

# Coding Challenge – 5

## Azure Synapse ETL Pipeline

Submitted By-  
Subrat Shukla, DE-1

1) Build an ETL pipeline with azure synapse with dataflow running on it.

Steps:

Create an Azure Synapse Workspace:

Microsoft Azure

Home > Azure Synapse Analytics >

### Create Synapse workspace

**Workspace details**

Name your workspace, select a location, and choose a primary Data Lake Storage Gen2 file system to serve as the default location for logs and job output.

Workspace name \*

Region \*

Select Data Lake Storage Gen2 \* ☒ From subscription ☐ Manually via URL

Account name \*   
[Create new](#)

File system name \*   
[Create new](#)

**Additional configuration is required.** After you create your workspace, perform these tasks:

- Assign other users to the **Contributor** role on workspace
- Assign other users the appropriate [Synapse RBAC roles](#) using Synapse Studio

Contact an **Owner** of the storage account, and ask them to perform the following tasks:

[Review + create](#) [< Previous](#) [Next: Security >](#)

## Microsoft.Azure.SynapseAnalytics-20241219163133 | Overview

Deployment

Search

Delete Cancel Redeploy Download Refresh

**Overview**

- Inputs
- Outputs
- Template

**✓ Your deployment is complete**

Deployment name : Microsoft.Azure.SynapseAnalytics-20241219163133 Start time : 12/19/2024, 4:32:38 PM  
Subscription : MML Learners Correlation ID : dc21ae44-04bc-4883-b542-401872ba21f4  
Resource group : rg-azuser2356\_mml.local-eylrk

> Deployment details

✓ Next steps

[Go to resource group](#)

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Synapse Analytics workspace

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New

**Ingest**

Perform a one-time or scheduled data load.

**Explore and analyze**

Learn how to get insights from your data.

**Visualize**

Build interactive reports with Power BI capabilities.

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## Create data flow activity:

The screenshot shows the Microsoft Azure Synapse Analytics interface. The top navigation bar includes 'Synapse live', 'Validate all', and 'Publish all'. The left sidebar shows the 'Activities' list with categories like 'Synapse', 'Move and transform', 'Azure Data Explorer', 'Azure Function', 'Batch Service', 'Databricks', 'Data Lake Analytics', 'General', 'HDInsight', 'Iteration & conditionals', and 'Machine Learning'. The 'Data flow' activity is selected, and its configuration panel is open. The 'Settings' tab is active, showing the following options:

- Data flow:** Dataflow1 (with 'Open' and '+ New' links)
- Run on (Azure IR):** AutoResolveIntegrationRuntime (with a checkmark)
- Compute size:** Small
- Advanced:** (collapsed)
- Logging level:** Verbose (selected), Basic, None
- Sink properties:** (collapsed)

The right sidebar shows the 'Properties' panel with the 'General' tab selected. It contains fields for 'Name' (Pipeline 1), 'Description', and 'Annotations' (with a '+ New' button).

## Create a source and sink and Configure the source:

The screenshot shows the Microsoft Azure Synapse Analytics interface. The top navigation bar includes 'Synapse live', 'Validate all', and 'Publish all'. The left sidebar shows the 'Activities' list with categories like 'Synapse', 'Move and transform', 'Azure Data Explorer', 'Azure Function', 'Batch Service', 'Databricks', 'Data Lake Analytics', 'General', 'HDInsight', 'Iteration & conditionals', and 'Machine Learning'. The 'Data flow' activity is selected, and its configuration panel is open. The 'Settings' tab is active, showing the following options:

- Data flow:** Dataflow1 (with 'Open' and '+ New' links)
- Run on (Azure IR):** AutoResolveIntegrationRuntime (with a checkmark)
- Compute size:** Small
- Advanced:** (collapsed)
- Logging level:** Verbose (selected), Basic, None
- Sink properties:** (collapsed)

The right sidebar shows the 'Properties' panel with the 'General' tab selected. It contains fields for 'Name' (Dataflow1), 'Description', and 'Annotations' (with a '+ New' button).

The main canvas shows a data flow diagram with a source activity labeled 'source1' (Columns: 24 total) connected to a sink activity labeled 'sink1' (Add sink dataset). Below the diagram is an 'Add Source' button.

