services: Configured to establish connections to on-premises data sources and Azure Blob Storage.

Datasets: Defined to represent the data structures and schemas of the data sources.

Pipelines: Created to define the sequence of activities for copying data from on-premises to Azure Blob Storage.

## • Azure Blob Storage:

Blob storage containers will be used as the destination for storing data transferred from onpremises sources.

The stored data will be accessed by Azure Databricks for data processing tasks.

### • Azure Databricks:

A Databricks workspace will be provisioned to perform data processing tasks using PySpark.

Databricks Notebooks: Utilized to write and execute PySpark code for data transformation, cleaning, and analysis.

## **Project Requirements**

### • Setting up Data Sources:

Begin by identifying the on-premises data sources from which you want to extract data. These could be databases, files, or any other structured or unstructured data repositories.

Ensure that the necessary connectivity options are available for accessing these on-premises data sources securely from the Azure cloud environment.

### • Azure Data Factory Configuration:

Create an Azure Data Factory (ADF) instance in your Azure subscription.

Configure linked services in ADF to establish connections to both the on-premises data sources and the Azure cloud environment. This involves providing authentication credentials and connection details.

Define datasets in ADF to represent the data structures and schemas of the data sources. This includes specifying the location, format, and schema of the data residing in both the onpremises and Azure cloud environments.

Create pipelines in ADF to orchestrate the data movement process. Pipelines define the sequence of activities required to copy data from the on-premises sources to the Azure cloud.

### • Data Movement with Azure Data Factory:

Use ADF activities such as Copy Data to copy data from the on-premises sources to Azure Blob Storage. This activity handles the movement of data securely and efficiently, with options for parallelism, fault tolerance, and monitoring.

Configure the Copy Data activity with appropriate settings such as source and destination datasets, data integration runtime, scheduling options, and error handling mechanisms.

### • Data Processing with Azure Databricks:

Provision an Azure Databricks workspace in your Azure subscription.

Define and implement data processing tasks using PySpark within the Databricks environment. PySpark provides a powerful framework for distributed data processing, enabling tasks such as data cleaning, transformation, aggregation, and analysis.

Use Databricks notebooks to write and execute PySpark code interactively, leveraging the scalability and performance of the Databricks runtime.

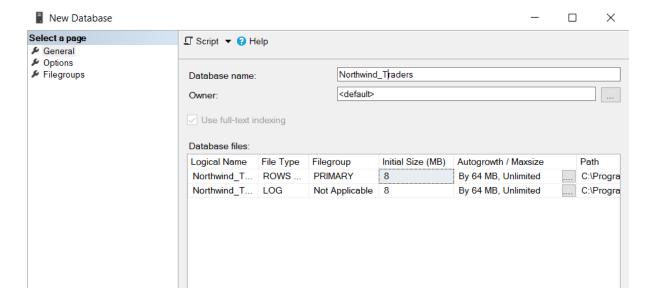
## Tasks performed:

- Set up SQL Server with a database, create schema, and add data.
- Create a new user with SQL Server Authentication as System Admin.
- Create Azure storage account and blob container.
- Set up an Azure Data Factory account and configure a data pipeline for data transfer.
- Install self-hosted integration runtime on-premises for SQL Server connection.
- Mount blob storage to Azure Databricks and read CSV file into dataframe.

- Perform data transformations including aggregation, sorting, and joining.
- Profile data to assess quality and characteristics.
- Visualize data using pie chart representation.
- Transfer manipulated dataframe to blob storage

# **Implementation**

• For On premise database, set it up in SQL server



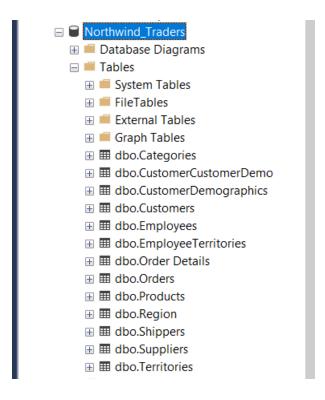
• Write query of the schema for the table and add data into it

```
INSERT "Order Details" VALUES(10622,2,19,20,0)
    INSERT "Order Details" VALUES(10622,68,12.5,18,0.2)
    INSERT "Order Details" VALUES(10623,14,23.25,21,0)
    INSERT "Order Details" VALUES(10623,19,9.2,15,0.1)
    INSERT "Order Details" VALUES(10623,21,10,25,0.1)
    INSERT "Order Details" VALUES(10623,24,4.5,3,0)
    INSERT "Order Details" VALUES(10623,35,18,30,0.1)
    INSERT "Order Details" VALUES(10624,28,45.6,10,0)
    INSERT "Order Details" VALUES(10624,29,123.79,6,0)
    INSERT "Order Details" VALUES(10624,44,19.45,10,0)
    INSERT "Order Details" VALUES(10625,14,23.25,3,0)
    INSERT "Order Details" VALUES(10625,42,14,5,0)
    INSERT "Order Details" VALUES(10625,60,34,10,0)
    INSERT "Order Details" VALUES(10626,53,32.8,12,0)
    INSERT "Order Details" VALUES(10626,60,34,20,0)
    INSERT "Order Details" VALUES(10626,71,21.5,20,0)
TNSERT "Ondon Dotails" WALLIES (18627 62 49 3 15 8)

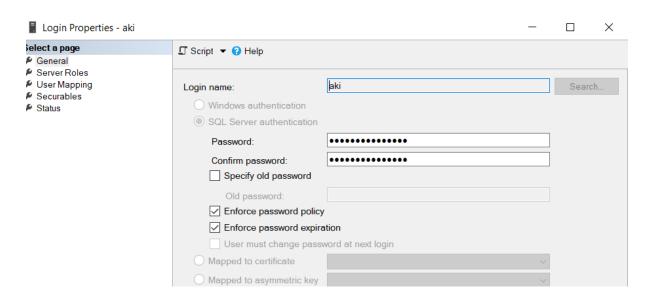
    Messages

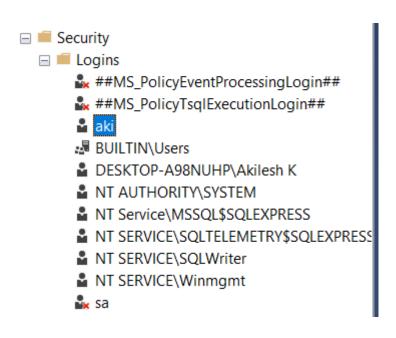
  Commands completed successfully.
  Completion time: 2024-02-25T15:20:44.9096926+05:30
```

