

Forecasting Ambient Temperature with RNNs

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- Linear vs Sequence Models

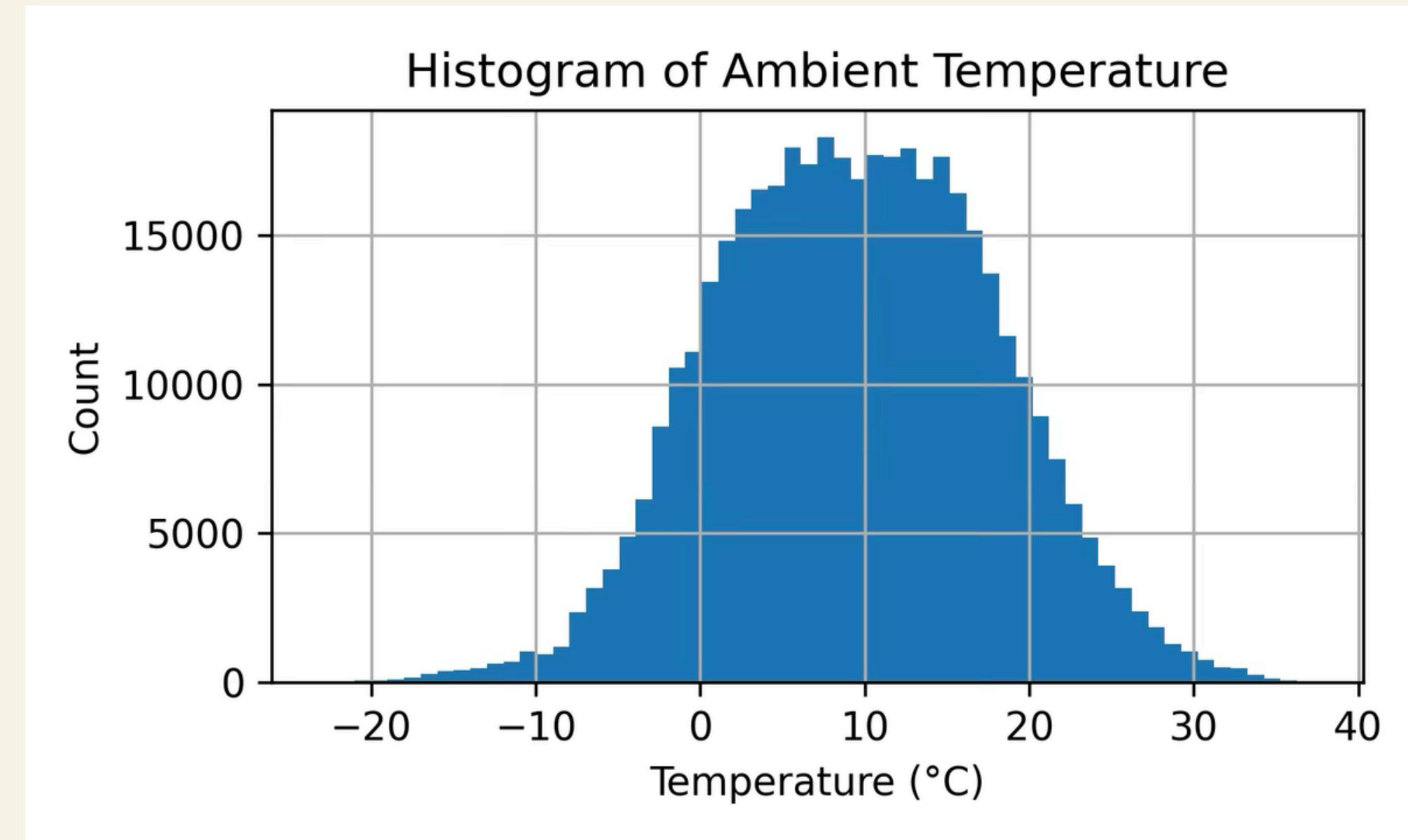
Mintesnot Kassa, Fordham University

Problem

- Need accurate short-horizon forecasts
- Applications: HVAC, energy, scheduling
- Temperature shows strong cycles

