

Retail Data Processing and Visualization Framework



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An End-to-End Process of Data Flow

We are creating pipelines to move data to the cloud and transform it using data flows and stored procedures. This process prepares the data for the Power BI team to use for their reports.

Data Description

1) **cashorder**: This table contains information about orders. It has the following columns:

- **storeid**: Identifier for the store
- **customernum**: Customer number
- **invoicenum**: Invoice number
- **invoicetype**: Type of invoice
- **itemnum**: Item number
- **localamt**: Local amount
- **cashierid**: Identifier for the cashier

2) **customer_data**: This table provides information about customers. It includes:

- **customernumber**: Customer number
- **name**: Customer's name
- **phone1**: Primary phone number
- **phone2**: Secondary phone number
- **email**: Email address
- **address**: Physical address

3) **products_data**: This table contains information about products. It has:

- **productid**: Product identifier
- **name**: Product name
- **price**: Product price

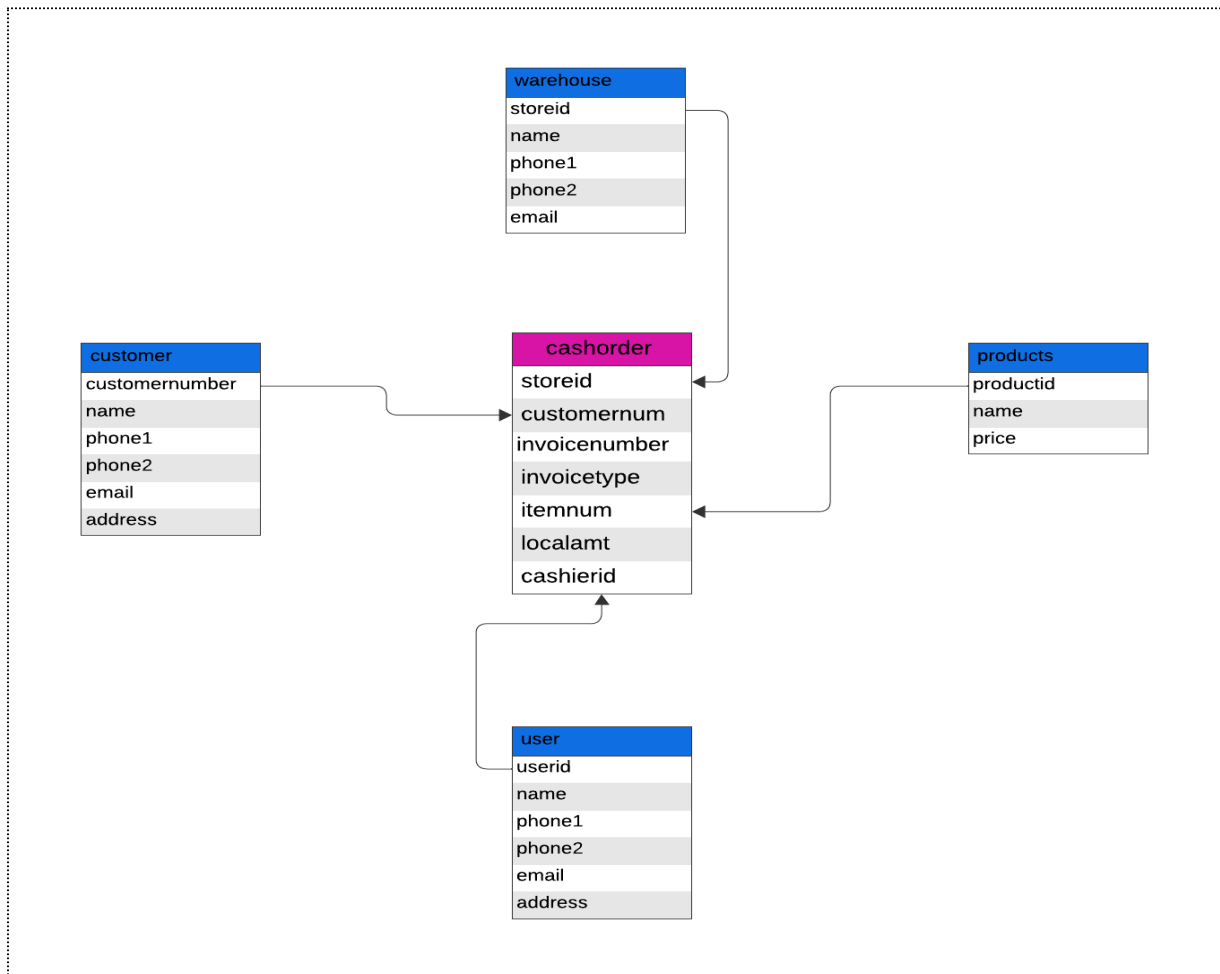
4) **store_data**: This table contains information about stores. It includes:

- **storeid**: Store identifier
- **name**: Store name
- **phone1**: Primary phone number
- **phone2**: Secondary phone number
- **email**: Email address

5) **cashier_data**: This table provides information about cashiers working at the billing desks. It has:

- **userid**: User identifier
- **name**: Cashier's name
- **phone1**: Primary phone number
- **phone2**: Secondary phone number
- **email**: Email address
- **address**: address

