# **Research Methodology**

This study employed a comparative analysis approach to evaluate the performance of Tableau and Power BI dashboards.

### **Data Collection**

The Superstore dataset was obtained from Kaggle and Tableau websites.

# **Data Preparation**

- 1. Data cleaning: Handling missing values and outliers.
- 2. Data transformation: Converting data types and aggregating data.
- 3. Data visualization: Creating dashboards using Tableau and Power BI.

### **Dashboard Creation**

- 1. Tableau: Utilized Tableau Desktop for dashboard creation.
- 2. Power BI: Used Power BI Desktop for dashboard creation.

### **Performance Evaluation**

The data sets offer details on meta data about the sales, profit, and other fascinating key factors of a massive superstore.

# **Data Warehousing and Dataset Overview**

The Superstore dataset contains sales, orders, and product information from various regions worldwide, including:

- Canada
- America
- Asia
- Africa
- India

- Cameroon
- Indonesia
- Mexico
- Japan
- Belgium
- France
- England
- United States

### **Dataset Characteristics**

- Comma-Separated Values (CSV) file
- 51,290 rows of data
- Flat file with separate tables for orders and returns
- Order details: 2 files
- Returns details: 1 file

# **Data Preparation and Warehousing**

To handle large datasets, proper cleaning and preparation are necessary. Tableau and Power BI may struggle with processing large amounts of data.

# **Proposed Research Methodology**

To facilitate data analysis, this study employs a data warehousing approach to clean and prepare the data. We utilize data mining techniques to extract relevant information, ensuring accurate analysis.

# **Data Processing and Manipulation**

Using a business scenario, we will demonstrate data importation, processing, and manipulation with Tableau Desktop and Power BI Desktop. Both tools offer free downloads and support live or desktop server connections.

# **Data Cleaning and Preparation**

Although Tableau offers automatic data cleaning, we opt for the data warehousing method (Extract, Transform, Load - ETL) to preserve crucial customer information. This approach ensures data integrity and relevance.

# **Research Methodology Steps**

Figure 12 illustrates the process:

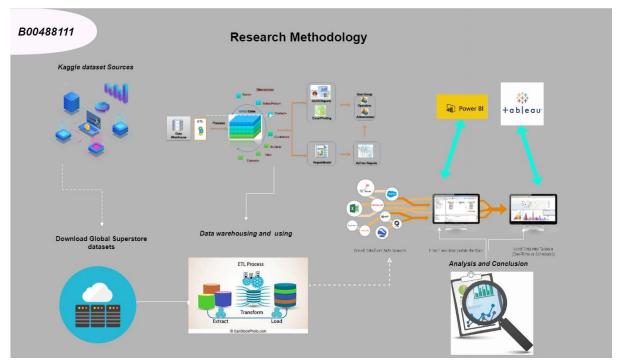


Figure 1: Research methodology

1. Data Extraction: Kaggle dataset download

2. Data Transformation: Data warehousing (ETL)

3. Data Loading: Into Tableau and Power BI

4. Data Analysis: Using business scenarios

5. Results Comparison: Tableau vs Power BI

Dashboard Performance Analysis

# **Data Preparation**

Cleaning and normalizing data is crucial before importing it into visualization tools for business performance monitoring. Although Tableau handles more datasets than Power BI, improper data cleaning can lead to errors.

# **Data Import and Analysis**

After data normalization, we import datasets from the DataMart created in our Datawarehouse processing. Extracted data is organized into sheets for analysis.

# **Research Objective**

Investigate key aspects of Tableau and Power BI using a business scenario.

# **Research Objective**

This study aims to investigate the differences in performance and data analysis capabilities between Tableau and Power BI, despite their similarities in presenting information on dashboards.

# **Literature Gap**

Existing research has explored these tools, but new features and updates from vendors remain unexamined. This study fills this gap.

