



मुंबई विद्यापीठ
University of Mumbai
Re-accredited with A++ Grade
(CGPA 3.65) by NAAC (3rd Cycle 2021)



University of Mumbai
DEPARTMENT OF COMPUTER SCIENCE

M. Sc. (Computer Science) (NEP) Semester-III

2025-2026

Data Visualization

Elective - II

Submitted by
Nayan Naresh Khuje

Seat No.



University of Mumbai

DEPARTMENT OF COMPUTER SCIENCE

CERTIFICATE

This is to certify that the work entered in this journal was done in the University Department of Computer Science laboratory by Mr. Nayan Naresh Khuje Seat No. _____ for the course of **M.Sc. (Computer Science) - Semester III (NEP 2020)** during the academic year **2025- 2026** in a satisfactory manner.

Subject In-charge
Department of Computer Science

Head
Department of Computer Science

External Examiner

INDEX

Sr.no	Name of the practical	Page No.	Date	Signature
1	Create Charts and Reports in Power BI	1-2		
2	Time Intelligence and data analysis Functions with DAX	3-6		
3	Operations on Pinned Reports and Visuals using Power BI	7-8		
4	Create one-dimensional data using series and perform various operations on it	9-12		
5	Perform Reshaping of the hierarchical data and pivoting data frame data	13-14		
6	Connecting and extracting with various data resources in tableau and Perform calculations and creating parameters in Tableau	15-18		
7	Designing Tableau Dashboards for different displays and devices.	19-21		
8	Create a Trend model using data, Analyse it and use it for forecasting	22-23		
9	Creating Geospatial feature maps in Tableau using Geospatial Data	24-25		
10	Create Dashboard and Storytelling using tableau	26-28		

Practical 1

Aim: Create Charts and Reports in Power BI.

Dataset Used: The dataset used for this practical is the Superstore dataset containing fields like Order Date, Sales, Profit, Category, Sub-Category, and more. It spans data from 2016 to 2019.

Chart 1: Stacked Bar Chart

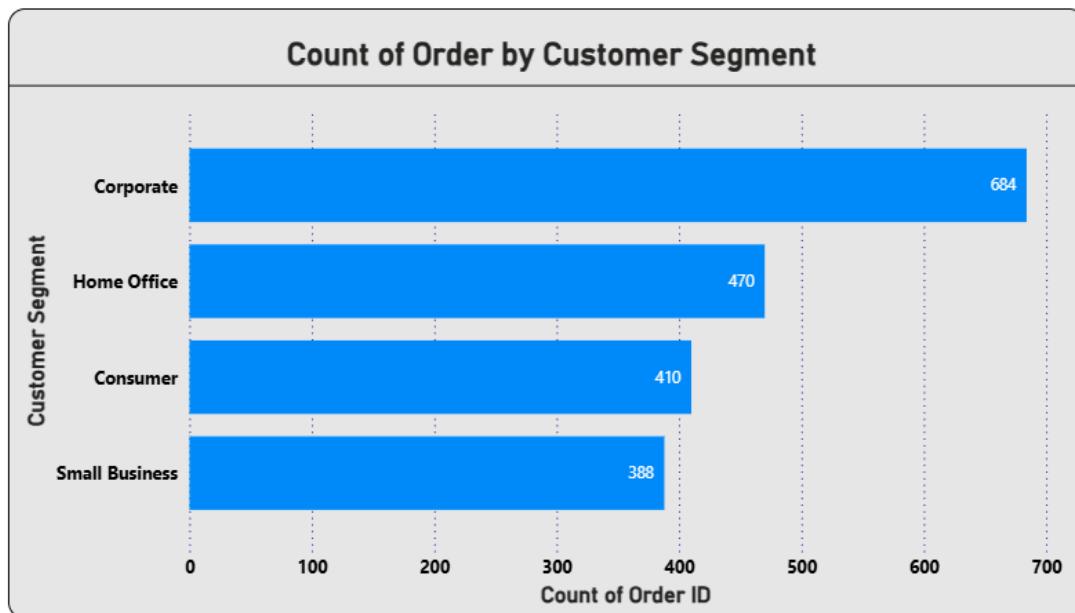


Chart 2: Funnel Chart

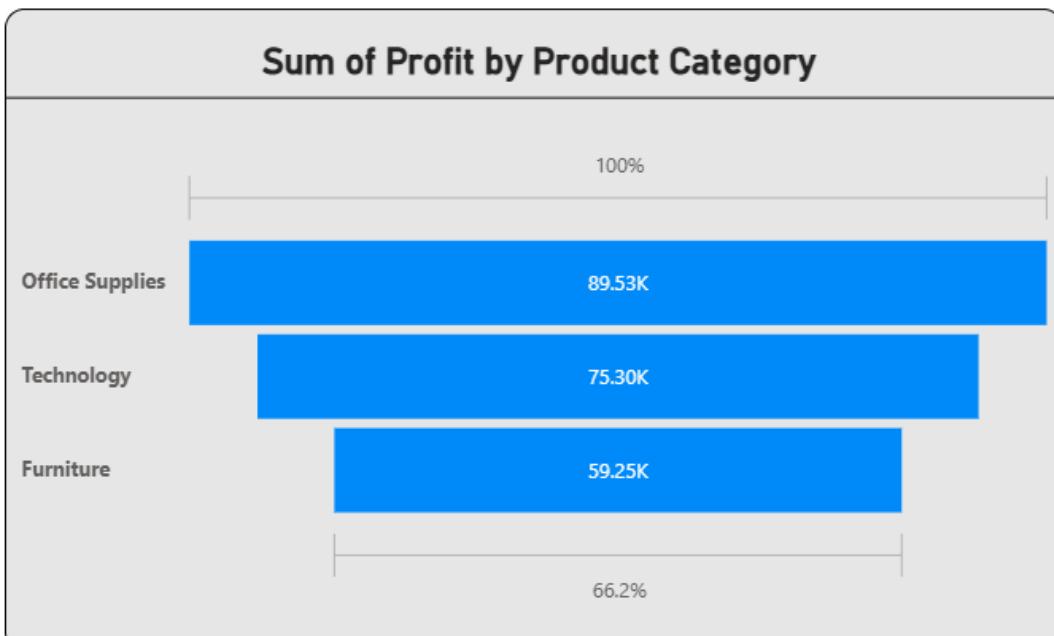


Chart 2: Donut Chart

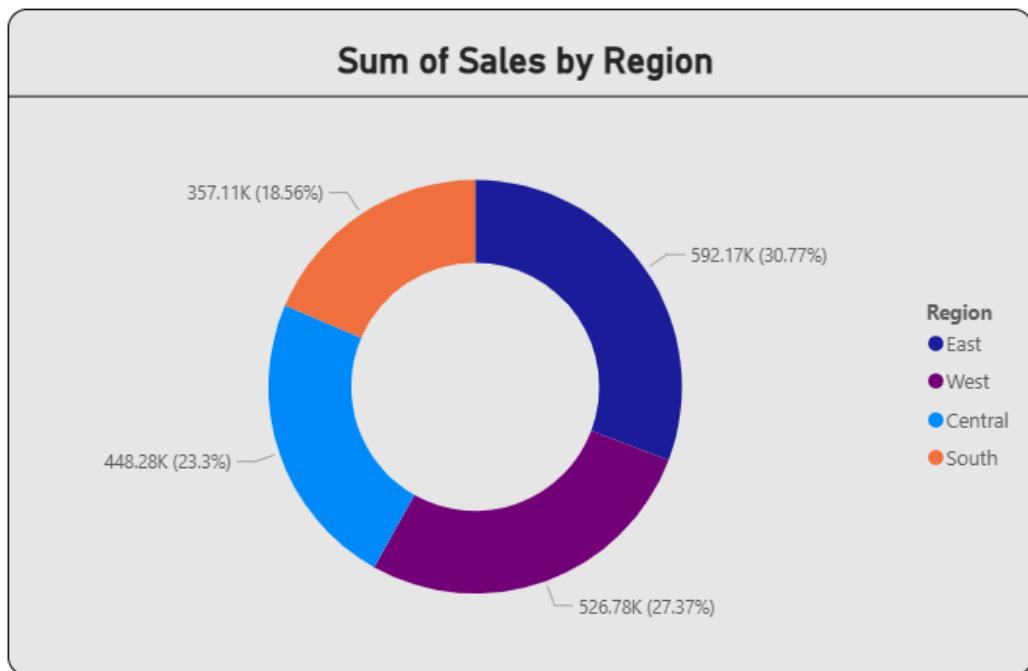


Chart 4: Stacked Column Chart



Practical 2

Aim: Time Intelligence and data analysis Functions with DAX

Dataset Used: The dataset used for this practical is the Superstore dataset containing fields like Order Date, Sales, Profit, Category, Sub-Category, and more. It spans data from 2016 to 2019.