ER Diagram



Python Code for Sample Data Creation

```
brands = ['Apple', 'Samsung', 'Google', 'Microsoft', 'Sony', 'LG', 'HP', 'Dell', 'Lenovo', 'Asus']
item_types = ['Phone', 'Laptop', 'Tablet', 'Smartwatch', 'Monitor', 'Keyboard', 'Mouse', 'Headphones', 'Speaker']

# Function to generate a random product name
def generate_product_name():
    brand = random.choice(brands)
    item_type = random.choice(item_types)
    return f"{brand} {item_type}"

# Function to generate a random price (between $100 and $2000)
def generate_price():
    return round(random.uniform(100, 2000), 2)

# Generate 100 products
data = []
for product_id in range(1, 101):
    name = generate_product_name()
    price = generate_price()

    data.append({
        'productid': product_id,
        'name': name,
        'price': price
    })

# Create a DataFrame from the generated data
product = pd.DataFrame(data)

# Display the first few rows of the DataFrame
```

Created SSMS tables using the sample data

Specify Input File

This operation will create a table from your input file.

Location of file to be imported

C:\courses_internship\internship\ssms\Trent_sample_data\sales data\cashorder.cs

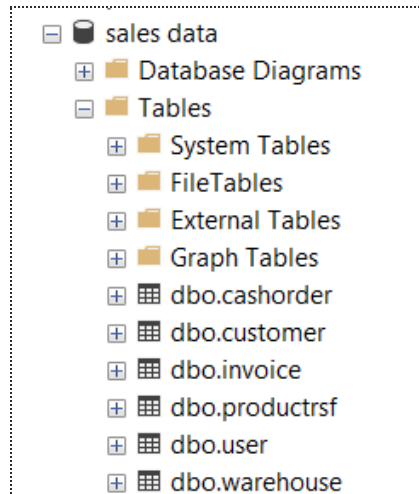
[Browse...](#)

New table name:

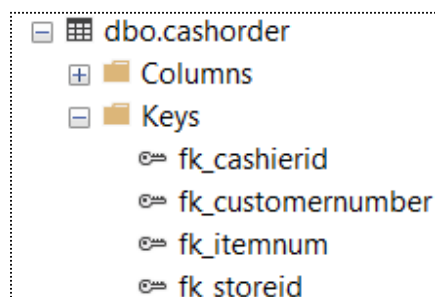
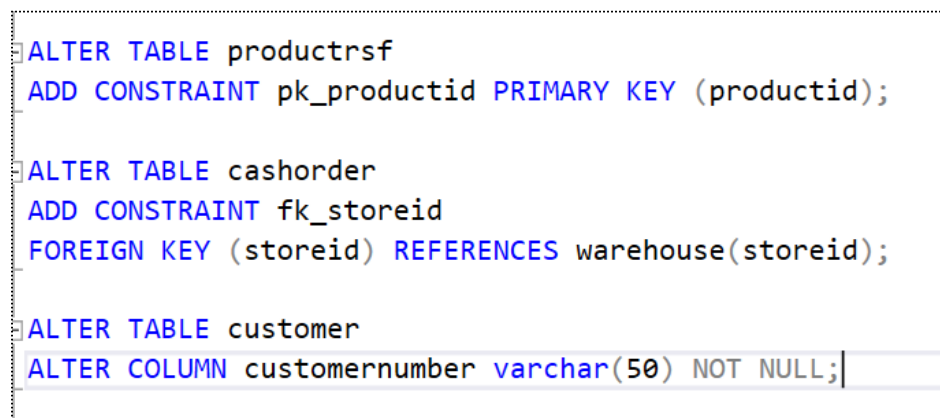
cashorder

Table schema:

