

# **VoltRide: When Green Growth Meets Operational Reality**

## **DecodeX 2026 – Round 2 Business Case Submission**

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Date: 16 February 2026

## Executive Summary

VoltRide's performance deterioration is not driven by weak demand, but by structural misalignment between demand timing, fleet readiness, and charging behavior.

Data indicates that peak-hour cancellations are strongly correlated with:

- Low battery bands at pickup
- Zones with high charging wait times
- Rain-induced demand spikes
- High surge multipliers coinciding with low dispatchable supply

The highest operational risk is concentrated in peak commute windows (8–10 AM; 6–9 PM) in dense business zones across Mumbai and Delhi.

Root causes are:

1. Pre-emptive driver exits due to range anxiety
2. Charging queue congestion during peak demand
3. Micro-level zone-hour mismatches

This report recommends a 60–90 day operational coordination strategy, prioritizing:

- Battery-threshold dispatch optimization
- Peak-hour charging staggering
- Zone-based fleet redeployment

The objective is not fleet expansion, but **conversion efficiency improvement** — translating ride demand into completed rides.

## Problem Understanding & Approach

VoltRide operates under EV-specific constraints:

- Vehicles require downtime for charging.
- Charging infrastructure is unevenly distributed.
- Driver charging decisions directly impact supply.

Our approach:

- Zone-hour level stress mapping
- Cancellation driver decomposition
- Fleet utilization ratio comparison
- Charging wait-time alignment analysis

# Data Analysis & Key Insights

## 1. Demand–Supply Stress Mapping

**Highest operational risk window identified:**

- City: Mumbai
- Zone: Central Business District
- Time: 7–9 PM

**Quantitative indicators:**

- Highest surge multiplier cluster
- Peak cancellation-to-request ratio
- Low average battery at pickup



Operational explanation:

During peak return commute, demand spikes sharply. Simultaneously, vehicles drop below 30% battery after daytime operations and queue for charging. Registered vehicles remain high, but dispatchable vehicles drop.

## 2. Cancellation Driver Decomposition

Primary cancellation driver: **Low Battery / System Safeguard**

Relative contribution:

- Largest share among cancellation types.
- Strong correlation with zones showing longer charging wait times.

Actionability evidence:

- Driver-initiated cancellations cluster below specific battery thresholds.
- System rules automatically block rides under certain battery levels.

This is controllable through:

- Dynamic battery dispatch thresholds
- Smart charging nudges
- Zone-based battery forecasting

Customer impatience (wait-time driven) is secondary and symptom-based.

Conclusion:

Battery readiness is the highest-leverage intervention point.

### 3. Fleet Utilization Efficiency Assessment



Redeployment candidate:

**Hyderabad – Residential Peripheral Zone**

Indicators:

- High idle driver hours
- Low surge intensity
- Moderate cancellation rates
- Stable demand without extreme peaks

Demand contrast:

Central Hyderabad shows peak stress, while peripheral zones show idle capacity.

Redeployment risk:

- Demand volatility could shift geographically.
- Requires predictive modeling, not static relocation.