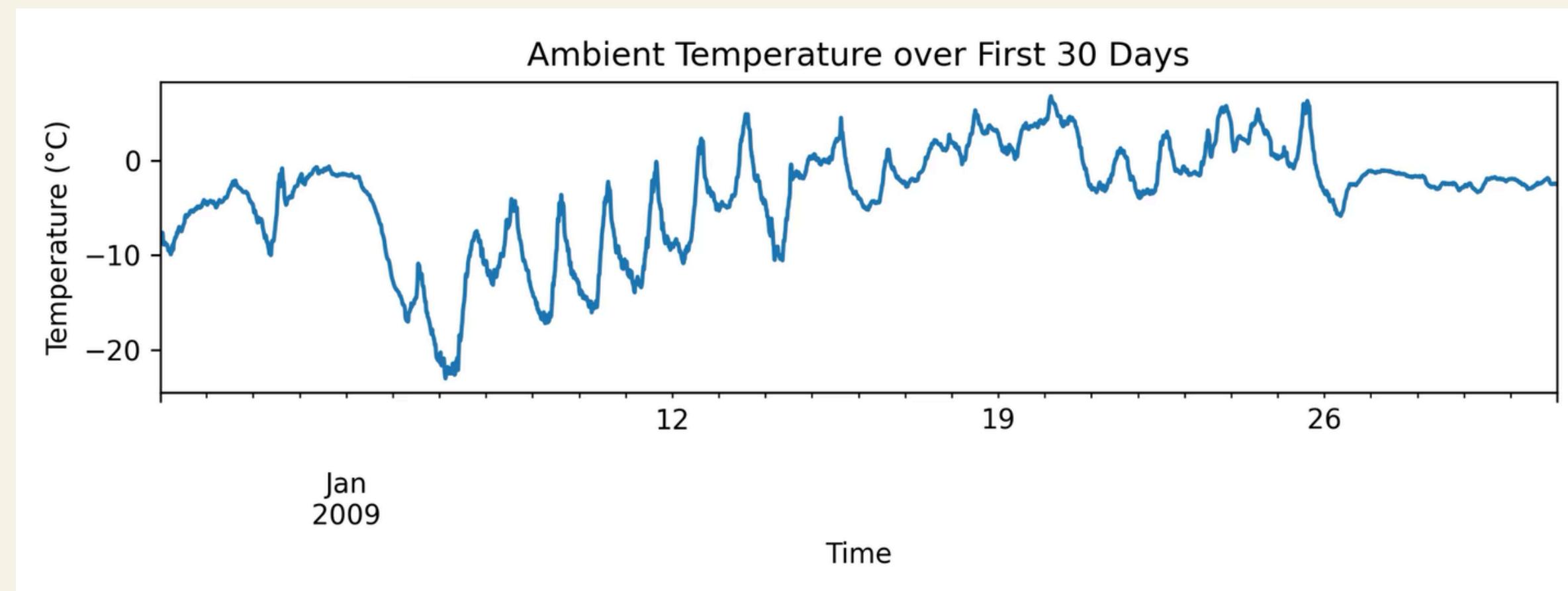
Dataset Overview

- Jena Climate Dataset (2009–2016)
- 10-minute resolution



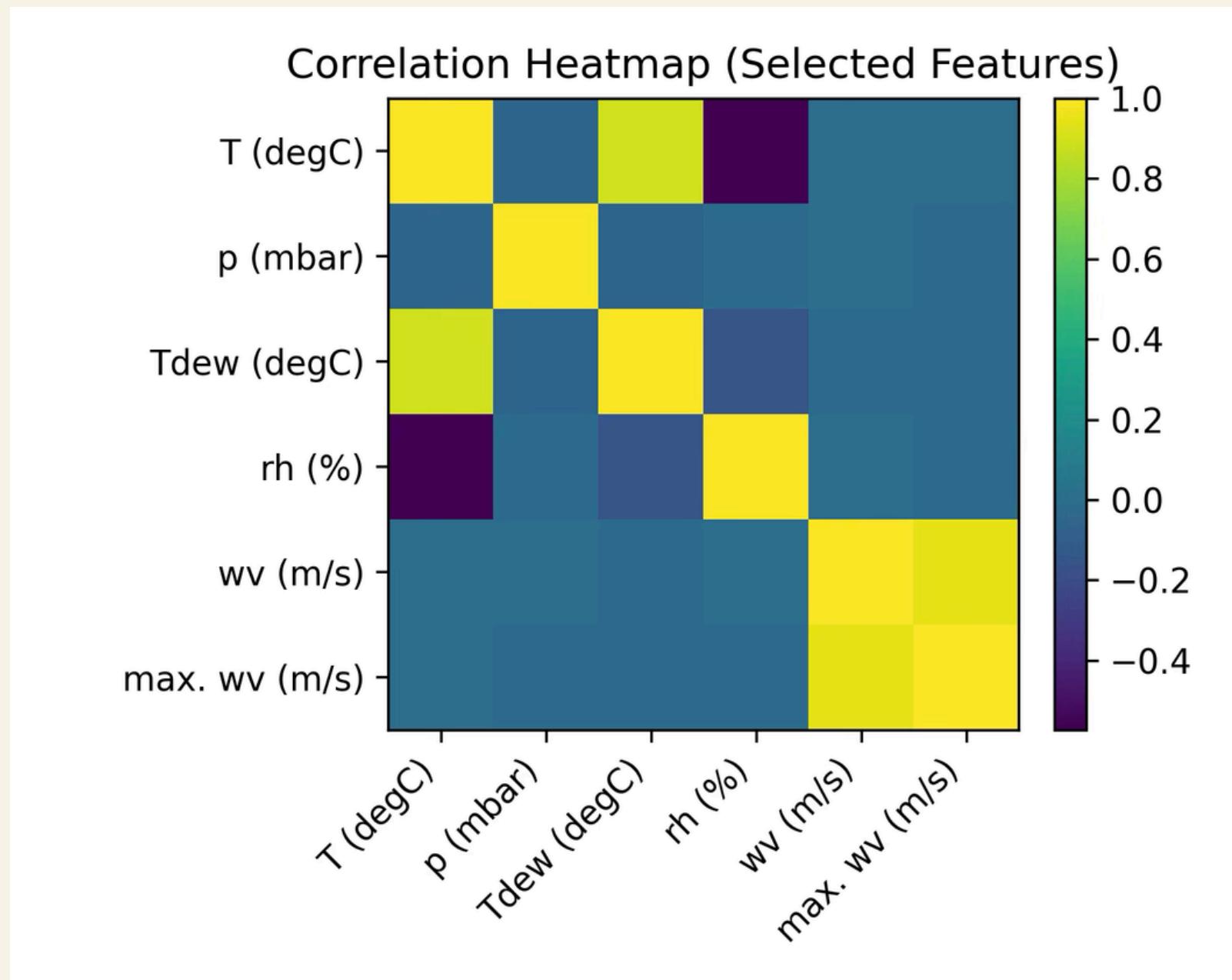
First 30 Days Visualization

- Strong diurnal structure and weather variability.