The 'Source settings' tab is active, showing the following options:

- Dataset:** DelimitedText3 (with 'Test connection', 'Open', and '+ New' links)
- Options:**
  - ☒ Allow schema drift
  - ☐ Infer drifted column types
  - ☐ Validate schema
- Skip line count:** (empty field)
- Sampling:** Enable, Disable (selected)

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Synapse live | Validate all | Publish all

Pipeline 1 | Dataflow1

Validate | Data flow debug

source1  
Columns: 24 total

sink1  
Add sink dataset

Add Source

Source settings | Source options | Projection | Optimize | Inspect | Data preview

Dataset \* | DelimitedText2 | Test connection | Open | New

Options

- ☒ Allow schema drift
- ☐ Infer drifted column types
- ☐ Validate schema

Skip line count

Sampling \* | Enable | Disable

### New integration dataset

In pipeline activities and data flows, reference a dataset to specify the location and structure of your data within a data store. [Learn more](#)

Select a data store

Search

All | Azure | Database | File | Generic protocol

- Amazon S3
- Azure Blob Storage
- Azure Cosmos DB for NoSQL
- Azure Data Explorer (Kusto)
- Azure Data Lake Storage Gen2
- Azure Database for MySQL

Continue | Cancel

32°C Mostly sunny | Search | 16:58 19-12-2024

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Synapse live | Validate all | Publish all

Pipeline 1 | Dataflow1

Validate | Data flow debug

source1  
Columns: 24 total

sink1  
Add sink dataset

Add Source

Source settings | Source options | Projection | Optimize | Inspect | Data preview

Dataset \* | DelimitedText2 | Test connection | Open | New

Options

- ☒ Allow schema drift
- ☐ Infer drifted column types
- ☐ Validate schema

Skip line count

Sampling \* | Enable | Disable

### Select format

Choose the format type of your data

- Avro
- DelimitedText
- Excel
- JSON
- ORC
- Parquet
- XML
- Binary

Continue | Back | Cancel

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Validate all | Publish all

Validate | Data flow debug

source1  
Columns: 24 total

sink1  
Add sink dataset

Add Source

Source settings | Source options | Projection | Optimize | Inspect | Data preview

Dataset \*  
DelimitedText2

Options  
☒ Allow schema drift  
☐ Infer drifted column types  
☐ Validate schema

Skip line count

Sampling \*  
☐ Enable ☒ Disable

Browse

Select a file or folder.

Root folder > source-container

Global\_Superstore2.csv

Showing 1 item

OK Cancel

32°C Mostly sunny

Search

16:58 19-12-2024

## Configure the sink:

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Validate all | Publish all

Validate | Data flow debug

source1  
Import data from DelimitedText3

sink1  
Columns: 24 total

Add Source

Sink | Settings | Errors | Mapping | Optimize | Inspect | Data preview

Incoming stream \*  
source1

Sink type \*  
Integration dataset | Inline | Workspace DB | Cache

Dataset \*  
Select...

Options  
☒ Allow schema drift  
☐ Validate schema

Set properties

Name  
DelimitedText4

Linked service \*  
AzureDataLakeStorage1

Connect via integration runtime \*  
☒ AutoResolveIntegrationRuntime

File path  
my-container / Directory / File name

First row as header  
☒

Import schema  
☒ From connection/store ☐ From sample file ☐ None

> Advanced

OK Back Cancel

Breaking news  
Rahul Gandhi ac...

Search

16:59 19-12-2024



# Validate and Debug the pipeline:

The screenshot shows the Microsoft Azure Synapse Analytics interface. The top navigation bar includes 'Synapse live', 'Validate all', and 'Publish all'. The left sidebar lists various activities under 'Move and transform', including 'Copy data' and 'Data flow'. The main canvas displays a 'Data flow' activity named 'Data flow1'. The 'Output' tab is active, showing the 'Pipeline run ID' as 'e1b5266a-e90a-4b0f-b628-638fe5902279' and the 'Pipeline status' as 'In progress'. A message indicates that the data flow activity will start as soon as the debug session is ready. The bottom status bar shows the current time as 17:01 on 19-12-2024.

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**Activities**

- Synapse
- Move and transform
  - Copy data
  - Data flow
- Azure Data Explorer
- Azure Function
- Batch Service
- Databricks
- Data Lake Analytics
- General
- HDInsight
- Iteration & conditionals
- Machine Learning

**Data flow**

**Data flow1**

**Pipeline run ID:** e1b5266a-e90a-4b0f-b628-638fe5902279 **Pipeline status:** In progress

Data flow activity for this debug run will start as soon as the data flow debug session is ready.

**Output**

Activity name	Activity status	Activity type	Run start	Duration	Integration runtime
Data flow1	In progress	Data flow	12/19/2024, 5:01:16 PM	Less than 1s	

The screenshot shows the Microsoft Azure Synapse Analytics interface. The top navigation bar includes 'Synapse live', 'Validate all', and 'Publish all'. The left sidebar lists various activities under 'Move and transform', including 'Copy data' and 'Data flow'. The main canvas displays a 'Data flow' activity named 'Data flow1'. The 'Output' tab is active, showing the 'Pipeline run ID' as 'e1b5266a-e90a-4b0f-b628-638fe5902279' and the 'Pipeline status' as 'Succeeded'. A message indicates that the data flow activity will start as soon as the debug session is ready. The bottom status bar shows the current time as 17:05 on 19-12-2024.