dbo



Created Keys and Constraints on the Table



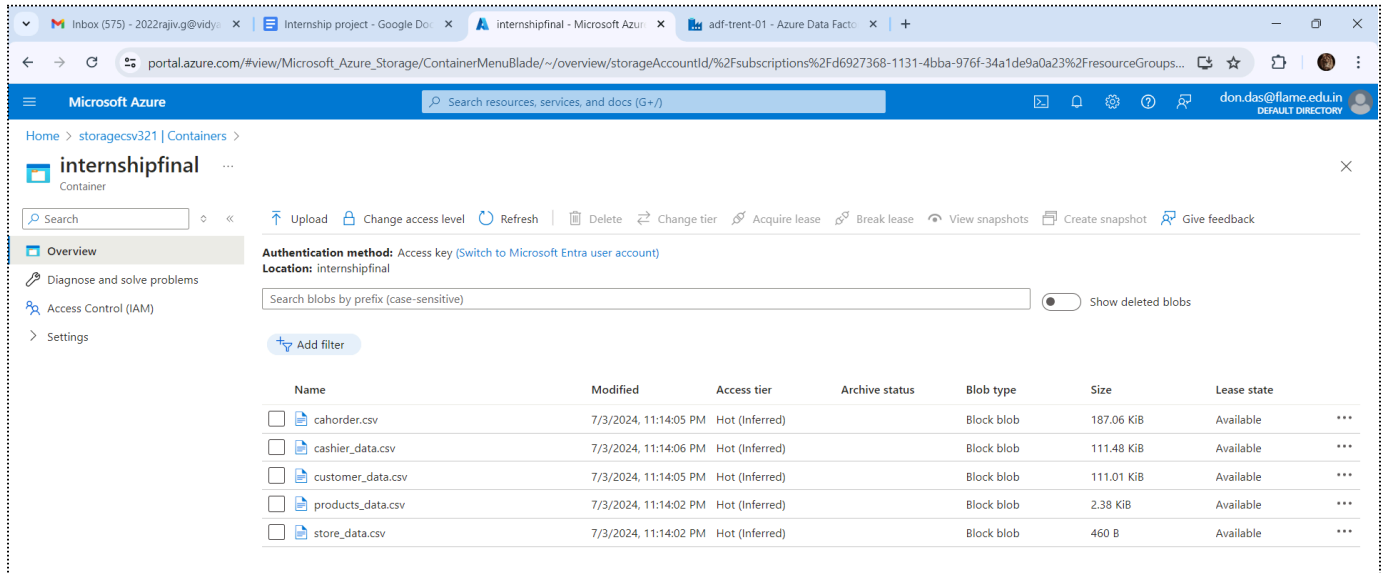
Retrieved Data using various report queries

```
select
    p.[name],
    round(sum(c.localamt),2) [Total Sales]
from
    cashorder c left join productrsf p on c.itemnum = p.productid
where
    c.invoicetype = 31
group by
    p.productid,p.[name]
order by
    [Total Sales] desc
```

```
select
    cu.[name],
    count(distinct c.invoicenum) [Total Visits]
from
    cashorder c left join customer cu on c.customernum = cu.customernumber
group by
    cu.customernumber,cu.[name]
order by
    [Total Visits] desc
```

Azure application

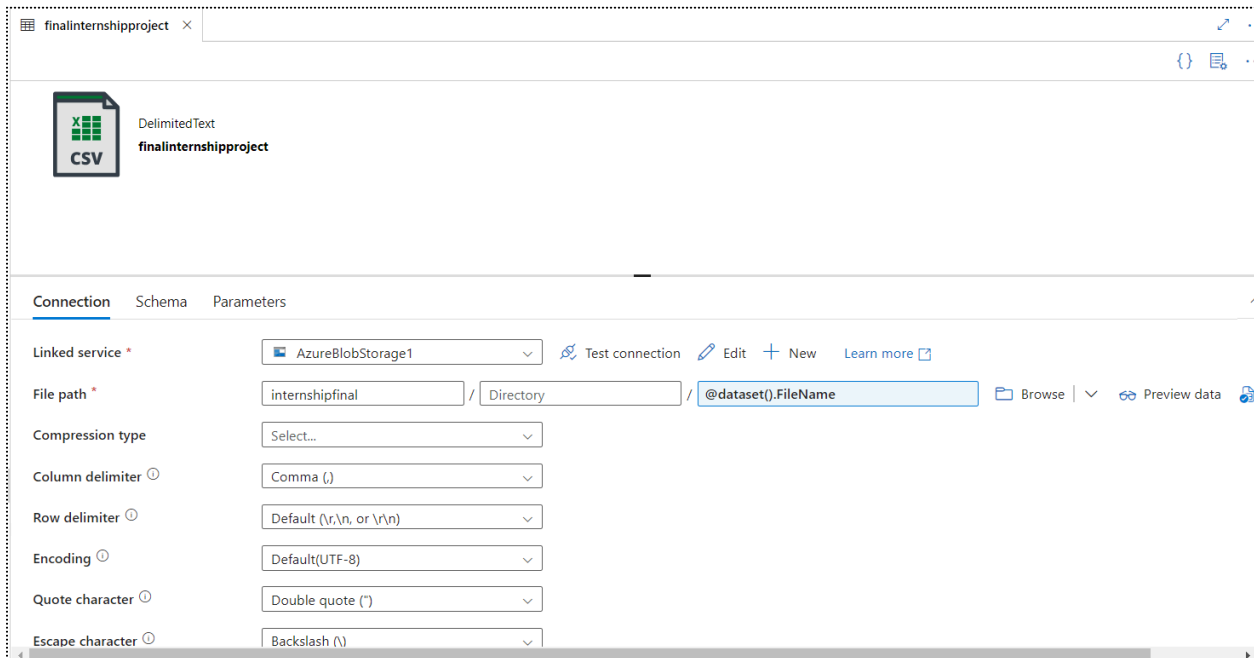
1. Load all these 5 files (on-premises data) in the container of the blob storage:
2. Upload the five data files from your local system to a designated folder (container) in Azure Blob Storage.



Microsoft Azure portal view of the 'internshipfinal' container. The container contains five CSV files:

Name	Modified	Access tier	Archive status	Blob type	Size	Lease state
cahorder.csv	7/3/2024, 11:14:05 PM	Hot (Inferred)		Block blob	187.06 KiB	Available ***
cashier_data.csv	7/3/2024, 11:14:06 PM	Hot (Inferred)		Block blob	111.48 KiB	Available ***
customer_data.csv	7/3/2024, 11:14:05 PM	Hot (Inferred)		Block blob	111.01 KiB	Available ***
products_data.csv	7/3/2024, 11:14:02 PM	Hot (Inferred)		Block blob	2.38 KiB	Available ***
store_data.csv	7/3/2024, 11:14:02 PM	Hot (Inferred)		Block blob	460 B	Available ***

3. Create dataset- finalinternshipproject



Azure Data Factory dataset configuration for 'finalInternshipproject' (DelimitedText format).

