

# NILMFormer: Non-Intrusive Load Monitoring that Accounts for Non-Stationarity

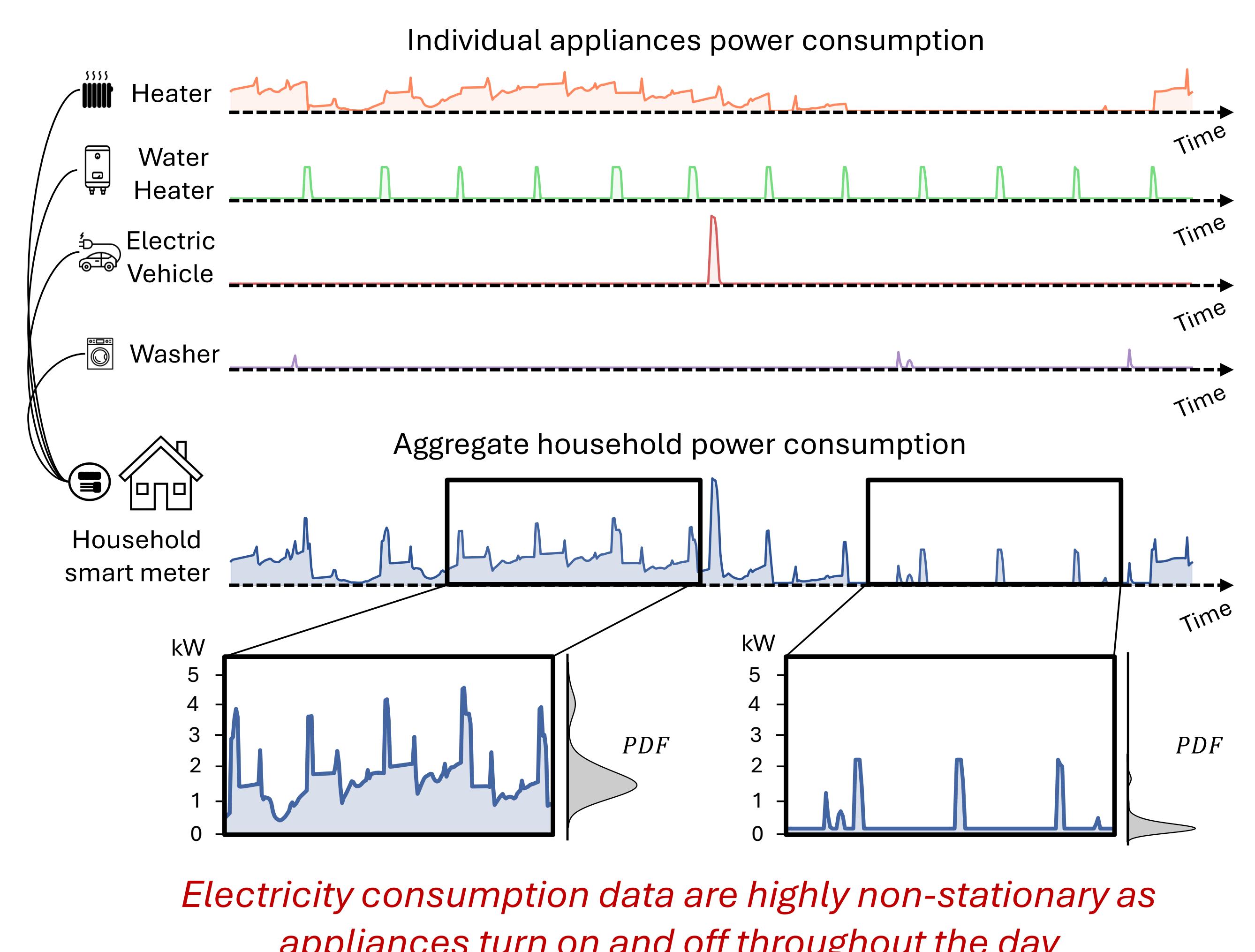


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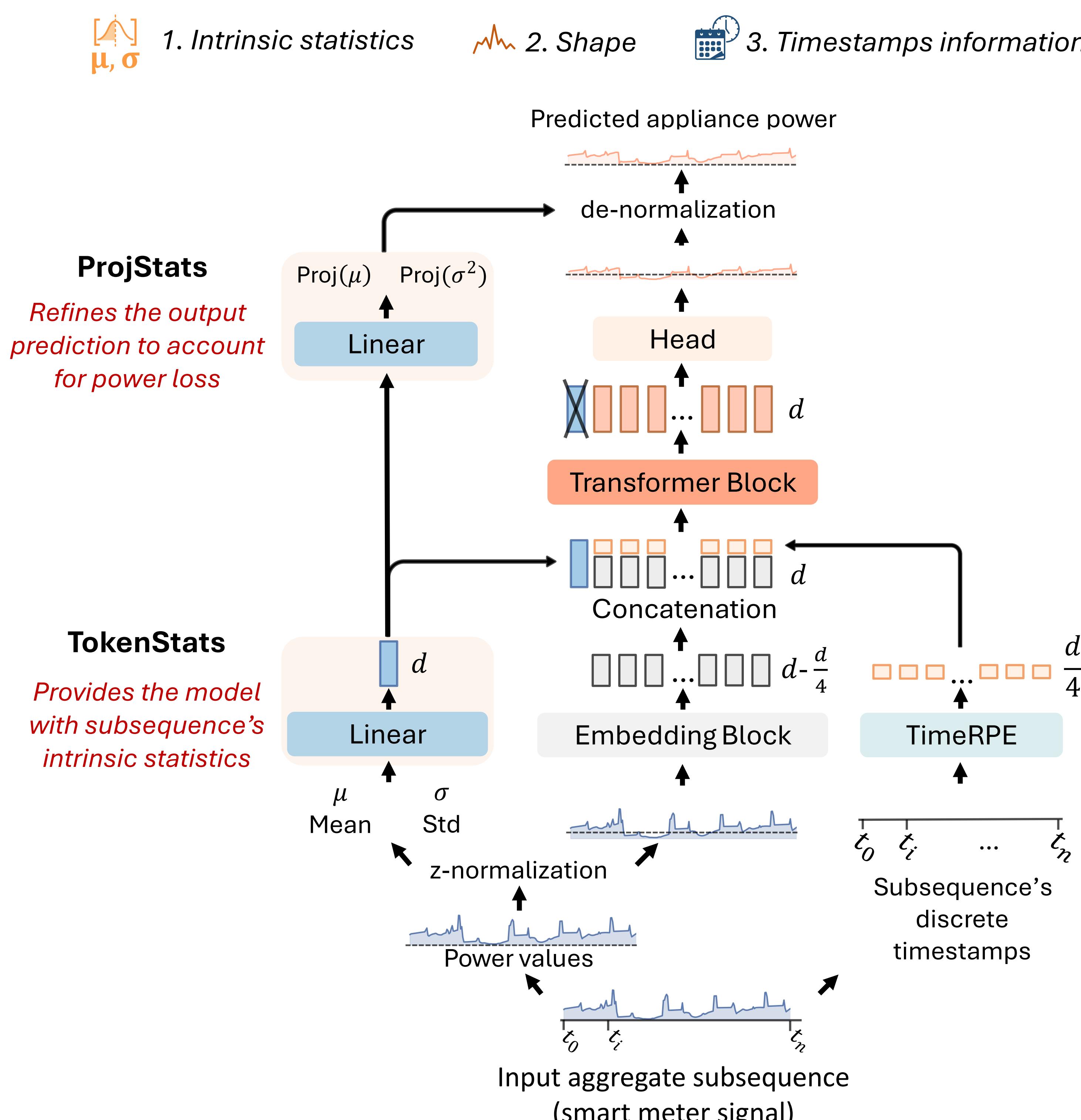
## Challenges & Motivation

- Smart meters ≠ insight: They only capture whole-home load, while decarbonization needs **appliance-level detail**.
- NILM in a nutshell: **Non-Intrusive Load Monitoring<sup>1</sup>** infers each appliance's usage from the **smart meter aggregate signal only**.
- Why it matters: Appliance-specific feedback cuts **consumption by ≈12%**<sup>2,3</sup>, yet utilities still report in monthly or yearly chunks.
- EDF's monitoring solution: **Electricité De France**, the main electricity supplier in Europe, proposed "Mon Suivi Conso"<sup>4</sup> which uses time-series regression for monthly estimates—**still too coarse** for real-time action.
- Current gap: State-of-the-art deep-learning NILM falters due to **electricity consumption patterns drift** (non-stationarity).
- Our contribution: First deployed (at scale) NILM solution that learns to **correct subsequence distribution drift** while preserving **temporal context**—unlocking robust, detailed, appliance-level insights.