• Tables are created in the database of the SQl server

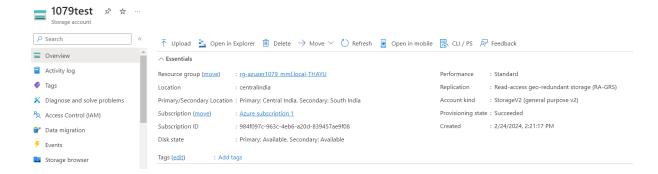


• Create a new user with SQL server Authentication with the role of System admin

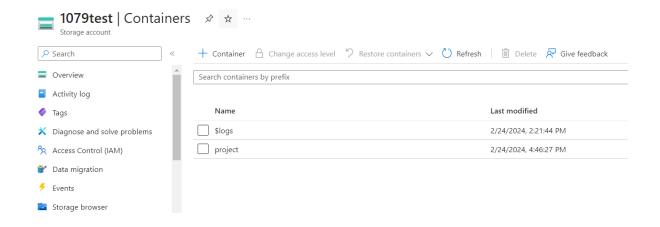




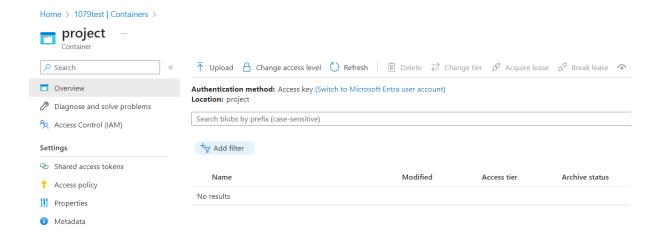
## • In azure ,create a storage account mentioning the location



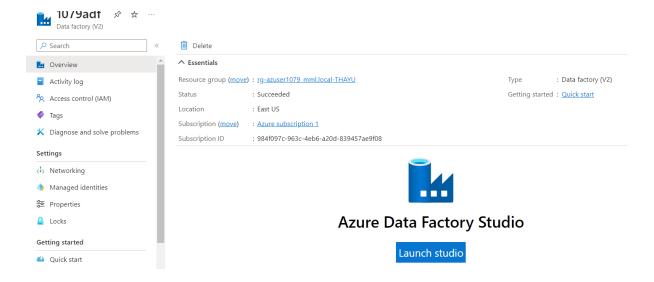
### • create a new blob container



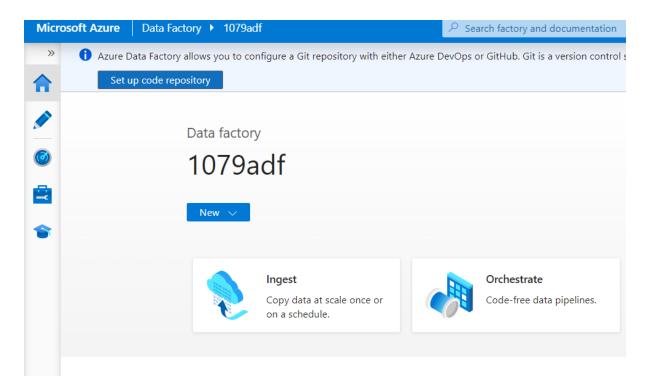
### • keep the container file empty



## • create a azure data factory account



## • select ingest data to create a new data pipeline



- In Properties set type as built in copy task.
- task schedule as run once now

### Copy Data tool Use Copy Data Tool to perform a one-time or scheduled data load from 90+ data sources. Properties Follow the wizard experience to specify your data loading settings, and let the Copy Data Tool generate the artifacts for you, including pipelines, **Properties** 2 Source Select copy data task type and configure task schedule Task type Built-in copy task Metadata-driven copy task You will get single pipeline to copy data You will get parameterized pipelines which (4) Settings from 90+ data source easily. can read metadata from an external store to load data at a large scale. (5) Review and finish You will get single pipeline to quickly copy objects from data source store to destination in a very intuitive manner. Task cadence or task schedule \* ■ Run once now Schedule Tumbling window

• In integration runtime setup, select self Hosted for running on premises activities

### Integration runtime setup

### Network environment:

Choose the network environment of the data source / destination or external compute to which the integration runtime will connect to for data flows, data movement or dispatch activities:



### Azure

Use this for running data flows, data movement, external and pipeline activities in a fully managed, serverless compute in Azure.

