

# Hotel Revenue Forecasting

Group 11

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# Feature Engineering

## Foundational Features:

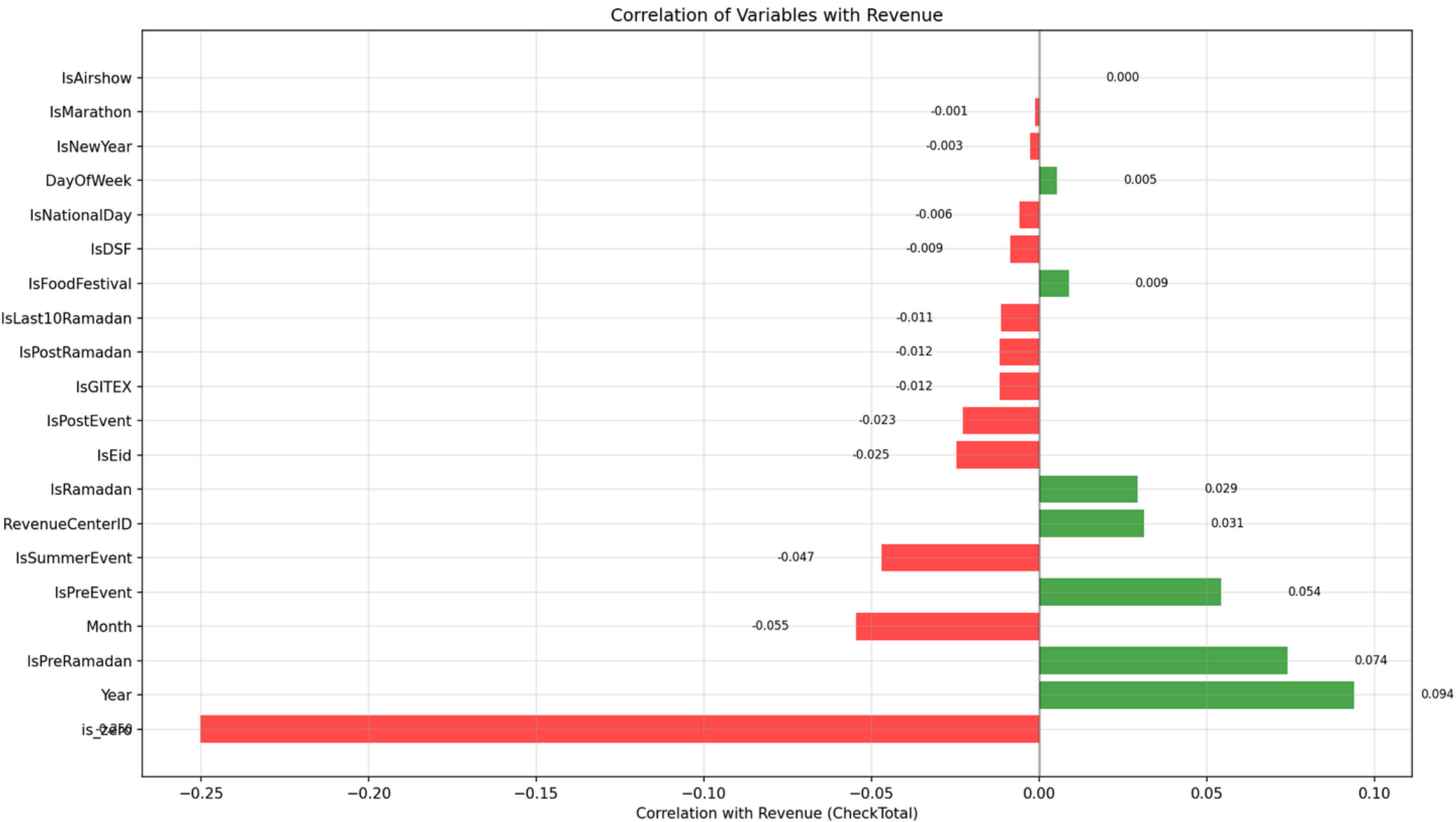