1. Profit Category Classification

Objective: To classify profits into High and Low categories based on a threshold.

```
Profit Category = IF(Orders[Profit] > 100, "High", "Low")
```

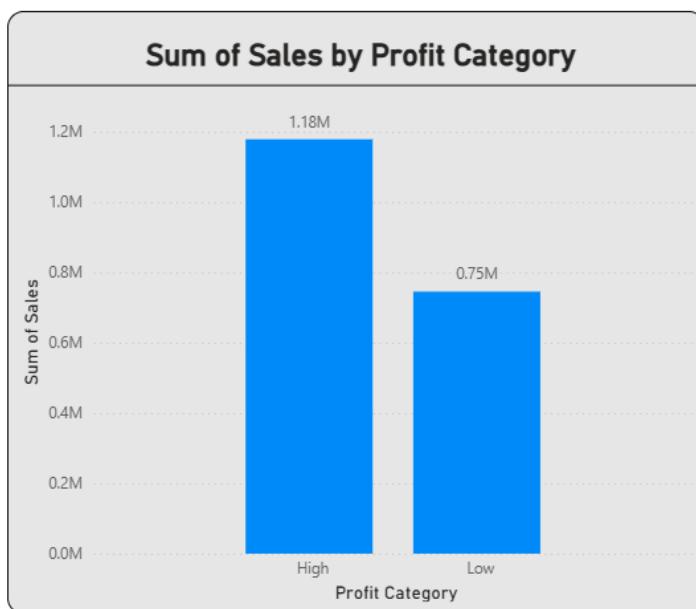
The screenshot shows the Power BI Data View interface. At the top, there are two tabs: 'Structure' and 'Formatting'. Below the tabs, there is a code editor window with two lines of DAX code:

```
1 Profit Category =
2 IF(Orders[Profit] > 100, "High", "Low")
```

Below the code editor is a list of columns from the Superstore dataset. The column 'Profit Category' is highlighted with a green selection bar on its left. The list includes:

- Product Name
- Product Sub-Category
- \sum Profit
- Profit Category
- \sum Quantity ordered new
- Region
- \sum Row ID

Column Chart: Profit Category:



2. Group By and Summarize

Objective: To group data by Category and Sub-Category and calculate total sales and profit.

```
Category Summary =  
SUMMARIZE(  
    Orders, Orders[Category], Orders[Sub-Category], "Total Sales",  
    SUM(Orders[Sales]), "Total Profit", SUM(Orders[Profit]))
```

Matrix: Sales and Profit by Category

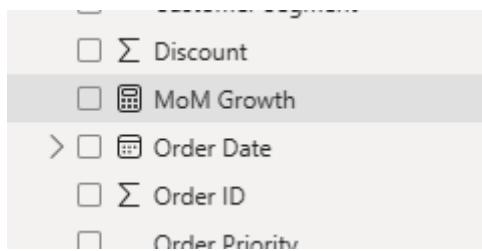
Product Category	Sum of Total Sales	Sum of Total Profit
Furniture	6,60,704.31	59,249.45
Bookcases	1,07,796.09	-930.44
Chairs & Chairmats	2,61,072.73	48,695.84
Office Furnishings	98,070.91	18,724.12
Tables	1,93,764.58	-7,240.07
Office Supplies	5,51,368.62	89,525.01
Appliances	82,201.15	12,594.82
Binders and Binder Accessories	1,85,928.14	59,296.39
Envelopes	10,479.77	-1,194.41
Labels	4,914.82	7,028.16
Paper	55,813.92	7,769.32
Pens & Art Supplies	26,071.61	-257.63
Rubber Bands	1,789.43	-1,544.83
Scissors, Rulers and Trimmers	6,752.18	-1,291.10
Storage & Organization	1,77,417.60	7,124.29
Technology	7,12,264.95	75,303.16
Computer Peripherals	96,261.30	1,698.04
Copiers and Fax	99,069.48	23,990.21
Office Machines	3,18,169.68	8,824.39
Telephones and Communication	1,98,764.49	40,790.51
Total	19,24,337.88	2,24,077.61

3. To analyze sales and profit performance over specific time periods.

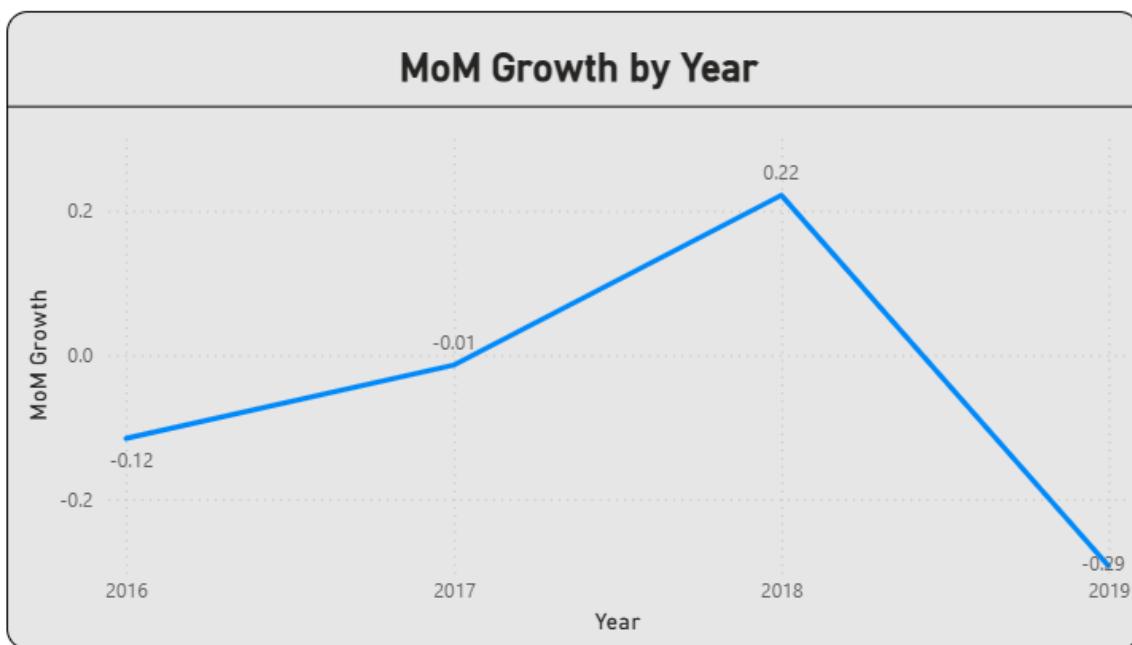
Objective: Month-to-Date (MTD), and Quarter-to-Date (QTD) metrics.

```
MoM Growth =  
VAR MaxDate = MAX(Orders[Order Date])  
VAR CurrentMonthSales =  
    CALCULATE(  
        SUM(Orders[Sales]),  
        MONTH(Orders[Order Date]) = MONTH(MaxDate) && YEAR(Orders[Order  
            Date]) = YEAR(MaxDate)  
    )  
VAR PreviousMonthSales =  
    CALCULATE(  
        SUM(Orders[Sales]),  
        MONTH(Orders[Order Date]) = MONTH(MaxDate) - 1 && YEAR(Orders[Order  
            Date]) = YEAR(MaxDate)  
    )  
RETURN  
DIVIDE(  
    CurrentMonthSales - PreviousMonthSales,  
    PreviousMonthSales,  
    0  
)
```

Structure	Formatting	Properties	measure measure	Calculations
✓ 1 MoM Growth = 2 VAR MaxDate = MAX(Orders[Order Date]) 3 4 VAR CurrentMonthSales = 5 CALCULATE(6 SUM(Orders[Sales]), 7 MONTH(Orders[Order Date]) = MONTH(MaxDate) && 8 YEAR(Orders[Order Date]) = YEAR(MaxDate) 9) 10 11 VAR PreviousMonthSales = 12 CALCULATE(13 SUM(Orders[Sales]), 14 MONTH(Orders[Order Date]) = MONTH(MaxDate) - 1 && 15 YEAR(Orders[Order Date]) = YEAR(MaxDate) 16) 17 18 RETURN 19 DIVIDE(20 CurrentMonthSales - PreviousMonthSales, 21 PreviousMonthSales, 22 0 23) 24	Sum of Sales			