# Methodology

# Data Mining using ETL Processes

Table 2 outlines our plan of action:

- 1. Search datasets on Kaggle
- 2. Download datasets as CSV files
- 3. Convert files to Excel 2003-2007 format
- 4. Connect to CSVSOCSQL01\SQL2012 database engine
- 5. Create B0048811SalesDataMartMSc data warehouse

- 6. Create Staging Area and DataMart schemas
- 7. Create tables with SQL queries in DataMart schema
- 8. Import data into Staging Area using Import Wizard
- 9. Analyse datasets and fetch information
- 10. Run SQL queries to export datasets

# **Data Visualization and Analysis**

### Tableau Desktop

- 1. Import data
- 2. Create new sheets
- 3. Create columns and measures
- 4. Create visuals
- 5. Format visuals
- 1. Import data
- 2. Edit in Power Query Editor
- 3. Create columns and measures
- 4. Create visuals
- 5. Format visuals

### **Results and Evaluation**

- 1. Compare results
- 2. Evaluate performance differences
- 3. Draw conclusions

### **Tools Used**

- 1. Tableau Desktop (Version 2022.1)
- 2. Power BI Desktop (Version 2.104.702.0 32-bit)

### 3. Microsoft SQL Server 2012

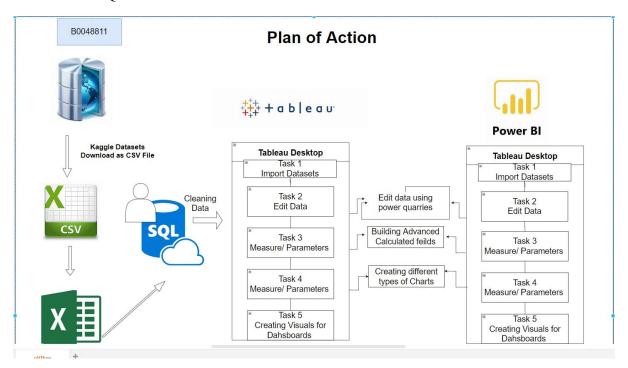


Table 1: Plan of Action

# **Proposed Methods**

Data mining, as defined by Friedman, involves extracting valuable patterns from large, complex datasets (Luo, 2022). This study employs data mining techniques to analyze sales data from a Global Superstore.

# **Dataset Description**

The dataset, sourced from a SQL database, is saved in CSV format and contains:

- Global Superstore sales data
- 2 sheets (Global Superstore returns 2016)
- 1 sheet (Global Superstore Returns 2016)

# **Data Cleaning and Preparation**

To manage large datasets, a single data source is essential. We utilize:

- SQL Server for data storage
- CSV files for data exchange
- Power BI and Tableau for visualization

Data Analysis and Visualization

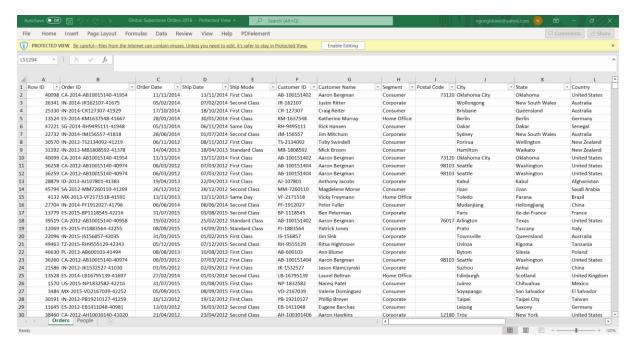
# Our methodology involves:

- 1. Data cleaning and preparation
- 2. Data analysis using advanced metrics and calculated fields
- 3. Visualization using Power BI and Tableau
- 4. Drawing valuable conclusions from dashboard insights

### **Tools Used**

- Tableau Desktop (Version 2022.1)
- Power BI Desktop (Version 2.104.702.0 32-bit)
- Microsoft SQL Server 2012

# **Changes made:**



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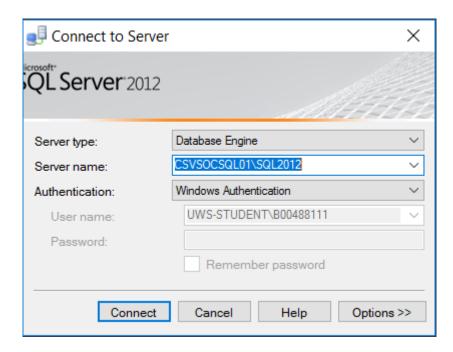


Figure 2: Screenshot of Database Engine Connection

Database Setup and Schema Design

After connecting to the database engine, I created:

- 1. Database: B0048811SalesDataMartMSc
- 2. Schemas:
  - Staging Area (for data import)
  - DataMart (for data transformation and loading)