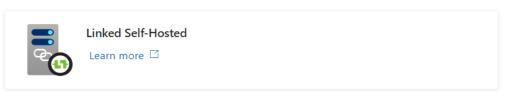


### Self-Hosted

Use this for running activities in an on-premises / private network View more  $\,^{\checkmark}$ 

### **External Resources:**

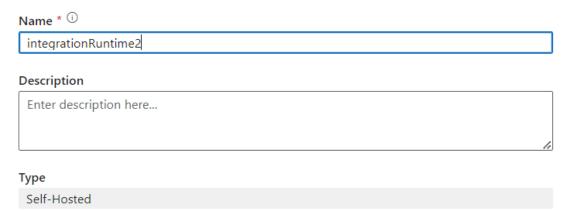
You can use an existing self-hosted integration runtime that exists in another resource. This way you can reuse your existing infrastructure where self-hosted integration runtime is setup.



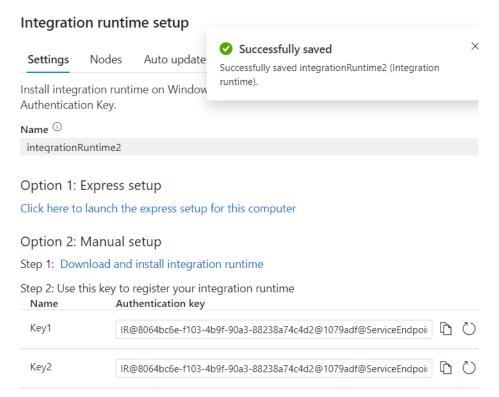
## • Give a name for the integration runtime setup

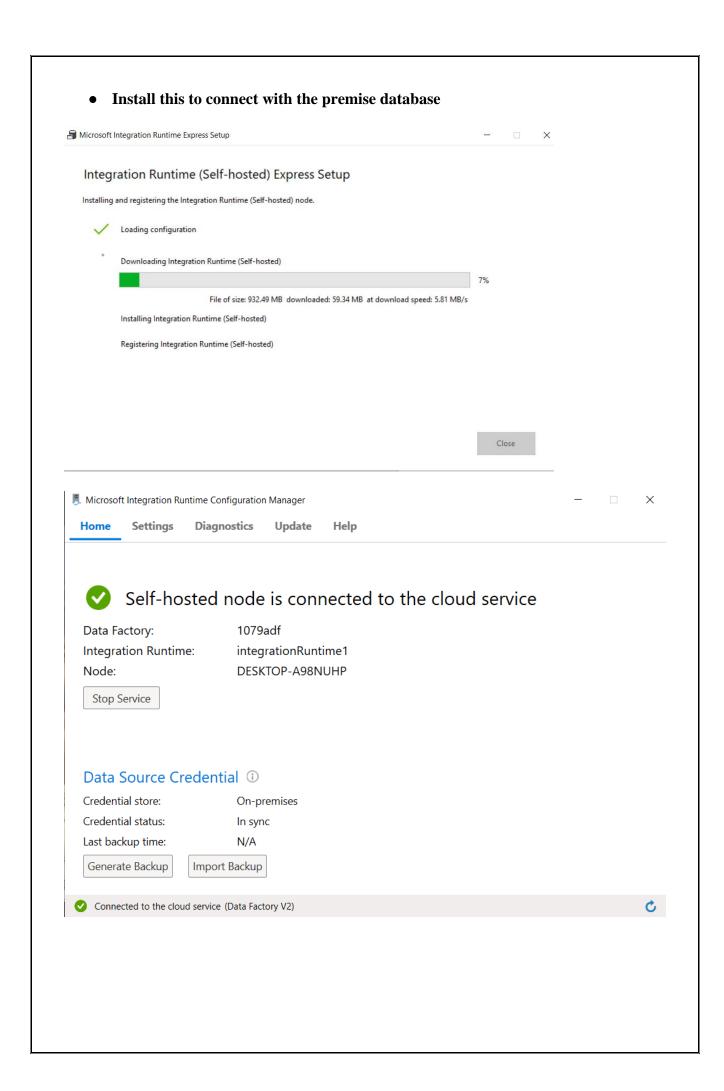
## Integration runtime setup

Private network support is realized by installing integration runtime to machines in the same on-premises network/VNET as the resource the integration runtime is connecting to. Follow below steps to register and install integration runtime on your self-hosted machines.

