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.

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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.

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Subject In-charge
Department of Computer Science

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Head
Department of Computer Science

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External Examiner

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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

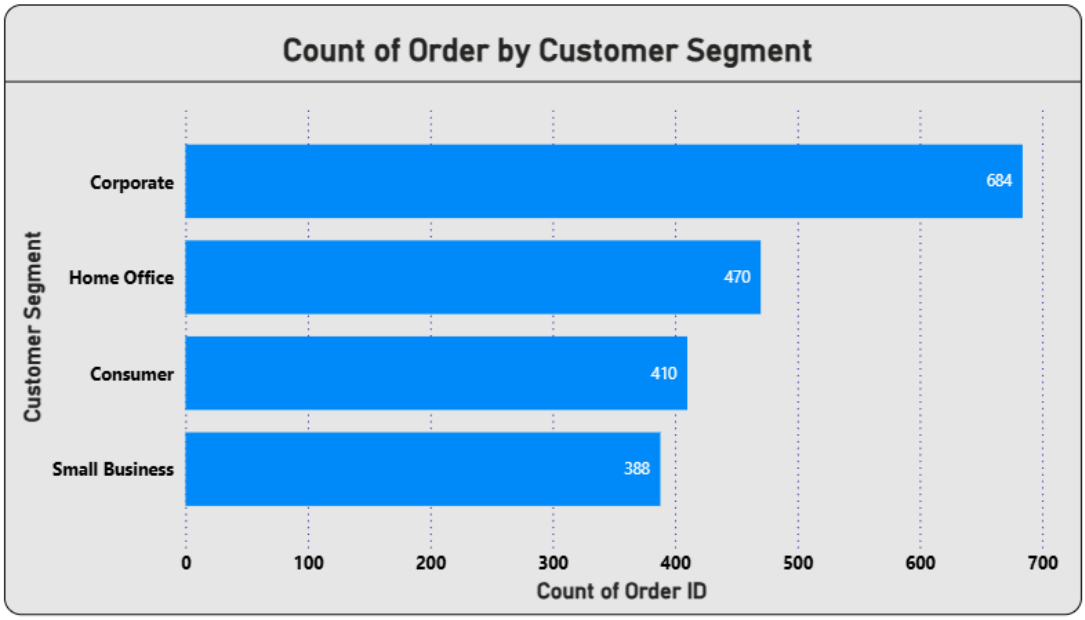


Chart 2: Funnel Chart

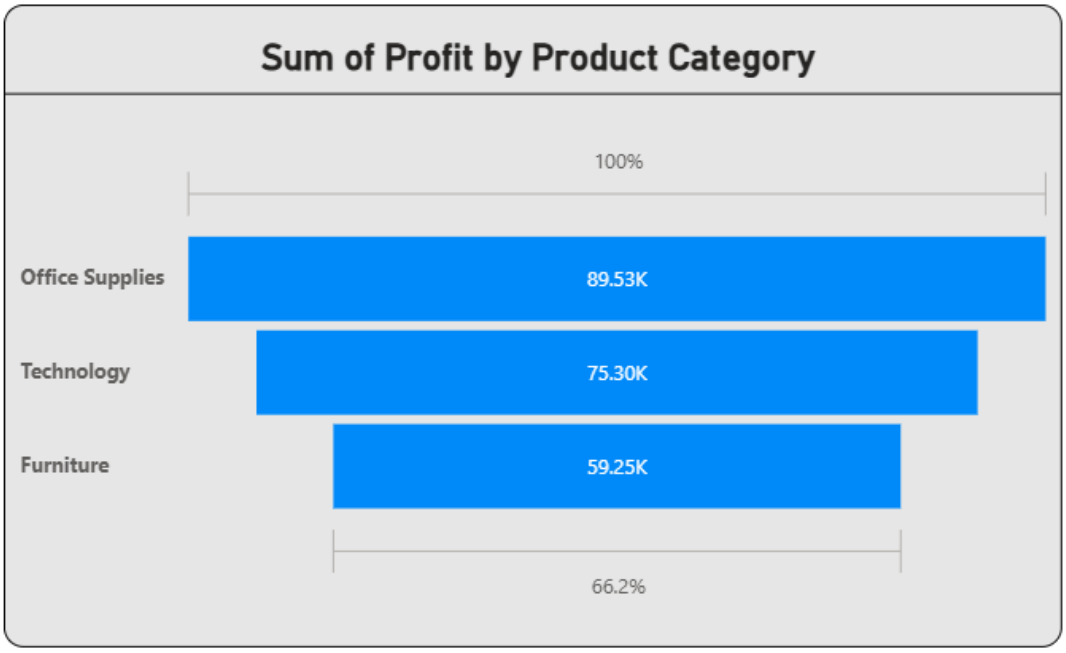


Chart 2: Donut Chart

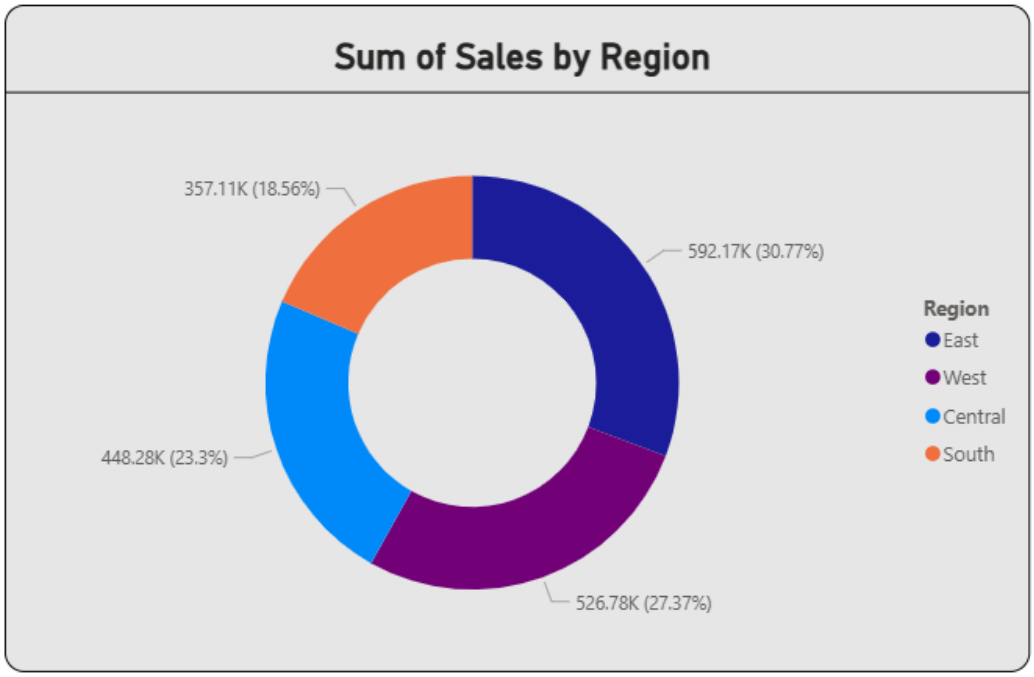
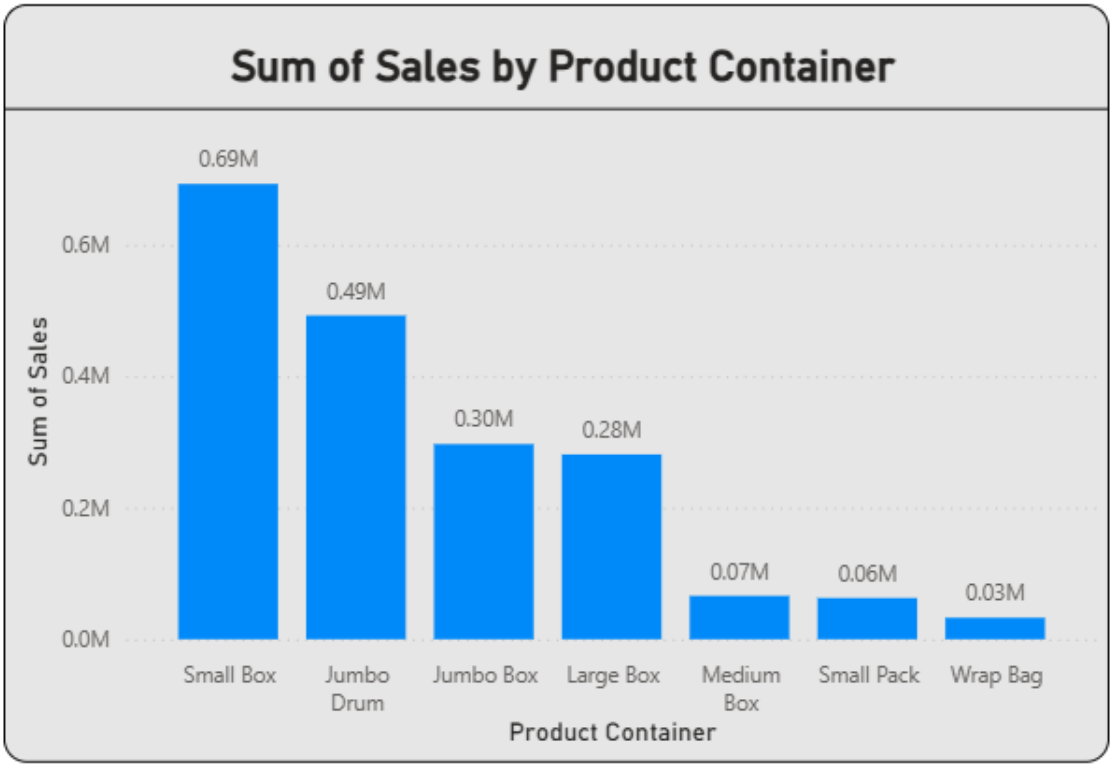


Chart 4: Stacked Column Chart



Practical 2

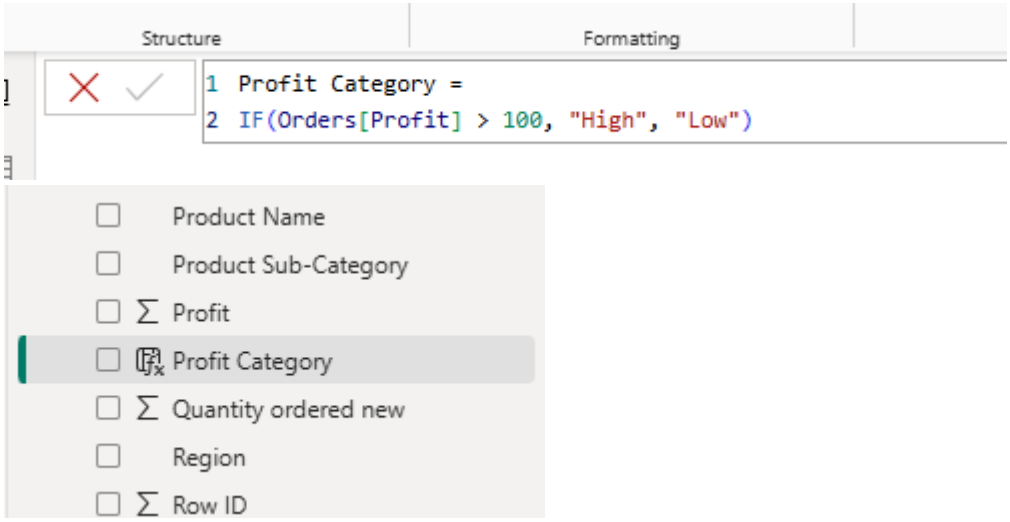
Aim: Time Intelligence and data analysis Functions with DAX

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.