**Dimension Creation** 

In the DataMart schema, I created four dimensions using SQL codes:

- 1. DimCustomer
- 2. DimProduct
- 3. DimOrders
- 4. DimFactsGlobalSales

**SQL** Statements

Below are the SQL statements used to create dimensions and populate tables with extracted data:

-- Create database

CREATE SCHEMA DataMart;

CREATE DATABASE B0048811SalesDataMartMSc USE B0048811SalesDataMartMSc; go CREATE SCHEMA StagingArea;

```
-- Create schemas
CREATE SCHEMA StagingArea;
CREATE SCHEMA DataMart;
-- Create dimensions
CREATE TABLE DataMart.DimCustomer (
 CustomerID INT PRIMARY KEY,
 CustomerName VARCHAR(255),
Address VARCHAR(255),
 City VARCHAR(100),
 Country VARCHAR(100)
);
CREATE TABLE DataMart.DimProduct (
 ProductID INT PRIMARY KEY,
 ProductName VARCHAR(255),
 Category VARCHAR(100),
 Price DECIMAL(10, 2)
);
CREATE TABLE DataMart.DimOrders (
 OrderID INT PRIMARY KEY,
 OrderDate DATE,
 CustomerID INT,
 ProductID INT,
 Quantity INT
);
CREATE TABLE DataMart.DimFactsGlobalSales (
 SalesID INT PRIMARY KEY,
```

```
OrderID INT,

ProductID INT,

SalesAmount DECIMAL(10, 2),

SalesDate DATE
);
```

-- Populate tables with extracted data

INSERT INTO DataMart.DimCustomer (CustomerID, CustomerName, Address, City, Country)
SELECT CustomerID, CustomerName, Address, City, Country

INSERT INTO DataMart.DimProduct (ProductID, ProductName, Category, Price)

SELECT ProductID, ProductName, Category, Price

FROM StagingArea.Products;

INSERT INTO DataMart.DimOrders (OrderID, OrderDate, CustomerID, ProductID, Quantity)

SELECT OrderID, OrderDate, CustomerID, ProductID, Quantity

FROM StagingArea.Orders;

INSERT INTO DataMart.DimFactsGlobalSales (SalesID, OrderID, ProductID, SalesAmount, SalesDate)

SELECT SalesID, OrderID, ProductID, SalesAmount, SalesDate

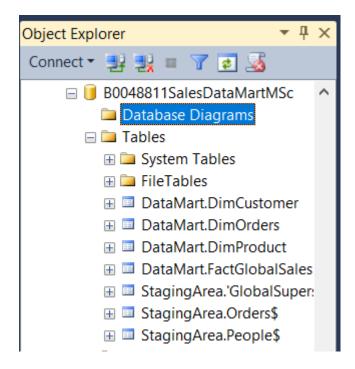


Figure 3: Database, DataMart, and Schema's

Importing Data into Staging Area

To import data into the Staging Area:

- 1. Right-click on the database (B0048811SalesDataMartMSc)
- 2. Select "Tasks" from the dropdown menu
- 3. Choose "Import Data"
- 4. The Import Wizard will appear

Selecting Data Source

In the Import Wizard:

- 1. Choose "Excel" as the data source
- 2. Select the Excel file containing the data
- 3. Proceed to the "Import Data" option

### Successful Importation

The image below illustrates the successful importation of datasets into the Staging Area:

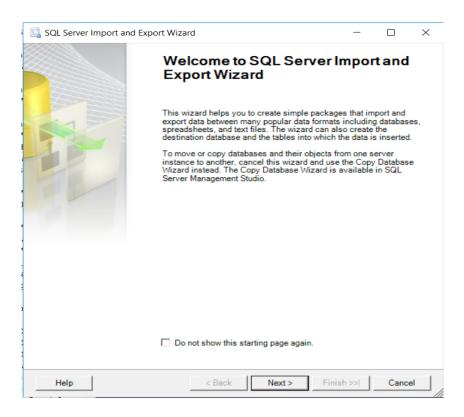


Figure 4: Data Importation Using Import Wizard

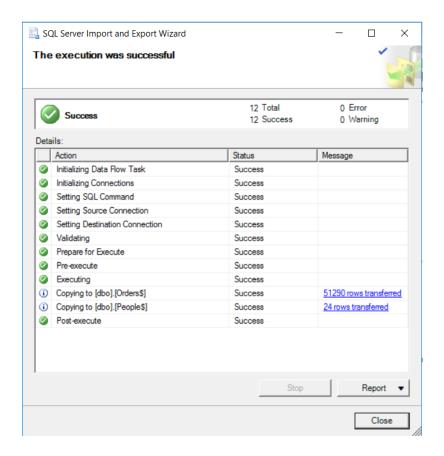


Figure 5: Successfully Importing Datasets

Transferring Data to Staging Area

To transfer imported data into the Staging Area schema:

-- Transfer table (orders\$) into schema (StagingArea)

USE B0048811SalesDataMartMSc

ALTER SCHEMA StagingArea TRANSFER dbo.SampleSuperstore\$;

GO

-- Transfer schema into schema (StagingArea)

ALTER SCHEMA StagingArea TRANSFER StagingArea.Orders\$;

GO

```
ALTER SCHEMA StagingArea TRANSFER dbo.People$;
GO
ALTER SCHEMA StagingArea TRANSFER StagingArea.People$;
GO
ALTER SCHEMA StagingArea TRANSFER dbo.['GlobalSuperstoreReturns2016$'];
GO
ALTER SCHEMA StagingArea TRANSFER StagingArea.['GlobalSuperstoreReturns2016$'];
GO
Creating Dimensions and Extraction
Dimension: Customer
-- Create DimCustomer table in DataMart schema
USE B0048811SalesDataMartMSc
CREATE TABLE DataMart.DimCustomer (
  customerID NVARCHAR(255),
  customerName NVARCHAR(255),
  country NVARCHAR(255),
  region NVARCHAR(255),
  segment NVARCHAR(255),
  CONSTRAINT pkcustomerID PRIMARY KEY (customerID)
);
GO
-- Insert values into DimCustomer table
INSERT INTO DataMart.DimCustomer
SELECT DISTINCT
```

[CustomerID],

```
[CustomerName],
[Country],
[region],
[Segment]

FROM
[B0048811SalesDataMartMSc].[StagingArea].[Orders$];

Viewing DimCustomer Table
-- View all records in DimCustomer table

USE B0048811SalesDataMartMSc

GO

SELECT *

FROM DataMart.DimCustomer;
```

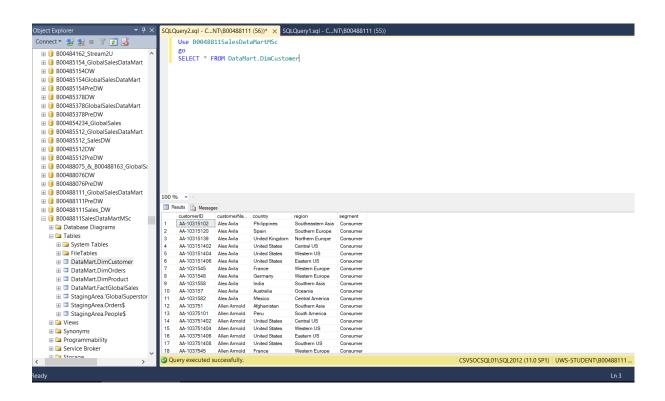


Figure 6: Dimension Customer Table

### **Dimension: Product**

```
-- Create DimProduct table in DataMart schema
USE B0048811SalesDataMartMSc
CREATE TABLE DataMart.DimProduct (
  productID NVARCHAR(255),
  productName NVARCHAR(255),
  category NVARCHAR(255),
  subCategory NVARCHAR(255),
  CONSTRAINT pkproductID PRIMARY KEY (productID)
);
GO
-- Insert values into DimProduct table
INSERT INTO DataMart.DimProduct
SELECT DISTINCT
  [ProductID],
  [ProductName],
  [Category],
  [SubCategory]
FROM
  [B0048811SalesDataMartMSc].[StagingArea].[Orders$];
GO
**Viewing DimProduct Table**
sql
-- View all records in DimProduct table
USE B0048811SalesDataMartMSc
```

#### SELECT \*

FROM DataMart.DimProduct;