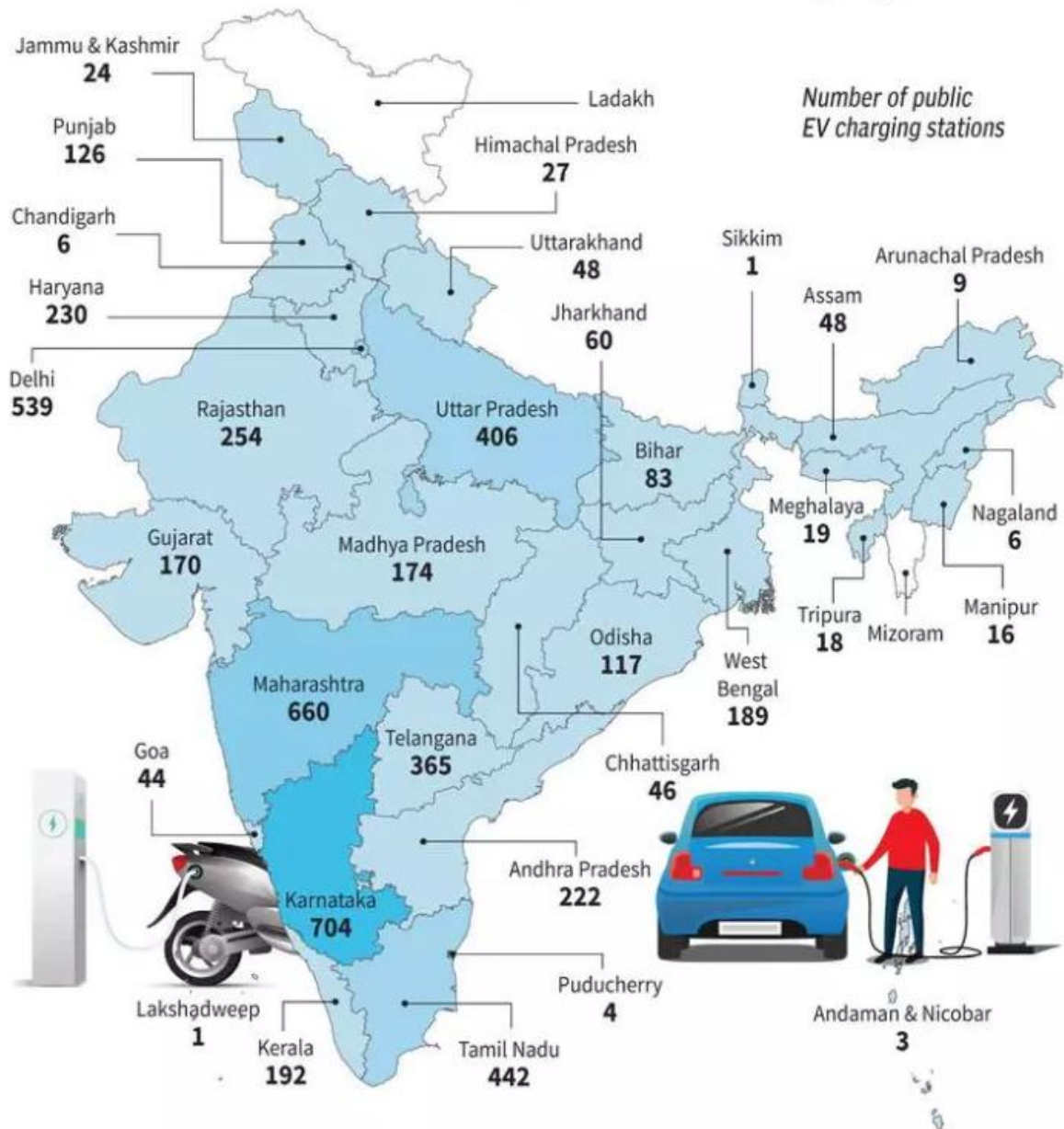
However, risk is moderate compared to Mumbai CBD or Delhi Connaught clusters.

Conclusion:

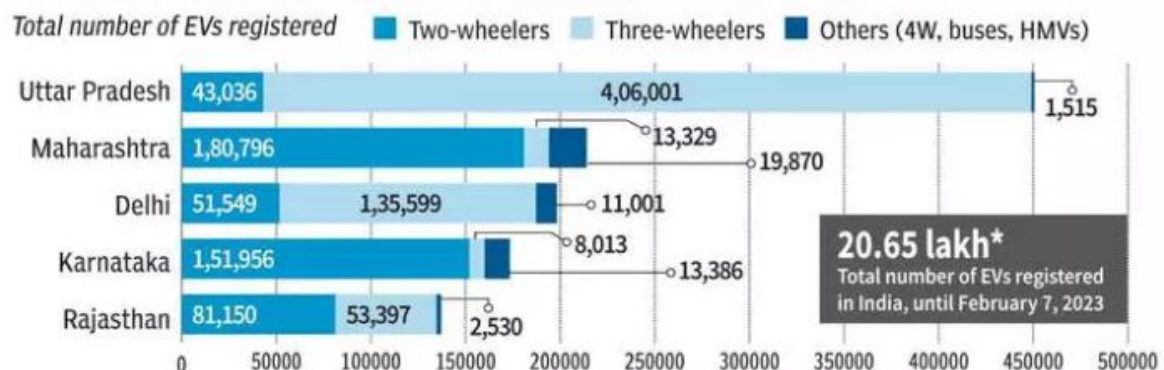
Hyderabad peripheral zones present redeployment opportunity without increasing fleet size.

#### 4. Charging Infrastructure Stress & Alignment

Karnataka has the most public EV charging stations



But UP and Maharashtra have the most number of electric vehicles



Key distinction:

The primary constraint is timing, not charger count.

Evidence:

- Charging wait peaks overlap with ride demand peaks.
- Off-peak charging capacity remains underutilized.
- Zones with sufficient chargers still experience peak congestion.

Structural vs Operational:

- Structural issue: Uneven geographic distribution (long-term CAPEX).
- Operational issue: Peak-hour charging clustering (short-term fixable).

From supplementary context:

India's charging expansion remains uneven and utilization-driven

## **Application of Analytical Frameworks**

### **1. Conversion Efficiency Framework**

Conversion Efficiency = Completed Rides / Requested Rides

High request growth without conversion efficiency improvement amplifies dissatisfaction.

### **2. Bottleneck Theory**

Primary bottleneck = Charge-ready fleet availability

Secondary bottleneck = Zone-hour misallocation

Optimizing bottleneck improves total system throughput.

### **3. EV Constraint Operating Model**

Unlike conventional ride-hailing:

- Charging introduces downtime.
- Drivers anticipate congestion.
- Supply is state-dependent (battery level).

The analysis applies a combination of operational reasoning and data-driven diagnostics to evaluate VoltRide's fulfillment performance. Rather than focusing solely on aggregate demand levels, .



## **Business Recommendations**

### **1. Dynamic Battery Threshold Dispatch .**

Instead of static minimum battery requirements:

- Adjust thresholds based on zone demand forecast.
- Allow low-battery rides in short-trip zones.
- Push predictive charging alerts during low-demand windows.

Expected benefit:

Immediate cancellation reduction.

Risk:

Battery depletion mismanagement if thresholds are too aggressive.

### **2. Peak-Hour Charging Staggering .**

Introduce:

- Incentives for off-peak charging
- Zone-based charging reservations
- Real-time congestion alerts

Expected benefit:

Reduced queue clustering.

Risk:

Driver compliance variability.

### **3. Smart Fleet Redeployment .**

Deploy idle fleet from:

- Low-utilization peripheral zones
- To high surge-intensity clusters

Supported by:

- Zone-hour historical demand heatmaps
- Battery-level distribution modeling

Expected benefit:

Improved supply-demand match without CAPEX.

Risk:

Temporary oversupply shifts.

## **4. Incentive Redesign**

Move from ride-count incentives to:

- Peak-hour availability bonuses
- Battery-ready rewards
- Zone compliance incentives

Aligns driver economics with platform conversion efficiency.

## **Assumptions Made**

The analysis is based on the provided dataset and assumes that the observed ride patterns represent typical weekday operational behavior in major urban markets. Battery percentage is treated as a proxy for vehicle readiness, assuming that lower battery levels reduce dispatch likelihood.

- Cancellation reasons accurately recorded.
- Battery levels reflect real readiness.
- Charging wait-time data is representative.

