

Forecasting Bicycle Rental Demand

Data Analysis & Model Implementation for
Optimized Bicycle Logistics

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Agenda



INTRODUCTION &
BUSINESS
OBJECTIVES



DATA ANALYSIS
OVERVIEW



FORECASTING
MODEL &
PERFORMANCE



IMPLEMENTATION
STRATEGY



REFRESH INTERVALS
& PLANNING
HORIZON



KEY BENEFITS & NEXT
STEPS

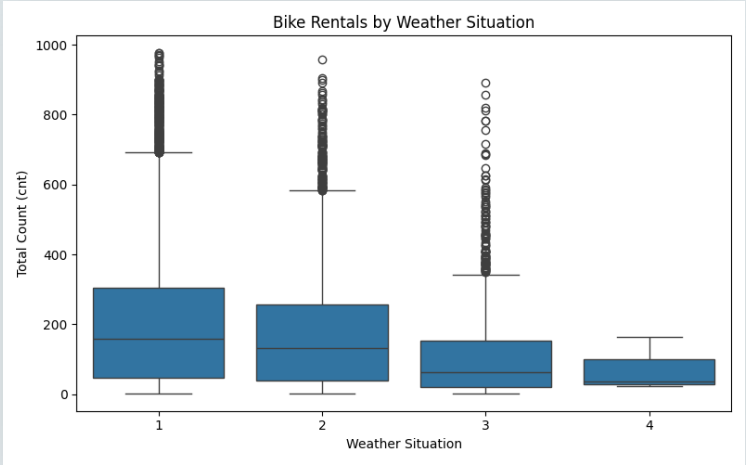
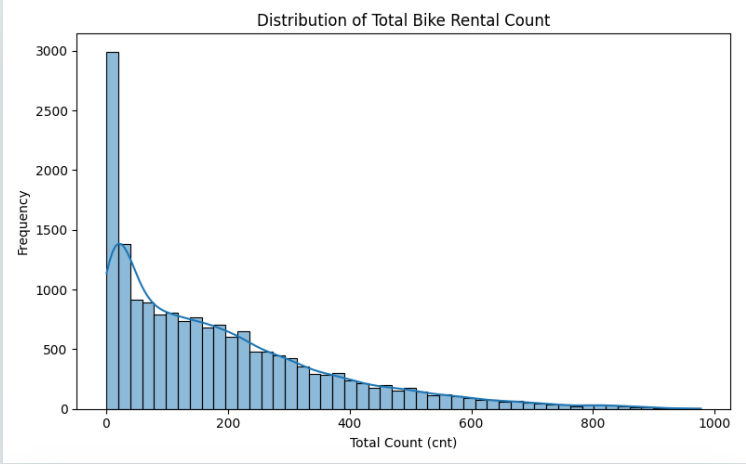
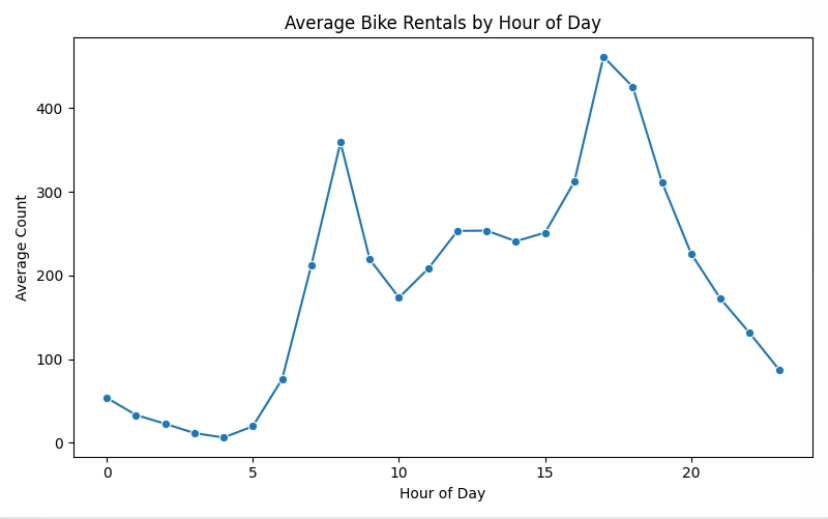
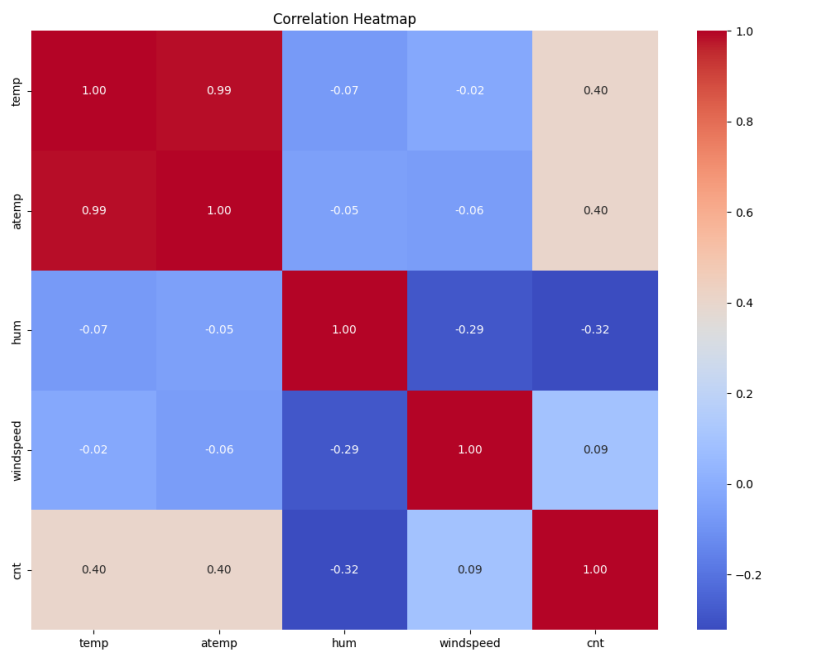
Business Objectives & Context

