

EECE 693 - Energy Demand Forecasting

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Problem Statement

Forecasting household energy demand is critical for grid stability, resource allocation, and sustainable planning. Modern consumption data exhibit non-linear patterns, multiple seasonalities, and long-range dependencies due to renewable integration and electric vehicle adoption, challenging traditional forecasting methods.

Dataset & Features

- **Source:** IHEPC dataset (UCI ML Repo)
- **Target:** Global Active Power (kW).
- **Exogenous:** Temp, Humidity, Holiday (France).
- **Features:** Lags (1h, 24h, 168h), Calendar (hour, day, month).

Model Architectures

Baselines & LSTMs:

- ARIMA(0,1,3): Classical baseline.
- Single-layer LSTM (1h look-back).
- Stacked Bi-LSTM + Attention (24h look-back).
- Stacked LSTM (12h look-back).

Standard Transformer:

- Encoder-decoder (MH=4, $d_{model} = 64$, L=3).

Evaluation Metrics

Key metrics quantify prediction error (kW):

- **RMSE:** Penalizes larger errors.
 $\sqrt{\frac{1}{N} \sum (\hat{y}_i - y_i)^2}$
- **MAE:** Average absolute difference.
 $\frac{1}{N} \sum |\hat{y}_i - y_i|$

Performance Comparison (Test Set)

Model	RMSE (kW)	MAE (kW)
ARIMA	0.750	0.620
Single-Layer LSTM	0.577	—
Bi-LSTM + Attn (24h)	0.520	0.380
Stacked LSTM (12h)	0.540	0.370
Transformer (Standard)	0.811	0.581
Hybrid LSTM–Transf.	0.0093	0.0058

Proposed Hybrid LSTM–Transformer Architecture

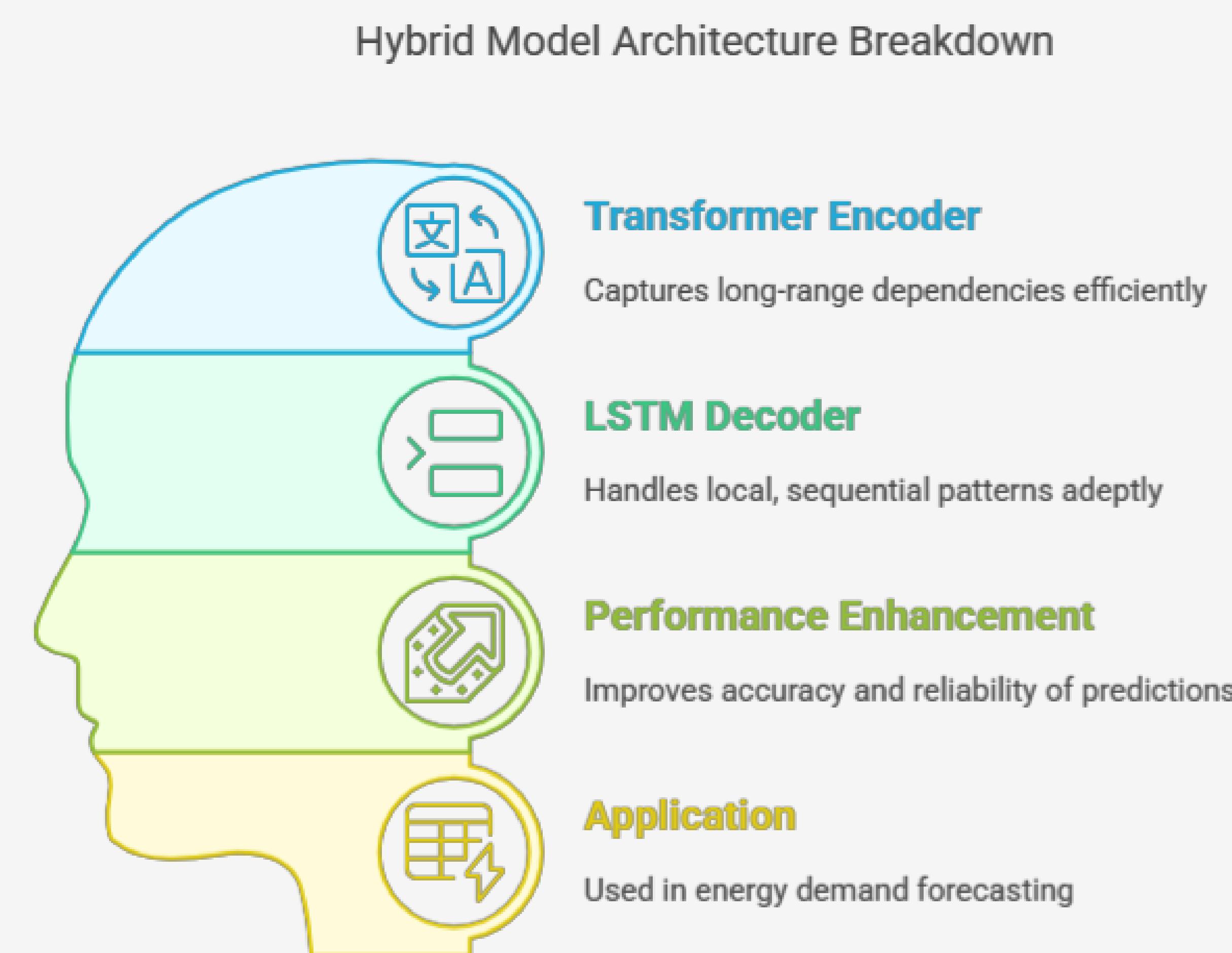


Figure 1: Hybrid model: parallel LSTM (local patterns) Transformer (global dependencies) branches fused for final forecast.