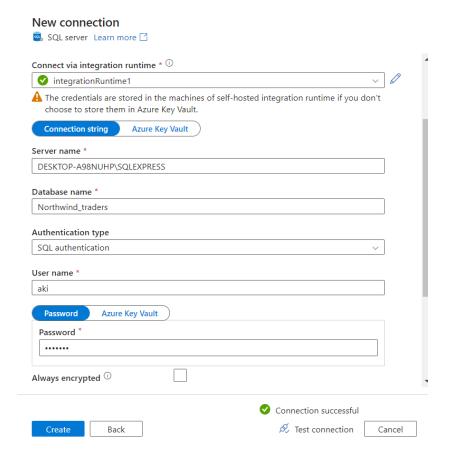


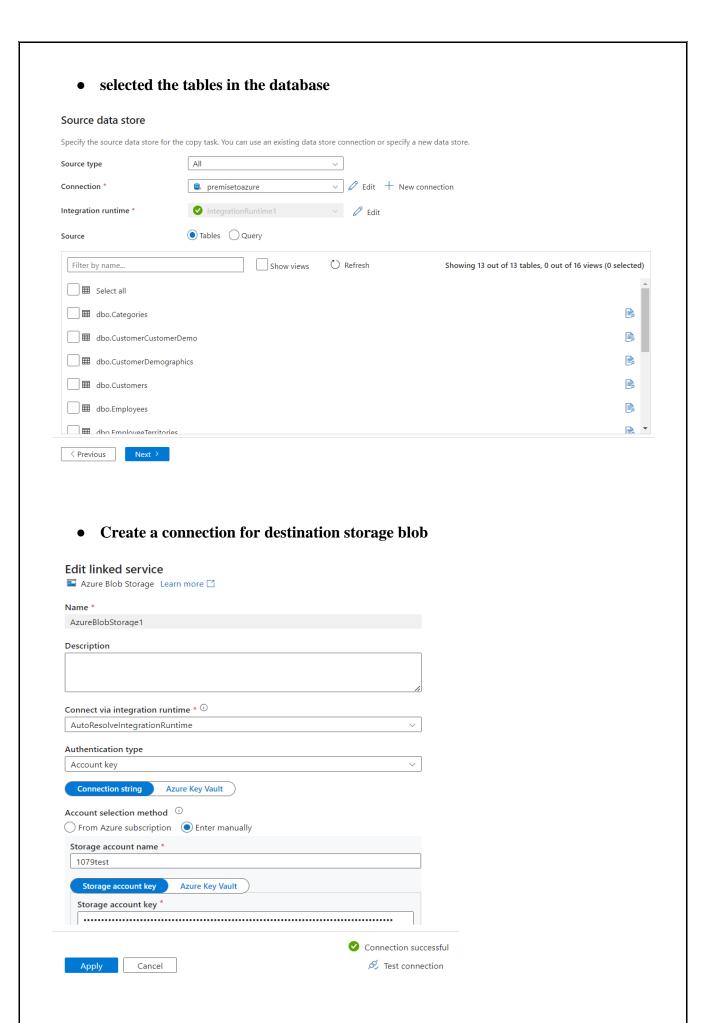
• In this setup, click to launch express setup for this computer.





- create the connection by mentioning server name and database name of the premise database
- select SQL authentication and enter user and password to access the database connection





# Select the container subfolder Destination data store Specify the destination data store for the copy task. You can use an existing data store connection or specify a new data store. Destination type ✓ Ø Edit + New connection AzureBlobStorage1 Connection \* Folder path \* If the identity you use to access the data store only has permission to subdirectory instead of the entire account, specify the path to browse. Browse project/ File name is defined by source table name > Advanced settings File name suffix Max concurrent connections $^{\scriptsize \bigcirc}$ Block size (MB) $^{\scriptsize \bigcirc}$ < Previous Enter the data pipeline name Settings Enter name and description for the copy data task, more options for data movement onpremisetoazure\_adf Task name \* Task description Data consistency verification <sup>(i)</sup> Fault tolerance (i) Enable logging ① Enable staging (i)

> Advanced

## • Summary of the deployment of from SQL server to azure blob storage

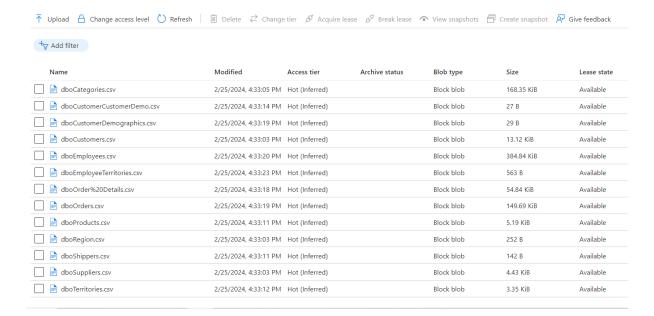


## Deployment complete

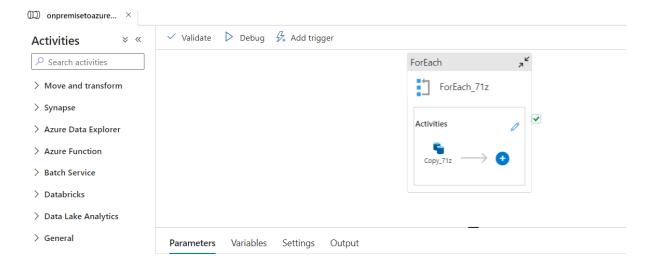
Deployment step	Status
Validating copy runtime environment	<b>⊘</b> Succeeded
> Creating datasets	✓ Succeeded
> Creating pipelines	✓ Succeeded
> Running pipelines	✓ Succeeded

Datasets and pipelines have been created. You can now monitor and edit the copy pipelines or click finish to close Copy Data Tool.

