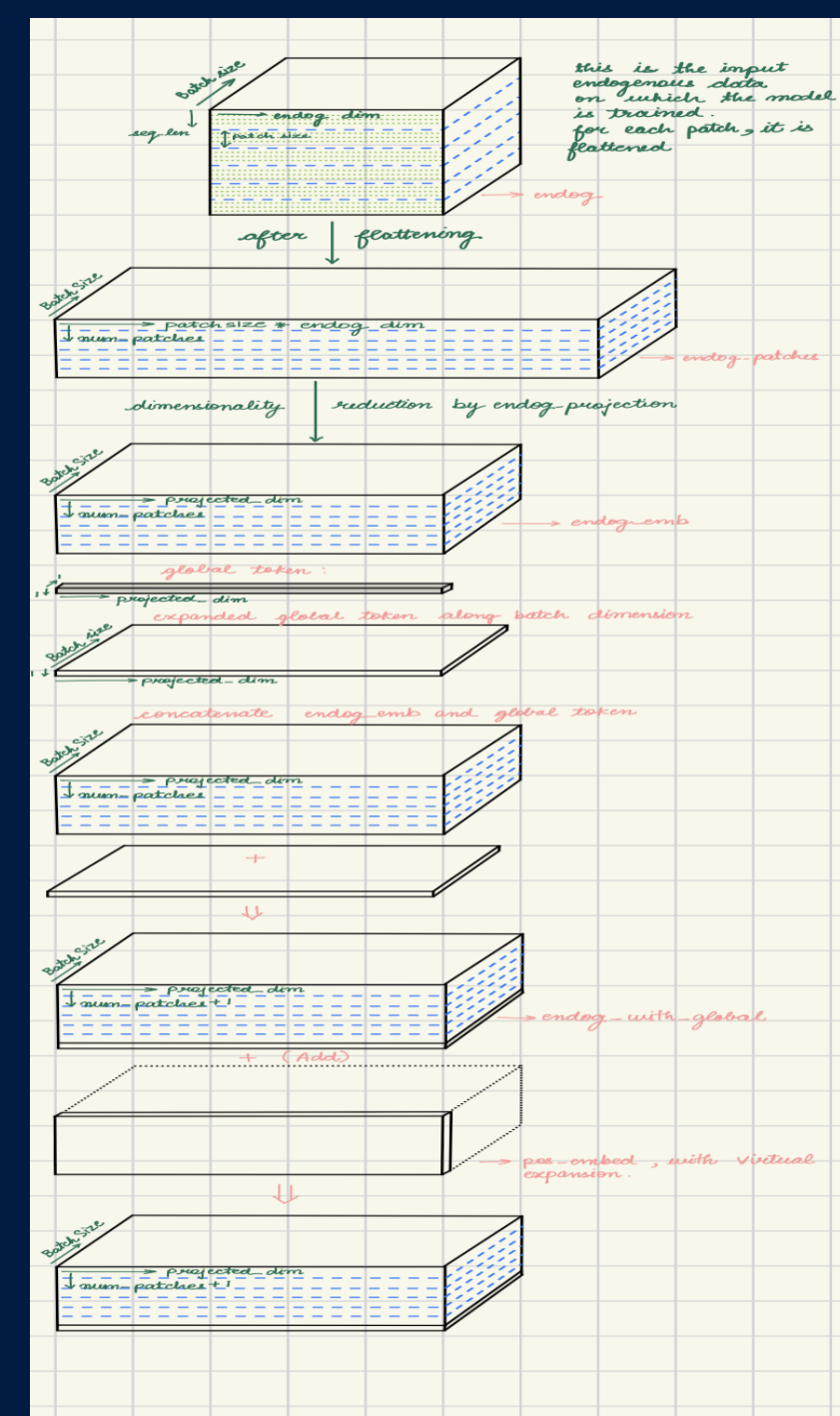
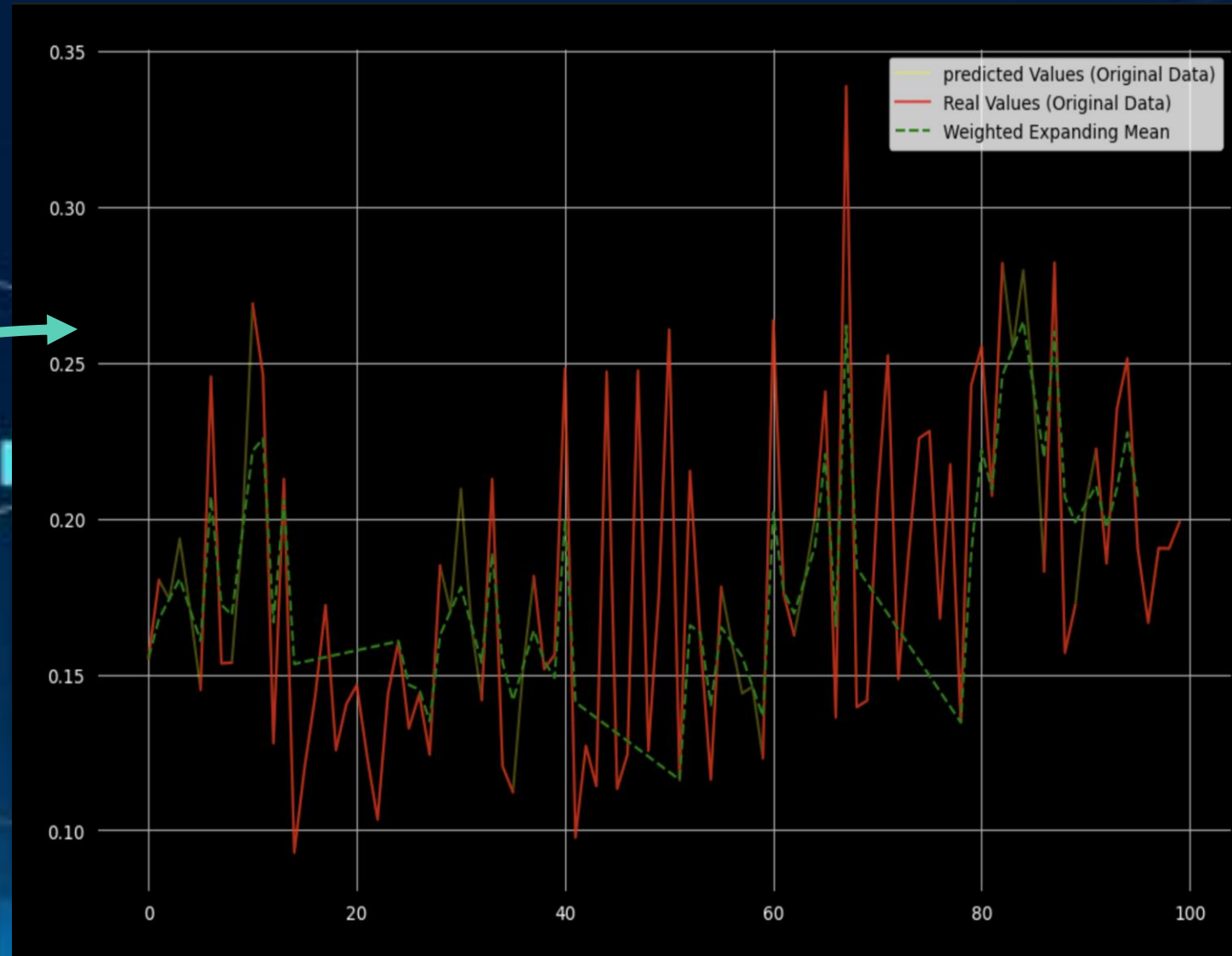