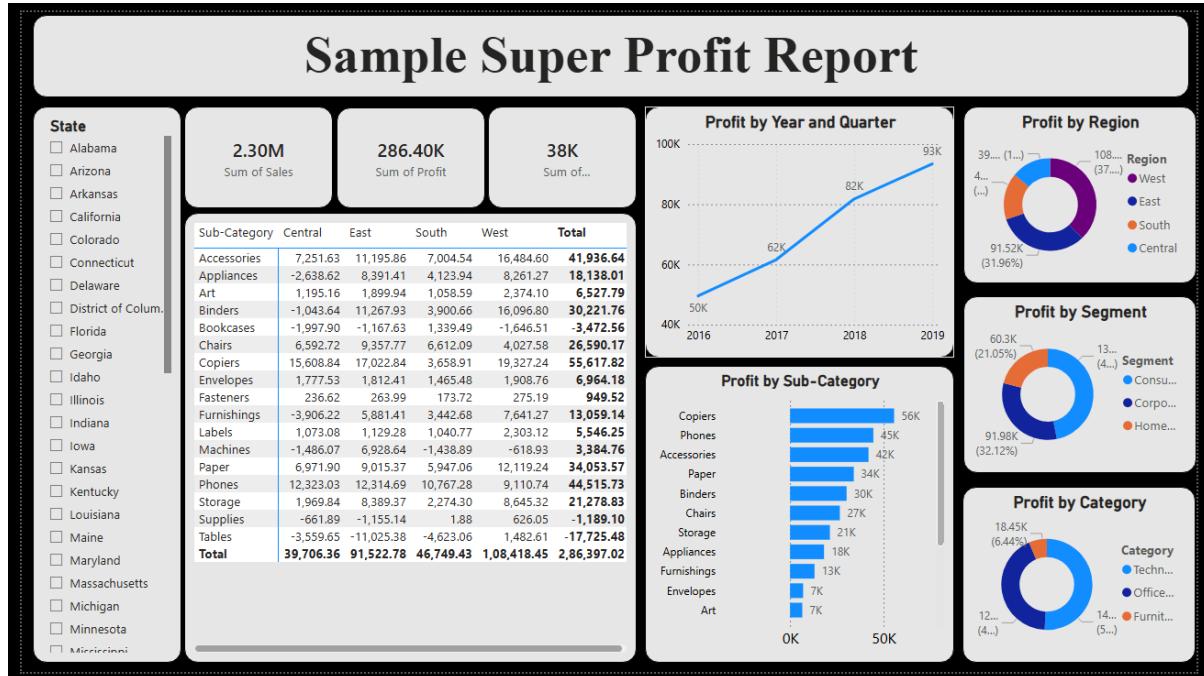
Line Chart: Month-over-Month Growth



Practical 3

Aim: Operations on Pinned Reports and Visuals using Power BI.

Sample superstore report:



Dataset Used: The dataset utilized for this practical is the Superstore Sample Dataset, which is used to design and analyze the Sample Super Profit Report dashboard.

The dataset includes the following fields:

- Order Information: Order ID, Order Date, Ship Date, Ship Mode
- Customer Information: Customer ID, Customer Name, Segment
- Geographical Information: Country/Region, City, State, Postal Code, Region
- Product Information: Product ID, Category, Sub-Category, Product Name
- Sales Metrics: Sales, Quantity, Discount, Profit

Dashboard-Based Steps and Visualizations

1. KPI Cards (Used in Dashboard)

Fields Used: Sales, Profit, Quantity

Purpose: To display key performance indicators at the top of the dashboard for quick overview.

Outcome (From My Dashboard)

- Sales: 2.30M, Profit: 286.40K, Quantity: 38K

2. Profit by Sub-Category

Visualization Type: Bar Chart

- Steps:
1. Drag Sub-Category to the Axis field.
 2. Drag Profit to the Values field.

Purpose: To compare profit performance across different product sub-categories.

3. Profit by Year and Quarter

Visualization Type: Line Chart

- Steps:
1. Drag Order Date to the Axis field.
 2. Set the date hierarchy to Year and Quarter.
 3. Drag Profit to the Values field.

Observation (From Dashboard): Profits were highest in specific quarters of 2019, indicating peak business periods.

4. Profit by Region

Visualization Type: Donut Chart

- Steps:
1. Drag Region to the Legend field.
 2. Drag Profit to the Values field.

Observation: West Region contributed the most profit (37.84%), Central Region had the least contribution (13.9%)

5. Profit by Segment

Visualization Type: Donut Chart

- Steps:
1. Drag Segment to the Legend field.
 2. Drag Profit to the Values field.

Observation: Consumer Segment performed best with 46.83% contribution, Home Office Segment contributed the least (21.05%)

6. Profit by Category

Visualization Type: Donut Chart

- Steps:
1. Drag Category to the Legend field.
 2. Drag Profit to the Values field.

Observation: Technology Category showed the highest profit (50.79%), Furniture Category had the lowest profit share (6.35%)

Practical No 4

Aim: Create one-dimensional data using series and perform various operations on it

To create a one-dimensional data structure using a Pandas Series and perform various operations on it such as: Mathematical operations, Aggregation functions, Indexing and slicing, Conditional selection, Applying custom functions

Software / Tools Used: Python, Pandas library, NumPy library

Theory: A Pandas Series is a one-dimensional labelled array capable of holding data of any type such as integers, floats, or strings. It supports various data manipulation operations which make it suitable for data analysis tasks.

```
[2]: import pandas as pd
import numpy as np

# Creating a Pandas Series
data = [10, 20, 30, 40, 50]
series = pd.Series(data)
```

```
[3]: # Displaying the created series
print("Original Series:")
print(series)
```

```
Original Series:
0    10
1    20
2    30
3    40
4    50
dtype: int64
```

```
[4]: # Accessing elements
print("\nAccessing elements at index 2 and 4:")
print(f"Element at index 2: {series[2]}")
print(f"Element at index 4: {series[4]}")
```

```
Accessing elements at index 2 and 4:
Element at index 2: 30
Element at index 4: 50
```

```
[5]: # Mathematical Operations
print("\nMathematical Operations")

# Adding a constant to each element
print(f"Adding 5 to each element:\n{series + 5}")
```

```
Mathematical Operations:
Adding 5 to each element:
0    15
1    25
2    35
3    45
4    55
dtype: int64
```

```
[6]:  
# Subtracting a constant from each element  
print(f"Subtracting 10 from each element:\n{series - 10}")
```

```
Subtracting 10 from each element:  
0    0  
1    10  
2    20  
3    30  
4    40  
dtype: int64
```

```
[7]: # Multiplying each element by 2  
print(f"Multiplying each element by 2:\n{series * 2}")
```

```
Multiplying each element by 2:  
0    20  
1    40  
2    60  
3    80  
4   100  
dtype: int64
```

```
[8]: # Dividing each element by 5  
print(f"Dividing each element by 5:\n{series / 5}")
```

```
Dividing each element by 5:  
0    2.0  
1    4.0  
2    6.0  
3    8.0  
4   10.0  
dtype: float64
```

```
[9]: # Applying a mathematical function (Square)  
print(f"Square of each element:\n{series**2}")
```

```
Square of each element:  
0    100  
1    400  
2    900  
3   1600  
4   2500  
dtype: int64
```

```
[10]: # Aggregation Operations
print("\nAggregation Operations:")

# Sum of all elements
print(f"Sum of elements: {series.sum()}")
# Mean of the elements
print(f"Mean of elements: {series.mean()}")
# Minimum element
print(f"Minimum element: {series.min()}")
# Maximum element
print(f"Maximum element: {series.max()}")
```

```
Aggregation Operations:
Sum of elements: 150
Mean of elements: 30.0
Minimum element: 10
Maximum element: 50
```

```
[11]: # Indexing and Slicing
print("\nIndexing and Slicing:")

# Slicing elements from index 1 to 3
print(f"Slicing from index 1 to 3:\n{series[1:4]}")
```

```
Indexing and Slicing:
Slicing from index 1 to 3:
1    20
2    30
3    40
dtype: int64
```

```
[12]: # Selecting a specific element with condition (greater than 20)
print(f"Elements greater than 20:\n{series[series > 20]}")
```

```
Elements greater than 20:
2    30
3    40
4    50
dtype: int64
```

```
[13]: # Applying a custom function
print("\nApplying Custom Function:")
# Define a custom function to subtract 3 from each element
def subtract_three(x):
    return x - 3
# Applying the function to each element
print(f"Subtracting 3 from each element using custom function:\n{series.apply(subtract_three)}")
```

```
Applying Custom Function:
Subtracting 3 from each element using custom function:
0    7
1   17
2   27
3   37
4   47
dtype: int64
```

```
[14]: # Sorting the elements
print("\nSorting the elements in ascending order:")
print(series.sort_values())
```

```
Sorting the elements in ascending order:
0    10
1    20
2    30
3    40
4    50
dtype: int64
```

```
[16]: # Checking for NaN values (No NaN here, but can be tested for)
print("\nChecking for NaN values:")
print(series.isna())
```

```
Checking for NaN values:
0    False
1    False
2    False
3    False
4    False
dtype: bool
```