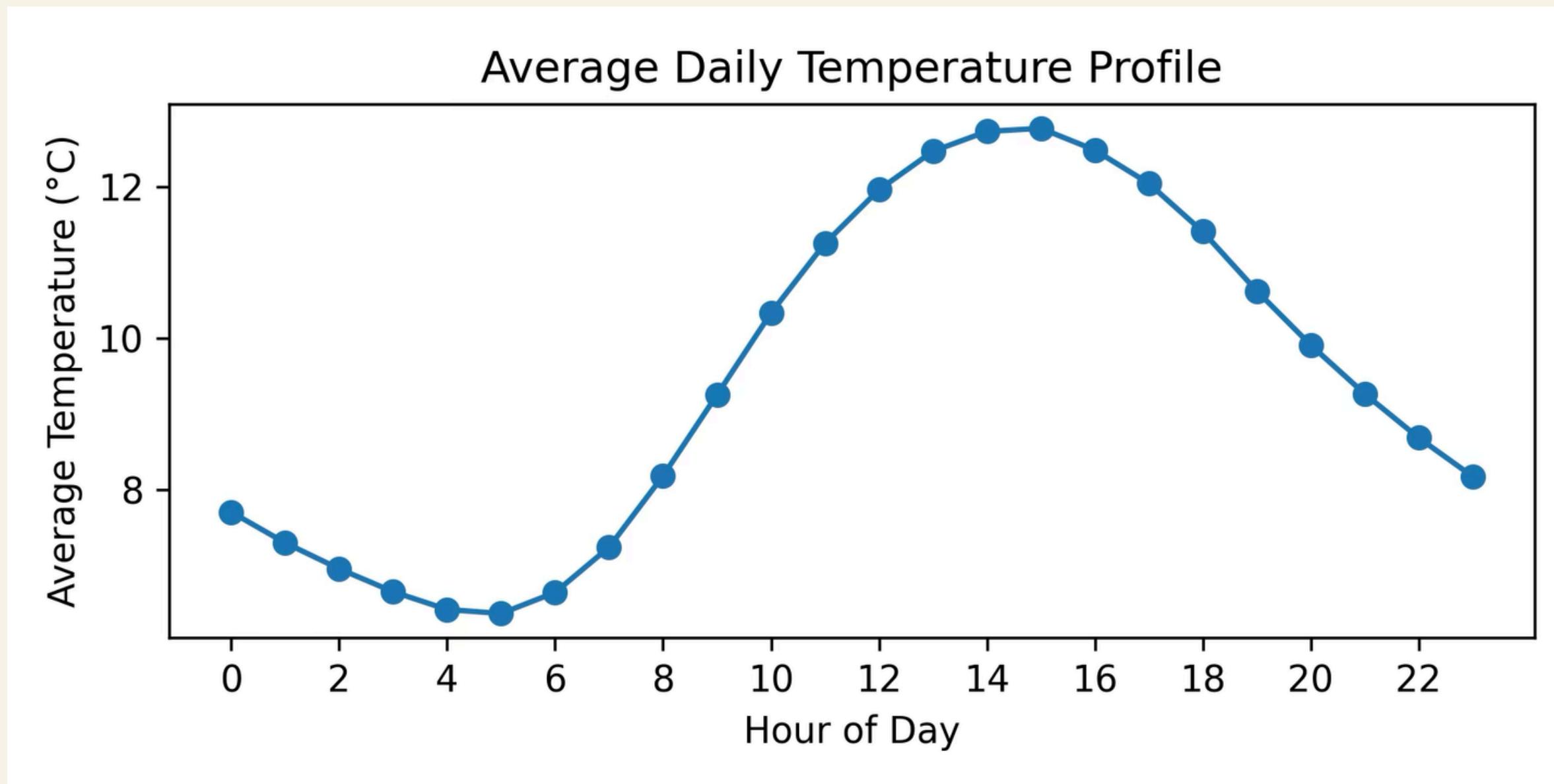
Correlation Analysis

- Correlations among key meteorological features.



Daily Temperature Profile

- Shows average diurnal cycle across dataset.