Connection | Schema | Parameters

Linked service: AzureBlobStorage1

File path: internshipfinal / Directory / @dataset().FileName

Compression type: Select...

Column delimiter: Comma (,)

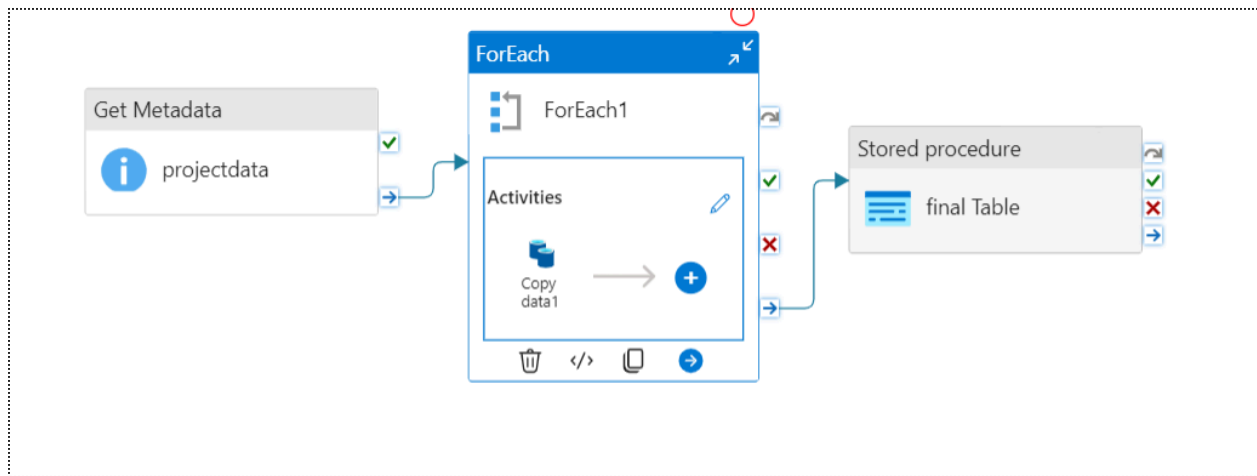
Row delimiter: Default (\r, \n, or \r\n)

Encoding: Default(UTF-8)

Quote character: Double quote (")

Escape character: Backslash (\)

4. Develop a pipeline which moves all data to an azure SQL database and then join tables with stored procedures.



- Open a new pipeline. Drag meta data activity which fetches all 5 tables. Add child items in a field list.

General **Settings** User properties

Dataset * finalinternshipproject [Open](#) [New](#) [Learn more](#)

Dataset properties

Name	Value
File Name	*.csv

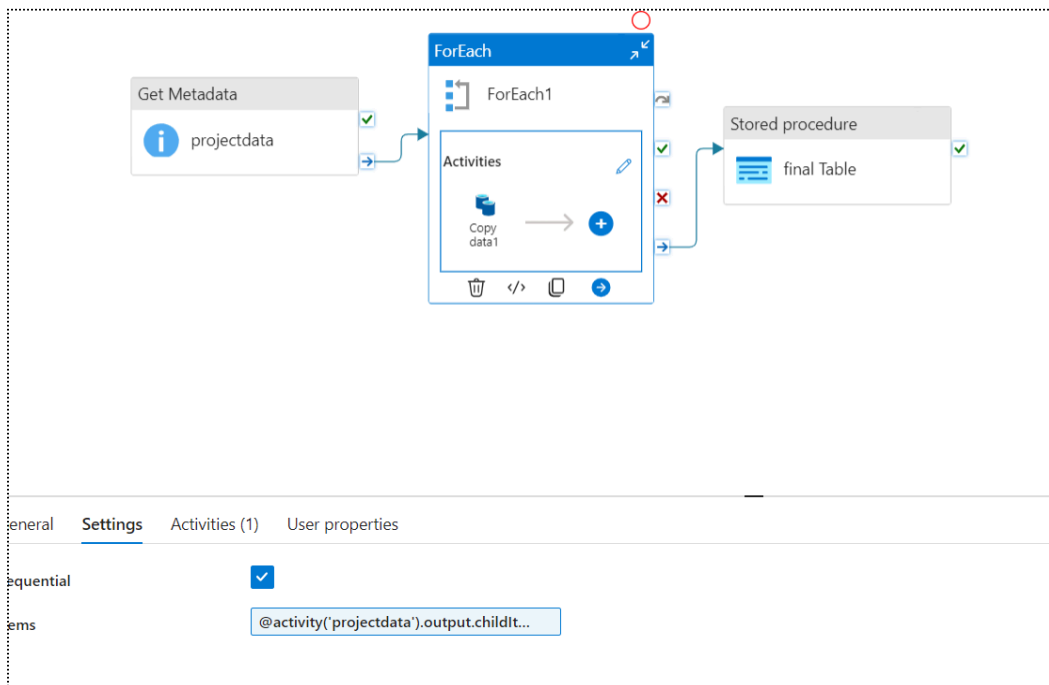
Field list * [New](#) [Delete](#)

☐ Argument

☐ Child items

Start time (UTC) End time (UTC)

- Get another activity ForEach and connect to Get Metadata activity. Give output of Get Metadata activity to the ForEach activity run the copy activity inside it which copies data to Azure SQL.