- Temporal
- Weekend

## Islamic Calendar & Holiday Analysis

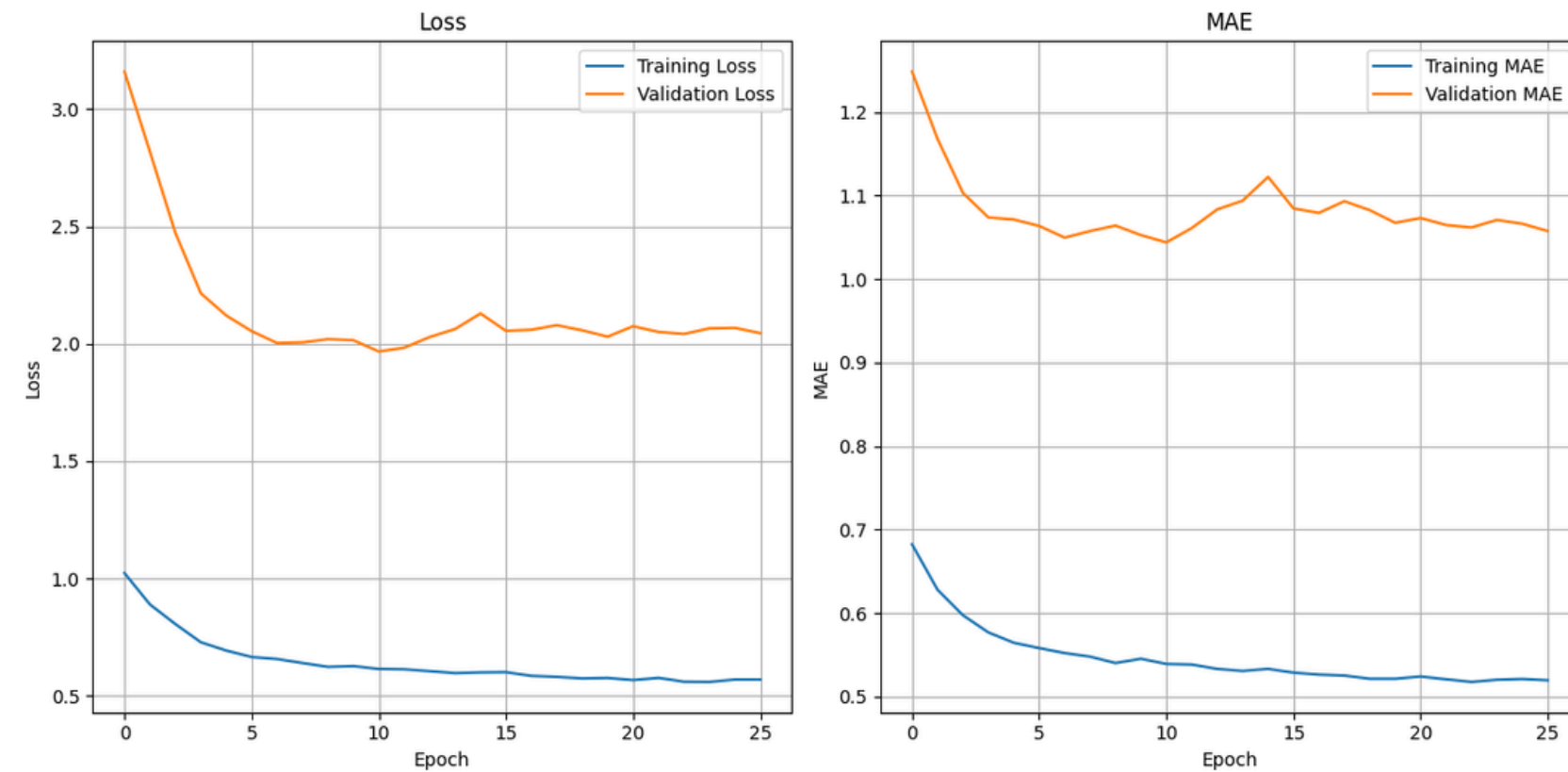
- Ramadan Segmentation
- Eid Cycles

## Behavioral Features

