

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

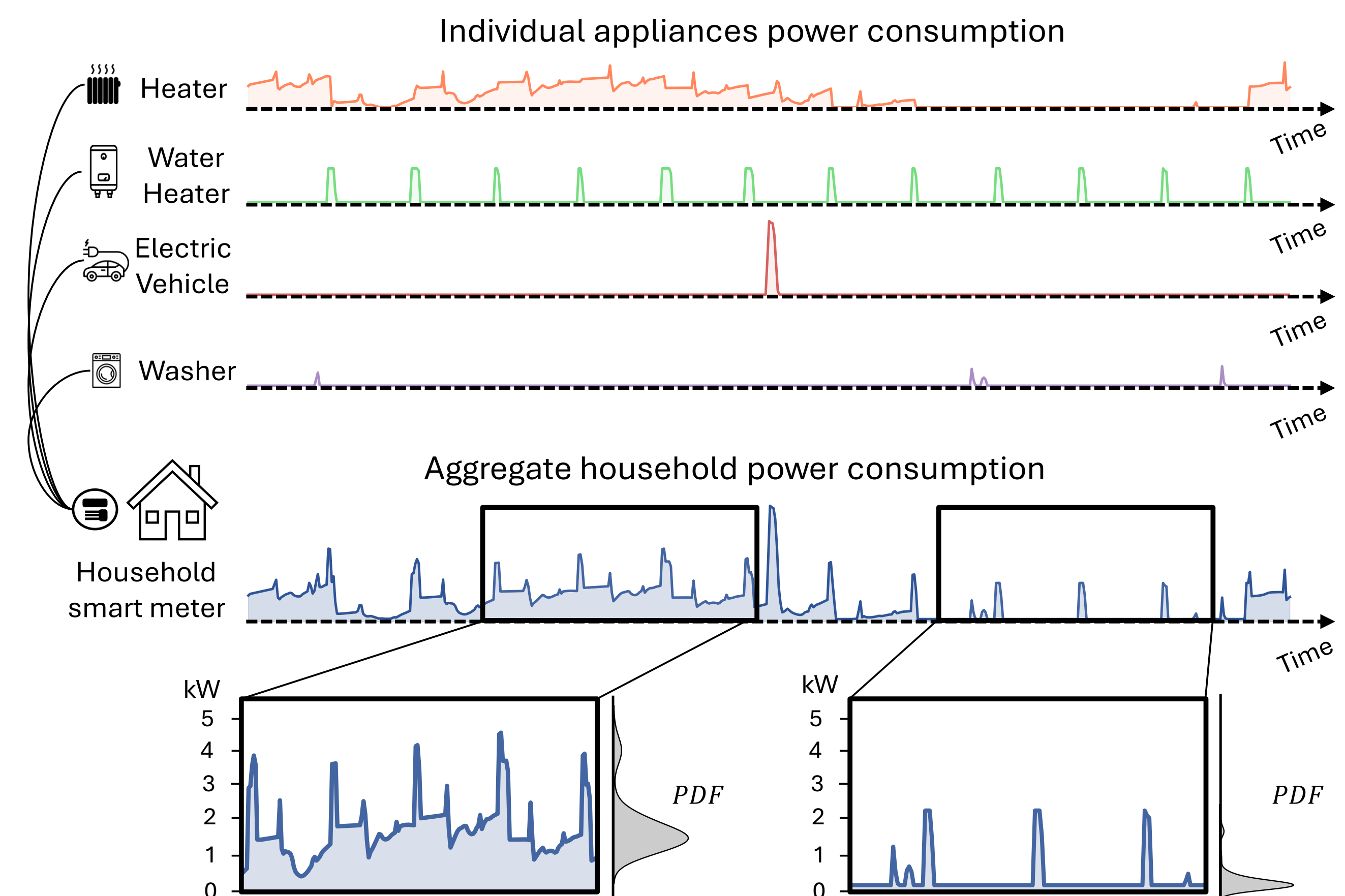


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Challenges & Motivation

- Smart meters \neq insight: They only capture whole-home load, while decarbonization needs **appliance-level detail**.
- NILM in a nutshell: **Non-Intrusive Load Monitoring**¹ infers each appliance's usage from the **smart meter aggregate signal only**.
- Why it matters: Appliance-specific feedback cuts **consumption by $\approx 12\%$** ^{2,3}, 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*"⁴ 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.