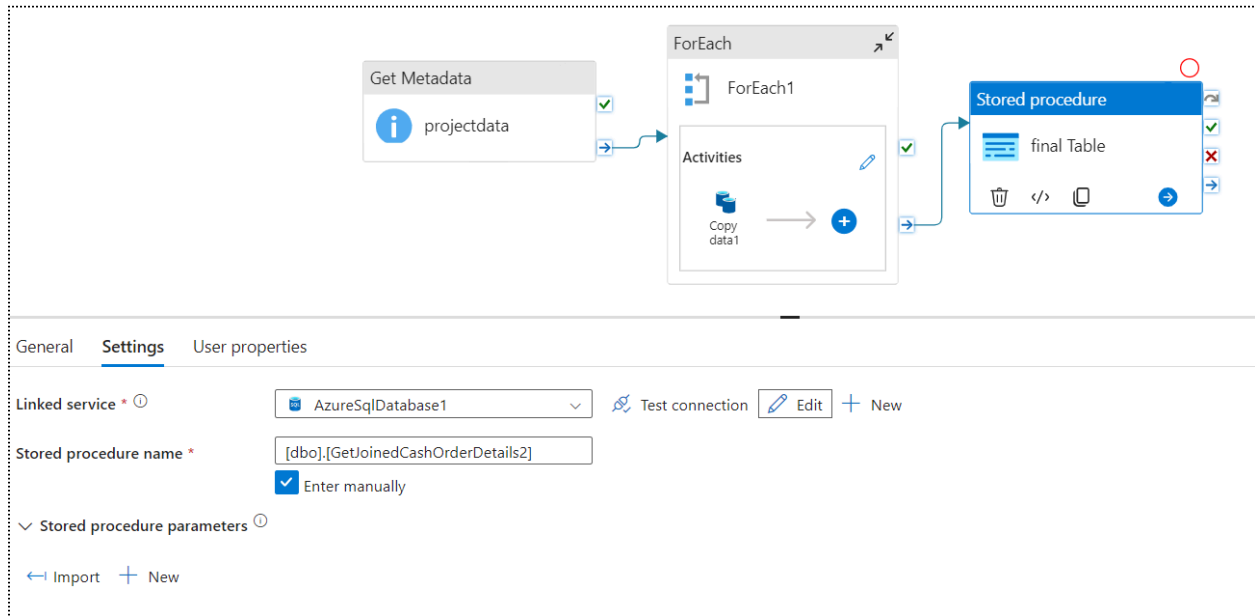


The screenshot shows the 'Sink' properties for a 'Copy data' activity (labeled 'Copy data1'). The 'Sink dataset' is set to 'projectoutput'. The 'Dataset properties' section is expanded, showing a table with the following properties:

Name	Value
filename	@item().name

The 'Write behavior' is set to 'Insert'. The 'Bulk insert table lock' is set to 'No'. The 'Table option' is set to 'Auto create table'.

5. Run the stored procedure which joins all these 5 tables.



```
CREATE PROCEDURE GetJoinedCashOrderDetails2
AS
BEGIN
    CREATE TABLE JoinedCashOrderDetails (
        storeid NVARCHAR(50),
        store_name NVARCHAR(255),
        store_phone1 NVARCHAR(50),
        store_phone2 NVARCHAR(50),
        store_email NVARCHAR(255),
        customernum NVARCHAR(255),
        customer_name NVARCHAR(255),
        customer_phone1 NVARCHAR(50),
        customer_phone2 NVARCHAR(50),
        customer_email NVARCHAR(255),
        customer_address NVARCHAR(255),
        invoicenum NVARCHAR(50),
        invoicetype NVARCHAR(50),
        itemnum NVARCHAR(50),
        product_name NVARCHAR(255),
        product_price NVARCHAR(255),
        localamt NVARCHAR(255),
        cashierid NVARCHAR(255),
        cashier_name NVARCHAR(255),
        cashier_phone1 NVARCHAR(50),
        cashier_phone2 NVARCHAR(50),
```

```

        cashier_email NVARCHAR(255),
        cashier_address NVARCHAR(255)
    );

insert into JoinedCashOrderDetails
SELECT
    co.storeid,
    s.name AS store_name,
    s.phone1 AS store_phone1,
    s.phone2 AS store_phone2,
    s.email AS store_email,
    co.customernum,
    c.name AS customer_name,
    c.phone1 AS customer_phone1,
    c.phone2 AS customer_phone2,
    c.email AS customer_email,
    c.address AS customer_address,
    co.invoicenum,
    co.invoicetype,
    co.itemnum,
    p.name AS product_name,
    p.price AS product_price,
    co.localamt,
    co.cashierid,
    ca.name AS cashier_name,
    ca.phone1 AS cashier_phone1,
    ca.phone2 AS cashier_phone2,
    ca.email AS cashier_email,
    ca.address AS cashier_address
FROM
    CASHORDER co
LEFT JOIN
    store s ON co.storeid = s.storeid
LEFT JOIN
    customer c ON co.customernum = c.customernumber
LEFT JOIN
    products p ON co.itemnum = p.productid
LEFT JOIN
    cashier ca ON co.cashierid = ca.userid;
END;

```

Join all these 5 tables based on following conditions:

cashorder.customernum== customer_data.customernumber

cashorder.productid== products_data.productid

cashorder.storeid== store_data.storeid

cashorder.cashiersid== cashier_data.userid

Final merged file has 6615 rows with 23 columns which get stored in the SQL database.