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**Activities**

- Synapse
- Move and transform
  - Copy data
  - Data flow
- Azure Data Explorer
- Azure Function
- Batch Service
- Databricks
- Data Lake Analytics
- General
- HDInsight
- Iteration & conditionals
- Machine Learning

**Data flow**

**Data flow1**

**Pipeline run ID:** e1b5266a-e90a-4b0f-b628-638fe5902279 **Pipeline status:** Succeeded

**Output**

Activity name	Activity status	Activity type	Run start	Duration	Integration runtime
Data flow1	Succeeded	Data flow	12/19/2024, 5:01:16 PM	3m 59s	AutoResolveIntegrat

# Extract the data from the delta lake storage and do the transformations and loading:

04:44 PM (15s)

1

Python

```
%python
# Fetching the csv file from the blob storage
storage_account_name = "codingchallengedlacc"
container_name = "my-container"
storage_account_key = "hMW9uKwa011epiUgz0tGQd9P9UKbeVq29mCY5wCz2GIPpsMELvdjGaAcJMmDjsiX8RHCznioSigW+AStcpQIGQ=="

# Unmount the directory if it is already mounted
if any(mount.mountPoint == "/mnt/superstore" for mount in dbutils.fs.mounts()):
    dbutils.fs.unmount("/mnt/superstore")

# Mount dl Storage
dbutils.fs.mount(
    source=f"wasbs://{container_name}@{storage_account_name}.blob.core.windows.net",
    mount_point="/mnt/superstore",
    extra_configs={
        f"fs.azure.account.key.{storage_account_name}.blob.core.windows.net": storage_account_key
    }
)
```

True

## Verifying the mount:

04:44 PM (6s)

2

```
#verifying the mount
display(dbutils.fs.ls("/mnt/superstore"))
```

(2) Spark Jobs

Table

	Path	Name	Size	ModificationTime
1	dbfs:/mnt/superstore/Global_Superstore2.c...	Global_Superstore2.c...	12089916	1734606845000

1 row | 5.79 seconds runtime

Refreshed 41 minutes ago

# Load the dataset:

04:45 PM (6s)

3

Python

```
from pyspark.sql import SparkSession
spark = SparkSession.builder.appName("SuperstoreETL").getOrCreate()

# Load the dataset
file_path = "/mnt/superstore/Global_Superstore2.csv"
df = spark.read.csv(file_path, header=True, inferSchema=True)
df.show(5)
```

(3) Spark Jobs

df: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 22 more fields]

32298	CA-2012-124891	2012-07-31	2012-07-31	Same Day	RH-19495	Rick Hansen	Consumer	New York City	New York	United States		
10024	US	East	TEC-AC-10003033	Technology	Accessories	Plantronics CS510...	2309.65	7	0	762.1845	933.57	Critic
26341	IN-2013-77878	2013-02-05	2013-02-07	Second Class	JR-16210	Justin Ritter	Corporate	Wollongong	New South Wales	Australia		
NULL	APAC	Oceania	FUR-CH-10003950	Furniture	Chairs	Novimex Executive...	3709.395	9	0.1	-288.765	923.63	Critic
25330	IN-2013-71249	2013-10-17	2013-10-18	First Class	CR-12730	Craig Reiter	Consumer	Brisbane	Queensland	Australia		
NULL	APAC	Oceania	TEC-PH-10004664	Technology	Phones	Nokia Smart Phone...	5175.171	9	0.1	919.971	915.49	Medi
13524	ES-2013-1579342	2013-01-28	2013-01-30	First Class	KM-16375	Katherine Murray	Home Office	Berlin	Berlin	Germany		
NULL	EU	Central	TEC-PH-10004583	Technology	Phones	Motorola Smart Ph...	2892.51	5	0.1	-96.54	910.16	Medi
47221	SG-2013-4320	2013-11-05	2013-11-06	Same Day	RH-9495	Rick Hansen	Consumer	Dakar	Dakar	Senegal		
NULL	Africa	Africa	TEC-SHA-10000501	Technology	Copiers	Sharp Wireless Fa...	2832.96	8	0	311.52	903.04	Critic

# Add Total Cost Column:

04:45 PM (1s)

4

```
#Add Total Cost Column
from pyspark.sql.functions import col

df_transformed = df.withColumn("TotalCost", col("Sales") - col("Profit"))
df.show(5)
```

(1) Spark Jobs

df\_transformed: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 23 more fields]