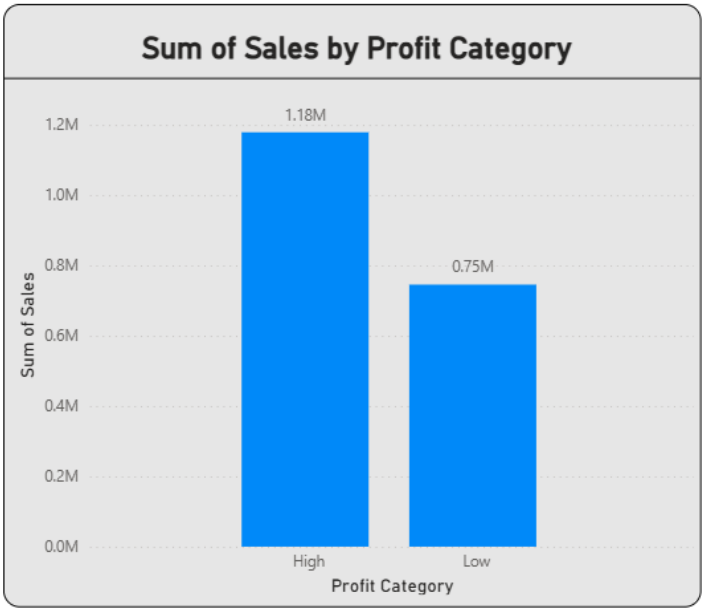
1. Profit Category Classification

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

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



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]) )
```

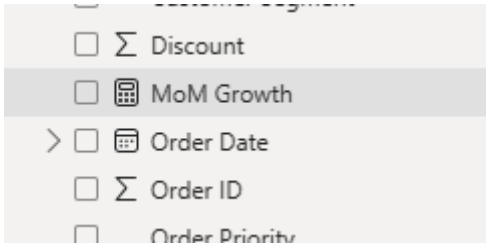
Matrix: Sales and Profit by Category

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**

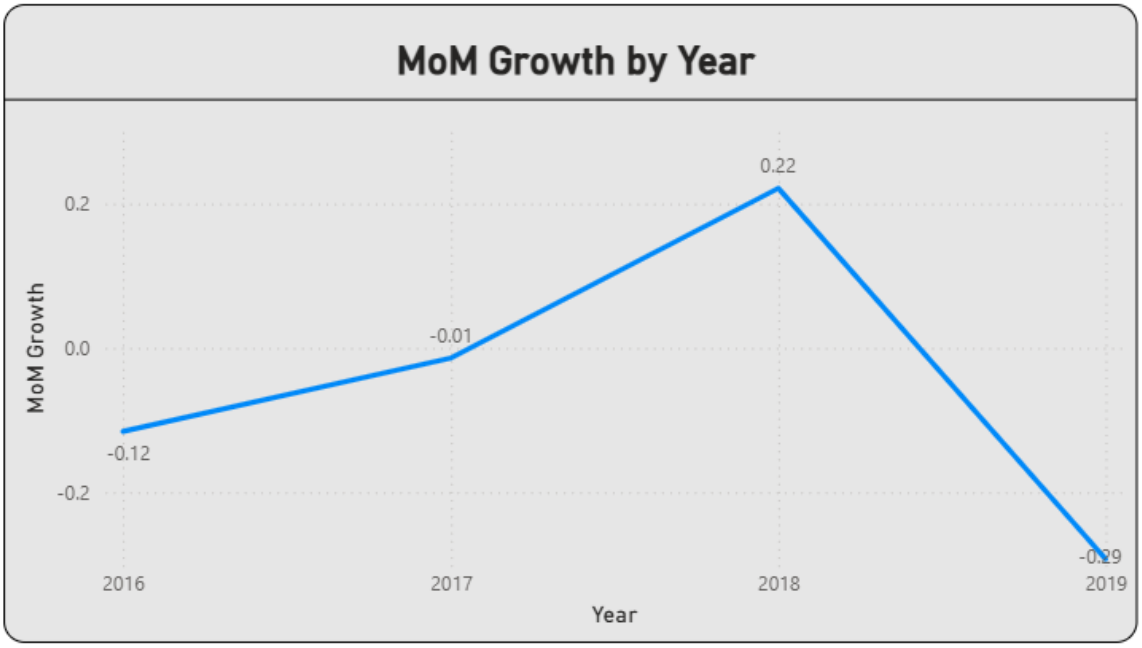
3. To analyze sales and profit performance over specific time periods.

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
        )
```



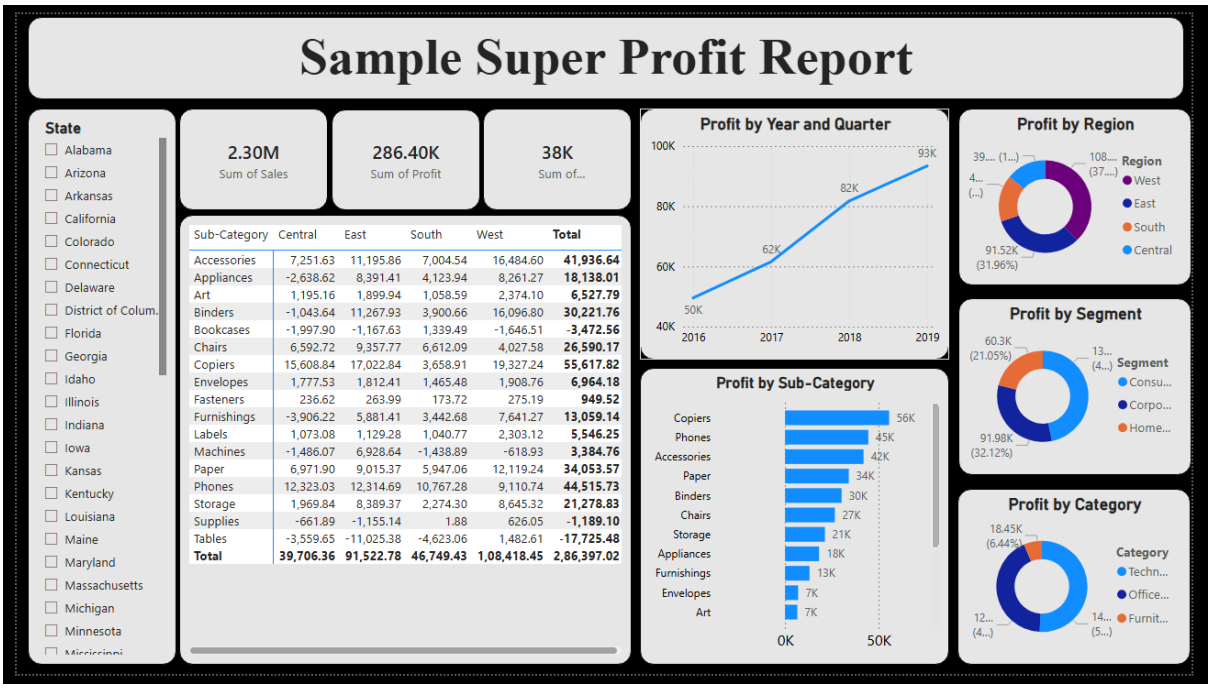
Line Chart: Month-over-Month Growth



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
```

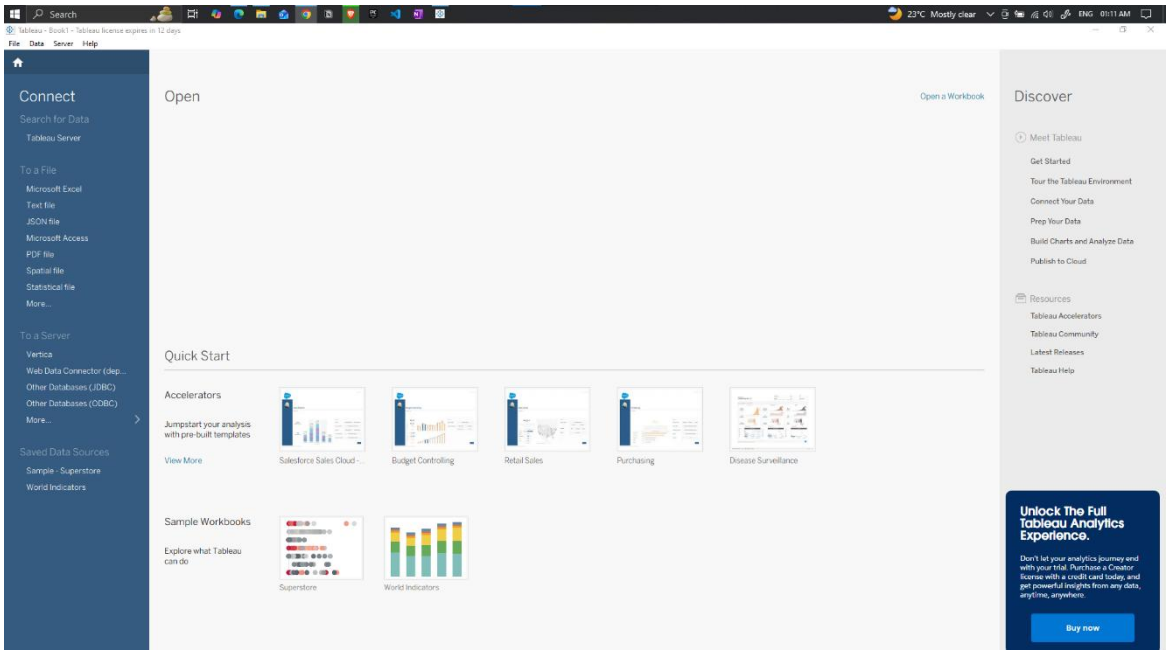
[8]:
```
# Unstack the DataFrame (reverse stacking, expand index back into columns)

unstacked_df = stacked_df.unstack()