Query 3

Run Cancel query Save query Export data as Show only Editor

```
1 SELECT TOP (1000) * FROM [dbo].[JoinedCashOrderDetails];
2 select count(*) From [dbo].[JoinedCashOrderDetails];
3
```

Results Messages

Search to filter items...

storeid	store_name	store_phone1	store_phone2	store_email	customernum	customer_name
3	DigitalSolutions Palace	658-865-1281	991-749-8380	cainderrick@example.com	409	Jonathan Andrews
3	DigitalSolutions Palace	658-865-1281	991-749-8380	cainderrick@example.com	98	Christina Allison
3	DigitalSolutions Palace	658-865-1281	991-749-8380	cainderrick@example.com	409	Jonathan Andrews
3	DigitalSolutions Palace	658-865-1281	991-749-8380	cainderrick@example.com	98	Christina Allison

```
1 SELECT TOP (1000) * FROM [dbo].[JoinedCashOrderDetails];
2 select count(*) From [dbo].[JoinedCashOrderDetails];
3
```

Results Messages

Search to filter items...

6615

As this data doesn't need any transformation, connect Power BI to the SQL server and start making reports.

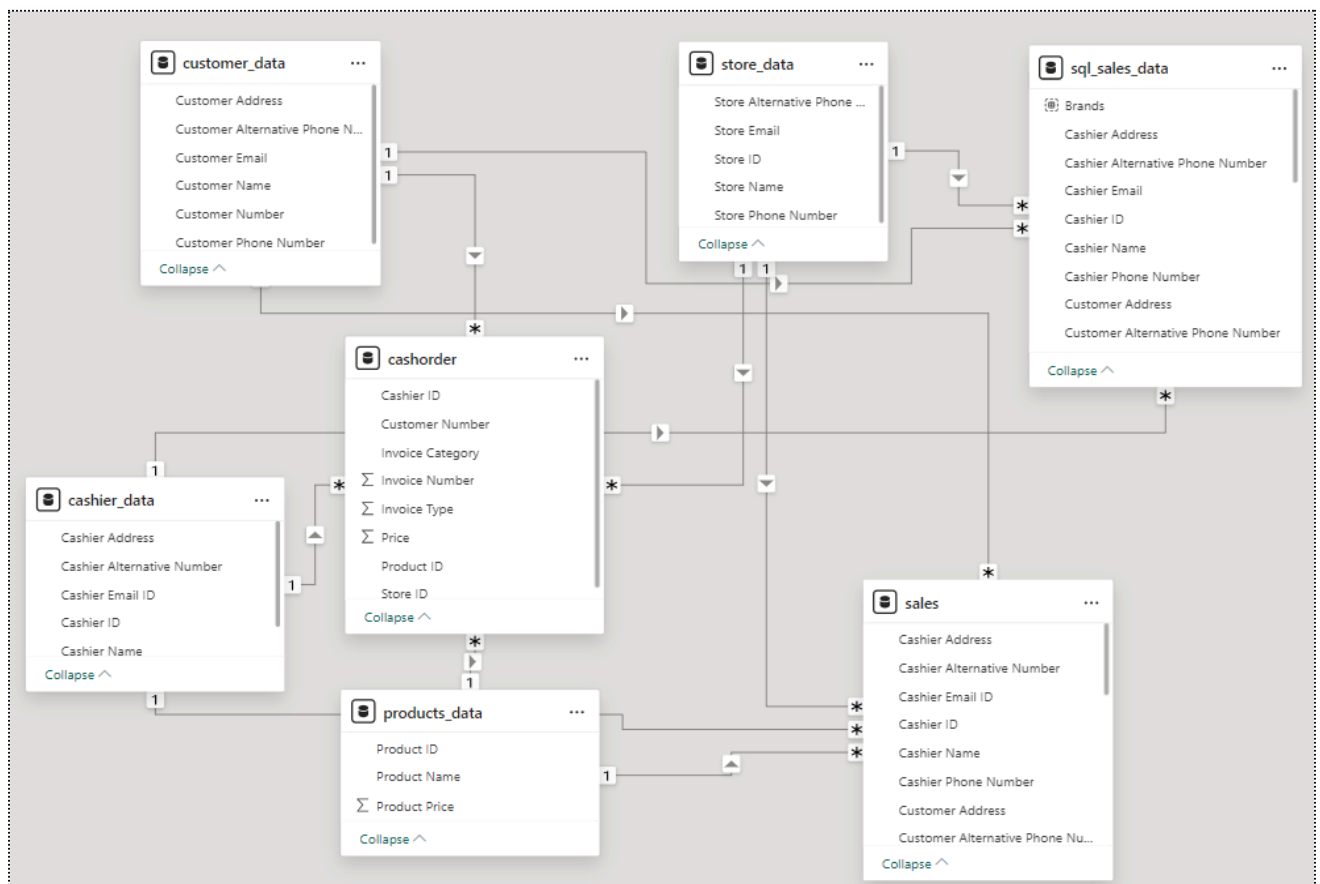
Challenges:

- Connection Issues and Data Source Configuration
- Data Quality Issues
- User Access Management
- Understanding Power BI Features
- SQL Queries Complexity
- Integrating with Other Tools such as Power BI

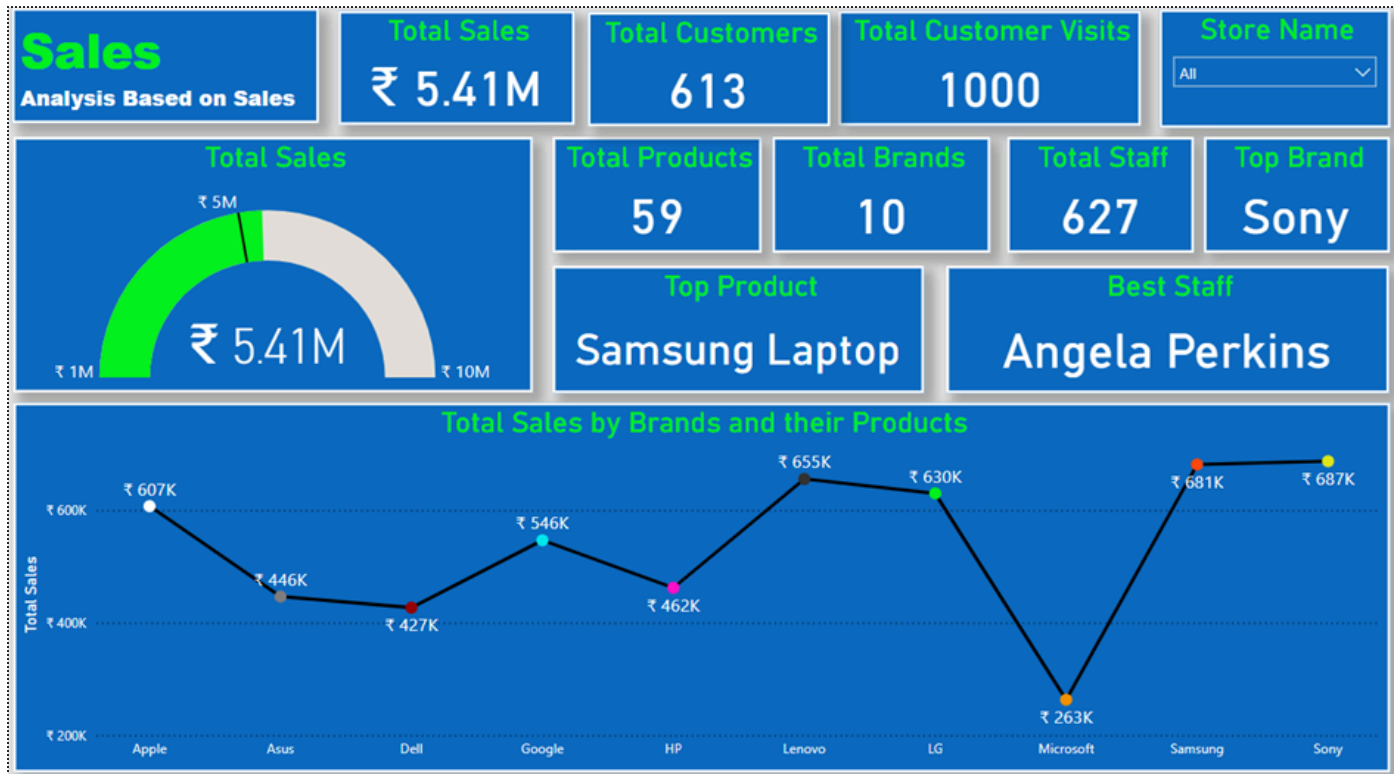
Power BI Application

In Power BI, we did the basic transformation in Power Query Editor like renaming the columns for better understanding, creating another column using the existing column where we have invoice type as a whole number (31 means Individual, 30 means final) and sorting of columns in ascending or descending to get proper insights. We used close and apply to move the final table from Power Query Editor to Power BI Desktop.

We grouped the data from one column to create another column for hierarchy visualization.