Windowing & Problem Setup

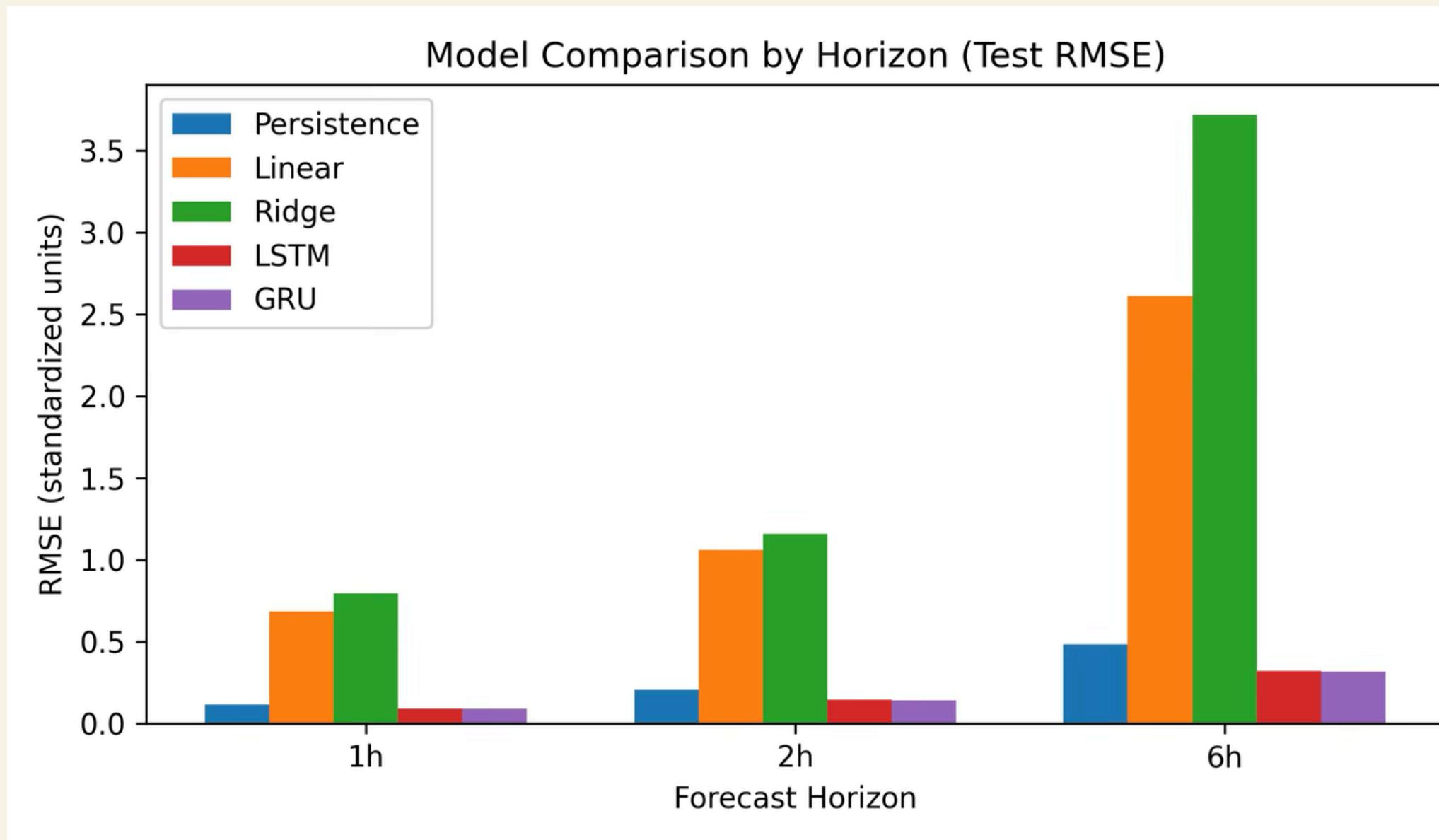
- History windows of 1,2,4,8 hours
- Forecast horizons: 1h, 2h, 6h
- Standardization applied