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

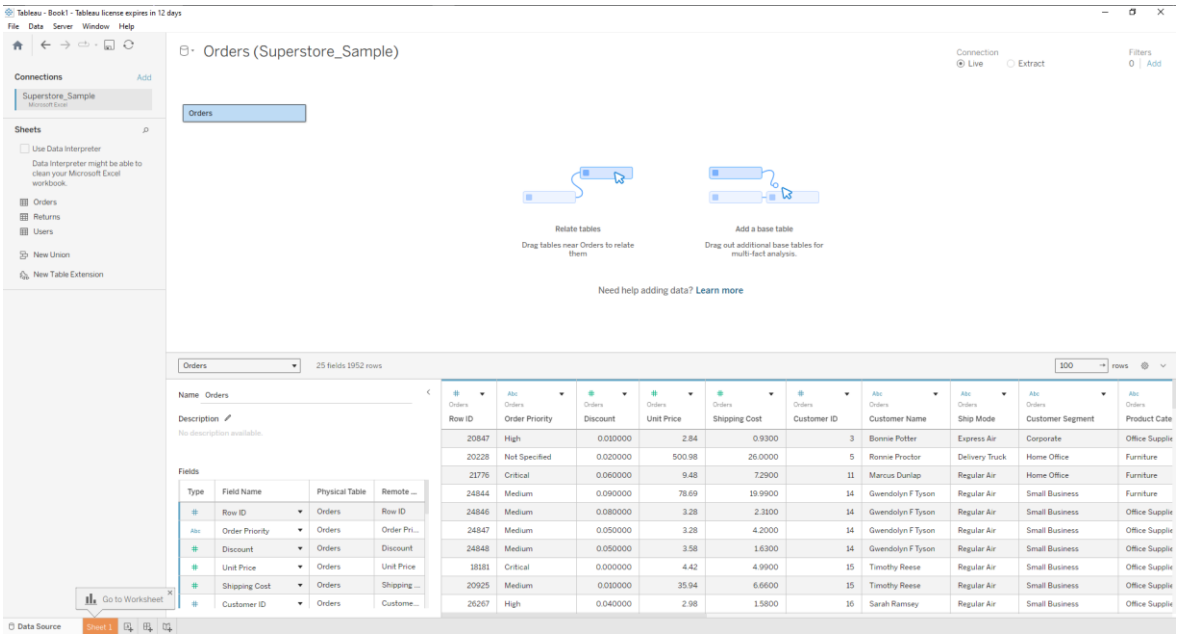
```
Unstacked DataFrame (Reverted to original columns):
                     Temperature  Humidity
Date       City
2024-01-01 Los Angeles          75        10
           New York             32        80
2024-01-02 Los Angeles          72        15
           New York             30        85
2024-01-03 Los Angeles          70        12
           New York             28        90
```

Practical 6

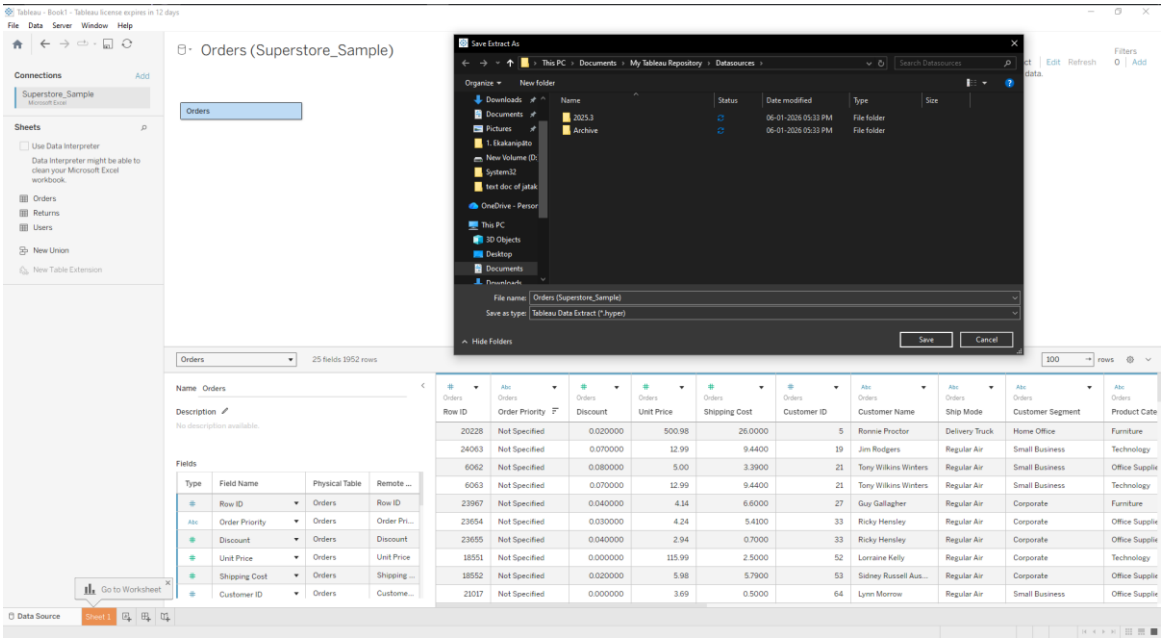
Aim: Connecting and extracting with various data resources in tableau and Perform calculations and creating parameters in Tableau.

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.).





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.



Step 3: Top N subcategories dynamic

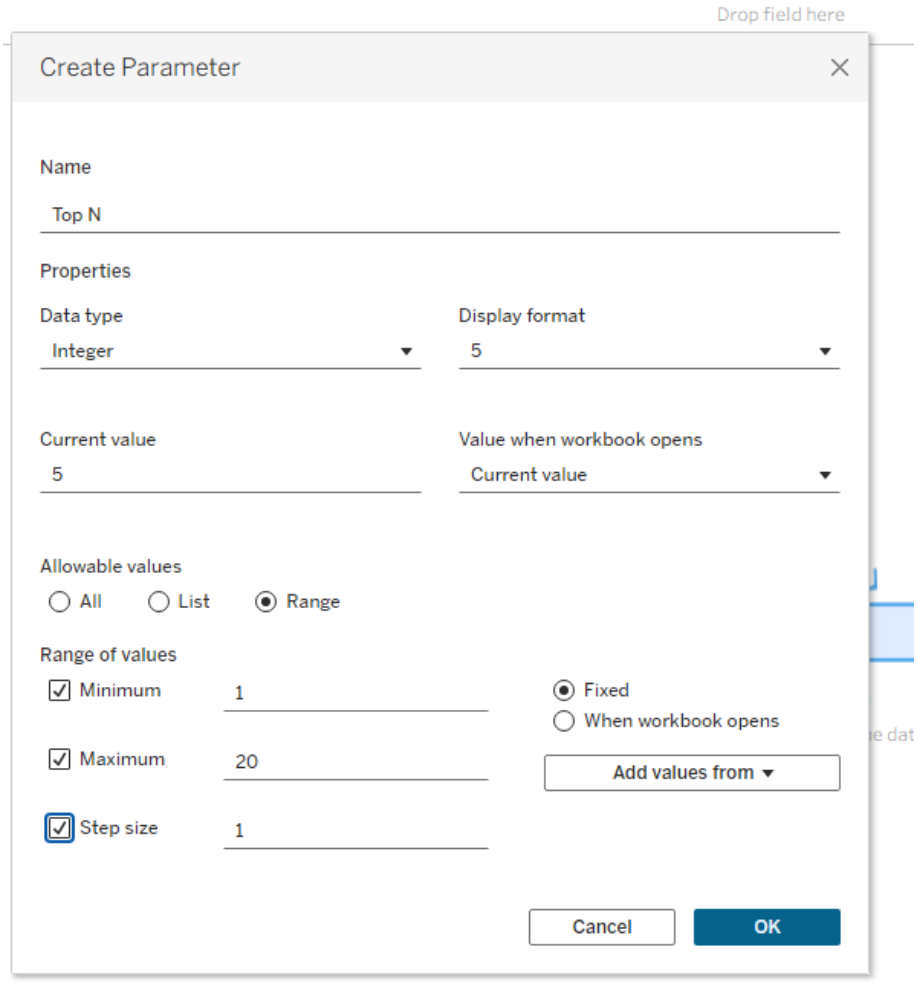
1. Create a parameter for N with current value 5:

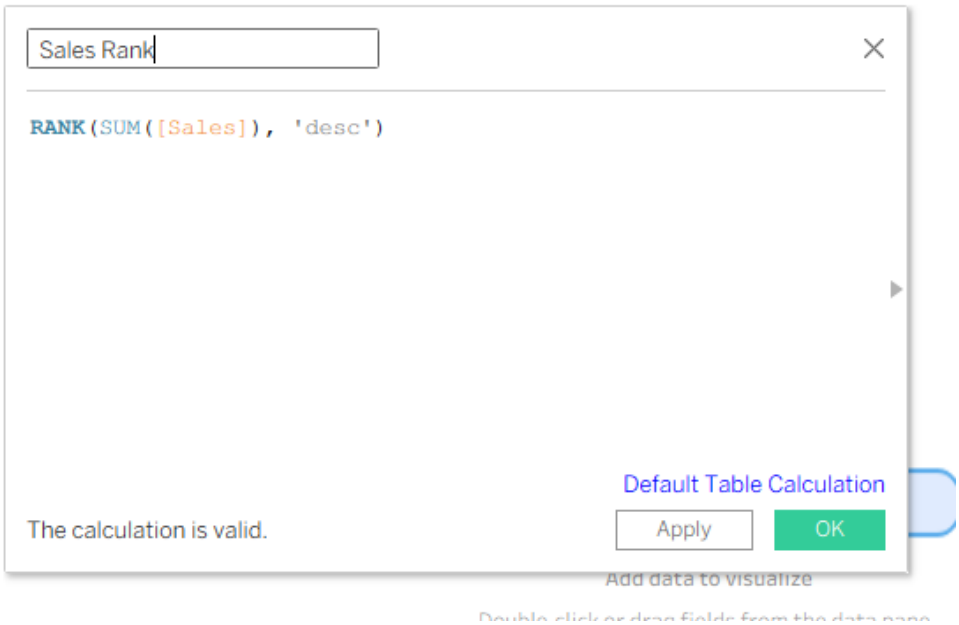
2. Create a calculated field for ranking:

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

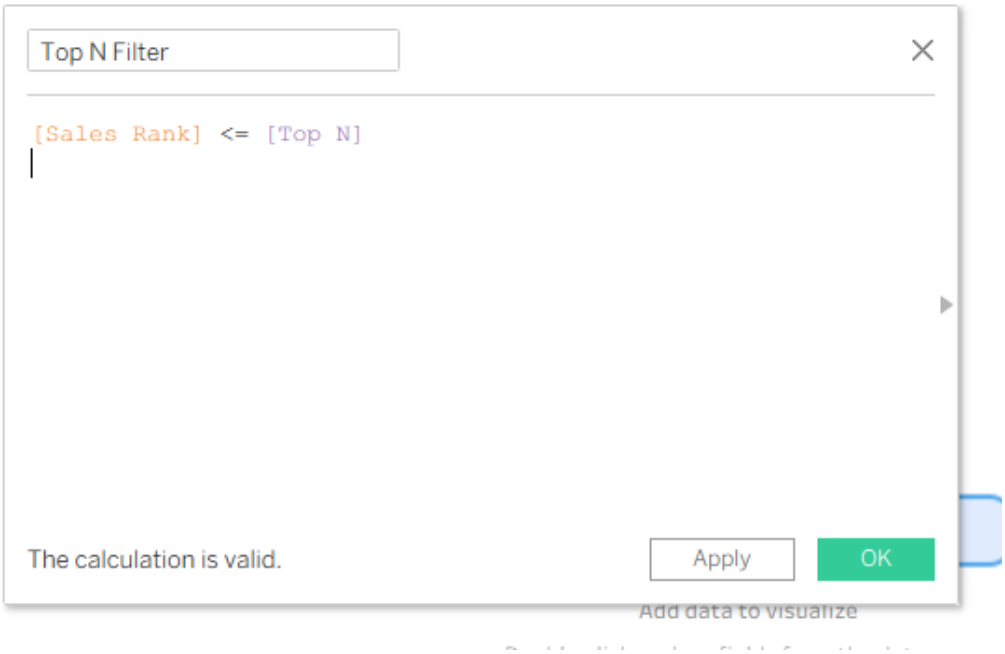
2. Create another calculated field to filter the top N:

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

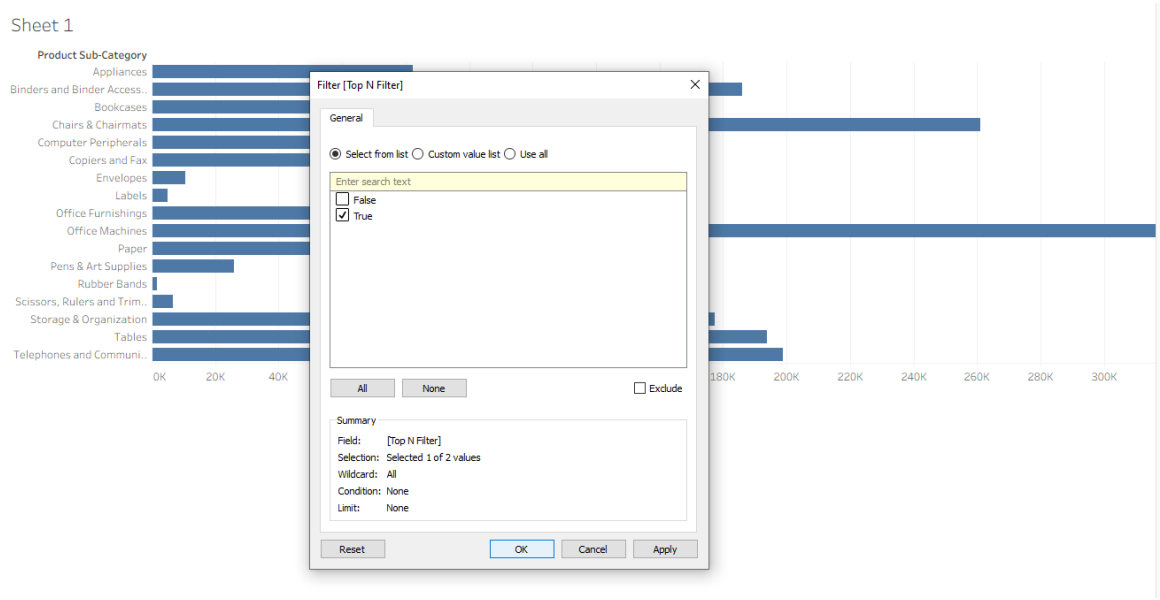


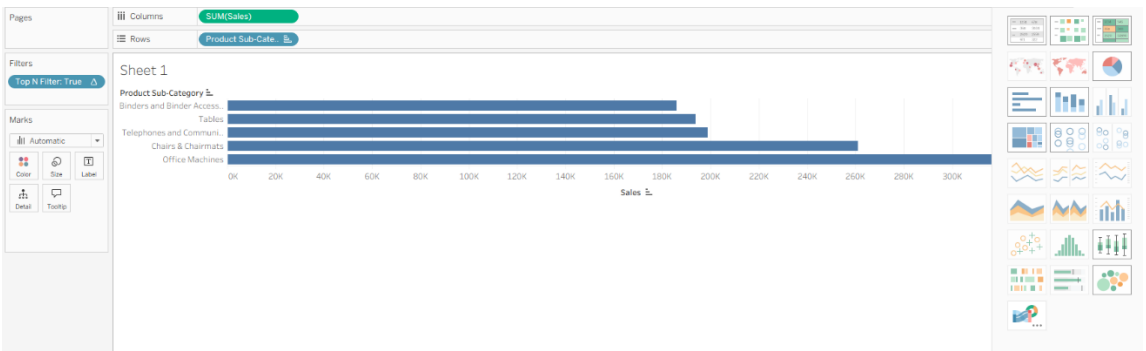


Step 3.: Create Calculated Field to Filter Top N



Step 3: Apply Top N Filter to Visualization

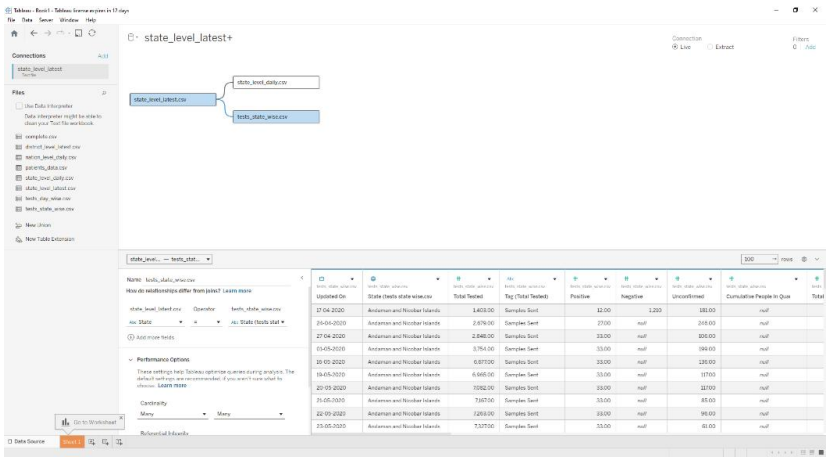




Practical 7

Aim: Designing Tableau Dashboards for different displays and devices.

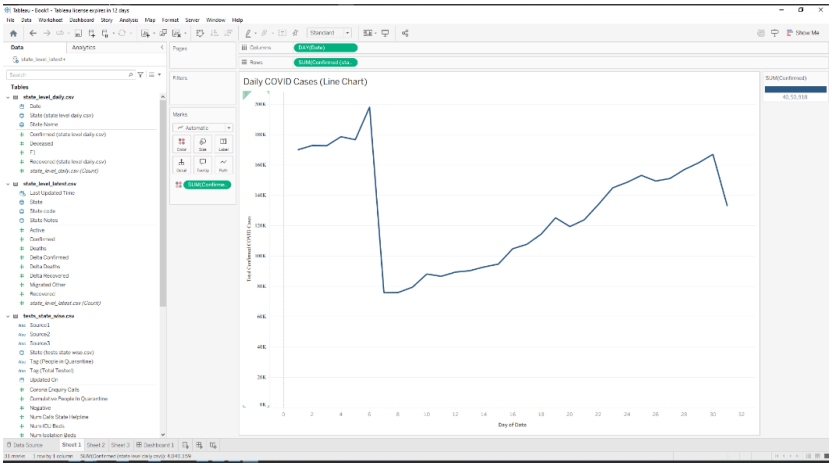
Integrated different dataset for making dashboard



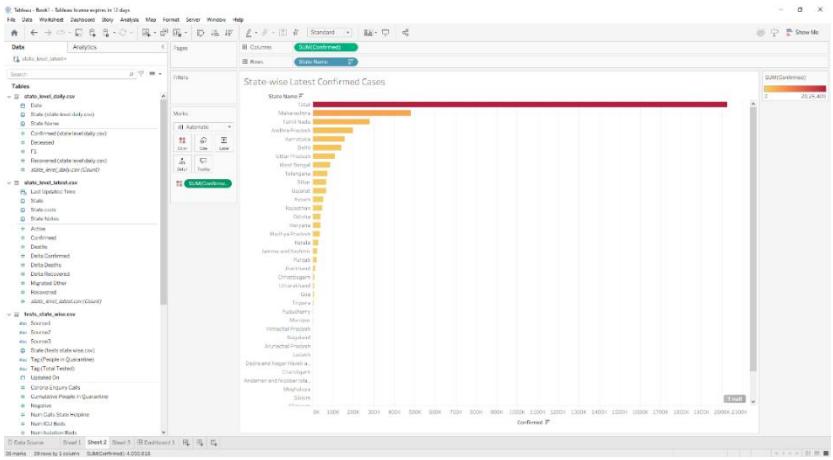
Create multiple sheets for dashboard:

Sheet 1: Daily COVID Cases (Line Chart)

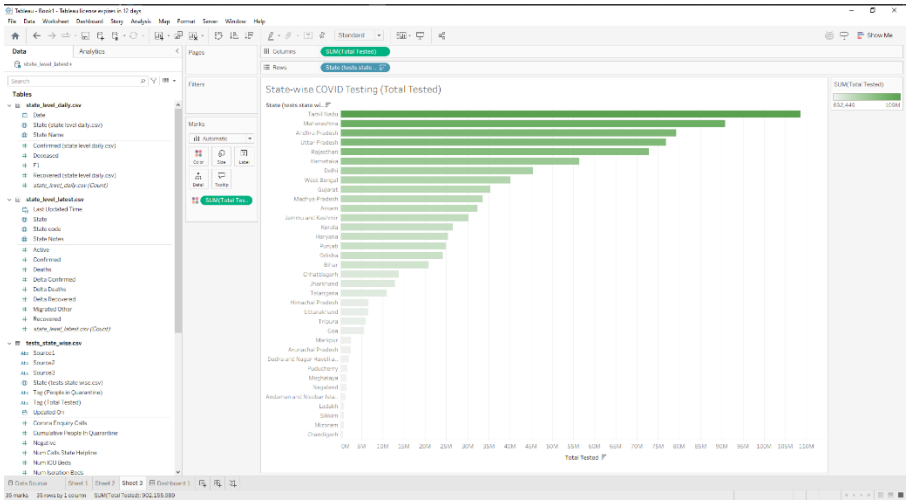
Purpose: To show Time-wise COVID case trend.