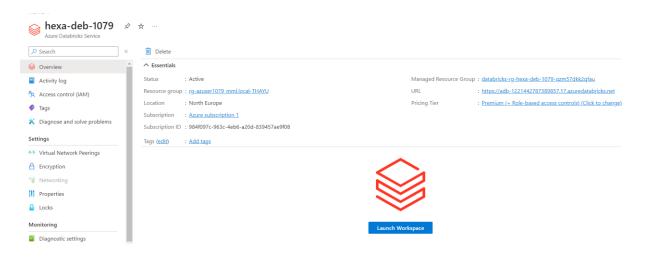
### • The table is converted to CSV file and stored in the container



## • Data pipeline of the activity



## Create a Azure databricks and launch it



## Mount blob storage to azure databricks

- Add source line of the blob storage
- enter the mount points to store in databricks
- Configure extra\_configs by mentioning the access key and pasting the key of the blob storage.

```
Python 

1 dbutils.fs.mount(source = 'wasbs://project@1079test.blob.core.windows.net',
2 | mount_point='/mnt/blobstrorage1',
3 | extra_configs = {'fs.azure.account.key.1079test.blob.core.windows.net':'ZOVqyFNZ+DoE5qfL8nYgTt409YgMem80Iitni0fsyJJVdrusQYWH/
V09vkfAldgEYBwlR4auSlbe+AStlLC3tQ=='})

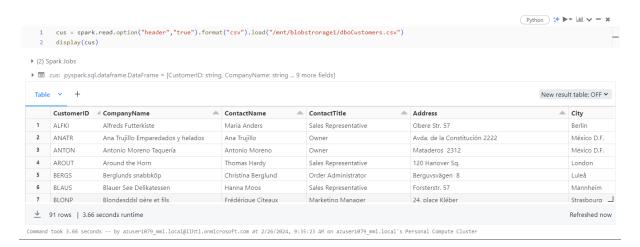
True

Command took 13.25 seconds -- by azuser1079_mml.local@iihtl.onmicrosoft.com at 2/26/2024, 9:28:10 AM on azuser1079_mml.local's Personal Compute Cluster
```

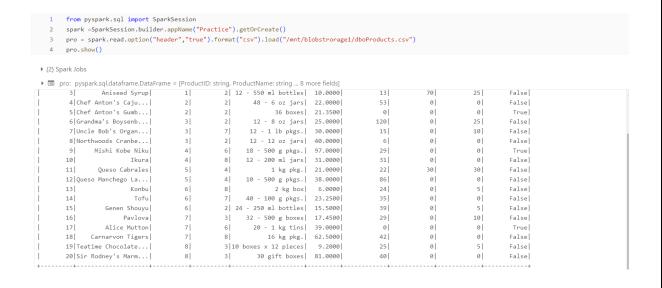
### List of the files mounted on Databricks

```
Pythor
dbutils.fs.ls('/mnt/blobstrorage1')
  [FileInfo(path='dbfs:/mnt/blobstrorage1/dboCategories.csv', name='dboCategories.csv', size=172394, modificationTime=1708858985000),
     FileInfo(path='dbfs:/mnt/blobstrorage1/dboCustomerCustomerDemo.csv', name='dboCustomerDemo.csv', size=27, modificationTime=1708858994000),
     FileInfo(path='dbfs:/mnt/blobstrorage1/dboCustomerDemographics.csv', name='dboCustomerDemographics.csv', size=29, modificationTime=1708858999000),
     File Info (path='dbfs:/mnt/blobstrorage1/dboCustomers.csv', name='dboCustomers.csv', size=13438, modificationTime=1708858983000), and the first of the first of
      FileInfo(path='dbfs:/mnt/blobstrorage1/dboEmployeeTerritories.csv', name='dboEmployeeTerritories.csv', size=563, modificationTime=1708859003000),
      FileInfo(path='dbfs:/mnt/blobstrorage1/dboEmployees.csv', name='dboEmployees.csv', size=394073, modificationTime=1708859000000),
      FileInfo(path='dbfs:/mnt/blobstrorage1/dboOrder%20Details.csv', name='dboOrder%20Details.csv', size=56160, modificationTime=1708858998000),
     FileInfo(path='dbfs:/mnt/blobstrorage1/dboOrders.csv', name='dboOrders.csv', size=153280, modificationTime=1708858999000)
     FileInfo(path='dbfs:/mnt/blobstrorage1/dboProducts.csv', name='dboProducts.csv', size=5315, modificationTime=1708858991000),
     FileInfo(path='dbfs:/mnt/blobstrorage1/dboRegion.csv', name='dboRegion.csv', size=252, modificationTime=1708858983000),
     File Info (path='dbfs:/mnt/blobstrorage1/dboShippers.csv', name='dboShippers.csv', size=142, modificationTime=1708858991000), and the file of the fi
      FileInfo(path='dbfs:/mnt/blobstrorage1/dboSuppliers.csv', name='dboSuppliers.csv', size=4537, modificationTime=1708858983000),
      FileInfo(path='dbfs:/mnt/blobstrorage1/dboTerritories.csv', name='dboTerritories.csv', size=3435, modificationTime=1708858992000)]
   Command took 0.97 seconds -- by azuser1079 mml.local@iihtl.onmicrosoft.com at 2/26/2024, 9:29:19 AM on azuser1079 mml.local" Personal Compute Cluster
```