Models Compared

- Persistence, Linear, Ridge
- LSTM & GRU
- Adam optimizer, early stopping

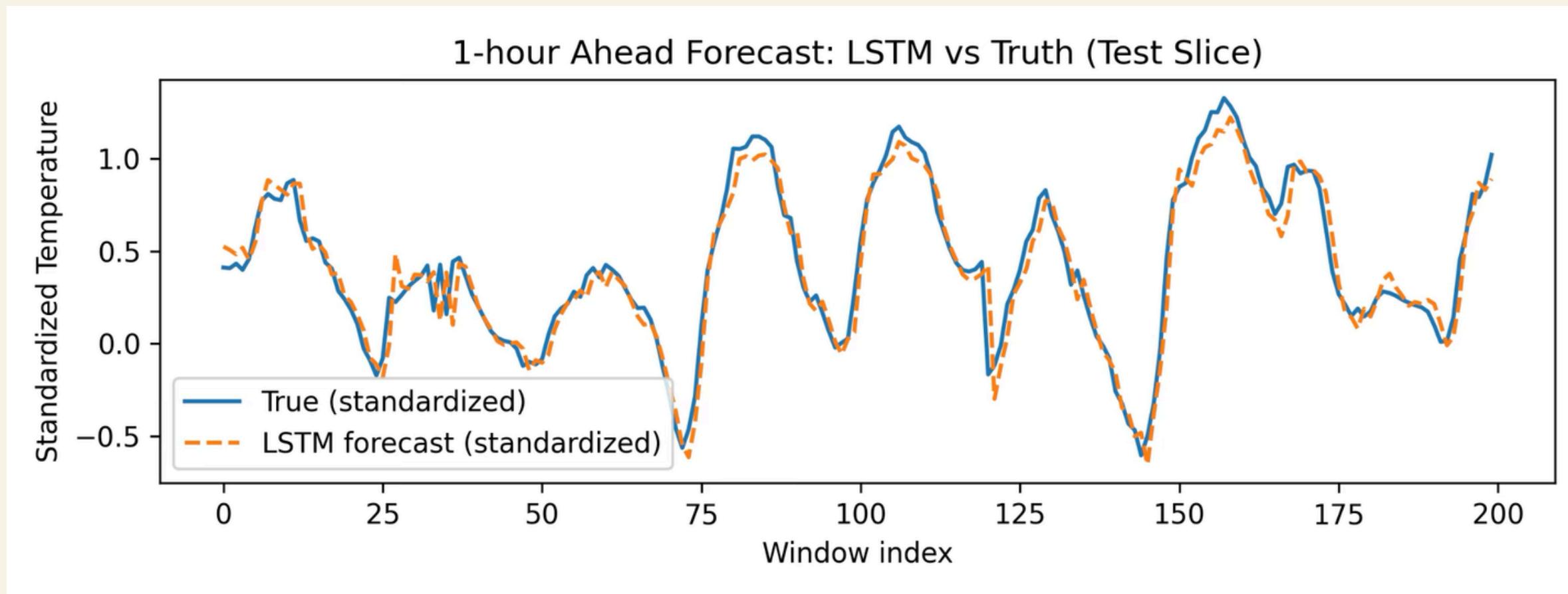
RMSE Comparison by Horizon