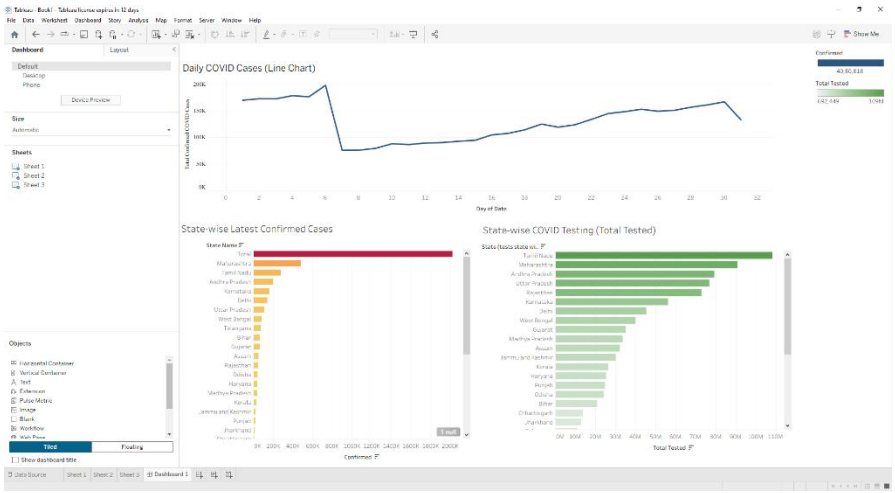
Sheet 2: State-wise Latest Confirmed Cases (Bar Chart)



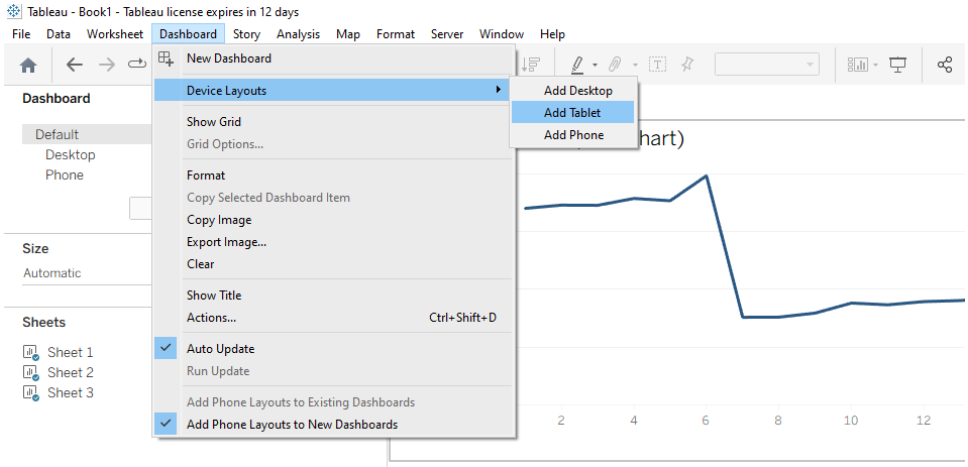
Sheet 3: State-wise COVID Testing (Bar Chart)



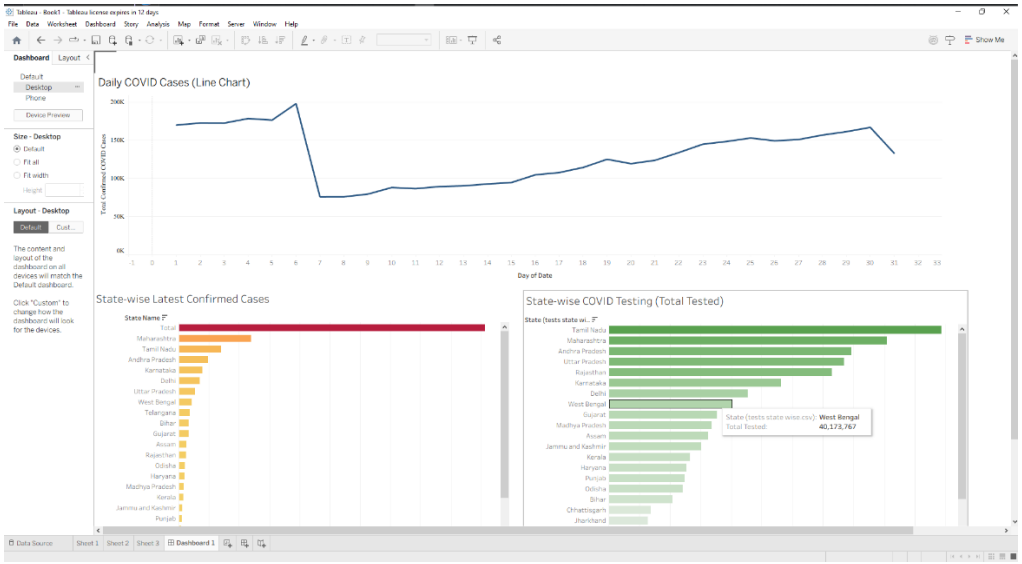
Combine all three sheets for dashboard



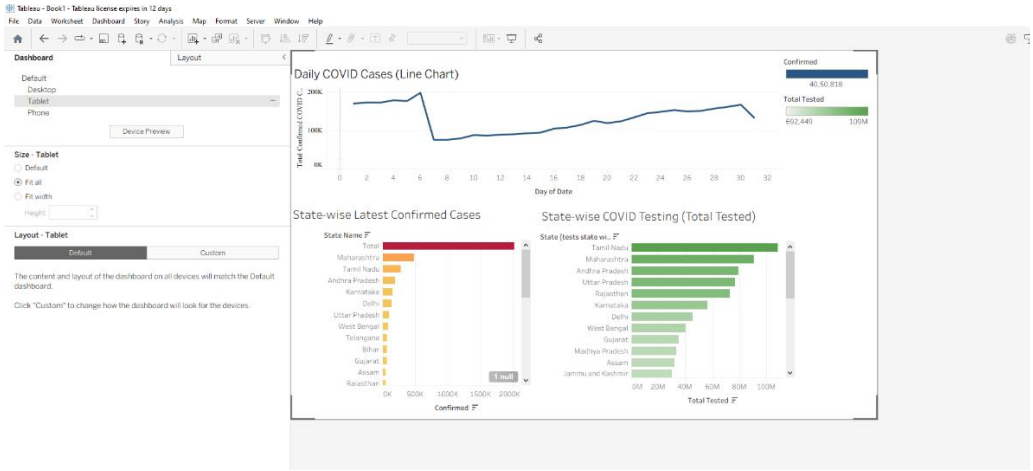
For selecting the dashboard into different layout go to the dashboard option in the navigator bar and select device layout