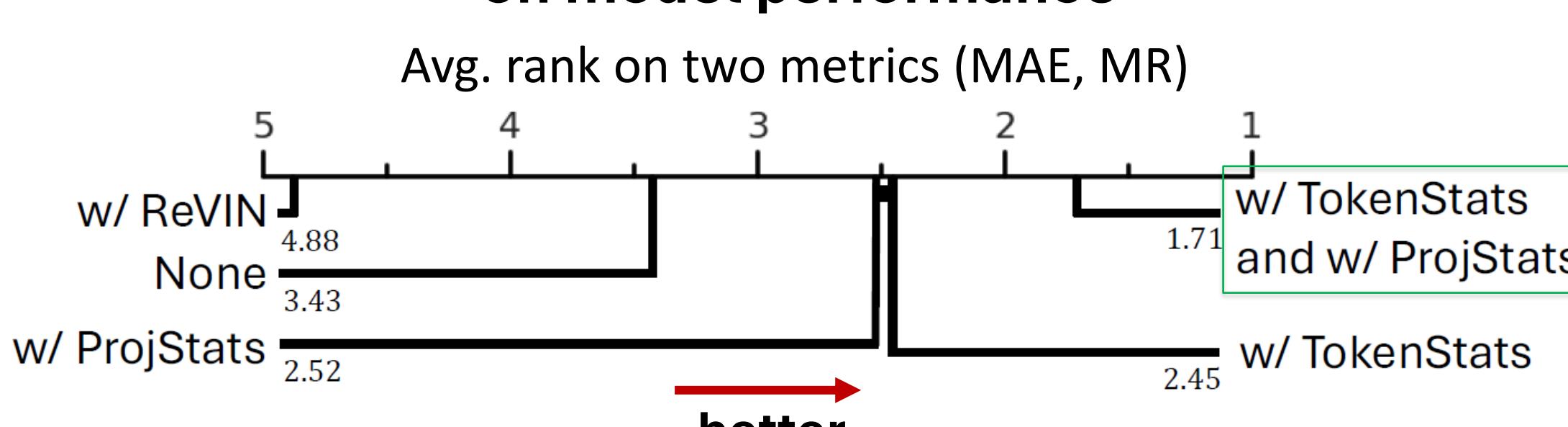


## NILMFormer architecture

Key idea—decompose the input **subsequence** to mitigate data drift in **3 different components**

