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Smart Meter Privacy: A Theoretical Framework

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Framework**

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Outline

- Introduction
- Model
- Utility-privacy tradeoff region
- Conclusion and future work

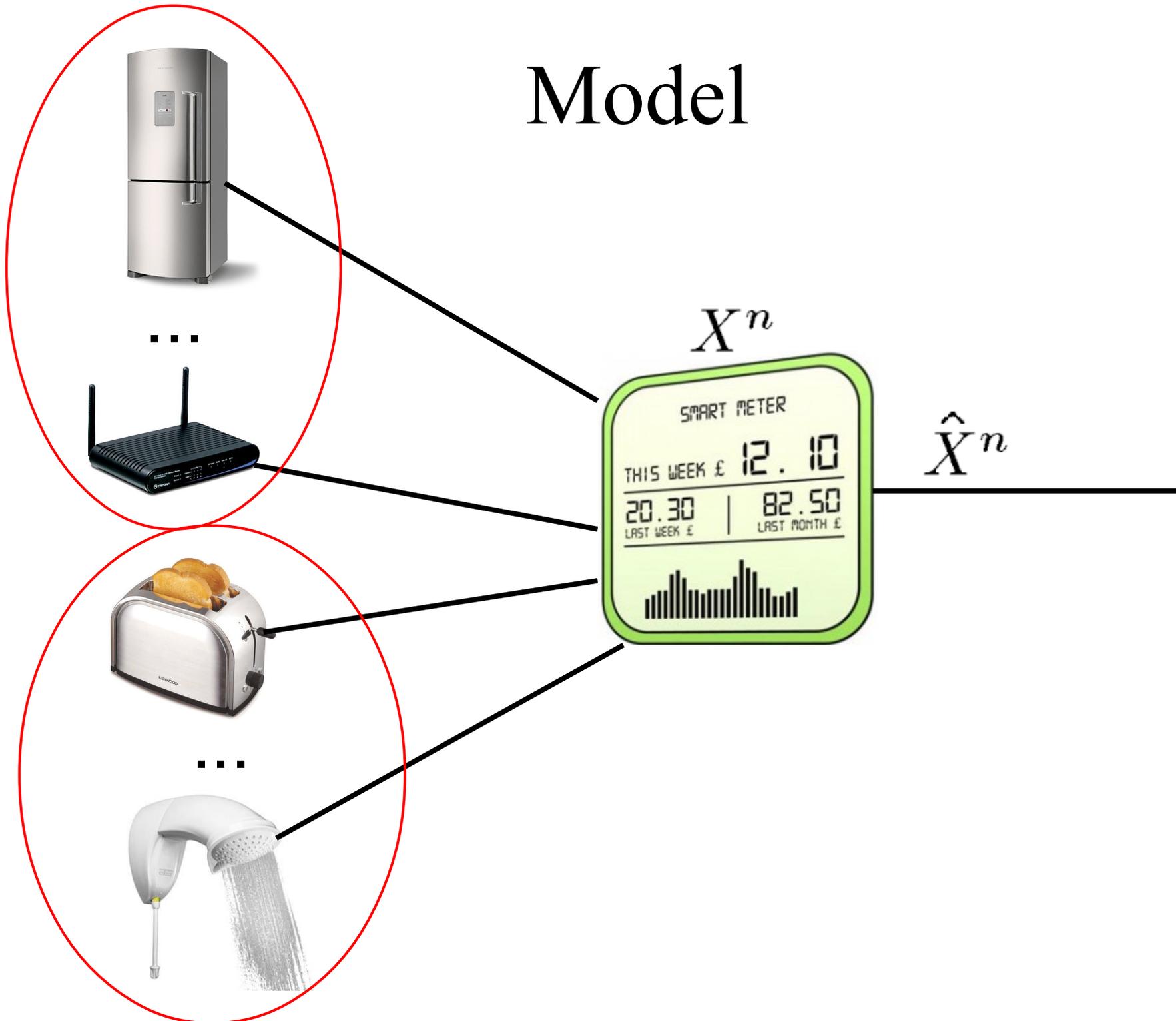
Introduction

- Current approaches lack a formal model of privacy and thus cannot answer some questions such as:
 - how much privacy is lost in such methods?
 - how much sensitive information can and should be left in the data so that it is still useful?
- In other words, current approaches to privacy only provide privacy assurances, but cannot provide any guarantees.

