Daily Public Transport Passenger Journeys by Service Type

NO OF ROWS

1,918

NO OF COLUMNS

7

The dataset titled "Daily Public Transport Passenger Journeys by Service Type" contains data about the number of passenger journeys made on different types of public transport services on various dates.

Data covers: From **JULY 1, 2019** to **SEPTEMBER 29, 2024** -USING POWERBI SORTING. **1.COVID-19 Impact**

- Notable dips in passenger journeys during:
 - March–May 2020
- Demonstrates how external events disrupt usage and how recovery trends can be measured.

2. School Transport Seasonality

 School journeys are highly seasonal, almost disappearing in school breaks and holidays.

3. Local Route Dominance

- Local Routes consistently contribute the highest share of total passengers across all service types.
- Often accounts for over 50% of the total daily journeys a critical backbone of public transport.

4. Total Daily Load Stability

Despite fluctuations in individual services, the overall total passenger count remains surprisingly stable on weekdays.

5. Weekend Dips

All service types show **notable drops in ridership on weekends**, presenting an opportunity to **adjust fleet operations** and reduce idle capacity

STATISTICAL INFORMATION

Detected missing values and outliers-using IQR.

Missing &	Outliers	Summary	:
	Missing	Values	Outliers
Date		0	0
Local Route		0	0
Light Rail		0	0
Peak Service		0	1
Rapid Route		0	0
School		0	0
Other		20	9

Algorithm to be used

The algorithm i have chosen is SARIMA

What is SARIMA?

SARIMA stands for Seasonal AutoRegressive Integrated Moving Average. It is a time series forecasting model that extends ARIMA to handle seasonality in data.

Why use SARIMA for this dataset?

The dataset contains **daily passenger counts** for different public transport service types. This kind of data usually exhibits:

- Trend (e.g., increasing/decreasing usage over time),
- Seasonality (e.g., more usage on weekdays vs. weekends),
- Autocorrelation (today's value is related to yesterday's).
 SARIMA is ideal because:
- It handles both trend and seasonality
- Captures weekly patterns (we set m=7 for weekly seasonality),
- Produces accurate short-term forecasts, such as 14-day predictions.