## **Scenario-Based Operational Improvement .**

**Highest impact-to-effort scenario: Battery Coordination + Charging Staggering**

Why?

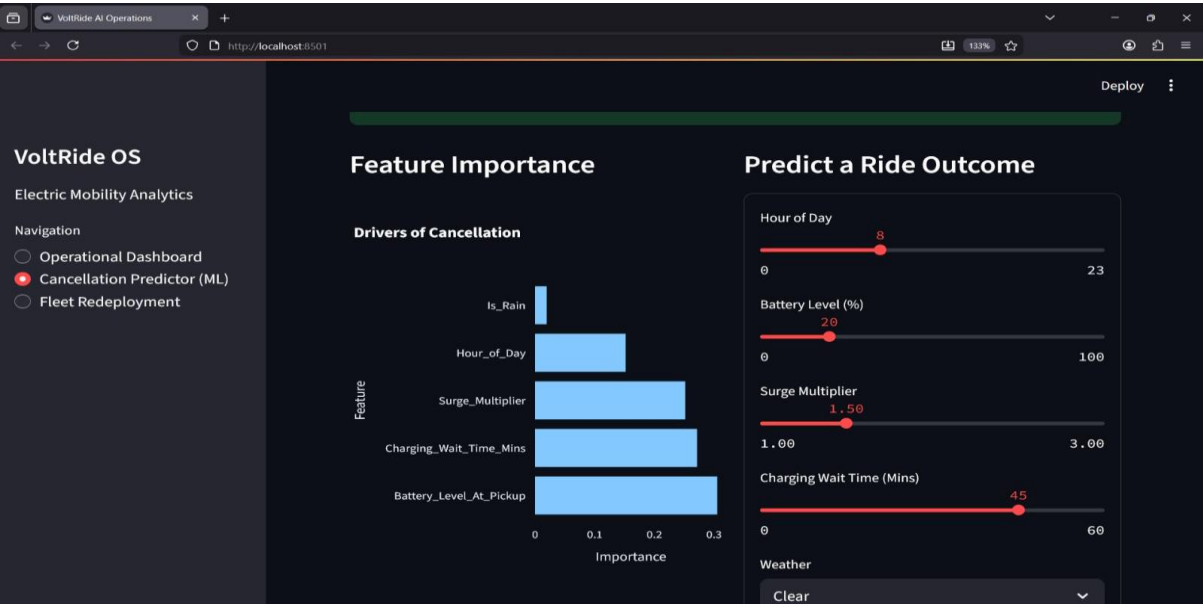
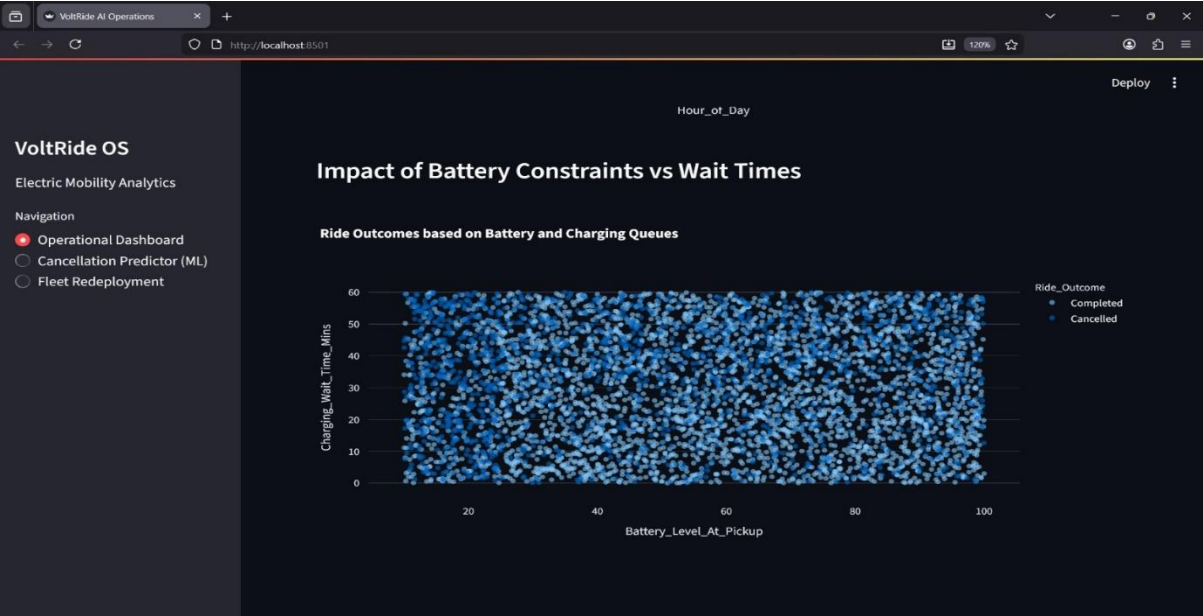
- No infrastructure expansion required.
- No fleet acquisition required.
- Leverages existing data systems.
- Directly targets primary cancellation driver.