- RNNs outperform at longer horizons, baselines strong at 1h.



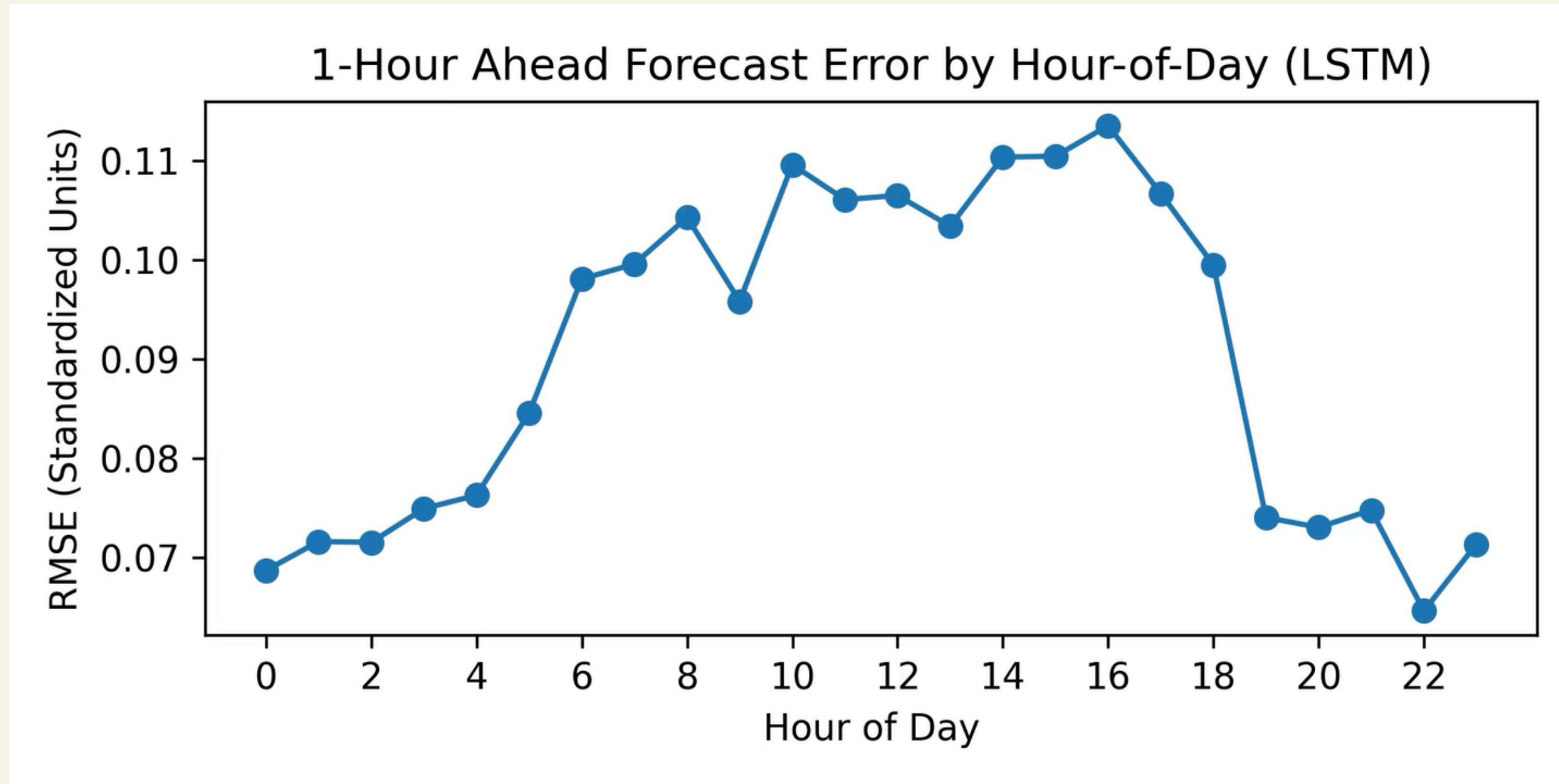
Forecast Example

- LSTM prediction vs true temperatures on test slice.