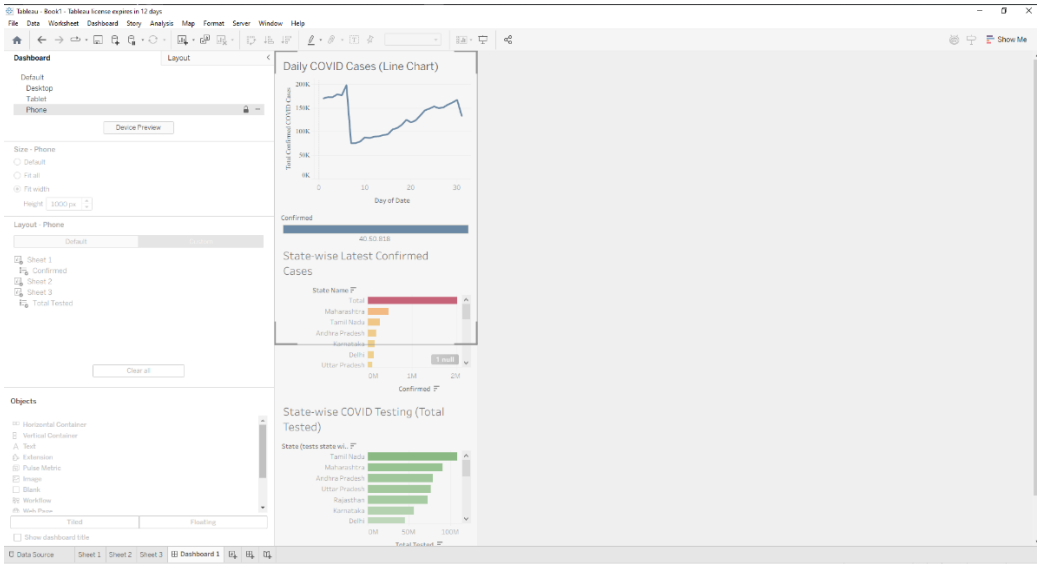
Dashboard in Desktop layout



Dashboard in Tablet layout



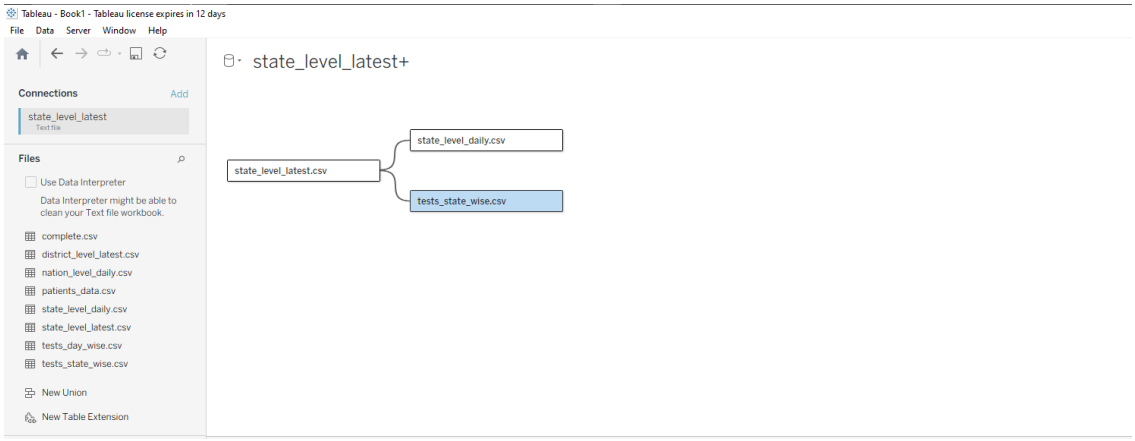
Dashboard in phone layout



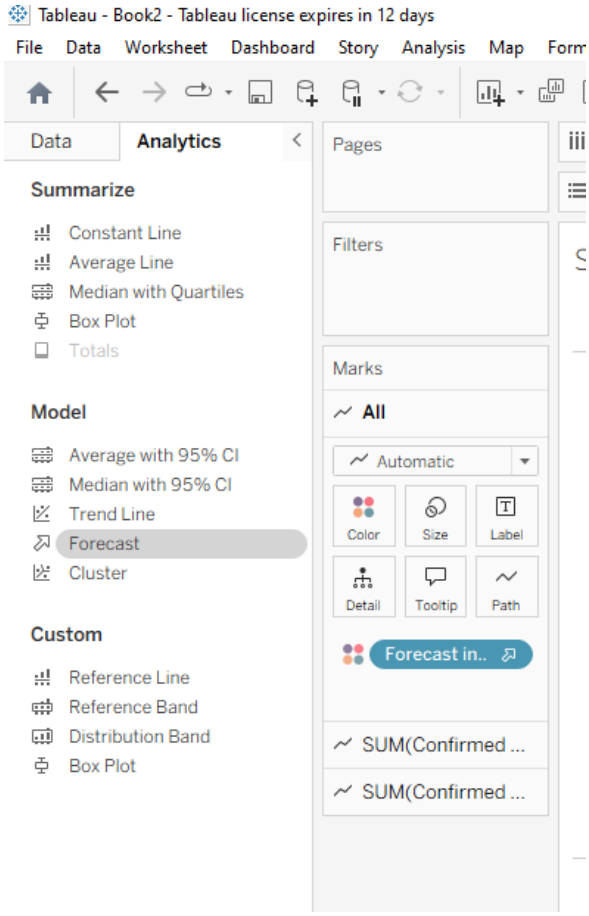
Practical 8

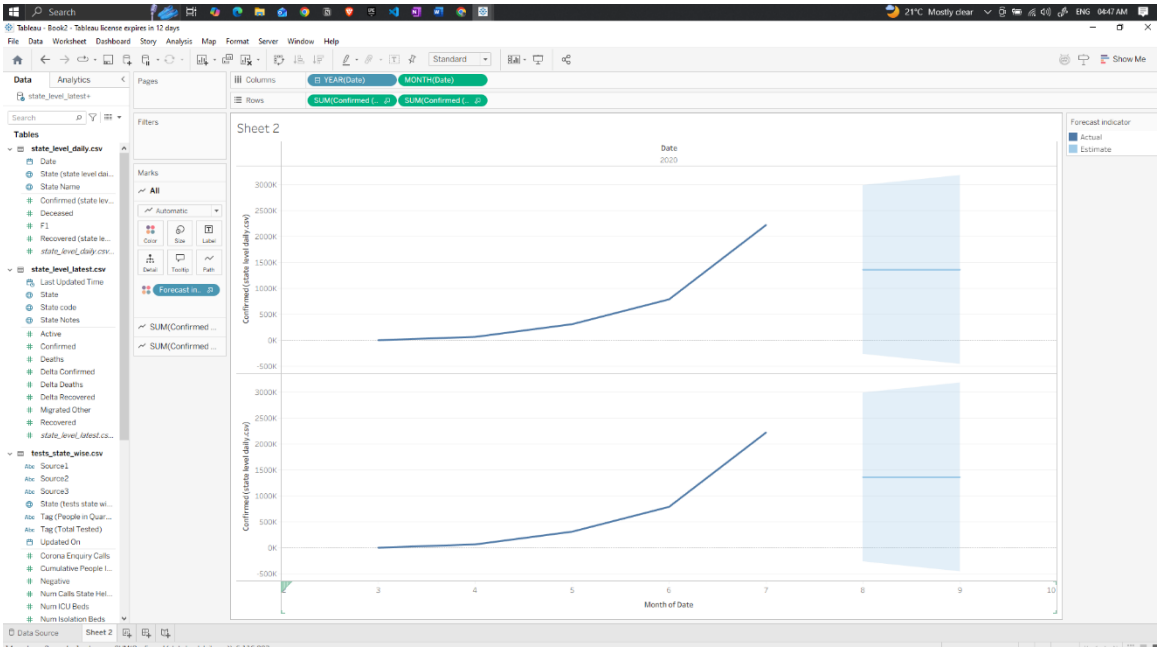
Aim: Create a Trend model using data, Analyse-it and use it for forecasting.

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



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

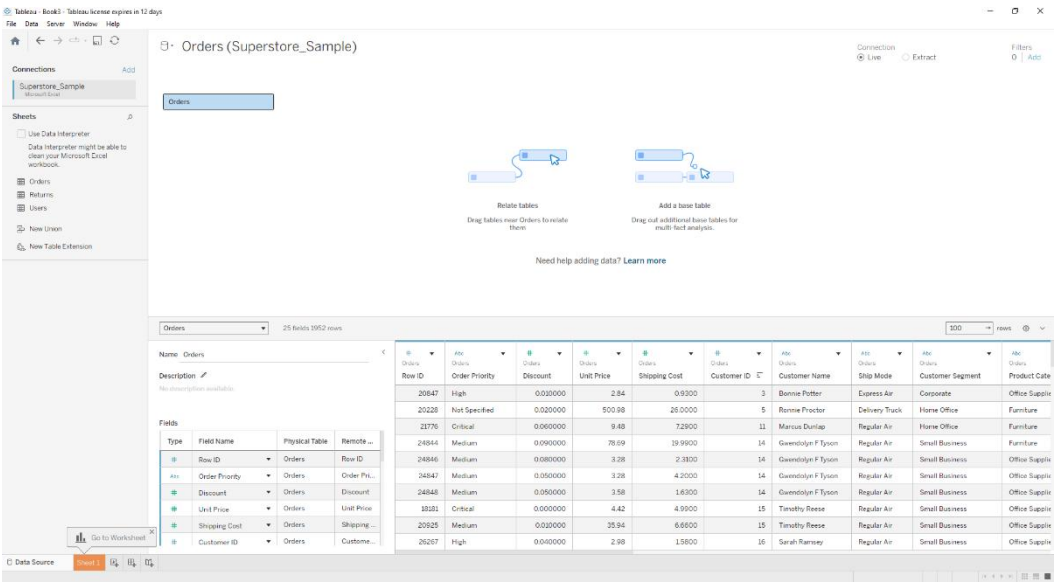



Sheet 2
Date
2020
Month of Date
Forecast indicator
Actual
Estimate

Practical 9

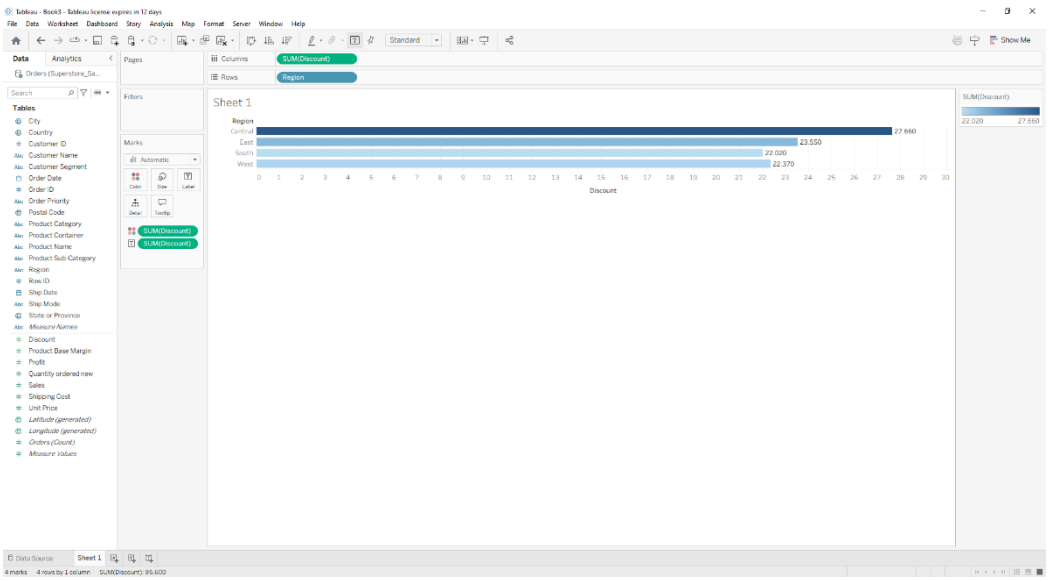
Aim: Creating Geospatial feature maps in Tableau using Geospatial Data.

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.

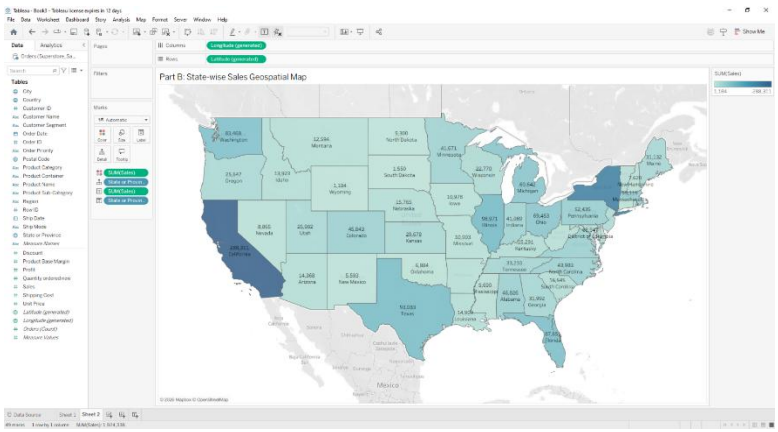
Part A: Region-wise Discount Visualization



Part B: State-wise Sales Geospatial Map

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

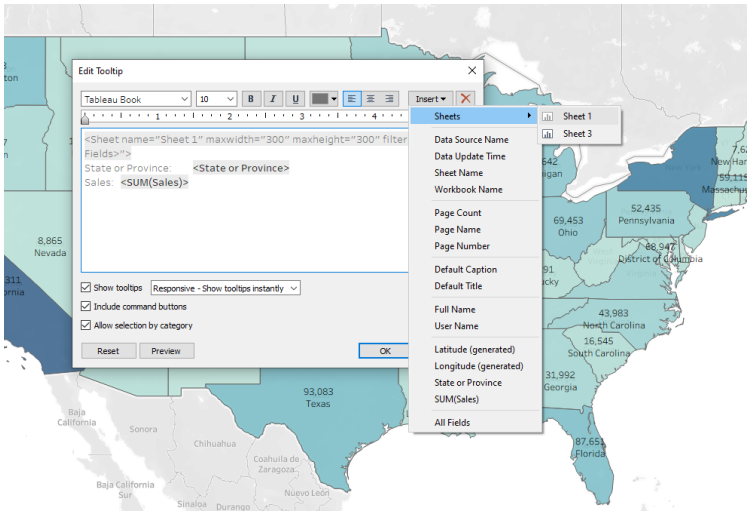
- Drag the 'Longitude' field into the Columns section.
- Drag the 'Latitude' field into the Rows section.
- Drag the sum of 'Sales' into both the Color and Label sections.
- Drag the 'State' field into both the Detail and Label sections.