Prediction Samples

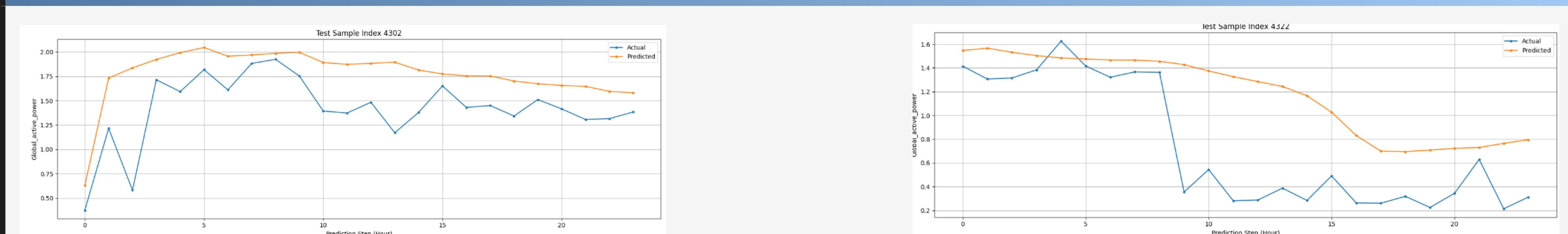


Figure 2: Actual vs. Predicted energy demand for two sample periods (Transformer model).

Conclusion

The proposed **Hybrid LSTM–Transformer** model effectively combines architecture strengths:

- Achieved lowest error (**RMSE 0.0093 kW, MAE 0.0058 kW**).
- Outperformed standalone LSTMs and standard Transformer.
- Captured local (LSTM) global (Transformer) dependencies.
- Promising for complex time series forecasting.

Full Series Comparison

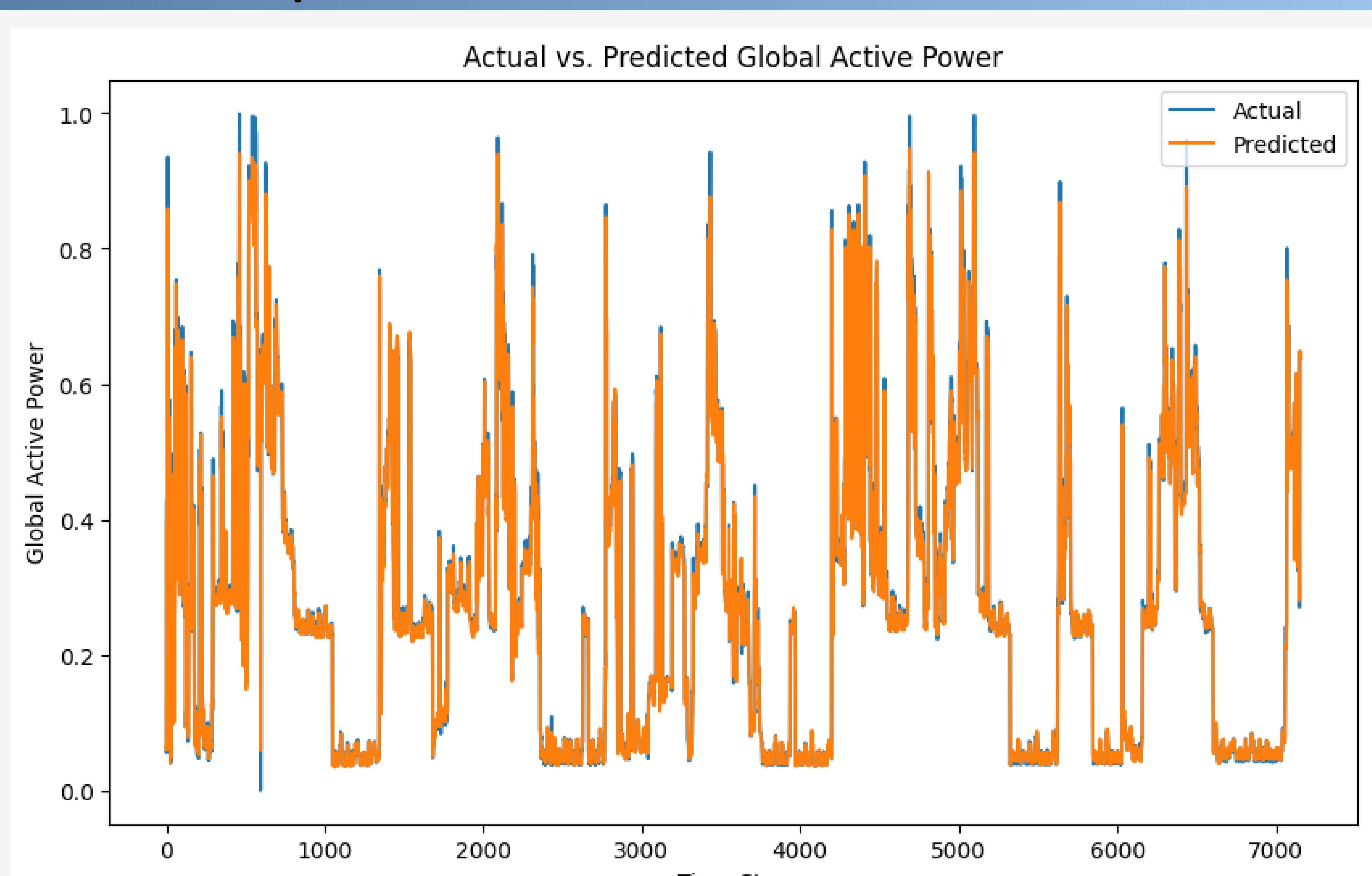


Figure 3: Hybrid model prediction across the entire test period.