32298	CA-2012-124891	2012-07-31	2012-07-31	Same Day	RH-19495	Rick Hansen	Consumer	New York City	New York	United States		
10024	US	East	TEC-AC-10003033	Technology	Accessories	Plantronics CS510...	2309.65	7	0	762.1845	933.57	Critic
26341	IN-2013-77878	2013-02-05	2013-02-07	Second Class	JR-16210	Justin Ritter	Corporate	Wollongong	New South Wales	Australia		
NULL	APAC	Oceania	FUR-CH-10003950	Furniture	Chairs	Novimex Executive...	3709.395	9	0.1	-288.765	923.63	Critic
25330	IN-2013-71249	2013-10-17	2013-10-18	First Class	CR-12730	Craig Reiter	Consumer	Brisbane	Queensland	Australia		
NULL	APAC	Oceania	TEC-PH-10004664	Technology	Phones	Nokia Smart Phone...	5175.171	9	0.1	919.971	915.49	Medi
13524	ES-2013-1579342	2013-01-28	2013-01-30	First Class	KM-16375	Katherine Murray	Home Office	Berlin	Berlin	Germany		
NULL	EU	Central	TEC-PH-10004583	Technology	Phones	Motorola Smart Ph...	2892.51	5	0.1	-96.54	910.16	Medi
47221	SG-2013-4320	2013-11-05	2013-11-06	Same Day	RH-9495	Rick Hansen	Consumer	Dakar	Dakar	Senegal		
NULL	Africa	Africa	TEC-SHA-10000501	Technology	Copiers	Sharp Wireless Fa...	2832.96	8	0	311.52	903.04	Critic



# Convert Sales and Profit to Float:

04:45 PM (1s)

5

Python

#Convert Sales and Profit to Float

df\_transformed = df\_transformed.withColumn("Sales", col("Sales").cast("float"))

df\_transformed = df\_transformed.withColumn("Profit", col("Profit").cast("float"))

df\_transformed.show(5)

(1) Spark Jobs

df\_transformed: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 23 more fields]

Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	City	State	Country	Postal Code	Market	Region	Product ID	Category	Sub-Category	Product Name	Sales	Quantity	Discount	Profit	Shipping Cost	Order Priority	Total Cost
32298	CA-2012-124891	2012-07-31	2012-07-31	Same Day	RH-19495	Rick Hansen	Consumer	New York City	New York	United States	10024	US	East	TEC-AC-10003033	Technology	Accessories	Plantronics CS510...	2309.65	7	0	762.1845	933.57	Critical	1547.4655000000002
26341	IN-2013-77878	2013-02-05	2013-02-07	Second Class	JR-16210	Justin Ritter	Corporate	Wollongong	New South Wales	Australia	NULL	APAC	Oceania	FUR-CH-10003950	Furniture	Chairs	Novimex Executive...	3709.395	9	0.1	-288.765	923.63	Critical	

# Filter Rows with Zero or Negative Profit:

04:45 PM (1s)

6

Python

```
#Filter Rows with Zero or Negative Profit

df_transformed = df_transformed.filter(col("Profit") > 0)
df_transformed.show(5)
```

(1) Spark Jobs

df\_transformed: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 23 more fields]

32298	CA-2012-124891	2012-07-31	2012-07-31	Same Day	RH-19495	Rick Hansen	Consumer	New York City	New York	United States	
10024	US	East	TEC-AC-10003033	Technology	Accessories	Plantronics CS510...	2309.65	7	0	762.1845	933.57
											Critic
25330	IN-2013-71249	2013-10-17	2013-10-18	First Class	CR-12730	Craig Reiter	Consumer	Brisbane	Queensland	Australia	
	APAC	Oceania	TEC-PH-10004664	Technology	Phones	Nokia Smart Phone...	5175.171	9	0.1	919.971	915.49
											Medi
47221	SG-2013-4320	2013-11-05	2013-11-06	Same Day	RH-9495	Rick Hansen	Consumer	Dakar	Dakar	Senegal	
	NULL	Africa	Africa	TEC-SHA-10000501	Technology	Copiers	Sharp Wireless Fa...	2832.96	8	0	311.52
											Critic
22732	IN-2013-42360	2013-06-28	2013-07-01	Second Class	JM-15655	Jim Mitchum	Corporate	Sydney	New South Wales	Australia	
	NULL	APAC	Oceania	TEC-PH-10000030	Technology	Phones	Samsung Smart Pho...	2862.675	5	0.1	763.275
											Critic
30570	IN-2011-81826	2011-11-07	2011-11-09	First Class	TS-21340	Toby Swindell	Consumer	Porirua	Wellington	New Zealand	
	NULL	APAC	Oceania	FUR-CH-10004050	Furniture	Chairs	Novimex Executive...	1822.08	4	0	564.84
											Critic

04:45 PM (1s)

6

Python

```
# Step 5: Show the transformed data
df_transformed.show(5)
```

(2) Spark Jobs

df\_clean: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 22 more fields]

df\_transformed: pyspark.sql.dataframe.DataFrame = [Category: string, Region: string ... 4 more fields]