### • Read the Csv file from the mount and convert into a dataframe



### Create a spark session and display the data frame



## Aggregation of the average of the unit price

```
1 pro.agg(({"UnitPrice":"avg"})).show()
  ▶ (2) Spark Jobs
 | avg(UnitPrice)|
 28.866363636363637
 Command took 2.81 seconds -- by azuser1079_mml.local@iihtl.onmicrosoft.com at 2/26/2024, 10:05:40 AM on azuser1079_mml.local's Personal Compute Cluster
```

## **Display Grouped by product name**

```
1 pro.groupBy("ProductName").count().show()
 ▶ (2) Spark Jobs
     Chocolade 1
          Filo Mix | 1|
     Longlife Tofu| 1|
|Wimmers gute Semm...| 1|
|Rhönbräu Klosterbier| 1|
         Tourtière 1
      Vegie-spread| 1|
    Mishi Kobe Niku
                    1
                    1
|Grandma's Boysenb...|
                    1
 |Laughing Lumberja...|
                    1
      Côte de Blaye
   Camembert Pierrot
       Pâté chinois
                     1
       Gula Malacca | 1|
| Boston Crab Meat| 1|
 | Queso Cabrales| 1|
            Konbu 1
only showing top 20 rows
```

Command took 3.52 seconds -- by azuser1079 mml.local@iihtl.onmicrosoft.com at 2/26/2024, 11:02:01 AM on azuser1079 mml

# • Drop rows with Null values

(1)	Spark Jobs							
	3  Aniseed Syrup	1	2 12 - 550 ml bottles	10.0000	13	70	25	False
	4 Chef Anton's Caju	2	2  48 - 6 oz jars	22.0000	53	0	0	False
	5 Chef Anton's Gumb	2	2   36 boxes	21.3500	0	0	0	True
	6 Grandma's Boysenb	3	2   12 - 8 oz jars	25.0000	120	0	25	False
	7 Uncle Bob's Organ	3	7  12 - 1 lb pkgs.	30.0000	15	0	10	False
	8 Northwoods Cranbe	3	2   12 - 12 oz jars	40.0000	6	0	0	False
	9  Mishi Kobe Niku	4	6   18 - 500 g pkgs.	97.0000	29	0	0	True
	10  Ikura	4	8  12 - 200 ml jars	31.0000	31	0	0	False
	11  Queso Cabrales	5	4 1 kg pkg.	21.0000	22	30	30	False
	12 Queso Manchego La	5	4  10 - 500 g pkgs.	38.0000	86	0	0	False
	13 Konbu	6	8 2 kg box	6.0000	24	0	5	False
	14  Tofu	6	7   40 - 100 g pkgs.	23.2500	35	0	0	False
	15 Genen Shouyu	6	2   24 - 250 ml bottles	15.5000	39	0	5	False
	16  Pavlova	7	3  32 - 500 g boxes	17.4500	29	0	10	False
	17 Alice Mutton	7	6  20 - 1 kg tins	39.0000	0	0	0	True
	18  Carnarvon Tigers	7	8   16 kg pkg.	62.5000	42	0	0	False
	19 Teatime Chocolate	8	3 10 boxes x 12 pieces	9.2000	25	0	5	False
	20 Sir Rodney's Marm	8	3   30 gift boxes	81.0000	40	0	0	False