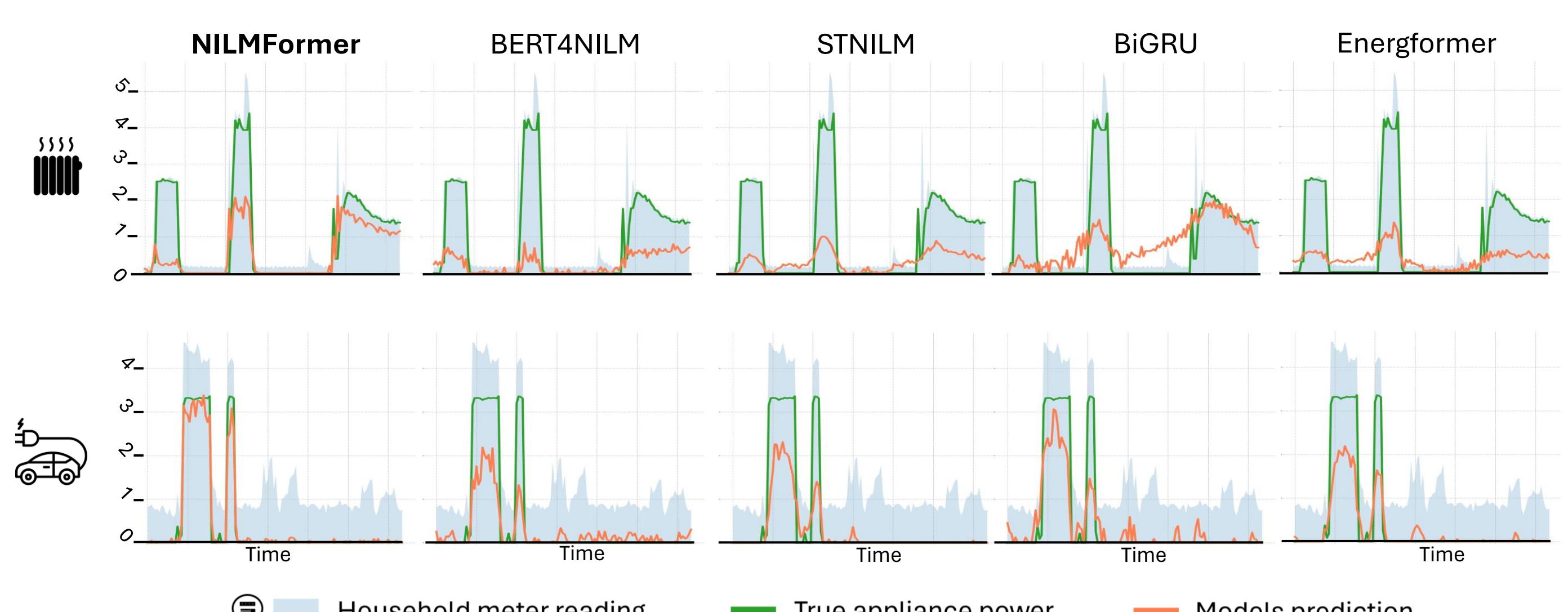
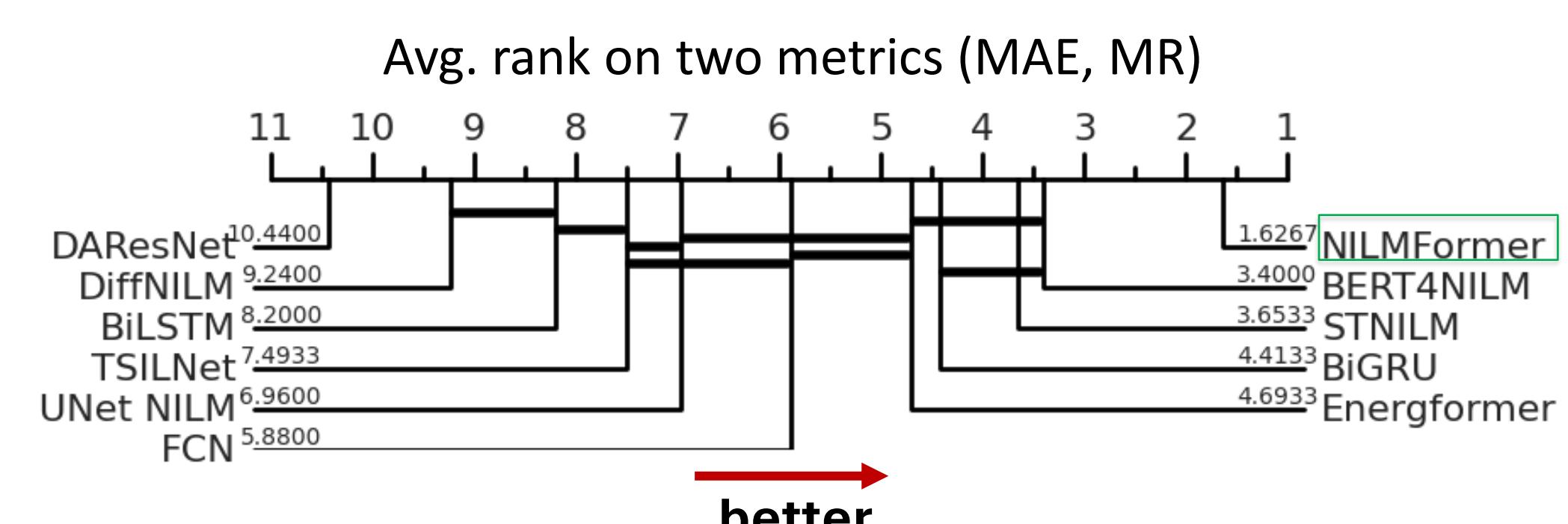