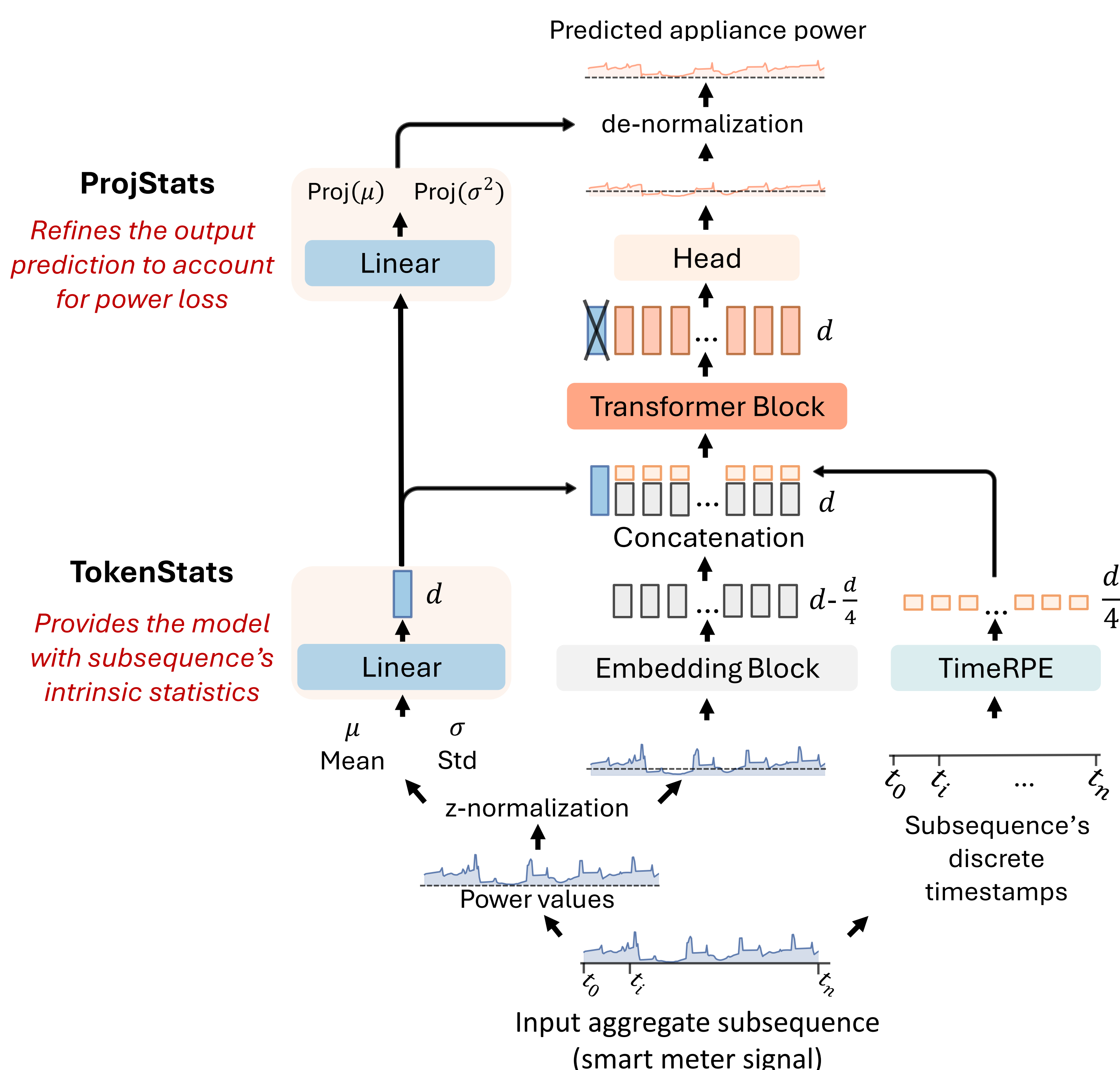


Electricity consumption data are highly non-stationary as appliances turn on and off throughout the day

