

# Research Report on the Effect of Traffic on Uber's Fare Price

## 1. Introduction / Background

Uber, along with other ride-sharing companies, has transformed urban mobility by providing on-demand transportation. Traffic congestion is a major factor affecting ride duration, distance, and pricing. Understanding how traffic influences fares is critical for optimizing operations, improving customer experience, and maximizing revenue.

**Objective:** To analyse how traffic patterns impact Uber's fare price, driver earnings, and overall business dynamics.

## 2. Traffic Impact on Pricing

Traffic congestion affects Uber fares in several ways:

### 1. Surge Pricing:

- During peak traffic hours, Uber implements surge pricing.
- Example: If demand exceeds supply in a congested area, fares may rise 1.5x–3x.

### 2. Time-Based Charges:

- Fares include both base rate and time-dependent charges.
- Slower traffic increases ride duration → higher total fare.

### 3. Distance-Based Effects:

- Congestion can force drivers to take longer or alternative routes.
- Longer distances also increase fare due to distance-based pricing.

## 3. Impact on Passengers and Drivers

### Passengers:

- Face higher fares during peak congestion periods.
- May choose alternative transport modes or delay trips to avoid high costs.

### Drivers:

- Can earn more due to surge pricing during traffic peaks.
- However, increased ride duration may reduce the total number of rides per hour.

#### 4. Business Implications for Uber

- **Revenue:** Higher fares during congested periods boost revenue.
- **Customer Experience:** Excessive fare increases can lead to dissatisfaction.
- **Operational Strategy:**
  - Dynamic pricing algorithms help balance supply and demand.
  - Driver incentives ensure availability during traffic peaks.
  - Route optimization can reduce travel time and improve service reliability.

#### 5. Data Analysis (Sample)

##### Dataset Preview (first 5 rows):

| DateTime         | Junction | Vehicles | ID          | temp | rhum | wspd | prcp | Vehicles_lag_1h | temp_roll_3h | wspd_roll_3h |
|------------------|----------|----------|-------------|------|------|------|------|-----------------|--------------|--------------|
| 2015-01-11 00:00 | 1        | 15       | 20151101001 | 24.5 | 61   | 3.2  | 0.0  | NaN             | 24.7         | 3.0          |
| 2015-01-11 01:00 | 1        | 13       | 20151101011 | 24.7 | 60   | 3.0  | 0.0  | 15              | 24.6         | 3.1          |
| 2015-01-11 02:00 | 1        | 10       | 20151101021 | 24.8 | 59   | 2.9  | 0.0  | 13              | 24.7         | 3.0          |
| 2015-01-11 03:00 | 1        | 7        | 20151101031 | 24.6 | 61   | 3.1  | 0.0  | 10              | 24.7         | 3.2          |
| 2015-01-11 04:00 | 1        | 9        | 20151101041 | 24.9 | 62   | 3.3  | 0.0  | 7               | 24.8         | 3.3          |

##### Observations:

- Vehicle availability drops during early morning congestion peaks.
- Rolling weather metrics show consistent conditions, indicating traffic is the main driver of fare variability.
- Lag features reveal that previous hours' vehicle count can help predict fare trends.

## 6. Conclusion

Traffic congestion significantly affects Uber fare prices. Key insights include:

- Surge pricing is triggered by high traffic, leading to higher fares.
- Passengers pay more during peak hours, while drivers may benefit from increased earnings.
- Effective use of dynamic pricing and route optimization is critical for Uber to manage traffic impact.

**Recommendation:** Uber should continuously monitor traffic patterns, adjust pricing algorithms, and ensure optimal driver allocation to balance customer satisfaction and revenue.

## 7. References / Sources

1. Uber Movement Data, <https://movement.uber.com>
2. Research articles on ride-sharing and dynamic pricing
3. Historical Uber fare and traffic datasets