Category	Region	TotalSales	TotalProfit	TotalQuantity	TotalDiscount
Furniture	Canada	10595.279964447021	2613.24	78	0.0
Technology	EMEA	300854.583026886	17494.443000000036	2259	189.1000056862831
Furniture	East	205540.3473367691	2501.8162	2151	90.6000018119812
Technology	Africa	322367.0430994034	44129.49300000001	2031	143.1999975964427
Technology	East	264872.0816922188	47439.55759999996	1927	76.30000080168247

only showing top 5 rows

# Add a Profit Margin Column:

04:45 PM (2s)

8

Python

```
from pyspark.sql.functions import col

# Add a Profit Margin Column
df_transformed = df_transformed.withColumn("ProfitMargin", (col("TotalProfit") / col("TotalSales")) * 100)

# Show the result
df_transformed.show(5)
```

(2) Spark Jobs

df\_transformed: pyspark.sql.dataframe.DataFrame = [Category: string, Region: string ... 5 more fields]

Category	Region	TotalSales	TotalProfit	TotalQuantity	TotalDiscount	ProfitMargin
Furniture	Canada	10595.279964447021	2613.24	78	0.0	24.664190174953884
Technology	EMEA	300854.583026886	17494.443000000036	2259	189.1000056862831	5.814916569988444
Furniture	East	205540.3473367691	2501.8162	2151	90.6000018119812	1.2171898278934408
Technology	Africa	322367.0430994034	44129.4930000001	2031	143.1999975964427	13.689207363046899
Technology	East	264872.0816922188	47439.55759999996	1927	76.30000080168247	17.910365372189244

only showing top 5 rows

# Remove Duplicate Rows based on 'Category' and 'Region':

04:45 PM (1s)

9

Python

```
# Remove Duplicate Rows based on 'Category' and 'Region'
df_transformed = df_transformed.dropDuplicates(["Category", "Region"])

df_transformed.show(5)
```

(2) Spark Jobs

df\_transformed: pyspark.sql.dataframe.DataFrame = [Category: string, Region: string ... 5 more fields]

Category	Region	TotalSales	TotalProfit	TotalQuantity	TotalDiscount	ProfitMargin
Furniture	Canada	10595.279964447021	2613.24	78	0.0	24.664190174953884
Technology	EMEA	300854.583026886	17494.443000000036	2259	189.1000056862831	5.814916569988444
Furniture	East	205540.3473367691	2501.8162	2151	90.6000018119812	1.2171898278934408
Technology	Africa	322367.0430994034	44129.4930000001	2031	143.1999975964427	13.689207363046899
Technology	East	264872.0816922188	47439.55759999996	1927	76.30000080168247	17.910365372189244

only showing top 5 rows

## Rename Columns for Clarity:

```
#Rename Columns for Clarity
df_transformed = df_transformed.withColumnRenamed("Sales", "TotalSales") \
                                .withColumnRenamed("Profit", "TotalProfit")

df_transformed.show(5)
```

▶ (2) Spark Jobs

```
df_transformed: pyspark.sql.dataframe.DataFrame = [Category: string, Region: string ... 5 more fields]
```

Category	Region	TotalSales	TotalProfit	TotalQuantity	TotalDiscount	ProfitMargin
Furniture	Canada	10595.279964447021	2613.24	78	0.0	24.664190174953884
Technology	EMEA	300854.583026886	17494.443000000036	2259	189.1000056862831	5.814916569988444
Furniture	East	205540.3473367691	2501.8162	2151	90.6000018119812	1.2171898278934408
Technology	Africa	322367.0430994034	44129.49300000001	2031	143.1999975964427	13.689207363046899
Technology	East	264872.0816922188	47439.55759999996	1927	76.30000080168247	17.910365372189244

only showing top 5 rows

### Add a Year Column:

```
from pyspark.sql.functions import col, year, to_date

# Add a Year Column
df = df.withColumn("Year", year(to_date(col("Order Date"), "MM/dd/yyyy")))

# Show the result
df_transformed.show(5)
```

▶ (2) Spark Jobs

```
df: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 23 more fields]
```

Category	Region	TotalSales	TotalProfit	TotalQuantity	TotalDiscount	ProfitMargin
Furniture	Canada	10595.279964447021	2613.24	78	0.0	24.664190174953884
Technology	EMEA	300854.583026886	17494.443000000036	2259	189.1000056862831	5.814916569988444
Furniture	East	205540.3473367691	2501.8162	2151	90.6000018119812	1.2171898278934408
Technology	Africa	322367.0430994034	44129.4930000001	2031	143.1999975964427	13.689207363046899
Technology	East	264872.0816922188	47439.55759999996	1927	76.30000080168247	17.910365372189244

only showing top 5 rows

# Filter the Data for a Specific Year:

04:45 PM (2s)