Objective: Ensure the right number of bicycles are available year-round.

Key Question: How many bikes should be ready each hour/day to meet demand?

Impact: Improve operational efficiency and customer satisfaction.

Data Analysis Overview



Forecasting Model – RandomForestRegressor

Model Choice:
RandomForest selected
based on best
performance (MAD: 32.86).

Strengths:


Handles nonlinear
patterns and feature
interactions.

Provides clear feature
importance for operational
insights.

Robust and scalable for
high-dimensional data.

Model Performance & Key Findings

Performance Metric: Mean Absolute Deviation
(MAD) = 32.86.



Interpretation: Lower MAD indicates more accurate hourly demand forecasts compared to alternative models.



Insights:

Strong seasonal and hourly
usage patterns.

Reliable performance for
operational decision-making.

Implementation Strategy

Integration: Deploy model into the existing IT infrastructure with regular data updates.

User Interface: Simple dashboard for planners showing hourly and aggregated forecasts.

Training & Support: Provide documentation and training sessions for staff.

Forecast Refresh Intervals & Planning Horizon



Refresh Intervals:

Operational Forecasts: Refresh forecasts daily to capture recent trends and weather updates.

Planning Adjustments: Weekly updates are recommended to account for changing patterns without overloading planning cycles.

Why: Daily updates ensure high responsiveness, while weekly refreshes balance operational agility and long-term planning.



Planning Horizon:

Short-term (up to 1 week): Reliable hourly forecasts support day-to-day operations.

Long-term: Use aggregated seasonal trends for annual planning, acknowledging that model accuracy decreases with longer horizons due to unforeseen factors.

Why: The model leverages strong seasonal patterns; however, real-time weather and local events can introduce variability beyond a week.

Key Benefits & Business Impact



Summary & Next Steps

Summary:

- Proven forecasting model with a MAD of 32.86.
- Actionable insights from data analysis support optimal bike availability.
- Practical implementation with daily to weekly forecast refresh intervals.

Next Steps:

- Pilot deployment and integration testing.
- Training sessions for planning staff.
- Regular review and iterative improvements based on feedback.