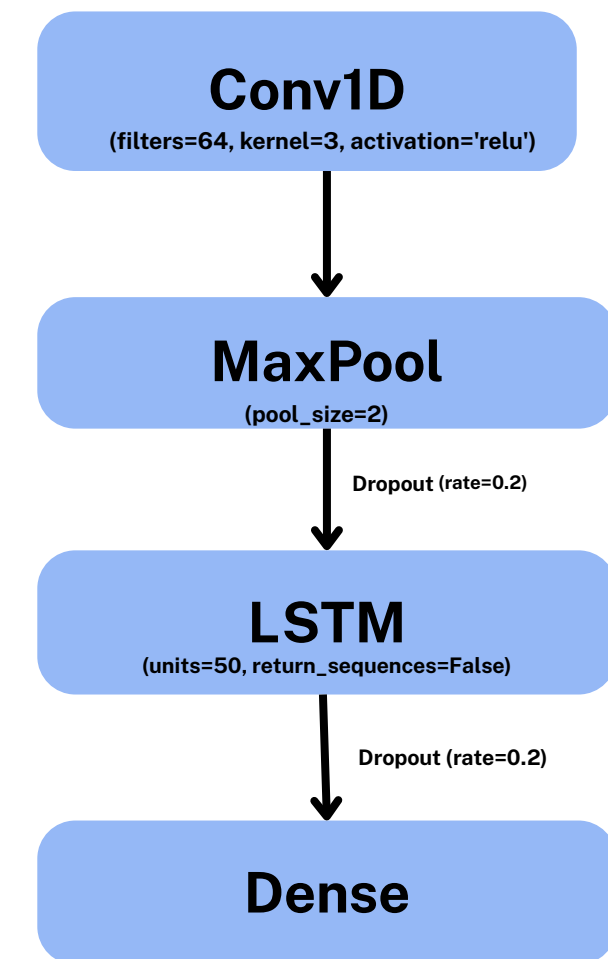
- Tourism Intensity



# Initial Approach: LSTM + CNN

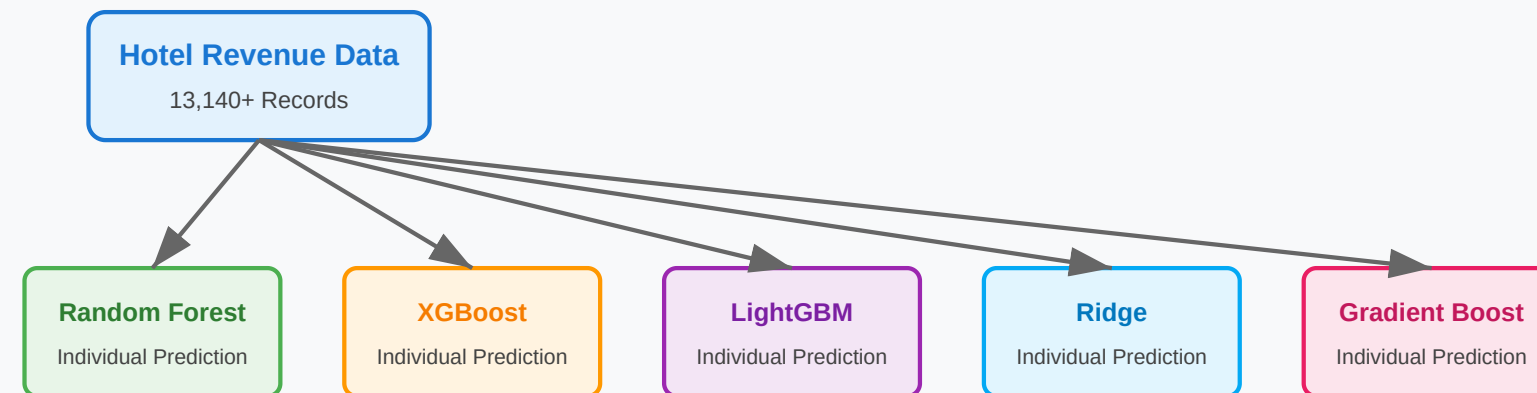


We observed a very high Mean Squared Error (MSE) and a low R-Squared value