### Impact of the proposed mechanisms for mitigating data non-stationarity on model performance



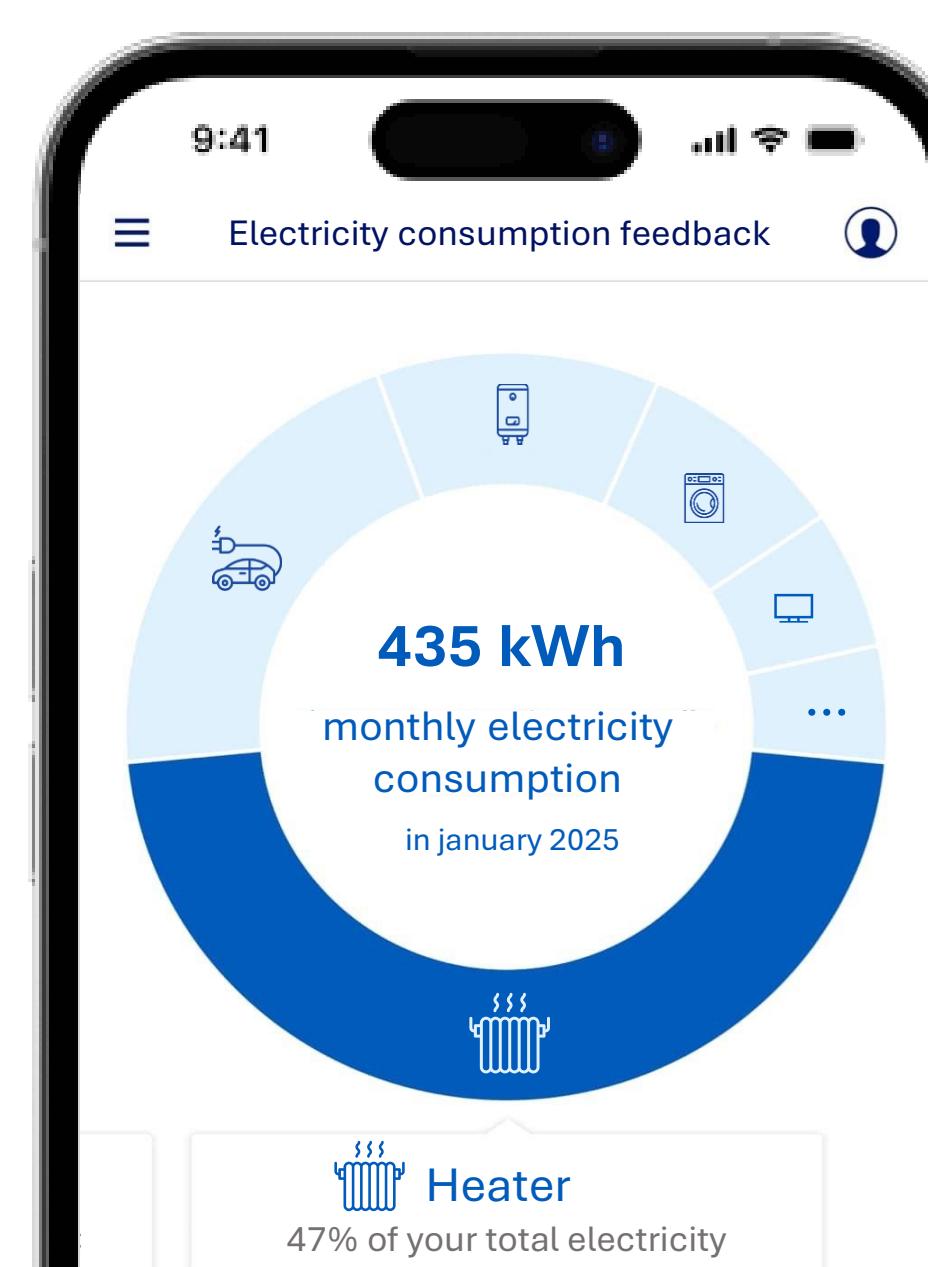
## Performances

Results averaged across **4 datasets** and **14 appliance-disaggregation scenarios** (REFIT, UKDALE, plus two private EDF datasets)



## Deployment

NILMFormer is currently deployed as the backbone algorithm in **Mon Suivi Conso<sup>4</sup>** (EDF's consumption monitoring solution)

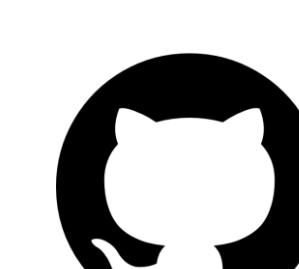


- Delivers appliance-level insights at **daily, weekly, and monthly** granularity to more than 4 million EDF customers.
- Raises **monthly-level accuracy** by **154 %** over EDF's previous deep-learning time-series regression backbone.
- Processes the entire EDF database (4M clients) in **11-hours**, demonstrating **industrial-grade scalability**.

## References

- [1] A review of current methods and challenges of advanced deep learning-based non-intrusive load monitoring (NILM) in residential context, Hasan Rafiq, Prajwal Manandhar, Edwin Rodriguez-Ubinas, Omer Ahmed Qureshi, Themis Palpanas, Energy&Building, 2024.  
[2] Martinez et al., Advanced Metering Initiatives and Residential Feedback Programs, 2010.  
[3] Allcott et al., Social norms and energy conservation, Journal of Public Economics, 2011.  
[4] Solution Mon Suivi Conso – EDF's Consumption Monitoring Solution, <https://particulier.edf.fr/fr/accueil/bilan-consommation/solution-suivi-conso.html>, 2025..

Github



Promotional Video