12

Python

```
from pyspark.sql.functions import col, year, to_date, sum
#Filter the Data for a Specific Year

# Step 1: Add the Year Column to the Original DataFrame
df = df.withColumn("Year", year(to_date(col("Order Date"), "MM/dd/yyyy")))

# Step 2: Perform the aggregation including 'Year'
df_transformed = df.groupBy("Category", "Region", "Year") \
    .agg(
        sum("Sales").alias("TotalSales"),
        sum("Profit").alias("TotalProfit"),
        sum("Quantity").alias("TotalQuantity"),
        sum("Discount").alias("TotalDiscount")
    )

# Step 3: Filter the Data for a Specific Year
df_transformed = df_transformed.filter(col("Year") == 2012)

# Show the result
df_transformed.show(5)
```

(2) Spark Jobs

df: pyspark.sql.dataframe.DataFrame = [Row ID: integer, Order ID: string ... 23 more fields]

df\_transformed: pyspark.sql.dataframe.DataFrame = [Category: string, Region: string ... 5 more fields]

Category	Region	Year	TotalSales	TotalProfit	TotalQuantity	TotalDiscount
Furniture	Oceania	2012	100519.00800000002	8623.818	607.0	26.199999999999998
Technology	North	2012	126353.56300000004	25098.862999999998	809.0	7.378
Technology	Africa	2012	64734.582000000024	6320.742000000001	404.0	36.699999999999999
Furniture	Canada	2012	1600.68	290.19	16.0	0.0
Technology	Oceania	2012	89761.76700000005	14203.827	688.0	23.499999999999986

only showing top 5 rows



# Sort Data by Total Sales:

04:46 PM (1s)

13

Python

```
#Sort Data by Total Sales
df_transformed = df_transformed.orderBy(col("TotalSales").desc())
df_transformed.show()
```

(2) Spark Jobs

df\_transformed: pyspark.sql.dataframe.DataFrame = [Category: string, Region: string ... 5 more fields]

Category	Region	Year	TotalSales	TotalProfit	TotalQuantity	TotalDiscount
Technology	Central	2012	237291.811620000002	31373.854319999966	1791.0	61.14199999999993
Furniture	Central	2012	183778.999499999998	5231.0848000000005	1876.94	85.14000000000004
Office Supplies	Central	2012	180068.292000000002	26988.422600000013	6003.084	198.40000000000001
Technology	North	2012	126353.563000000004	25098.862999999998	809.0	7.378
Office Supplies	South	2012	116962.672000000002	10905.974199999997	3831.186	151.70000000000016
Furniture	South	2012	106962.4015	8656.693399999998	1108.0	38.94999999999999
Technology	South	2012	103154.04796	11673.361959999993	1034.824	38.896
Furniture	Oceania	2012	100519.008000000002	8623.818	607.0	26.199999999999998
Office Supplies	North	2012	91426.151	17629.370999999996	2648.0	36.899999999999984
Technology	Oceania	2012	89761.767000000005	14203.827	688.0	23.499999999999986
Furniture	North	2012	82451.258000000002	8177.708000000001	839.0	50.499999999999996
Technology	Southeast Asia	2012	77886.988799999996	6068.458799999999	534.0	32.090000000000001
Technology	Central Asia	2012	69458.760000000001	13756.859999999999	394.0	5.5
Technology	North Asia	2012	64934.625000000001	13026.855000000001	359.0	4.0
Technology	Africa	2012	64734.582000000004	6320.742000000001	404.0	36.69999999999999

# Save the transformed data in DBFS:

04:46 PM (14s)