٠.,

### Query Results:

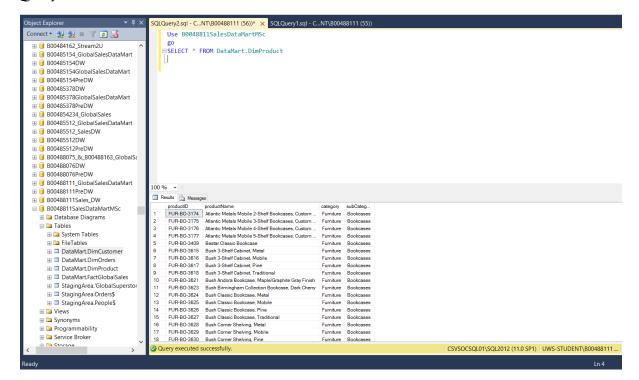


Figure 7: Dimension Product Table content

Here's a rewritten version with improved clarity, structure, and formatting:

#### **Dimension: Orders**

orderDate DATETIME

-- Create DimOrders table in DataMart schema
USE B0048811SalesDataMartMSc
CREATE TABLE DataMart.DimOrders (
orderID NVARCHAR(255) PRIMARY KEY,

```
);
GO
-- Insert values into DimOrders table
INSERT INTO DataMart.DimOrders
SELECT DISTINCT
  [OrderID],
  [OrderDate]
FROM
  [B0048811SalesDataMartMSc].[StagingArea].[Orders$];
GO
Viewing DimOrders Table
-- View all records in DimOrders table
USE B0048811SalesDataMartMSc
GO
SELECT *
FROM DataMart.DimOrders;
Query Results:
```

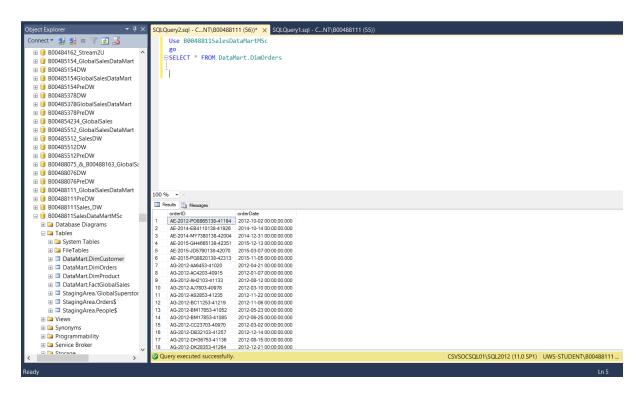


Figure 8: Dimension Orders table details

Facts Table: Global Sales

```
-- Create FactGlobalSales table in DataMart schema
```

USE B0048811SalesDataMartMSc

CREATE TABLE DataMart.FactGlobalSales (

customerID NVARCHAR(255),

productID NVARCHAR(255),

orderID NVARCHAR(255),

Sales MONEY,

Quantity FLOAT,

Discount FLOAT,

Profit MONEY,

ShippingCost FLOAT,

CONSTRAINT FK\_CustomerID FOREIGN KEY (CustomerID) REFERENCES DataMart.DimCustomer(CustomerID),

CONSTRAINT FK\_ProductID FOREIGN KEY (ProductID) REFERENCES DataMart.DimProduct(ProductID),

# CONSTRAINT FK\_OrderID FOREIGN KEY (OrderID) REFERENCES DataMart.DimOrders(OrderID)

);

GO

- View content of FactGlobalSales table

USE B0048811SalesDataMartMSc

GO

**SELECT\*** 

FROM DataMart.FactGlobalSales;

` ' !!

### **Query Results:**

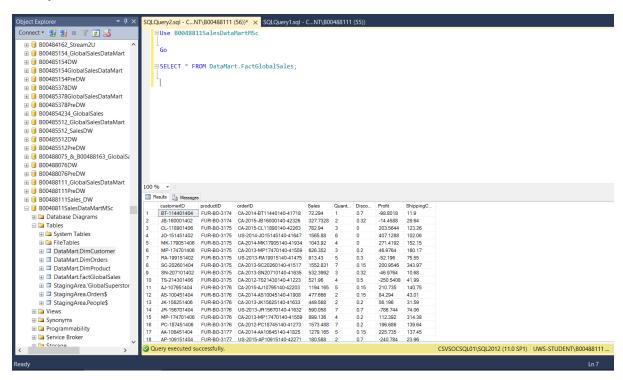


Figure 9: Facts Table Content with Extracted Data

Final Steps: Star Schema Diagram and Data Export

To visualize the DataMart's schema and prepare data for analysis, I:

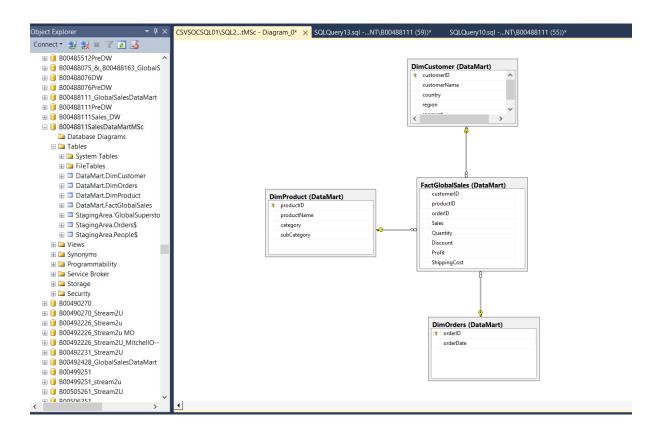
- 1. Created a Star Schema diagram
- 2. Exported data to an Excel file for analysis in Tableau Desktop and Power BI Desktop

### Creating Star Schema Diagram

To create the diagram:

- 1. Selected "Database Diagram" option on the database
- 2. Chose "New Database Diagram" option
- 3. Selected required tables from the pop-up menu
- 4. Added referential integrity option to ensure proper key administration

### Star Schema Diagram:



**Exporting Data to Excel** 

I exported the DataMart data to an Excel file:

- 1. Selected the database
- 2. Right-clicked and chose "Tasks" > "Export Data"
- 3. Chose Excel file format and destination (desktop)

Data Ready for Analysis

The exported data is now ready for analysis in:

- 1. Tableau Desktop
- 2. Power BI Desktop

### **Next steps:**

- 1. Connect to the Excel file in Tableau Desktop and Power BI Desktop
- 2. Create visualizations and dashboards to gain insights from the data

OLAP Queries and Cubes

The following OLAP (Online Analytical Processing) queries utilize relational aggregations to demonstrate rollup relationships on the newly created tables from the extracted datasets.

Query 1: Profit and Sales Analysis

```
SELECT
```

C.CustomerName,

SUM(F.Sales) AS TotalSales,

SUM(F.Profit) AS TotalProfit

*FROM* 

#### DataMart.FactGlobalSales F

#### **INNER JOIN**

DataMart.DimCustomer C ON F.CustomerID = C.CustomerID

GROUP BY

C.CustomerName

WITH ROLLUP

### This query:

- Joins FactGlobalSales and DimCustomer tables
- Groups sales and profit data by customer name
- Uses ROLLUP to display subtotals and grand totals

**OLAP Concepts** 

OLAP queries utilize various dimensions to analyze data:

- ROLLUP: aggregates data to display subtotals and grand totals
- CUBES: multidimensional data structures for analysis
- SLICING: creates a new cube by selecting specific dimensions
- DICING: creates a sub-cube by filtering specific dimensions

Benefits

OLAP queries enable:

- Multidimensional analytical queries
- Financial reporting and forecasting
- Data analysis and visualization

OLAP Query: Profit and Sales Analysis by Customer

USE B0048811SalesDataMartMSc

GO

#### **SELECT**

F.Profit,

F.Sales

**FROM** 

DataMart.DimCustomer C

**INNER JOIN** 

DataMart.FactGlobalSales F ON C.customerID = F.customerID

GROUP BY

CUBE (F.Profit, F.Sales);

**\\**||

### **Query Explanation:**

### This OLAP query:

- Joins DimCustomer and FactGlobalSales tables
- Groups profit and sales data by customer using CUBE aggregation
- Displays aggregated values for profit and sales

### **CUBE Aggregation**:

The CUBE aggregation generates all possible combinations of the specified columns, including:

- Total profit and sales
- Profit by customer

- Sales by customer
- Grand total

Benefits:

# This query enables:

- Multidimensional analysis of profit and sales
- Customer-level insights
- Rolled-up aggregations