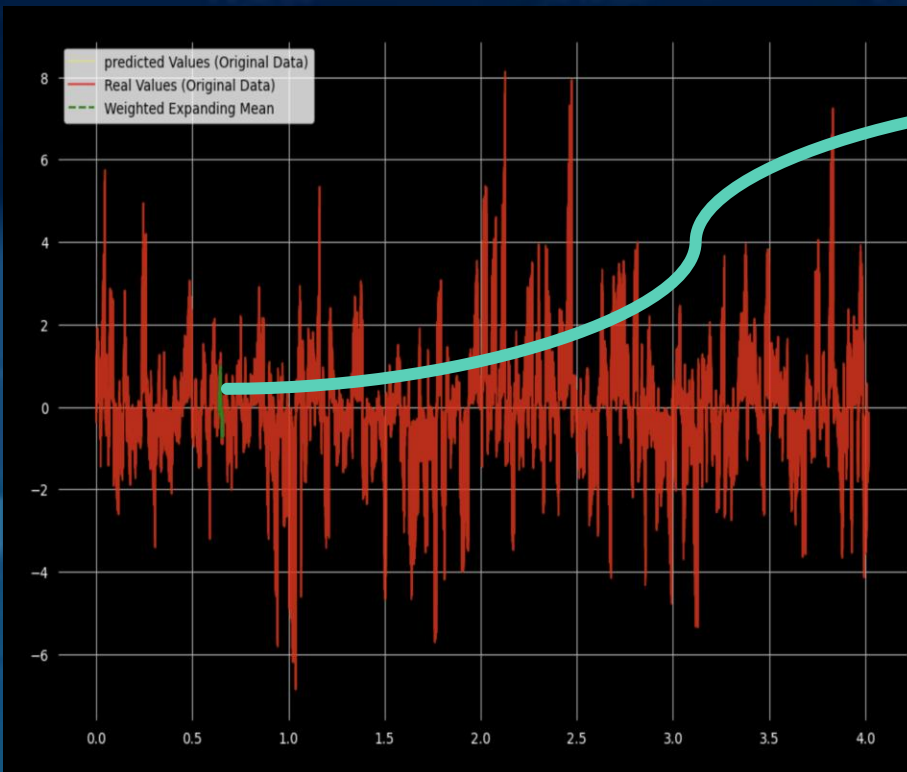