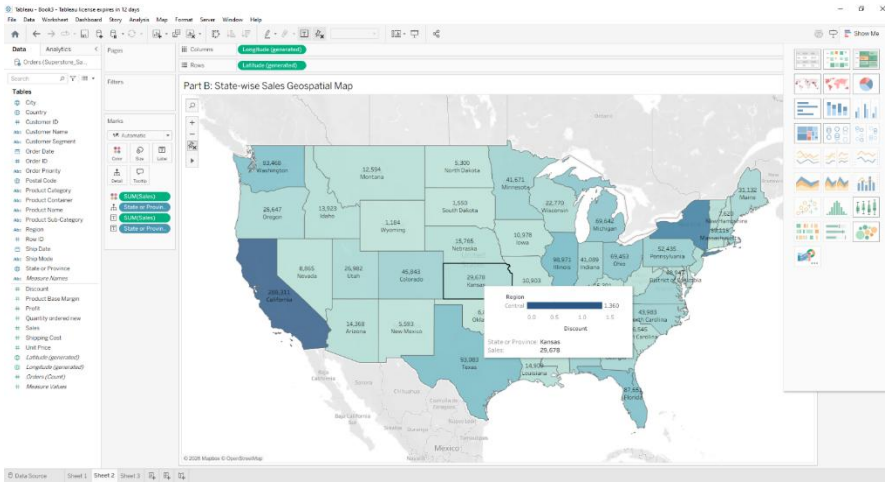


Part C: Integrating Region-wise Discounts using Tooltips

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.

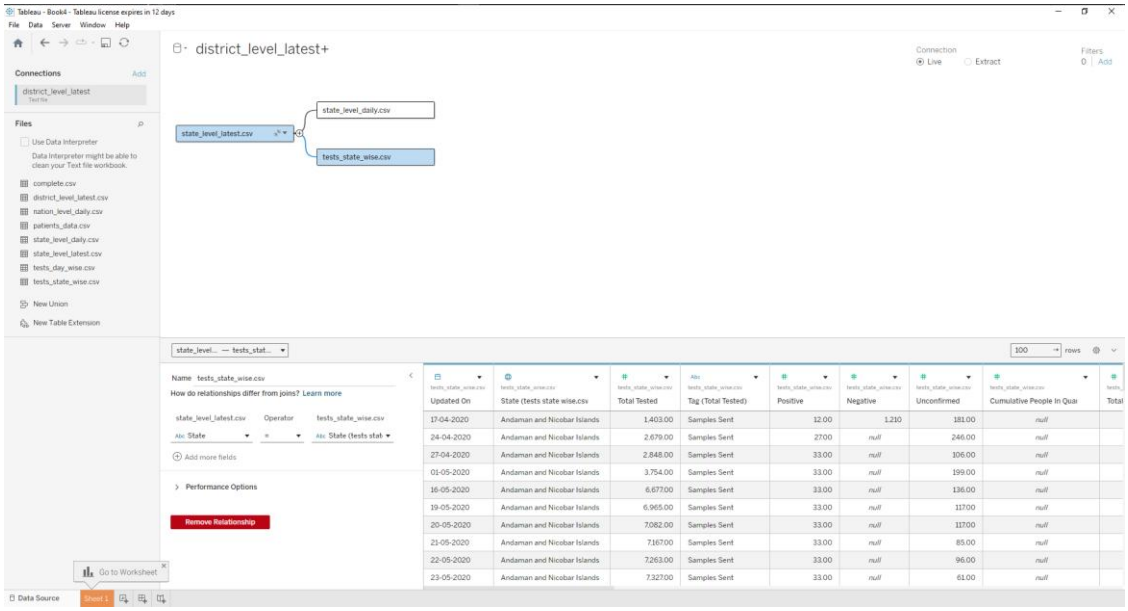




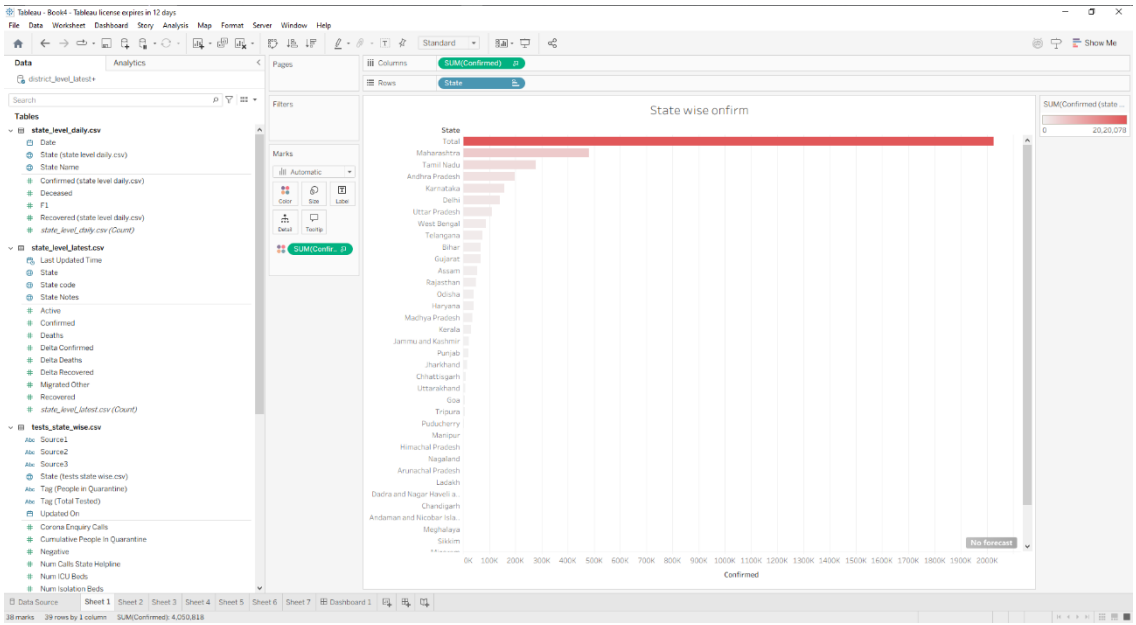
Practical 10

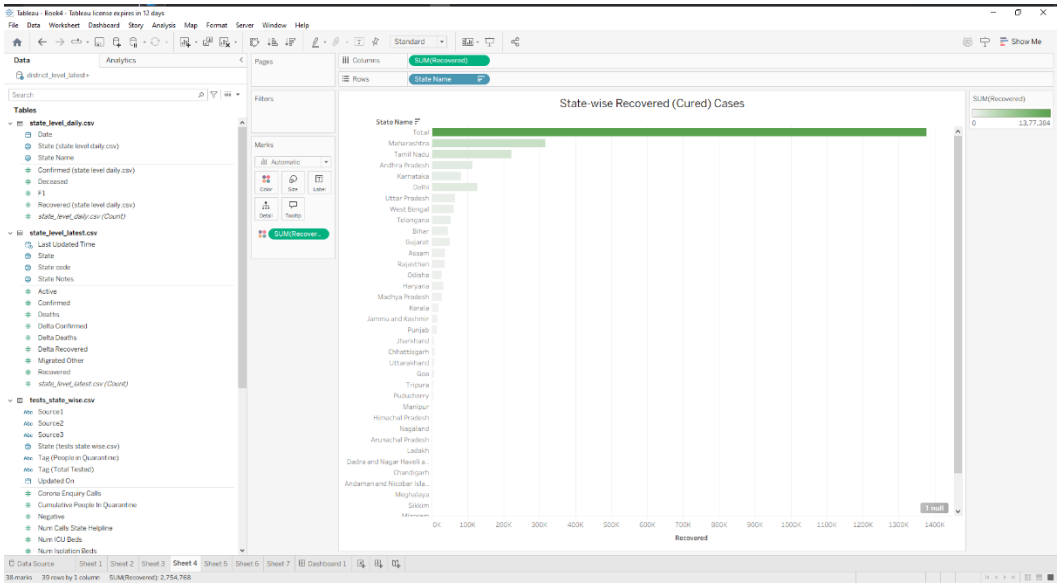
Aim: Create Dashboard and Storytelling using tableau.