Practical 5

Aim: Perform Reshaping of the hierarchical data and pivoting data frame data

To perform reshaping operations such as pivoting, melting, stacking, and unstacking using a hierarchical Data Frame structure in Pandas.

```
[1]: import pandas as pd
import numpy as np

[2]: # Create a sample DataFrame for demonstration
data = {
    'Date': ['2024-01-01', '2024-01-01', '2024-01-02', '2024-01-02', '2024-01-03', '2024-01-03'],
    'City': ['New York', 'Los Angeles', 'New York', 'Los Angeles', 'New York', 'Los Angeles'],
    'Temperature': [32, 75, 30, 72, 28, 70],
    'Humidity': [80, 10, 85, 15, 90, 12]
}

[3]: # Create DataFrame
df = pd.DataFrame(data)
print("Original DataFrame:")
print(df)

Original DataFrame:
   Date        City  Temperature  Humidity
0 2024-01-01  New York        32       80
1 2024-01-01  Los Angeles      75       10
2 2024-01-02  New York        30       85
3 2024-01-02  Los Angeles      72       15
4 2024-01-03  New York        28       90
5 2024-01-03  Los Angeles      70       12
```

1. Pivoting Data Frame using pivot()

```
[4]: # Pivot the DataFrame to reshape it, setting 'Date' as index and 'City' as columns
pivot_df = df.pivot(index='Date', columns='City', values=['Temperature', 'Humidity'])
print("\nPivoted DataFrame:")
print(pivot_df)

Pivoted DataFrame:
              Temperature          Humidity
City        Los Angeles  New York  Los Angeles  New York
Date
2024-01-01        75        32        10       80
2024-01-02        72        30        15       85
2024-01-03        70        28        12       90
```

2. Pivoting DataFrame using pivot_table() with aggregation

```
[5]: # Use pivot_table() to aggregate the data if multiple values exist for the same index/column combination
data_agg = {
    'Date': ['2024-01-01', '2024-01-01', '2024-01-02', '2024-01-02', '2024-01-03', '2024-01-03'],
    'City': ['New York', 'New York', 'Los Angeles', 'Los Angeles', 'New York', 'New York'],
    'Temperature': [32, 35, 72, 74, 28, 30],
    'Humidity': [80, 78, 15, 18, 90, 92]
}
df_agg = pd.DataFrame(data_agg)
pivot_table_df = df_agg.pivot_table(index='Date', columns='City', values=['Temperature', 'Humidity'],
aggfunc='mean')

print("\nPivot Table DataFrame (with aggregation):")
print(pivot_table_df)

Pivot Table DataFrame (with aggregation):
              Humidity          Temperature
City        Los Angeles  New York  Los Angeles  New York
Date
2024-01-01        NaN        79.0        NaN       33.5
2024-01-02        16.5       NaN        73.0       NaN
2024-01-03        NaN        91.0       NaN       29.0
```

3. Melting DataFrame (Unpivoting)

```
[6]: # Melt the DataFrame to long format, turning columns into rows
melted_df = df.melt(id_vars=['Date', 'City'], value_vars=['Temperature', 'Humidity'],
                     var_name='Metric', value_name='Value')
print("\nMelted DataFrame:")
print(melted_df)
```

Melted DataFrame:

	Date	City	Metric	Value
0	2024-01-01	New York	Temperature	32
1	2024-01-01	Los Angeles	Temperature	75
2	2024-01-02	New York	Temperature	30
3	2024-01-02	Los Angeles	Temperature	72
4	2024-01-03	New York	Temperature	28
5	2024-01-03	Los Angeles	Temperature	70
6	2024-01-01	New York	Humidity	80
7	2024-01-01	Los Angeles	Humidity	10
8	2024-01-02	New York	Humidity	85
9	2024-01-02	Los Angeles	Humidity	15
10	2024-01-03	New York	Humidity	90
11	2024-01-03	Los Angeles	Humidity	12

1. Stacking and Unstacking DataFrame (Hierarchical Reshaping)

```
[7]: # First, set 'Date' and 'City' as a MultiIndex for hierarchical structure
df_stacked = df.set_index(['Date', 'City'])

# Stack the DataFrame (compress columns into a single column level)
stacked_df = df_stacked.stack()
print("\nStacked DataFrame (Hierarchical):")
print(stacked_df)
```

Stacked DataFrame (Hierarchical):

Date	City		
2024-01-01	New York	Temperature	32
		Humidity	80
	Los Angeles	Temperature	75
		Humidity	10
2024-01-02	New York	Temperature	30
		Humidity	85
	Los Angeles	Temperature	72
		Humidity	15
2024-01-03	New York	Temperature	28
		Humidity	90
	Los Angeles	Temperature	70
		Humidity	12

dtype: int64

```
[8]: # Unstack the DataFrame (reverse stacking, expand index back into columns)
unstacked_df = stacked_df.unstack()

print("\nUnstacked DataFrame (Reverted to original columns):")
print(unstacked_df)
```

Unstacked DataFrame (Reverted to original columns):

Date	City	Temperature	Humidity
2024-01-01	Los Angeles	75	10
	New York	32	80
2024-01-02	Los Angeles	72	15
	New York	30	85
2024-01-03	Los Angeles	70	12
	New York	28	90

Practical 6

Aim: Connecting and extracting with various data resources in tableau and Perform calculations and creating parameters in Tableau.

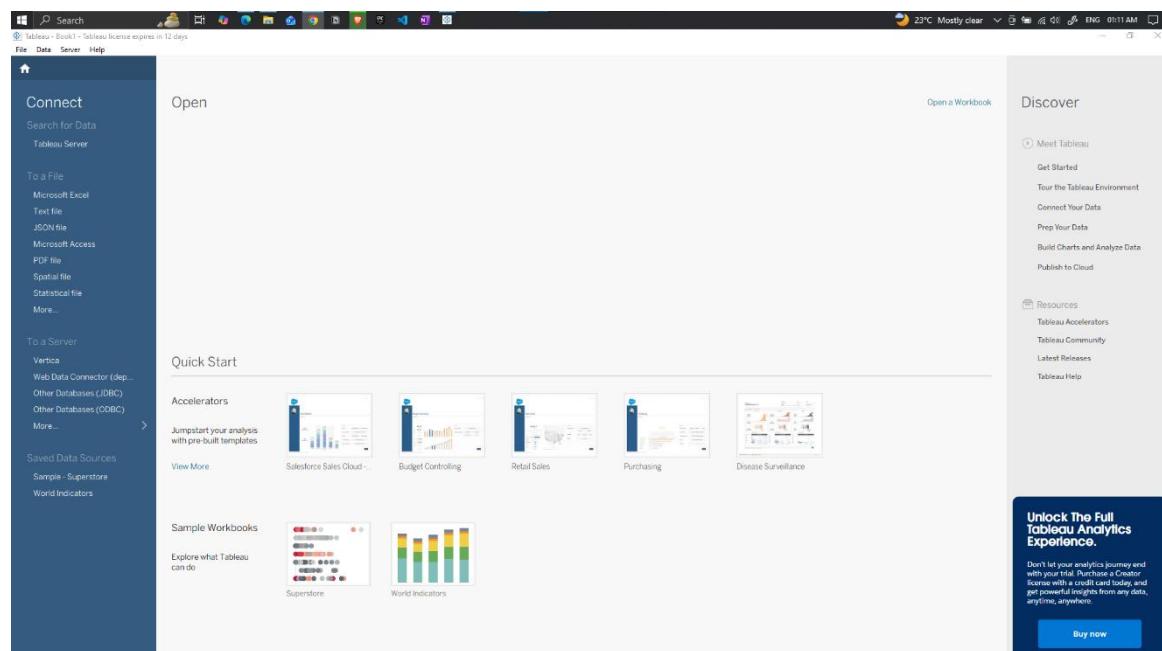
Step 1: Connect to Data Source 1. Open Tableau Desktop.

2. Click on "Connect" on the left-hand side and choose the desired data source:

For a local file: Choose Excel, CSV, or other file types. For a database: Choose from MySQL, PostgreSQL, or other supported databases.