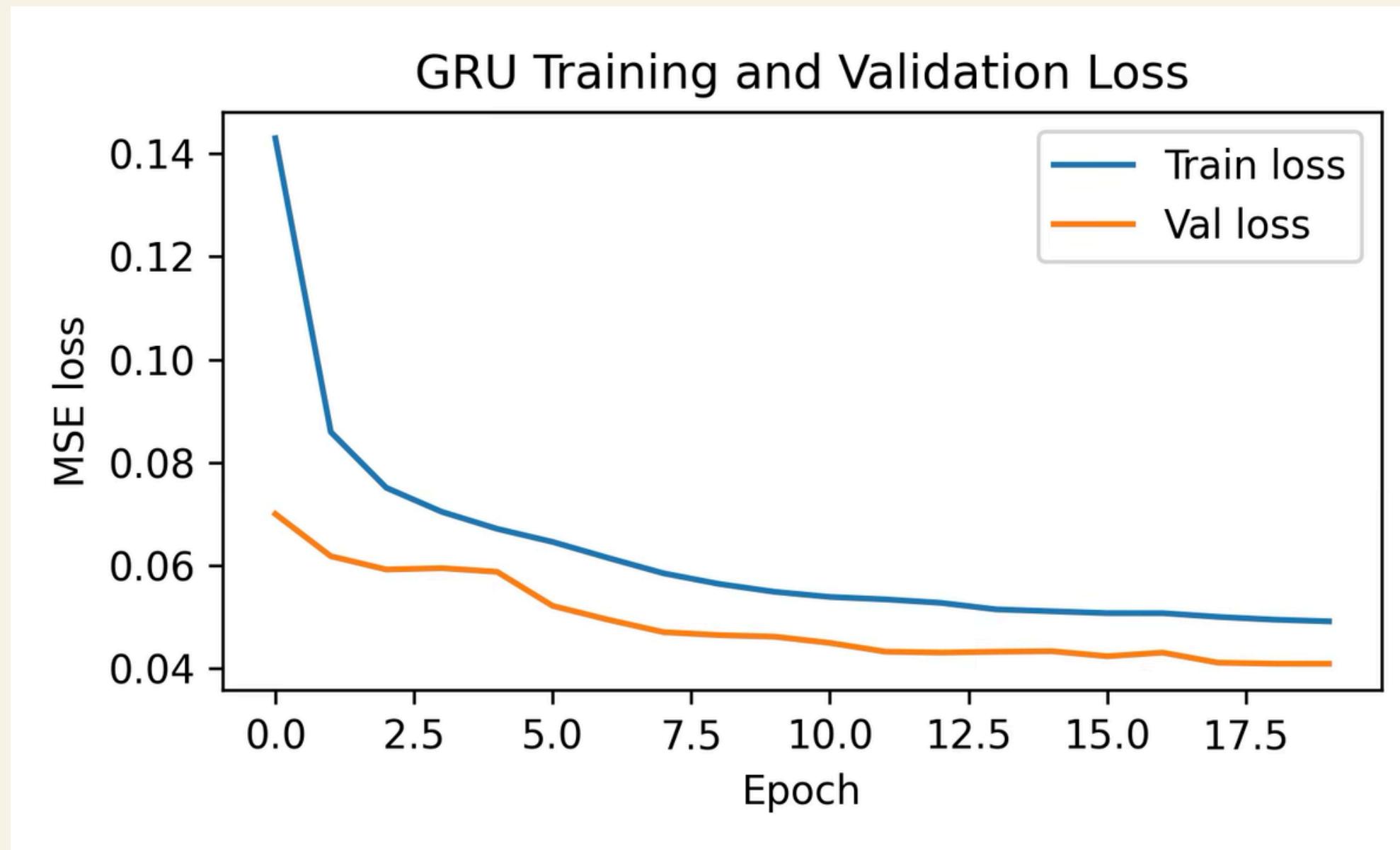
Error by Hour of Day

- Higher error at dawn/dusk; smoother at midday.