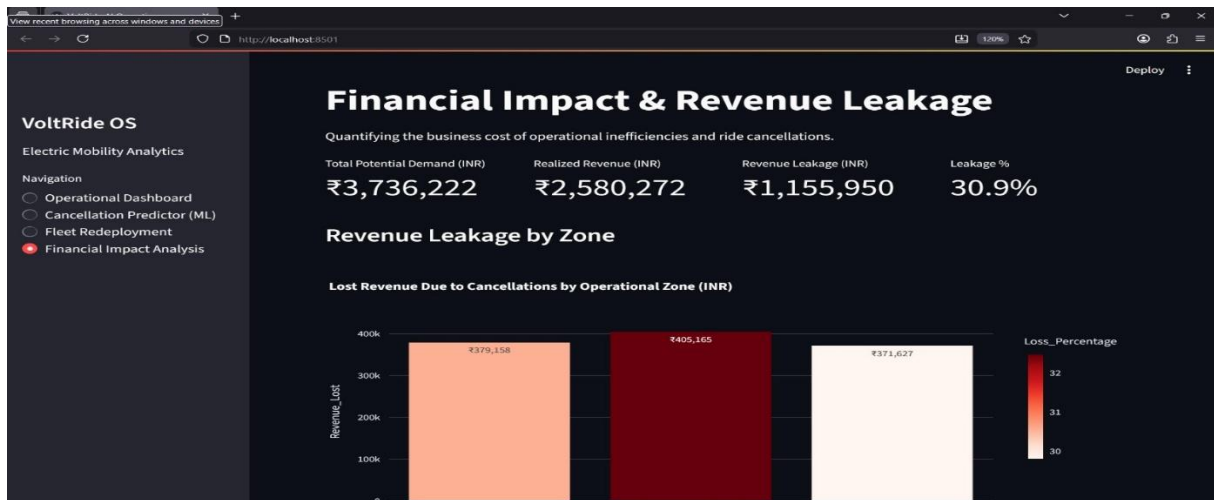
Expected impact:

- 8–15% improvement in ride completion during peak windows.
- Reduced surge volatility.



# Outputs :





**Git hub link** <https://github.com/codex826/Streamlit-EV-Operations-Analytics-Project>

## Conclusion

VoltRide's challenge is not growth. It is coordination.

The data reveals that:

- Peak-hour stress is predictable.
- Battery-driven cancellations are dominant.
- Charging congestion is temporal.
- Fleet misallocation exists across zones.

Scaling an EV-only fleet demands structured anticipation, not reactive dispatch.

VoltRide must shift focus from:

If managed correctly, EV constraints can become competitive advantage:

- Lower operating cost
- Lower maintenance
- Higher reliability