3. Once connected, drag the required table or sheet into the Data Pane.

4. Preview the data to ensure the table contains relevant columns (e.g., Category, Sub-Category, Sales, etc.).



This screenshot shows a specific worksheet titled 'Orders (Superstore_Sample)'. The top navigation bar includes 'File', 'Data', 'Server', 'Window', 'Help', 'Connections' (with 'Superstore_Sample' selected), 'Sheets' (with 'Orders' selected), and 'Filters' (with 'Live'). The main workspace displays a data grid for the 'Orders' table with 25 fields and 1952 rows. The columns include Row ID, Order Priority, Discount, Unit Price, Shipping Cost, Customer ID, Customer Name, Ship Mode, and Customer Segment. Below the grid, there are two callout boxes: 'Relate tables' (drag tables near Orders to relate them) and 'Add a base table' (drag out additional base tables for multi-fact analysis). The bottom left shows the 'Fields' pane with columns for Type, Field Name, Physical Table, and Remote. The bottom right shows the 'Data Source' pane with a 'Go to Worksheet' button and a list of sheets: 'Orders' (selected), 'Returns', 'Users', 'New Union', and 'New Table Extension'.

Step 2: Extract Data 1. Go to the Data Source tab.

2. Click on the Extract button in the top-right corner.

3. Select the required columns to optimize performance (e.g., Category, Sub-Category, Sales, Profit). Save the extract file (.hyper) locally by clicking Extract.

The screenshot shows the Tableau interface with the 'Data' tab selected. On the left, the 'Connections' pane shows 'Superstore_Sample Microsoft Excel'. The 'Sheets' pane lists 'Orders', 'Returns', 'Users', 'New Union', and 'New Table Extension'. The main area displays the 'Orders' sheet with 25 fields and 1952 rows. A 'Save Extract As' dialog box is open over the interface, showing the save location as 'This PC > Documents > My Tableau Repository > Datasources'. The file name is 'Orders (Superstore_Sample)' and the type is 'Tableau Data Extract (*.hyper)'. The 'Save' button is highlighted.

Step 3: Top N subcategories dynamic

1. Create a parameter for N with current value 5:

2. Create a calculated field for ranking:

```
RANK(SUM([Sales]), 'desc')
```

2. Create another calculated field to filter the top N:

```
[Sales Rank] <= [Top N]
```

Drop field here

Create Parameter

Name
Top N

Properties

Data type Integer Display format 5

Current value 5 Value when workbook opens Current value

Allowable values

All List Range

Range of values

Minimum 1 Fixed
 Maximum 20 When workbook opens

Step size 1 Add values from ▾

Cancel **OK**

Sales Rank

RANK(SUM([Sales]), 'desc')

The calculation is valid.

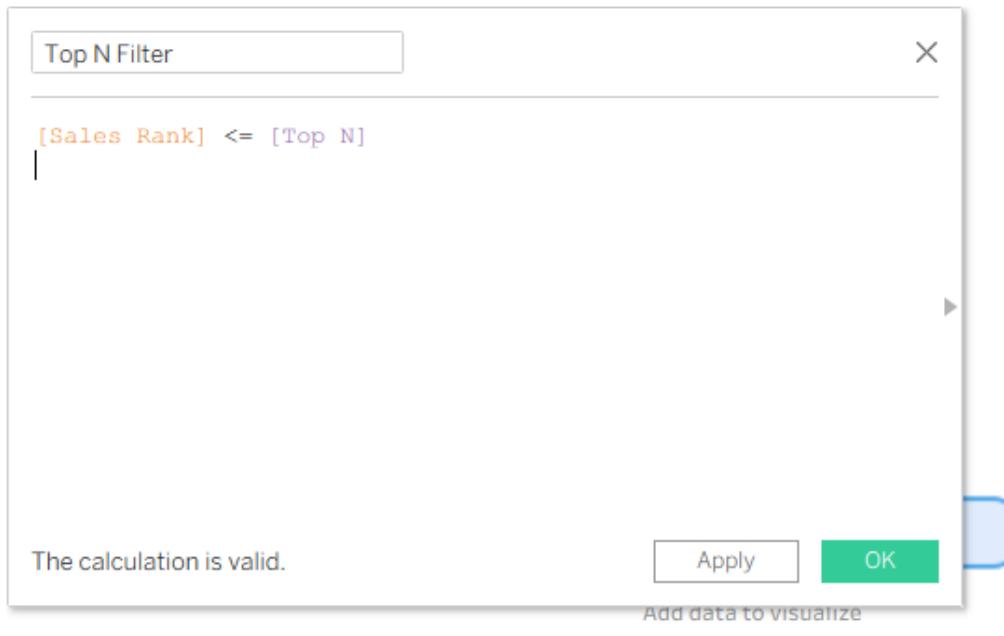
Default Table Calculation

Apply **OK**

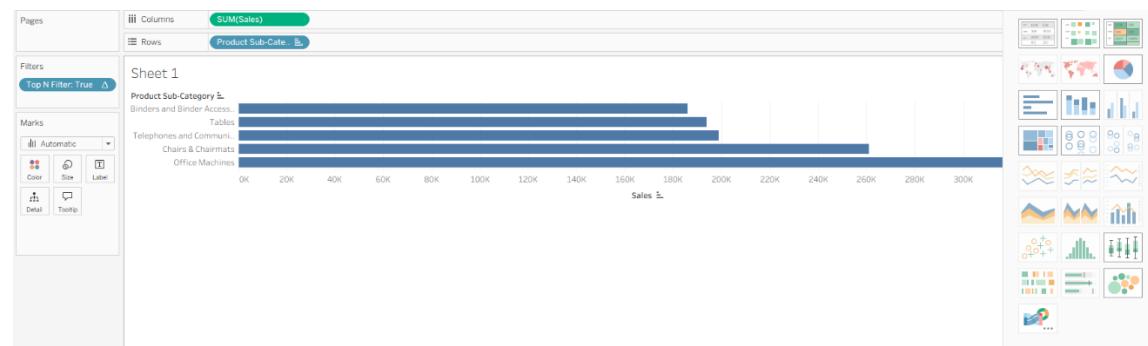
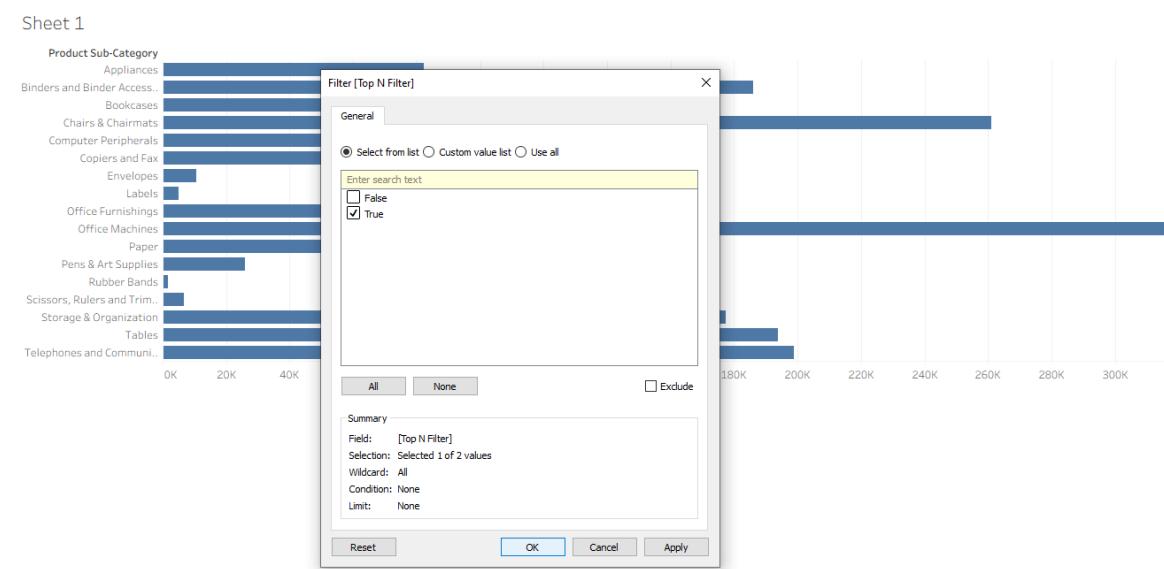
Add data to visualize

Double-click or drag fields from the data pane

Step 3.: Create Calculated Field to Filter Top N



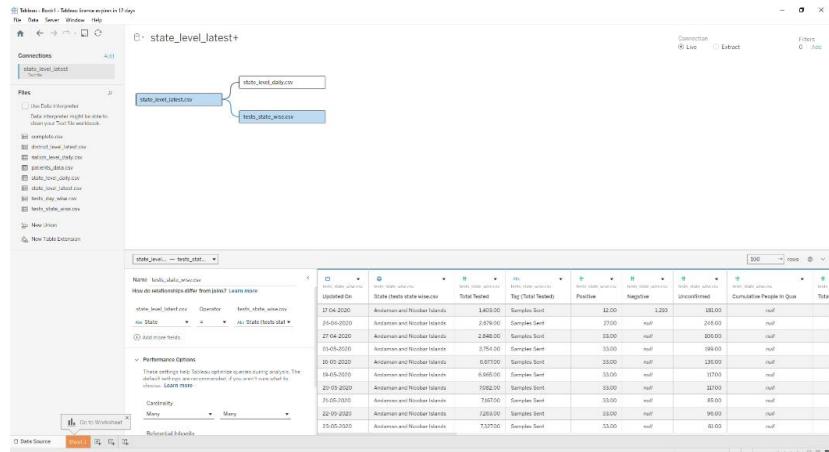
Step 3: Apply Top N Filter to Visualization



Practical 7

Aim: Designing Tableau Dashboards for different displays and devices.