Display data in Ascending order

11)	Spark Jobs							
')	74 Longlife Tofu	4	7 5 kg pkg.	10.0000	4	20	5	Fal
	46 Spegesild	21	8 4 - 450 g glasses		95	0	0	Fa
	31 Gorgonzola Telino	14	4 12 - 100 g pkgs	12.5000	0	70	20	Fa
	68   Scottish Longbreads	8	3 10 boxes x 8 pieces	12.5000	6	10	15	Fa:
	48 Chocolade	22	3 10 pkgs.	12.7500	15	70	25	Fa
	29 Thüringer Rostbra	12	6 50 bags x 30 sausgs.	123.7900	0	0	0	T
	77 Original Frankfur	12	2 12 boxes	13.0000	32	0	15	Fa
	58 Escargots de Bour	27	8 24 pieces	13.2500	62	0	20	Fa
	42 Singaporean Hokki	20	5  32 - 1 kg pkgs.	14.0000	26	0	0	Т
	25 NuNuCa Nuß-Nougat	11	3  20 - 450 g glasses	14.0000	76	0	30	Fa
	34  Sasquatch Ale	16	1   24 - 12 oz bottles	14.0000	111	0	15	Fa
	67 Laughing Lumberja	16	1   24 - 12 oz bottles	14.0000	52	0	10	Fa
	70 Outback Lager	7	1  24 - 355 ml bottles	15.0000	15	10	30	Fa
	73  Röd Kaviar	17	8  24 - 150 g jars	15.0000	101	0	5	Fa
	15 Genen Shouyu	6	2  24 - 250 ml bottles	15.5000	39	0	5	Fa
	50  Valkoinen suklaa	23	3   12 - 100 g bars	16.2500	65	0	30	Fa
	66 Louisiana Hot Spi	2	2  24 - 8 oz jars	17.0000	4	100	20	Fa
	16 Pavlova	7	3 32 - 500 g boxes	17.4500	29	0	10	Fa

Command took 1.96 seconds -- by azuser1079\_mml.local@iihtl.onmicrosoft.com at 2/26/2024, 11:02:17 AM on azuser1079\_mml.local's Personal Compute Cluster

### • Sort the data based on unit in stock

(1)	Spark Jobs							
	29 Thüringer Rostbra	12	6 50 bags x 30 sausgs.	123.7900	0	0	0	True
	31  Gorgonzola Telino	14	4   12 - 100 g pkgs	12.5000	0	70	20	False
	53  Perth Pasties	24	6 48 pieces	32.8000	0	0	0	True
	30 Nord-Ost Matjeshe	13	8  10 - 200 g glasses	25.8900	10	0	15	False
	49  Maxilaku	23	3 24 - 50 g pkgs.	20.0000	10	60	15	False
	73  Röd Kaviar	17	8  24 - 150 g jars	15.0000	101	0	5	False
	22  Gustaf's Knäckebröd	9	5  24 - 500 g pkgs.	21.0000	104	0	25	False
	37  Gravad lax	17	8  12 - 500 g pkgs.	26.0000	11	50	25	False
	34  Sasquatch Ale	16	1   24 - 12 oz bottles	14.0000	111	0	15	False
	33  Geitost	15	4  500 g	2.5000	112	0	20	False
	36  Inlagd Sill	17	8  24 - 250 g jars	19.0000	112	0	20	False
	61  Sirop d'érable	29	2  24 - 500 ml bottles	28.5000	113	0	25	False
	55  Pâté chinois	25	6 24 boxes x 2 pies	24.0000	115	0	20	False
	6 Grandma's Boysenb	3	2   12 - 8 oz jars	25.0000	120	0	25	False
	40  Boston Crab Meat	19	8 24 - 4 oz tins	18.4000	123	0	30	False
	75 Rhönbräu Klosterbier	12	1   24 - 0.5 l bottles	7.7500	125	0	25	False
	3  Aniseed Syrup	1	2  12 - 550 ml bottles	10.0000	13	70	25	False
	72 Mozzarella di Gio	14	4 24 - 200 g pkgs.	34.8000	14	0	0	False

Command took 1.69 seconds -- by azuser1079\_mml.local@iihtl.onmicrosoft.com at 2/26/2024, 11:02:24 AM on azuser1079\_mml.local's Personal Compute Cluster

## • Inner join of customer and order table based on customer ID



# • Display the table with a changed column name

31	Aniseed Syrup	1	2   1	12 - 550 ml bottles	10.0000	13	70	25	False
4   Ch	ef Anton's Caju	2	2	48 - 6 oz jars	22.0000	53	0	0	False
	ef Anton's Gumb	2	2	36 boxes	21.3500	0	0	0	True
6 Gr	andma's Boysenb	3	2	12 - 8 oz jars	25.0000	120	øİ	25	False
	cle Bob's Organ	3	7	12 - 1 lb pkgs.	30.0000	15	0	10	False
8 No	rthwoods Cranbe	3	2	12 - 12 oz jars	40.0000	6	0	0	False
9	Mishi Kobe Niku	4	6	18 - 500 g pkgs.	97.0000	29	0	0	True
10	Ikura	4	8	12 - 200 ml jars	31.0000	31	0	0	False
11	Queso Cabrales	5	4	1 kg pkg.	21.0000	22	30	30	False
12 Qu	eso Manchego La	5	4	10 - 500 g pkgs.	38.0000	86	0	0	False
13	Konbu	6	8	2 kg box	6.0000	24	0	5	False
14	Tofu	6	7	40 - 100 g pkgs.	23.2500	35	0	0	False
15	Genen Shouyu	6	2 2	24 - 250 ml bottles	15.5000	39	0	5	False
16	Pavlova	7	3	32 - 500 g boxes	17.4500	29	0	10	False
17	Alice Mutton	7	6	20 - 1 kg tins	39.0000	0	0	0	True
18	Carnarvon Tigers	7	8	16 kg pkg.	62.5000	42	0	0	False
19 Te	atime Chocolate	8	3 10	boxes x 12 pieces	9.2000	25	0	5	False
20 Si	r Rodney's Marm	8	3	30 gift boxes	81.0000	40	0	0	False