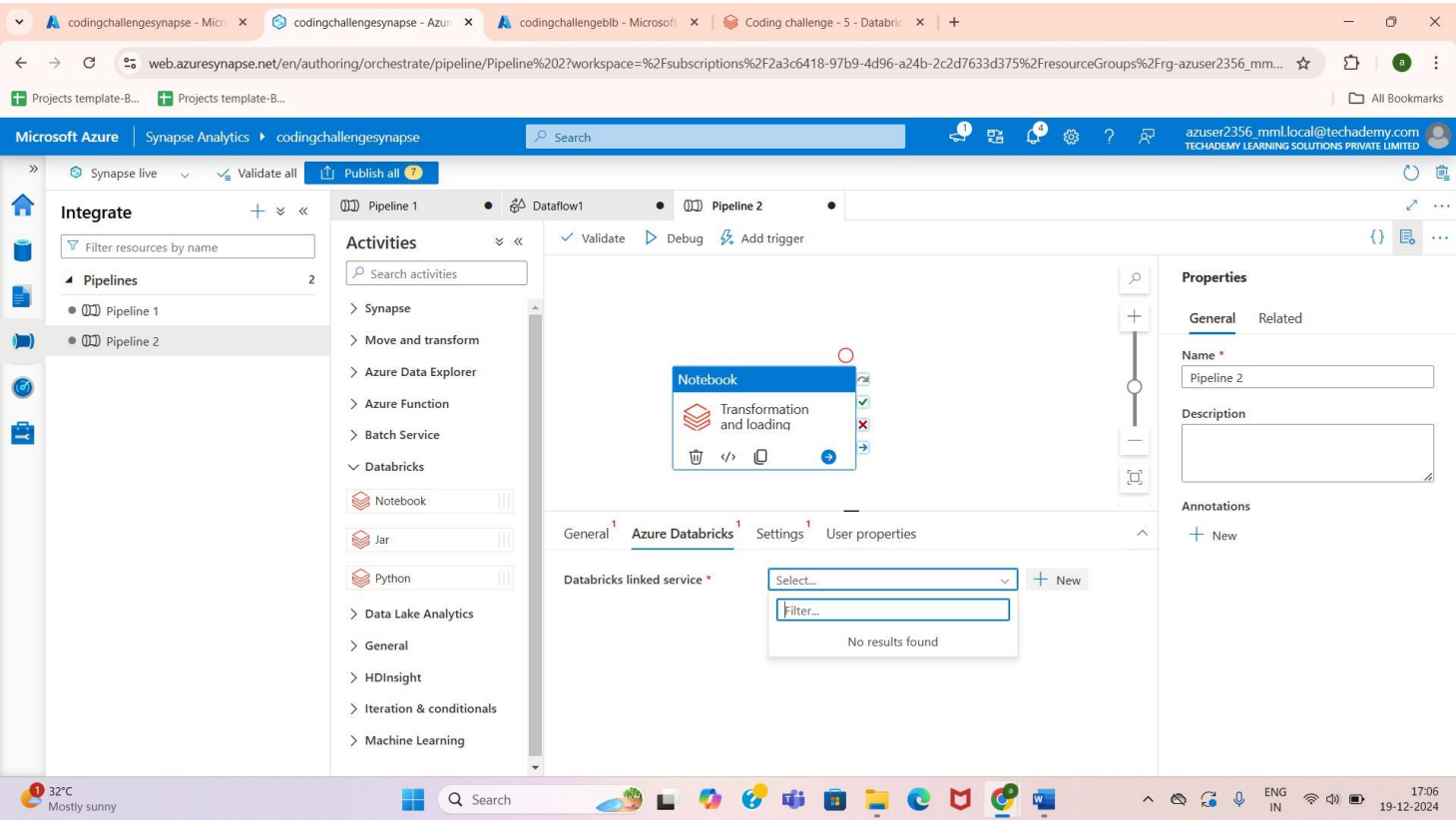
14

Python

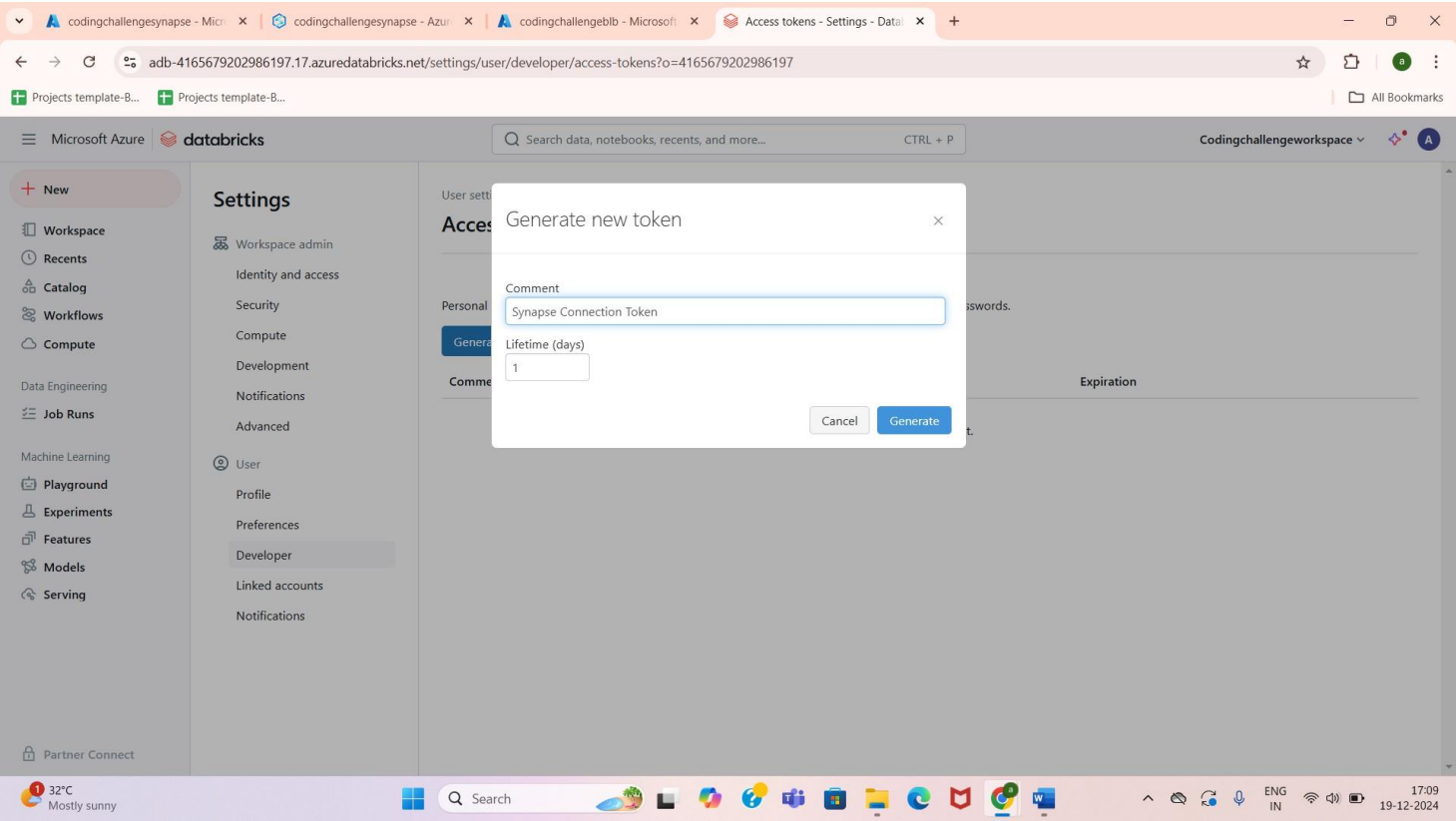
```
# Save the transformed data in DBFS
df_transformed.write.format("delta").mode("overwrite").saveAsTable("transformed_data")
```