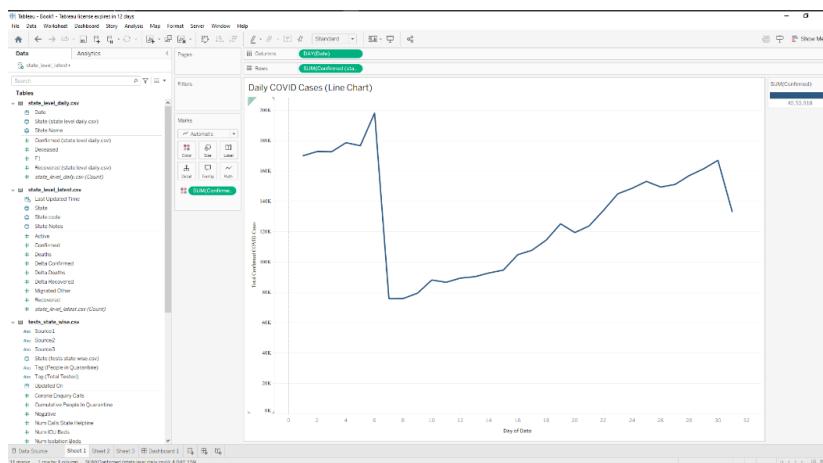
Integrated different dataset for making dashboard



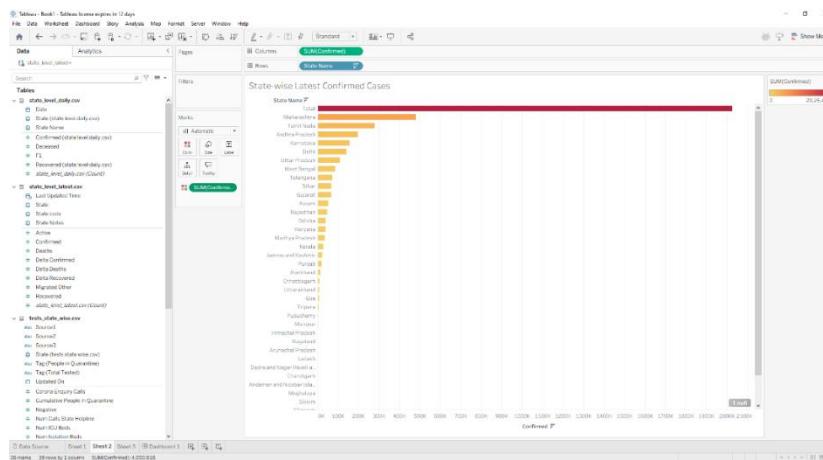
Create multiple sheets for dashboard:

Sheet 1: Daily COVID Cases (Line Chart)

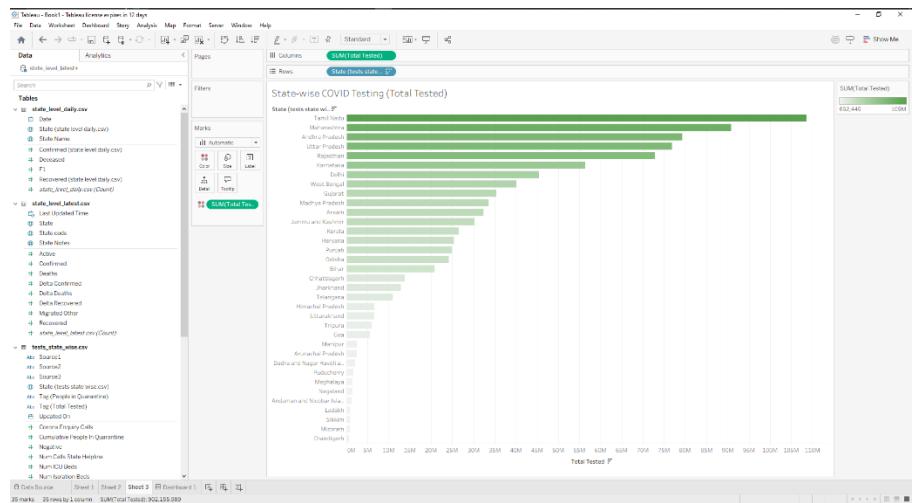
Purpose: To show Time-wise COVID case trend.



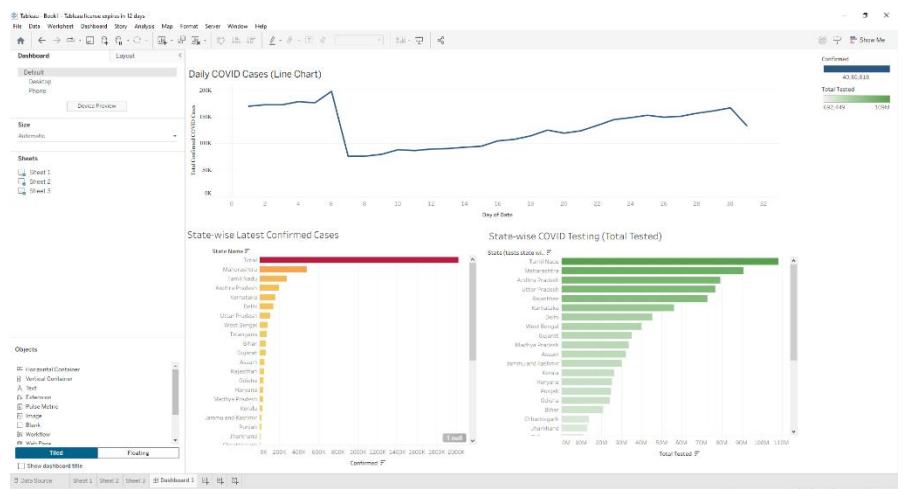
Sheet 2: State-wise Latest Confirmed Cases (Bar Chart)



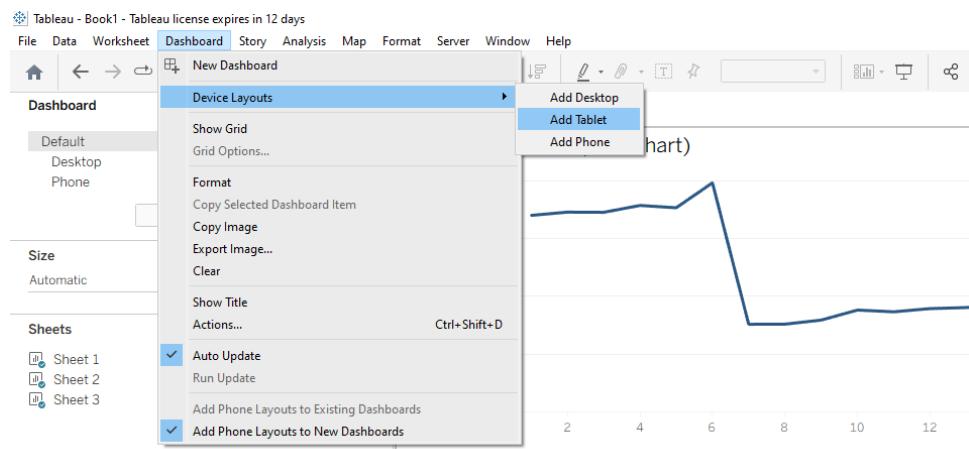
Sheet 3: State-wise COVID Testing (Bar Chart)