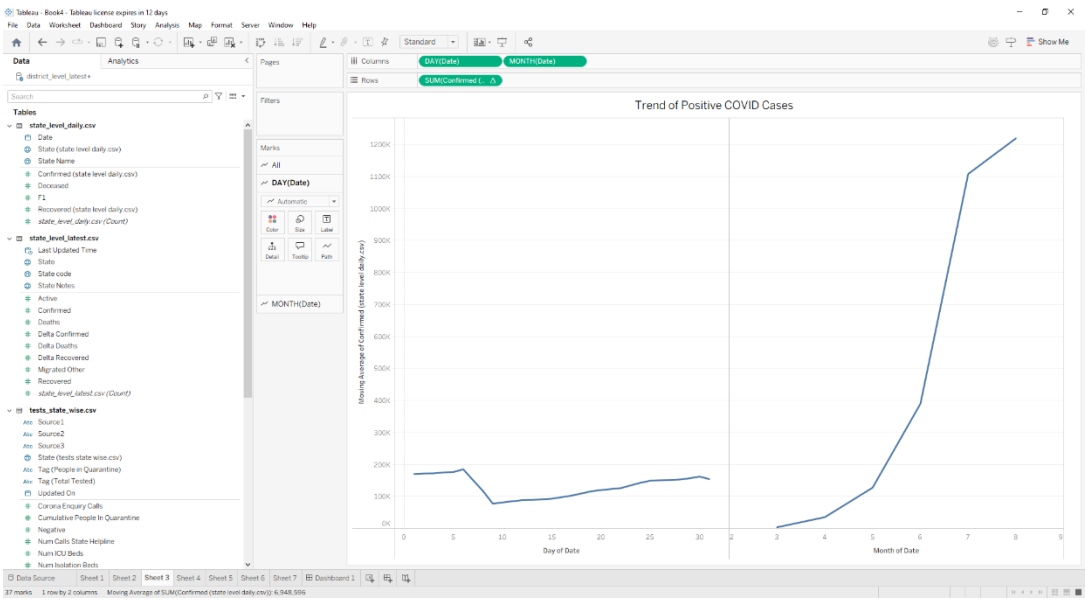
Open Tableau and integrate the dataset of covid 19 cases and create sheet for Forecasting of Positive cases

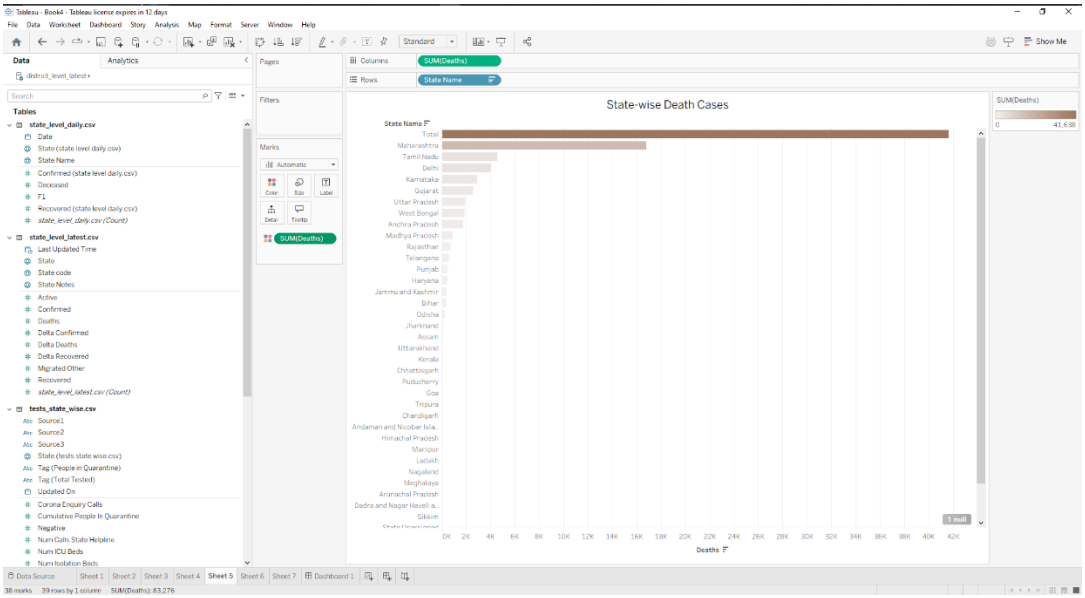


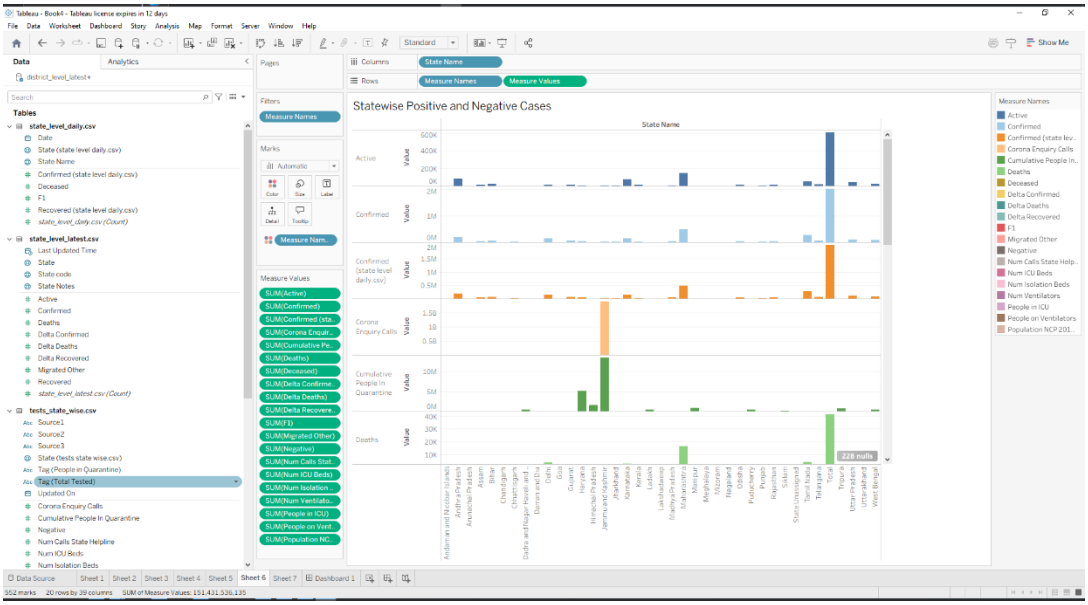
Create a multiple sheet reports like State wise confirmed covid cases and charts, Sate Wise Cured Covid Cases, Sate wise Covid Death cases, State wise Positive and Negative cases State wise positive Graph etc …

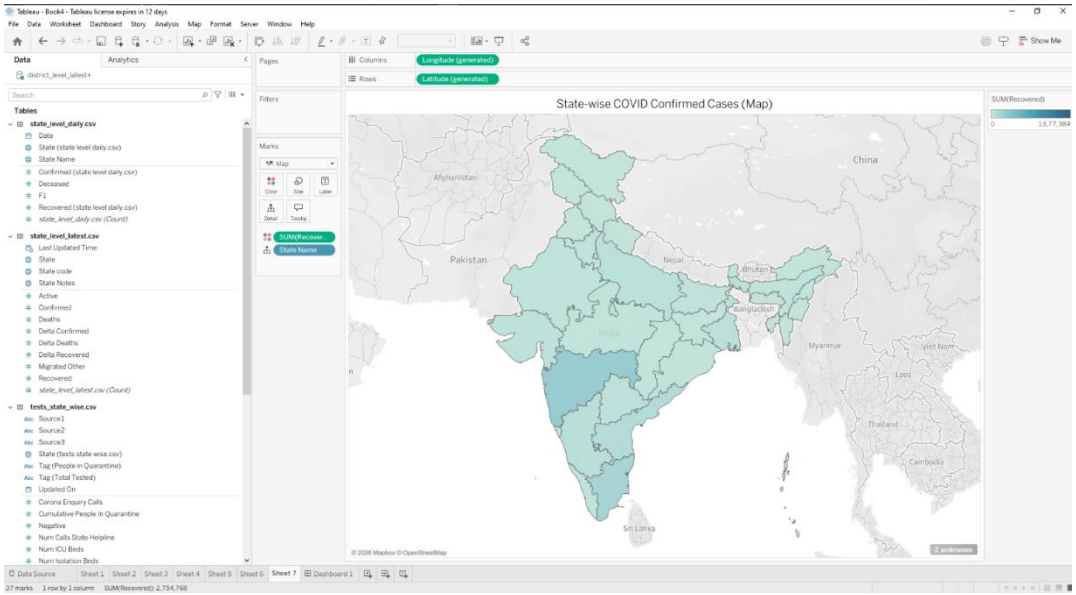


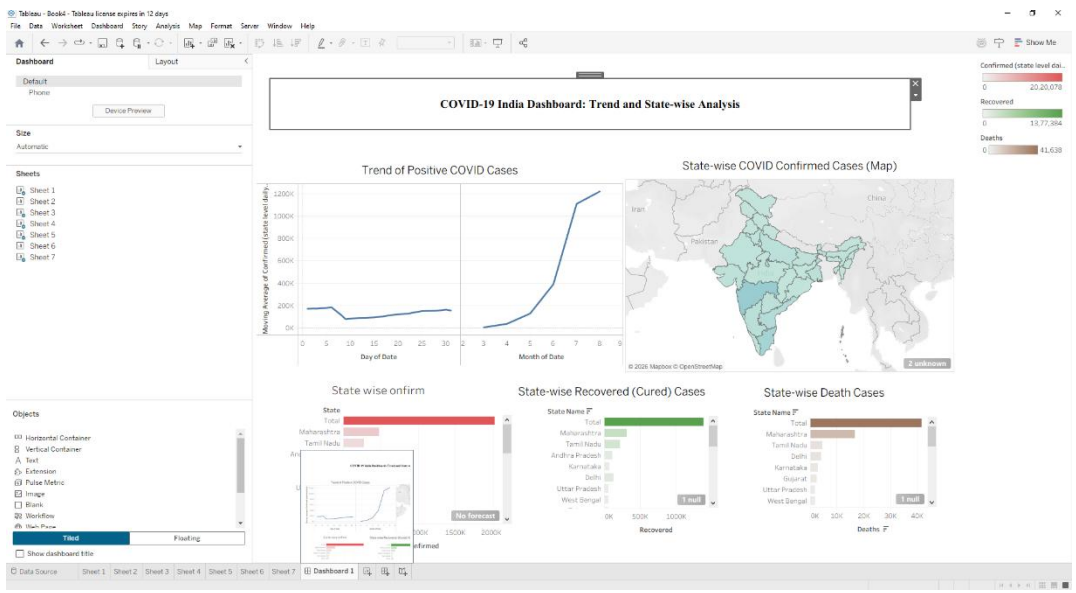

State-wise Recovered (Cured) Cases
Recovered


Trend of Positive COVID Cases
Day of Date
Month of Date


State-wise Death Cases
Deaths


Statewise Positive and Negative Cases
State Name
Active
Confirmed
Confirmed (state level daily.csv)
Corona Enquiry Calls
Cumulative People In Quarantine
Deaths


State-wise COVID Confirmed Cases (Map)
China
Afghanistan
Pakistan
Nepal
Bhutan
Bangladesh
India
Myanmar
Viet Nam
Laos
Thailand
Cambodia
Sri Lanka


COVID-19 India Dashboard: Trend and State-wise Analysis
Trend of Positive COVID Cases
State-wise COVID Confirmed Cases (Map)
State wise onfirm
State-wise Recovered (Cured) Cases
State-wise Death Cases