Key Findings and Recommendations:

1. Data Understanding:

- The shape of Delhivery data is (144867, 24).
- Identified 5 unknown fields, which can be dropped.
- Observed missing area names in both source_name and destination_name.
- Categorized columns into numerical and categorical fields.

2. Categorical Columns:

- data and route_type identified as main categorical columns.
- Mode of data is 'training', with frequent route_type as 'FTL'.
- Top source and destinations are 'Gurgaon_Bilaspur_HB (Haryana)'.

3. Data Imbalance:

- Training data is more than testing data.
- 60% of transportation type is carting, and 40% is Full Truck Load (FTL).

4. Outliers:

 Identified outliers in numerical columns, but left them untreated.

5. Correlations:

- Positive correlations observed between actual_time and segment_osrm_time_sum.
- High correlation (100%) between actual_time and segment_actual_time_sum.
- High correlations (98%) between osrm_distance and segment_osrm_distance_sum, and osrm_time and segment_osrm_time_sum.
- Possible glitches in ORSM navigation system.

6. Temporal Patterns:

- More trips created in September.
- Higher trip creation during night hours.

7. Geographical Patterns:

- Maharashtra is the top state with the highest number of trips.
- Gurgaon is the top city for trip creation, followed by Mumbai and Delhi.

Recommendations:

- Address missing values in source_name and destination_name.
- Consider treating outliers in numerical columns based on the specific context.
- Investigate and improve the accuracy of the ORSM navigation system.
- Monitor and optimize logistics operations during night hours.
- Explore further insights into geographical patterns for targeted marketing and resource allocation.