# • Display dataframe of order table

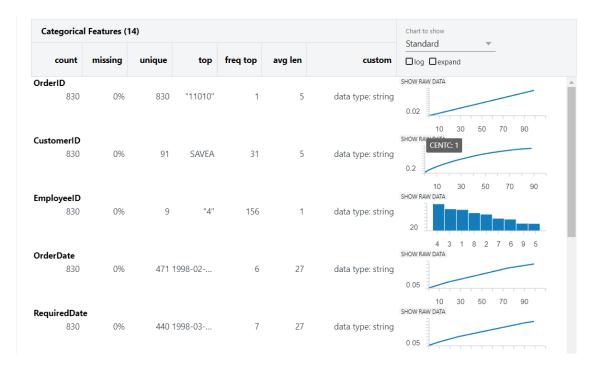
- order = spark.read.option("header","true").format("csv").load("/mnt/blobstrorage1/dboOrders.csv")
  display(order)
- ▶ (2) Spark Jobs
- Im order: pyspark.sql.dataframe.DataFrame = [OrderID: string, CustomerID: string ... 12 more fields]

	OrderID 📥	CustomerID 📤	EmployeeID 📤	OrderDate	_	RequiredDate	ShippedDate	$\triangle$	ShipVia	_	Freight
1	10248	VINET	5	1996-07-04 00:00:00.0000000		1996-08-01 00:00:00.0000000	1996-07-16 00:00:00.0000000		3		32.3800
2	10249	TOMSP	6	1996-07-05 00:00:00.0000000		1996-08-16 00:00:00.0000000	1996-07-10 00:00:00.0000000		1		11.6100
3	10250	HANAR	4	1996-07-08 00:00:00.0000000		1996-08-05 00:00:00.0000000	1996-07-12 00:00:00.0000000		2		65.8300
4	10251	VICTE	3	1996-07-08 00:00:00.0000000		1996-08-05 00:00:00.0000000	1996-07-15 00:00:00.0000000		1		41.3400
5	10252	SUPRD	4	1996-07-09 00:00:00.0000000		1996-08-06 00:00:00.0000000	1996-07-11 00:00:00.0000000		2		51.3000
6	10253	HANAR	3	1996-07-10 00:00:00.0000000		1996-07-24 00:00:00.0000000	1996-07-16 00:00:00.0000000		2		58.1700
7	10254	CHOPS	5	1996-07-11 00:00:00.0000000		1996-08-08 00:00:00.0000000	1996-07-23 00:00:00.0000000		2		22.9800

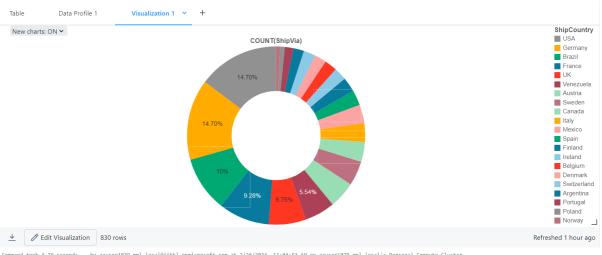
Command took 4.76 seconds -- by azuser1079\_mml.local@iihtl.onmicrosoft.com at 2/26/2024, 11:04:53 AM on azuser1079\_mml.local's Personal Compute Cluster

## Display the profiling of the data in the table

Data profiling is the process of collecting statistics and summaries of data to assess its quality and other characteristics



Visualization of the country data represented in the form of the pie chart.

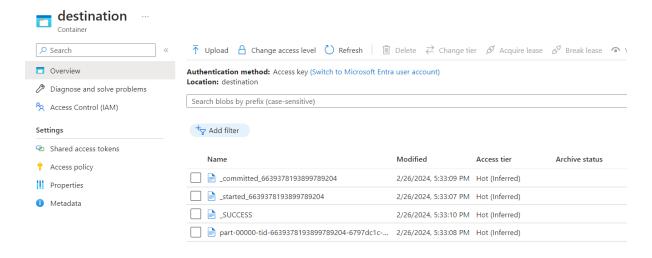


d took 4.76 seconds -- by azuser1079 mml.local@iihtl.onmicrosoft.com at 2/26/2024, 11:04:53 AM on azuser1079 mml.local's Personal Compute Cluster

• Transfer of the dataframe which we manipulated to a blob storage thorough access key.

```
purple of the state of the
```

• The Dataframe gets committed to a new storage



## **Conclusion**

In conclusion, the project successfully implemented a robust data pipeline leveraging various Azure services. By setting up a SQL server with a database, data was securely stored onpremises. Azure Data Factory was then employed to seamlessly transfer data from the SQL database to Blob storage in the Azure cloud, ensuring scalability and reliability. Additionally, Azure Databricks played a crucial role in performing data transformations and enabling visualization using PySpark, empowering data analysts and engineers to derive insights and make informed decisions. Overall, this project demonstrates the power of Azure services in building end-to-end data solutions, from ingestion to transformation and visualization, paving the way for efficient data-driven workflows and analytics.