# Data Pipeline

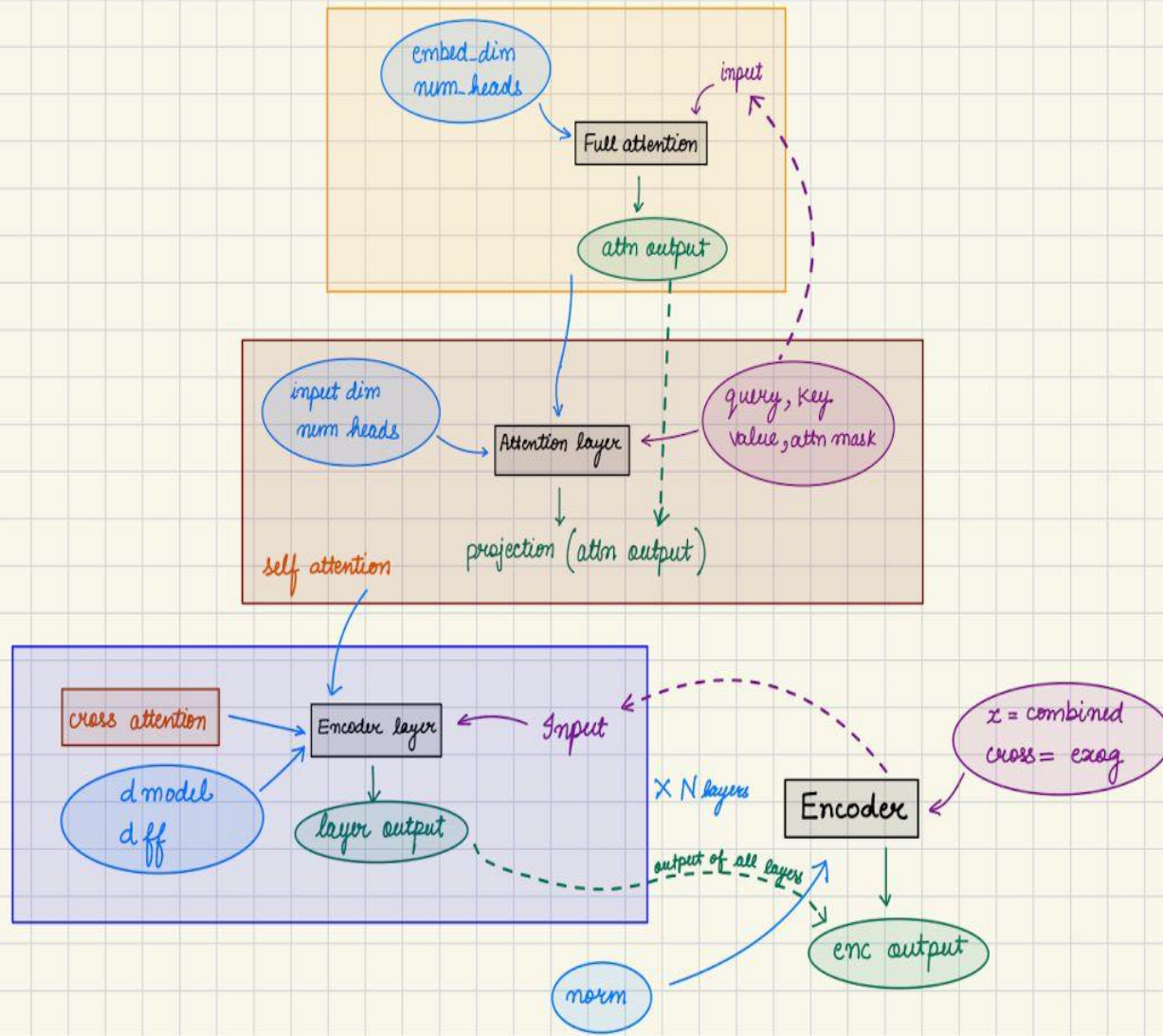
- **Input:** Multivariate Time-series of shape  $\rightarrow [\text{batch\_size}, \text{num\_features}, \text{seq\_len}]$
- **Flattening:** Endogenous features are flattened for each patch  $\rightarrow [\text{batch\_size}, \text{seq\_len} \times \text{num\_features}]$
- **Exogenous join:** Append learned embeddings (e.g., 8D) for each feature  $\rightarrow$  richer context per feature
- **Linear projection:** Combined input is projected to a shared projected\_dim using a feedforward layer
- **Global token extraction:** One token is cropped from each sequence to represent global temporal context
- **Token fusion:** Global token is concatenated back with the sequence for joint learning
- **Positional embedding:** Fixed sinusoidal positions are added to the input sequence
- **Transformer encoder:** 2 stacked blocks with:
  - Multi-head self-attention + cross-attention
  - Feed-forward MLP + layernorm + dropout
- **Forecast head:** sequence is flattened  $\rightarrow$  MLP head predicts future values (e.g., Next 1 step)
- **Output:** final shape  $\rightarrow [\text{batch\_size}, \text{output\_dim}]$ , trained with **weighted huber loss**