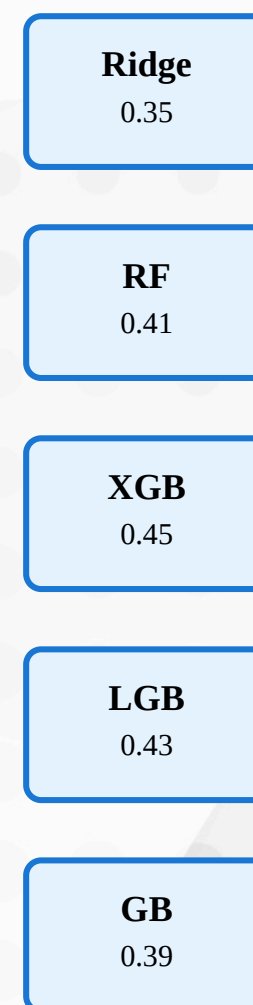


# The Ensemble Model

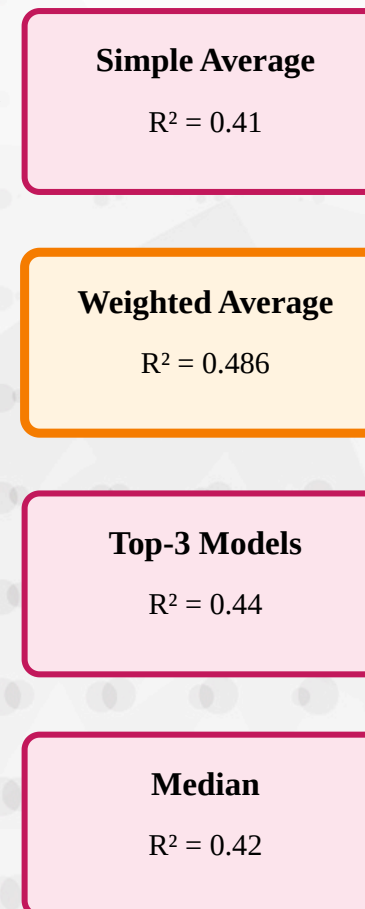
## Ensemble Methods for Hotel Revenue Prediction