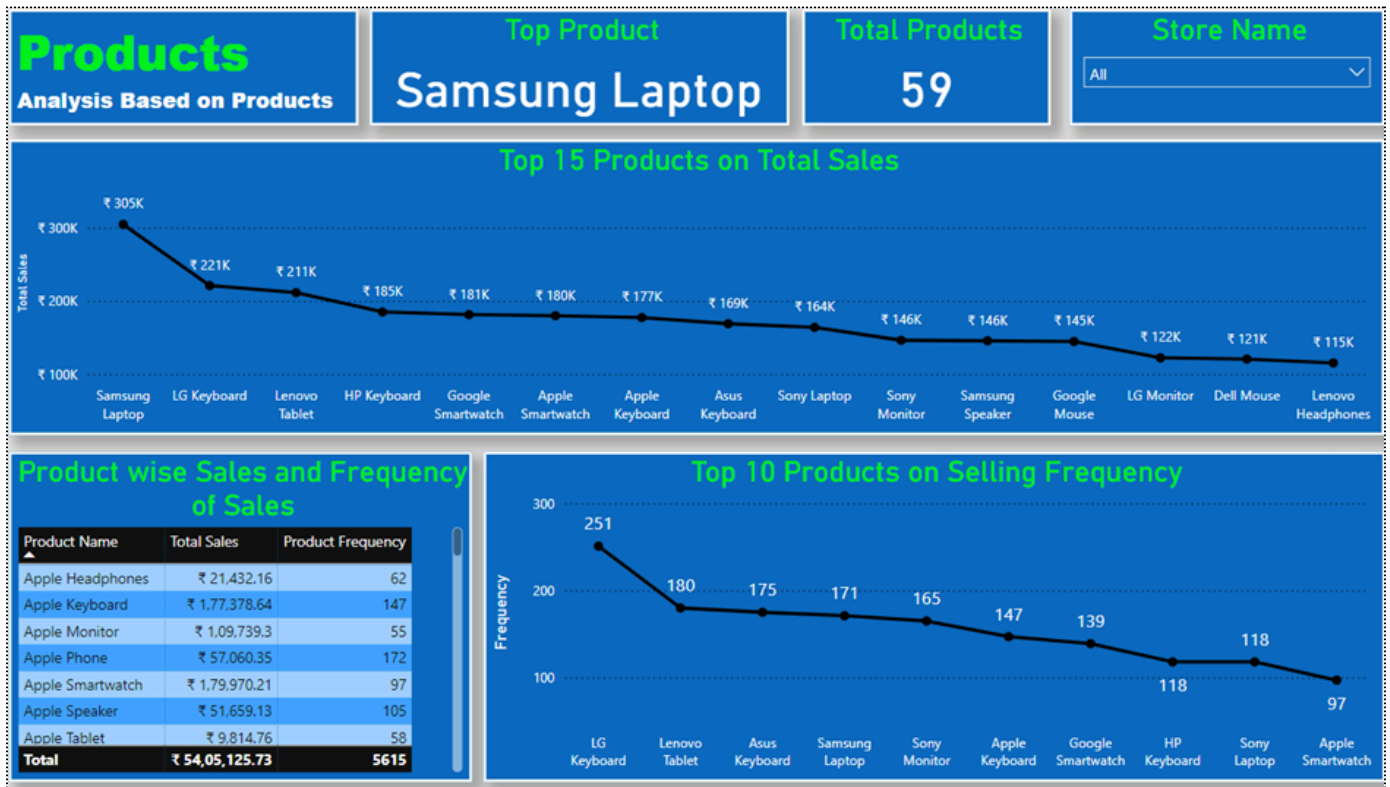


- Sales Overview



This is showing the total sales: ₹5.41 million. There are 613 customers and 1000 customer visits which shows there are some customers who visited the store multiple times, with 59 total products and 10 brands available. The top product is a Samsung Laptop, and the best staff member is Angela Perkins. The total sales by brands are plotted on a graph, with Sony leading at ₹687,000.

- Products



It displays product analysis for a store, highlighting the sales and frequency of various products. The top-selling product is a Samsung Laptop, and there are 59 products in total. The top 15 products by total sales are plotted on a graph, with samsung laptops leading at ₹3,05,000. Additionally, the product-wise sales and frequency of sales are listed, with LG Keyboards having the highest selling frequency at 251 units.

- Customers



This report is focusing on customer analysis for a store, displaying a total of 1000 customer visits and 613 total customers. The top customers by total visits are listed, with Elizabeth Clark, Kathryn Odom, and Rebecca Deleon having the highest number of visits at 5 each. Customer-wise total earnings show Elizabeth Thomas leading with ₹32K, followed closely by Sandra Clayton with ₹31K. It also includes a map for customer city-wise total sales to show the regions.

- Brands



This report is showing the analyzing of brand performance in a retail context. It shows there are 10 total brands, with Sony being the top brand. The line chart illustrates brand-wise customer visits, where Sony leads with around 580 visits, followed by a declining trend for other brands. The table provides details of product sales and customer visits. And a bar graph showing total sales by brand, with Sony again leading in sales volume.

- Staffs



This represents an analysis of staff performance. It identifies Angela Perkins as the best staff member out of a total of 627 employees. The line graph showing the top 15 staff members based on total sales, with Angela Perkins leading at around ₹33K. And the below line chart displaying the top 15 staff based on customers attended, where Mark Lloyd, Michael Glover, Sean Schultz, Thomas Martinez, and Brian Livingston are tied for the highest number of customers. Each one attended a total of 5 customers.

Conclusion

This project effectively illustrates a data pipeline process from beginning to end for moving on-premises data to the cloud, transforming it with Azure Data Factory, and enabling Power BI reporting on that data. This project successfully demonstrates how data ingestion, transformation, and reporting procedures can be streamlined with cloud-based data solutions. . By leveraging Azure Data Factory and Power BI, we created a scalable and efficient pipeline that provides valuable business insights. The comprehensive reports generated in Power BI offer a clear view of sales, product performance, customer behavior, brand success, and staff efficiency, empowering decision-makers with actionable data.