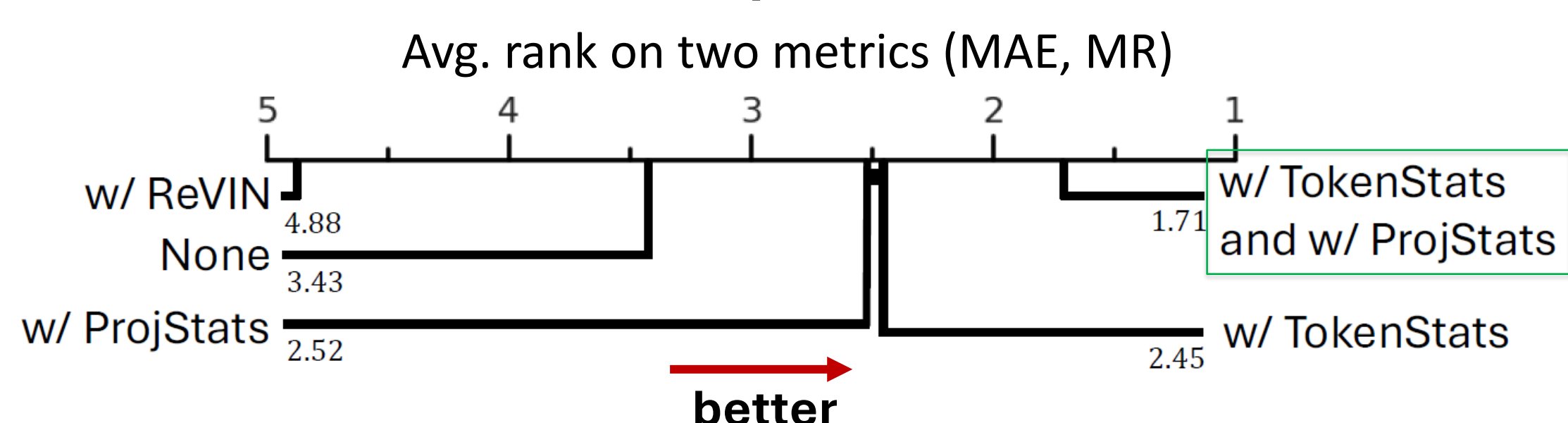
NILMFormer architecture

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

1. Intrinsic statistics 2. Shape 3. Timestamps information

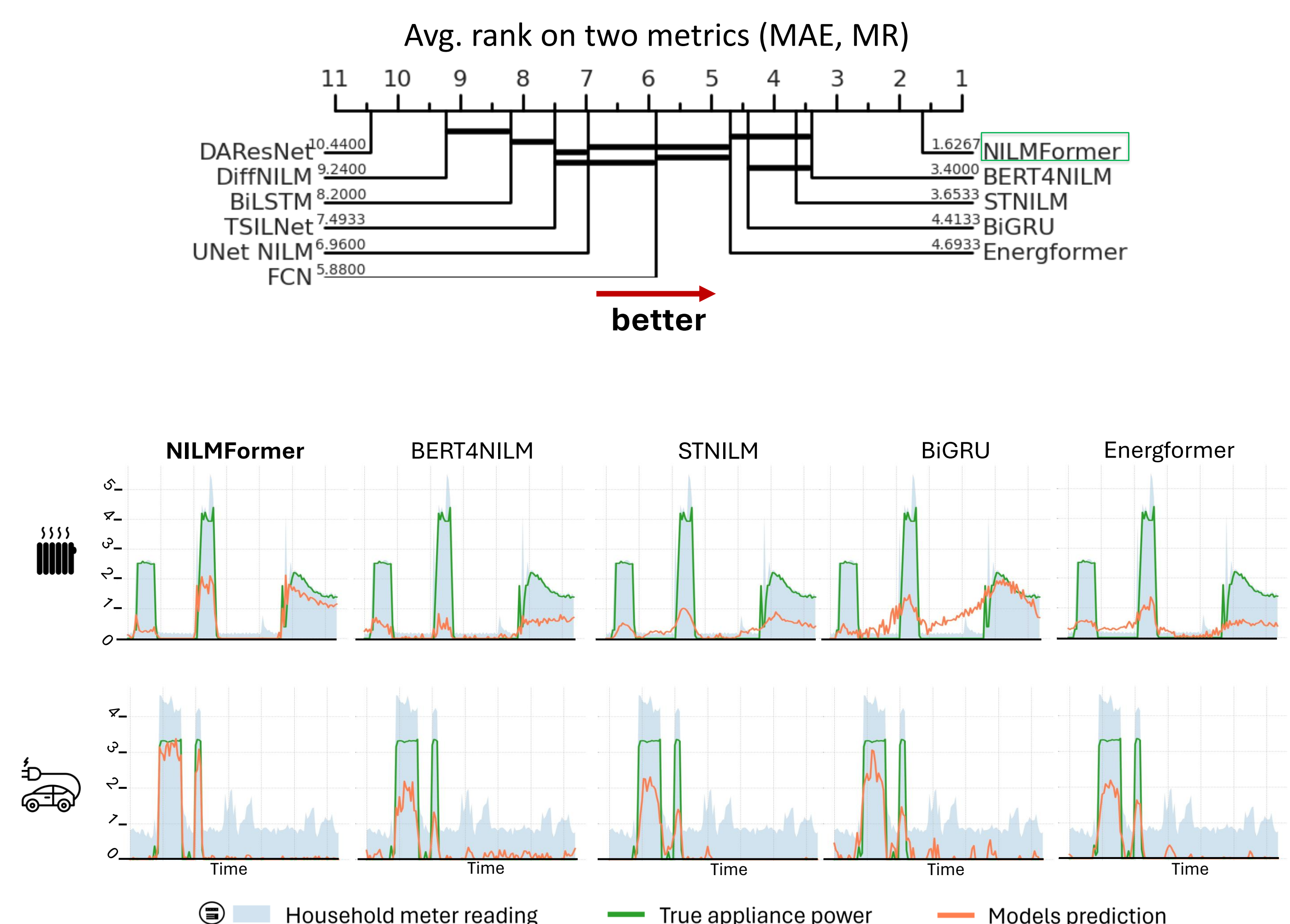


Impact of the proposed mechanisms for mitigating data non-stationarity on model performance