## Data Pre-Processing



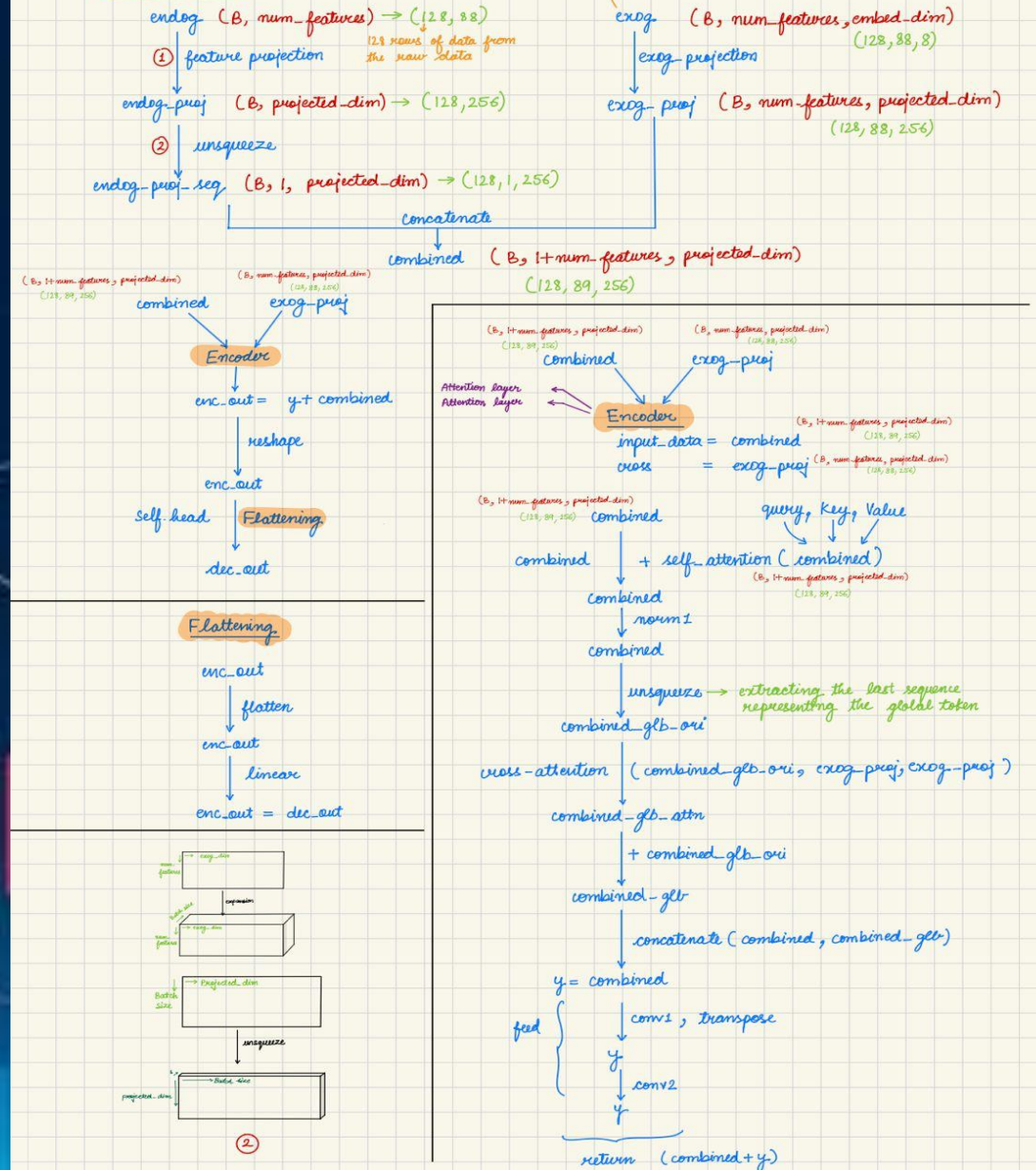
## Encoder



Version 2.7: Few fixes in 2.6

Code Flow:

Timexer Model:





# FORECAST ILLUSTRATION