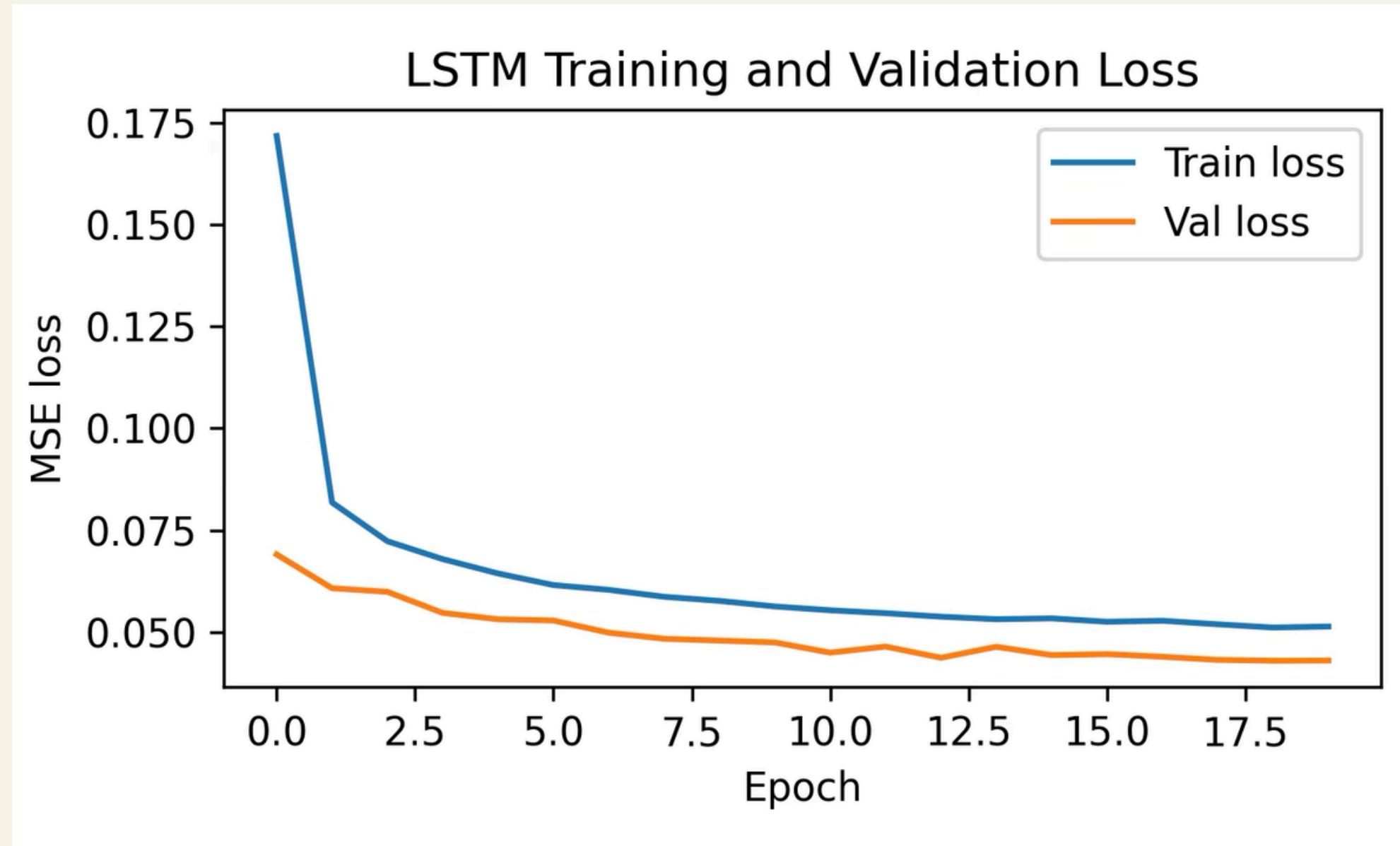
GRU Training Curve

- Validation loss stabilizes after ~10 epochs.



LSTM Training Curve

- Shows similar convergence properties with slight variance.



Key Observations

- 1h: persistence nearly optimal
- 6h: RNNs clearly better
- Ridge reduces overfitting

Conclusion & Future Work

- Extend to TCN/Transformers
- Add seasonal variables
- Probabilistic forecasting