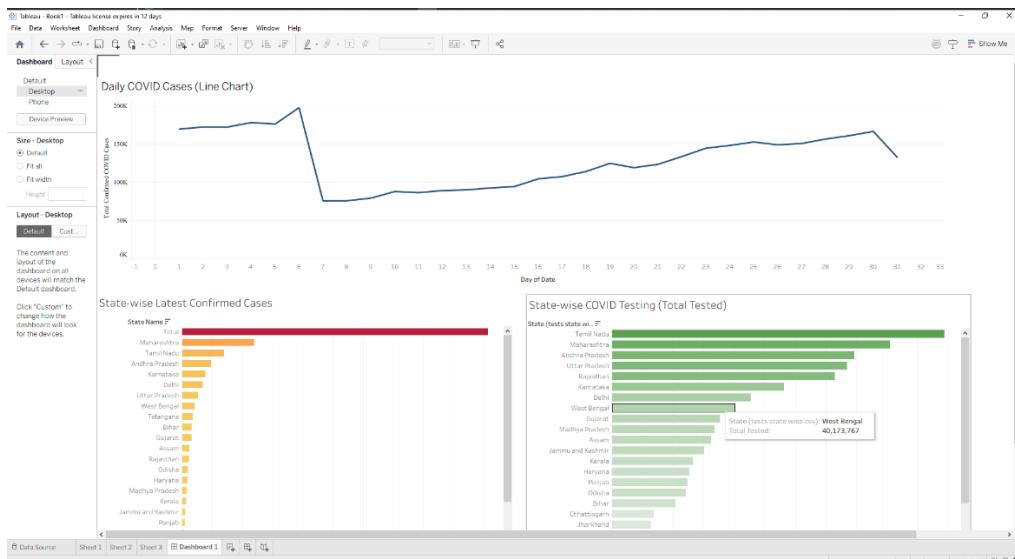
Combine all three sheets for dashboard



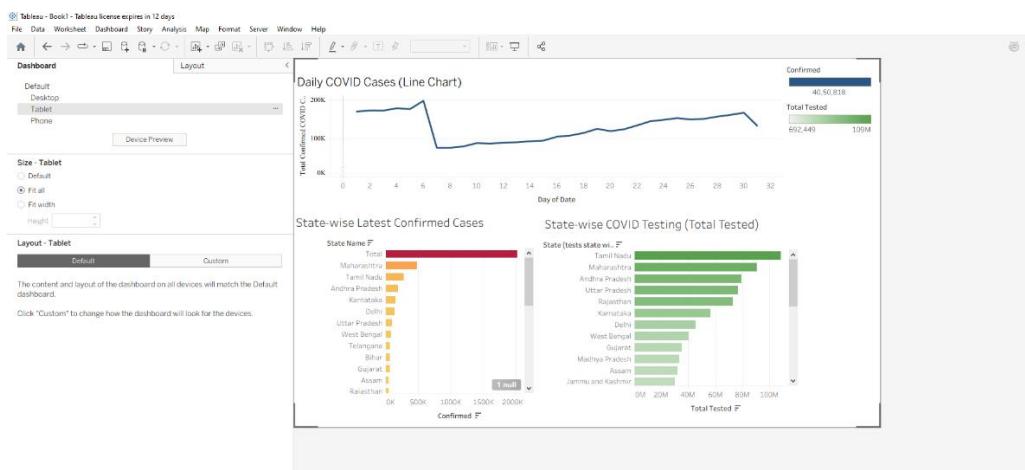
For selecting the dashboard into different layout go to the dashboard option in the navigator bar and select device layout



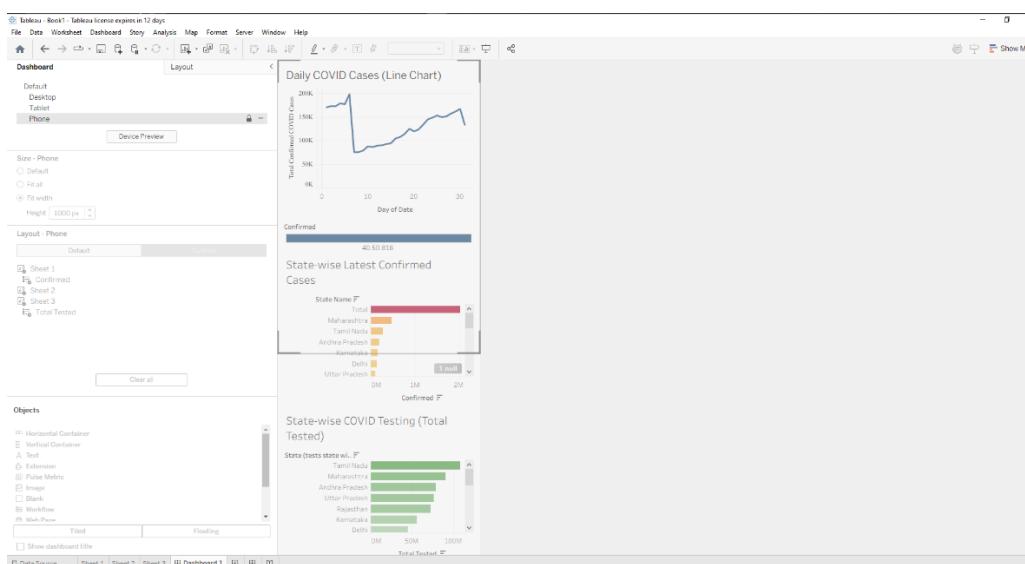
Dashboard in Desktop layout



Dashboard in Tablet layout



Dashboard in phone layout



Practical 8

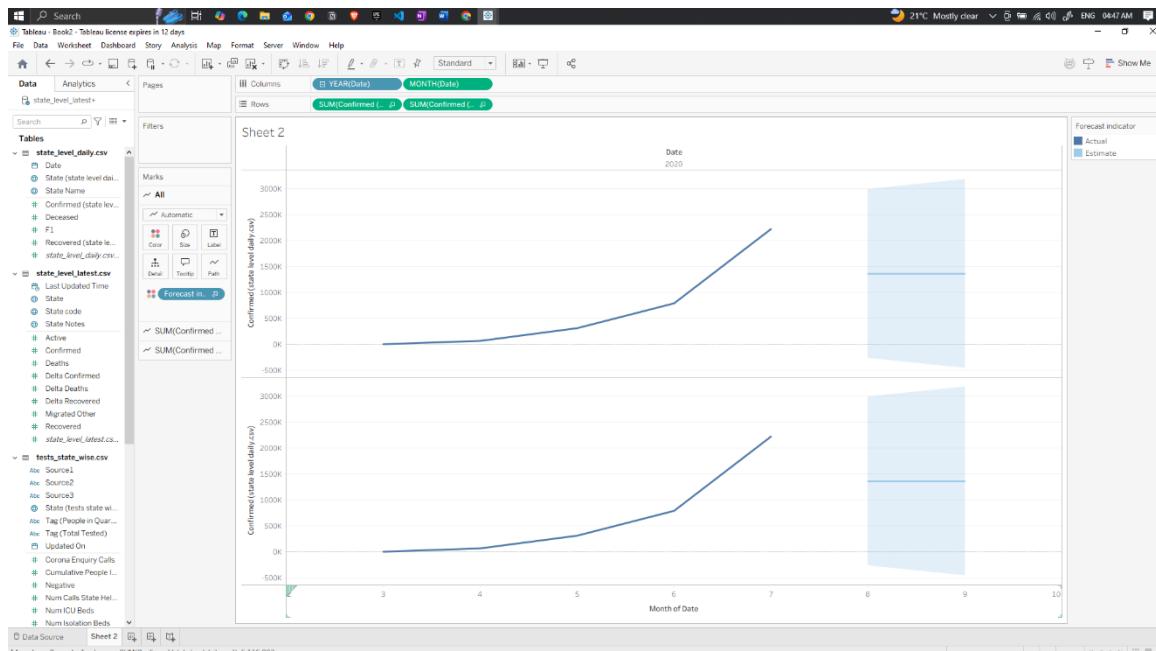
Aim: Create a Trend model using data, Analyse-it and use it for forecasting.

Open Tableau and integrate the dataset of covid 19 cases and create sheet for Forecasting of Positive cases

The screenshot shows the Tableau Data Source interface. A connection named "state_level_latest" is selected. The "Files" section lists several CSV files: complete.csv, district_level_latest.csv, nation_level_daily.csv, patients_data.csv, state_level_daily.csv, state_level_latest.csv, tests_day_wise.csv, and tests_state_wise.csv. A network diagram shows "state_level_latest.csv" connected to "state_level_daily.csv" and "tests_state_wise.csv".

Select all the files from the data plan and for the network mode go to analytics and select furcating option

The screenshot shows the Tableau Analytics shelf. The "Model" section is expanded, and the "Forecast" option is highlighted with a gray background. Other options in the Model section include Average with 95% CI, Median with 95% CI, Trend Line, Cluster, and Box Plot. The "Custom" section includes Reference Line, Reference Band, Distribution Band, and Box Plot. The "Summarize" section includes Constant Line, Average Line, Median with Quartiles, and Totals.