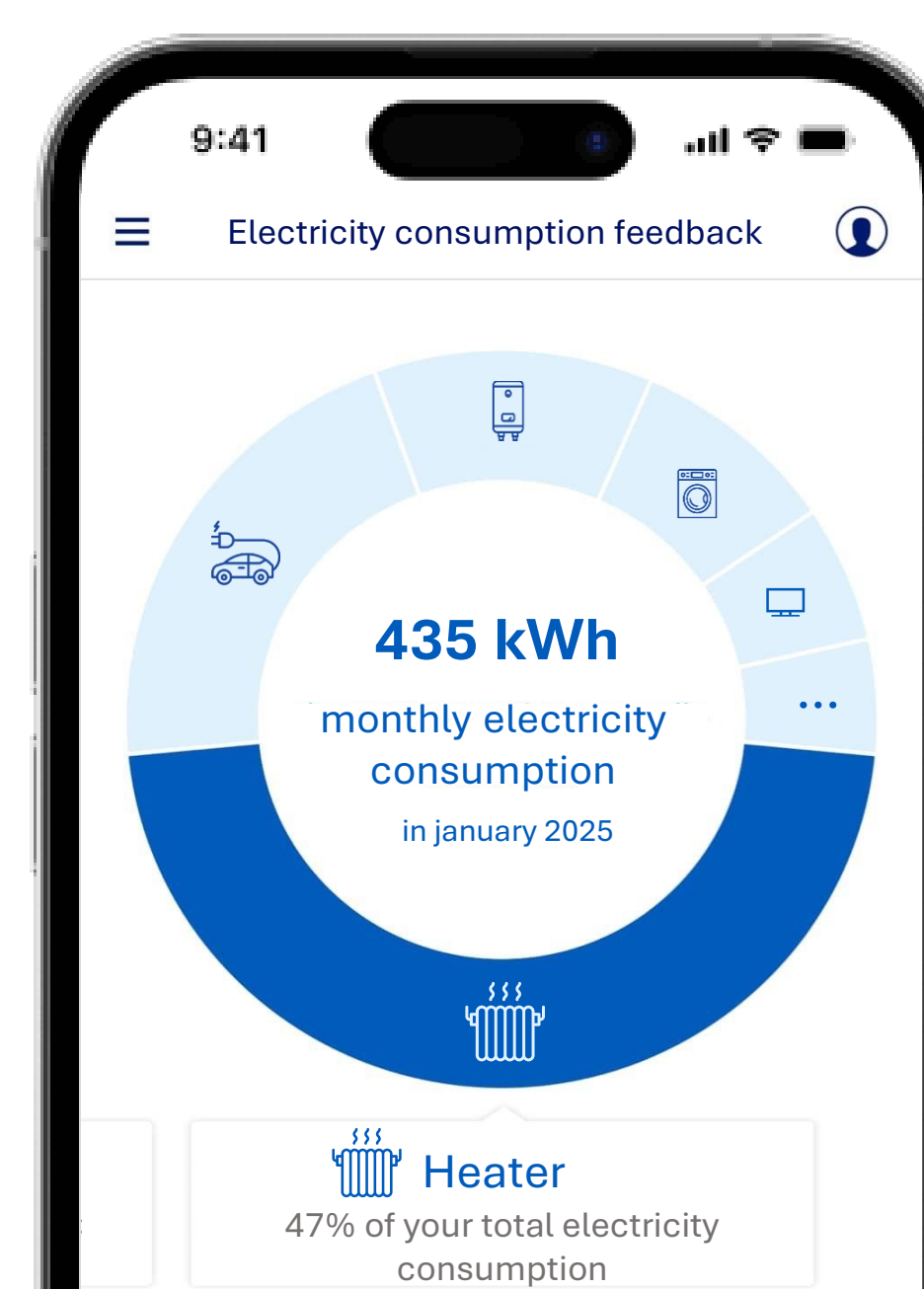
Performances

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*⁴ (EDF's consumption monitoring solution)

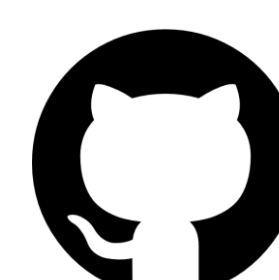


- Delivers appliance-level insights** at **daily, weekly, and monthly** granularity to more than 4 million EDF customers.
- Raises monthly-level accuracy by 154 %** over EDF's previous deep-learning time-series regression backbone.
- Processes the entire EDF database (4M clients) in **11-hours**, demonstrating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