### Base Models



### Ensemble Strategies



### Selection



# Results

Best Individual Model

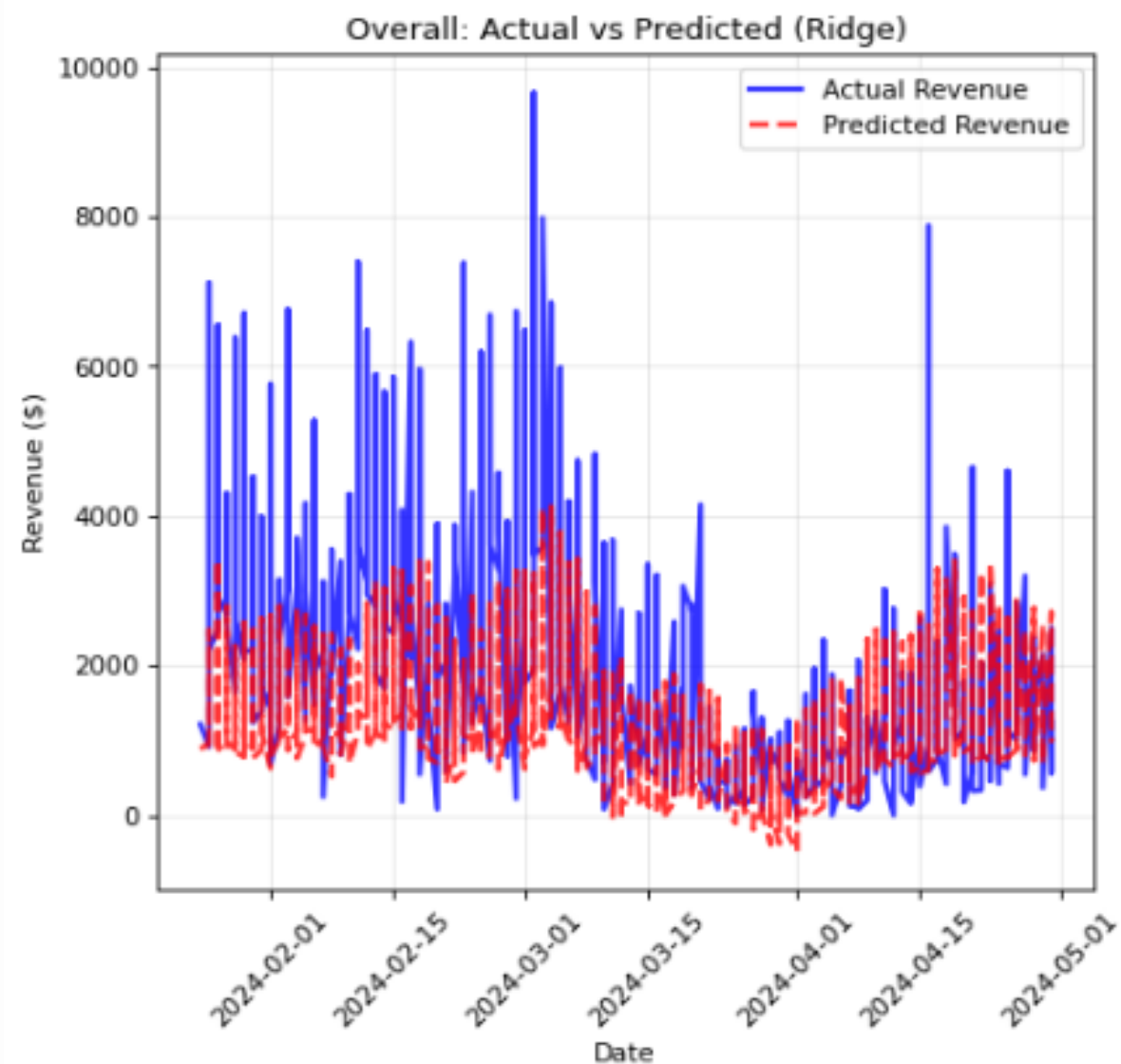
**ridge**  $R^2 = 0.503$

Best Ensemble Strategy

**top3\_average**  $R^2 = 0.471$

Total Models Trained

**5** Base models + ensemble strategies



## Performance Highlights:

- ✓ Ensemble approach provides stable predictions across time periods
- ✓ Strong correlation between actual and predicted values observed
- ✓ Model identifies high-revenue periods and some seasonal dips

# Challenges

- ◆ **Limited Data Availability**
  - Insufficient historical records for robust training
  - Reduced model generalization capability
- ◆ **Deep Learning Implementation Barriers**
  - Insufficient data volume for neural networks
  - Traditional ML outperformed due to dataset size
- ◆ **Model Selection Complexity**
  - No single optimal model found
  - Required ensemble approach for best results
- ◆ **Zero Revenue Frequency**
  - High zero-value entries created imbalance
  - Skewed learning patterns
- ◆ **Data Quality Issues**
  - Inconsistent patterns and missing values
  - Extensive preprocessing required





**Thank You**