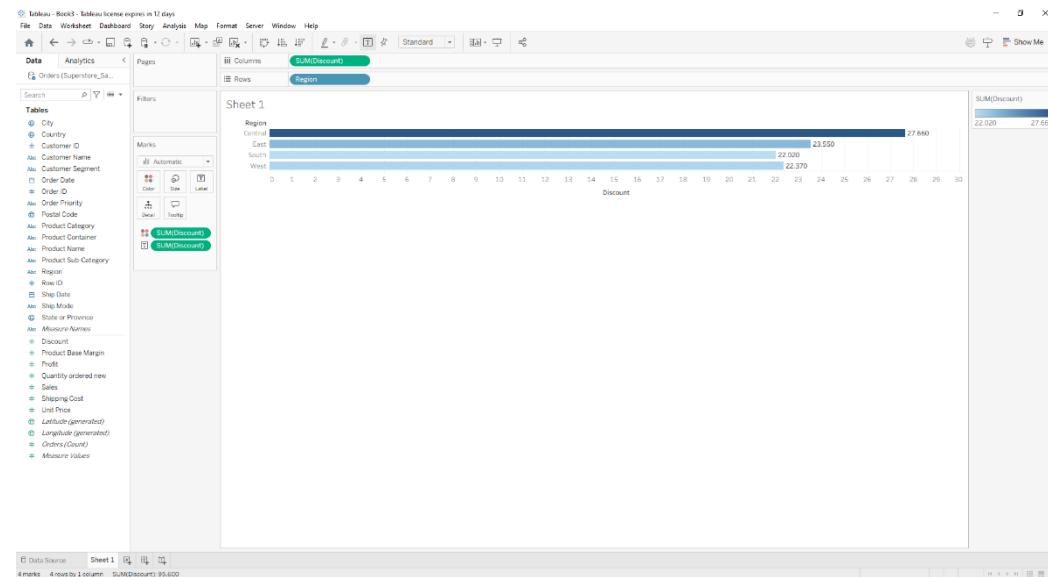
Practical 9

Aim: Creating Geospatial feature maps in Tableau using Geospatial Data.

Open the tableau and integrate sample-superstore dataset into the data panel

Drag the 'Discount' field into the Columns section, the 'Region' field into the Rows section, and add the sum of 'Discount' into both the Color and Label sections.

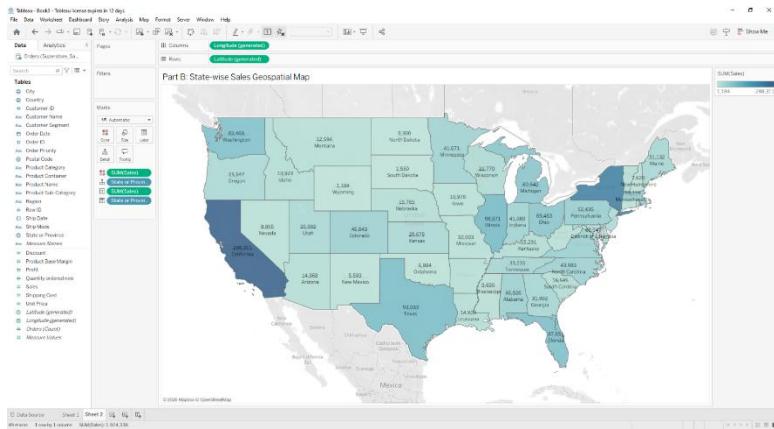
Part A: Region-wise Discount Visualization



Part B: State-wise Sales Geospatial Map

Open a new sheet. Create a state-wise sales map by following these steps:

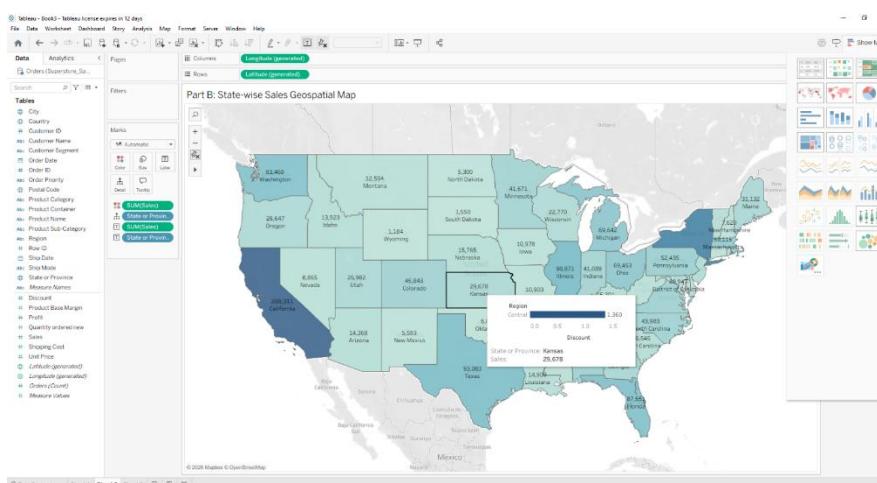
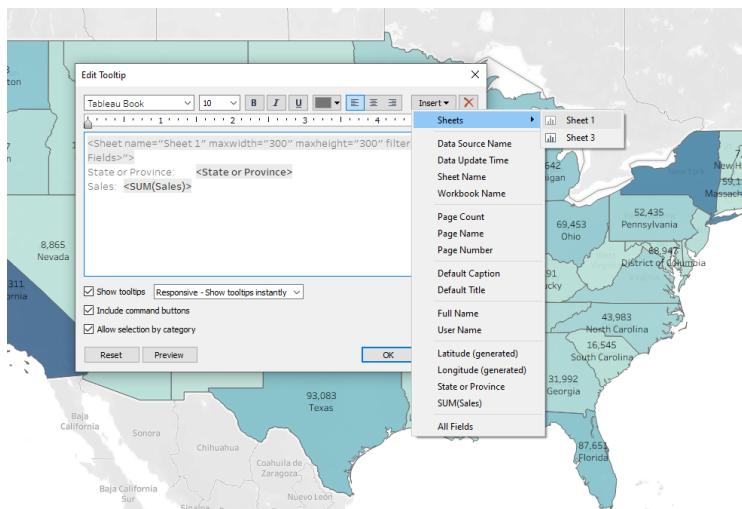
- Drag the 'Longitude' field into the Columns section.
- Drag the 'Latitude' field into the Rows section.
- Drag the sum of 'Sales' into both the Color and Label sections.
- Drag the 'State' field into both the Detail and Label sections.



Part C: Integrating Region-wise Discounts using Tooltips

To integrate region-wise discounts into the map graph using tooltips, follow these steps:

1. Select the Tooltips option in the Marks section.
2. Click on the Insert option.
3. Select the region-wise discount sheet for integration. This should help you display the region-wise discounts when you hover over the map.



Practical 10

Aim: Create Dashboard and Storytelling using tableau.

Open Tableau and integrate the dataset of covid 19 cases and create sheet for Forecasting of Positive cases

Updated On	State	Total Tested	Samples Sent	Positive	Negative	Unconfirmed	Cumulative People In Quarantine
17-04-2020	Andaman and Nicobar Islands	1403.00	Samples Sent	32.00	1,210	181.00	null
24-04-2020	Andaman and Nicobar Islands	2,679.00	Samples Sent	270.00	null	246.00	null
27-04-2020	Andaman and Nicobar Islands	2,848.00	Samples Sent	33.00	null	106.00	null
01-05-2020	Andaman and Nicobar Islands	3,754.00	Samples Sent	33.00	null	199.00	null
16-05-2020	Andaman and Nicobar Islands	6,677.00	Samples Sent	33.00	null	136.00	null
19-05-2020	Andaman and Nicobar Islands	6,965.00	Samples Sent	33.00	null	117.00	null
20-05-2020	Andaman and Nicobar Islands	7,082.00	Samples Sent	33.00	null	117.00	null
21-05-2020	Andaman and Nicobar Islands	7,167.00	Samples Sent	33.00	null	85.00	null
22-05-2020	Andaman and Nicobar Islands	7,263.00	Samples Sent	33.00	null	96.00	null
23-05-2020	Andaman and Nicobar Islands	7,327.00	Samples Sent	33.00	null	61.00	null

Create a multiple sheet reports like State wise confirmed covid cases and charts, State Wise Cured Covid Cases, State wise Covid Death cases, State wise Positive and Negative cases State wise positive Graph etc ...