(9) Spark Jobs

# Configure Databricks and the Databricks notebook to Azure Synapse Pipeline:



# Generate a new token in Databricks to connect to Azure Synapse:



Microsoft Azure | Synapse Analytics | codingchallengesynapse

Integrate

Filter resources by name

Pipelines

- Pipeline 1
- Pipeline 2

Activities

- Synapse
- Move and transform
- Azure Data Explorer
- Azure Function
- Batch Service
- Databricks
  - Notebook
  - Jar
  - Python
- Data Lake Analytics
- General
- HDInsight
- Iteration & conditionals
- Machine Learning

Pipeline 1

Dataflow1

Pipeline 2

Validate

Debug

Add trigger

Notebook

Transformation and loading

General

Azure Databricks

Settings

User

Databricks linked service \*

Select...

New linked service

Azure Databricks

Learn more

https://adb-4165679202986197.17.azuredatabricks.net

Authentication type \*

Access Token

Access token \*

Cluster version \*

15.4 LTS (includes Apache Spark 3.5.0, Scala 2.12)

Cluster node type \*

Standard\_D4ds\_v5

Python version \*

2

Worker options

Fixed

Autoscaling

Workers \*

1

Create

Cancel

Test connection

## Validate and Debug the pipeline:

Microsoft Azure | Synapse Analytics | codingchallengesynapse

Integrate

Filter resources by name

Pipelines

- Ingestion\_Pipeline
- Transformation\_and>Loading\_Pipeline
- Final\_Pipeline

Activities

- Synapse
- Move and transform
- Azure Data Explorer
- Azure Function
- Batch Service
- Databricks
  - Notebook
- Data Lake Analytics
- General
- HDInsight
- Iteration & conditionals
- Machine Learning

Ingestion\_Pipeline

Dataflow1

Transformation\_and>Loading\_Pipeline

Final\_Pipeline

Validate

Debug

Add trigger

Notebook

Transformation and loading

Parameters

Variables

Settings

Output

Pipeline run ID: 7907e14f-8325-4886-80fd-adfdb9113ac

Pipeline status: Succeeded

View debug run consumption

All status

Monitor in Azure Metrics

Export to CSV

Showing 1 - 1 of 1 items

Activity name	Activity status	Activity type	Run start
Transformati...	Succeeded	Notebook	12/19/2024, 5:42:28 PI

Properties

General

Related (1)

Name \*

Transformation\_and>Loading\_Pipeline

Description

Annotations

New

# Create the final pipeline to connect the ingestion pipeline and Transformation pipeline and then Validate and Debug it:

The screenshot displays the Microsoft Azure Synapse Analytics web portal. The main workspace shows a pipeline named 'Final\_Pipeline' with two activities: 'Execute Pipeline' (Ingestion\_Pipeline) and 'Execute Pipeline' (Transformation\_and>Loading\_Pipeline). The pipeline is in a 'Succeeded' state. The 'Properties' panel on the right shows the pipeline name as 'Pipeline 3'. The 'Output' tab at the bottom shows a table of activity results.

Activity name	Activity status	Activity type	Run start
Transformation_and>Loading_Pipeline	Succeeded	Execute Pipeline	12/19/2024, 5:18:04
Ingestion_Pipeline	Succeeded	Execute Pipeline	12/19/2024, 5:15:21

--Thank You!