Introduction

- The paper presents a formal model
 - Hidden Markov model
 - Stationary Gaussian model
- Appliances are classified as:
 - Continuously (less revealing of personal details)
 - Intermittently (more revealing)

Model



Model

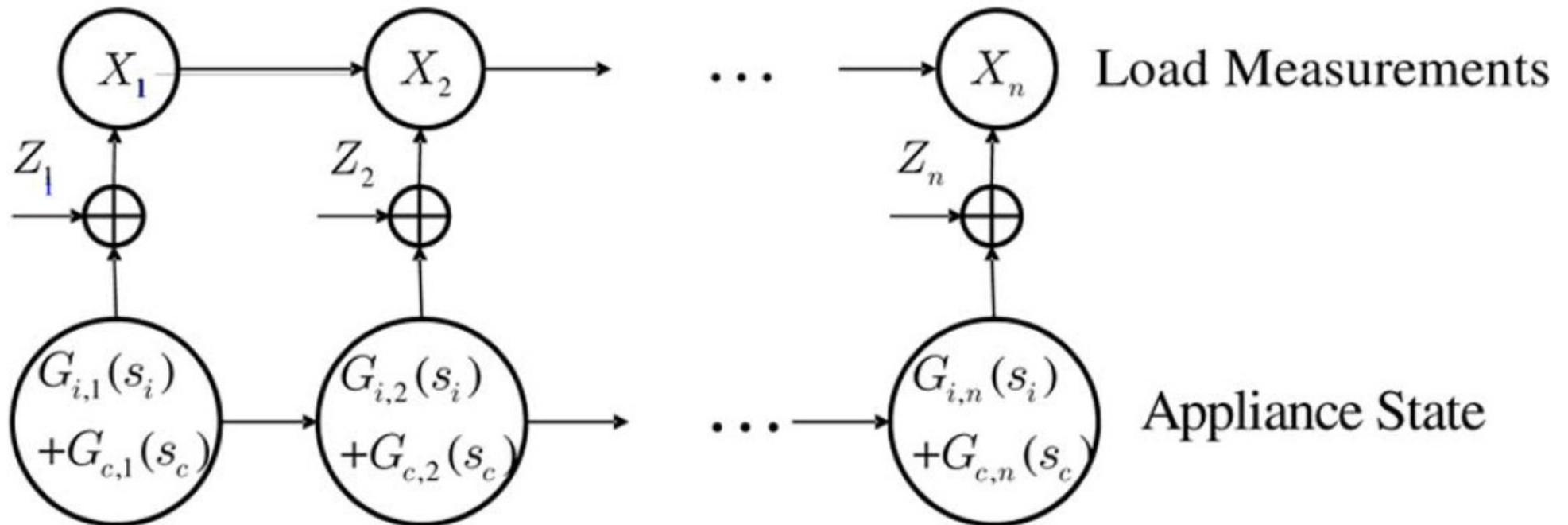


Fig. 2. Meter measurements obtained as a noisy sum of two Gaussian processes, corresponding to the intermittent and continuous appliances, respectively over a time window of length n .

$$X^n(S) = G_c^n(S_c) + G_i^n(S_i) + Z^n$$

Utility and Privacy Metrics

- Utility metric is an average “distance” distortion function between the original and the perturbed data
 - e.g., mean-square error
- Privacy metric measures the difficulty of inferring private information leaked by the appliance state via the meter measurements
 - e.g., mutual information

Privacy-Preserving Mapping

- Formally, an (n, M, D, L) code involves an encoder and a decoder described below:
 - Encoding: $F_E : \mathcal{X}^n(\mathcal{S}) \rightarrow \mathcal{M} = \{1, 2, \dots, 2^{nR}\}$
 - Decoding: $F_D : \mathcal{M} \rightarrow \hat{\mathcal{X}}^n.$
 - Desired utility: $D = \frac{1}{n} \sum_{k=1}^n \mathbb{E}[(X_k - \hat{X}_k)^2]$
 - Information leakage: $L = \frac{1}{n} I(Y^n; \hat{X}^n)$

Utility-Privacy Tradeoff Region

- The smart meter utility-privacy tradeoff region T : is the set of all pairs (D, L) such that, for every $\epsilon > 0$ and all sufficiently large n , there exists a coding scheme with parameters $(n, M, D + \epsilon, L + \epsilon)$
- for a desired D we seek to minimize the average number of bits-per-entry of the correlated sequence Y^n (that we wish to hide) that are leaked from the revealed sequence \hat{X}^n

Utility-Privacy Tradeoff Region

- The minimal leakage $\lambda(D)$ achievable for a desired distortion D for a source with memory subject to distortion and leakage constraints is given by

$$\lambda(D) = \lim_{n \rightarrow \infty} \inf_{p(x^n, y^n) p(\hat{x}^n | x^n)} \frac{1}{n} I(Y^n; \hat{X}^n).$$

Privacy Preserving Spectral Waterfilling

- Specially, we seek to hide the intermittently used appliances:

$$L = \frac{1}{n} I(G_i^n; \hat{X}^n)$$

- Filtering out all frequencies that have power below a certain threshold (determined directly by λ)

Privacy Preserving Spectral Waterfilling

$$\Delta(f) = \begin{cases} 0; & F_i(f) < F_c(f) + \sigma^2; \text{ (or)} \\ & \lambda < F_c(f) + \sigma^2 < F_i(f) \\ D_1(f); & F_c(f) + \sigma^2 < \lambda < F_i(f) \\ D_2(f); & \lambda > F_i(f) > F_c(f) + \sigma^2 \end{cases}$$

where

$$D_1(f) = (\lambda - F_c(f) - \sigma^2)^+$$

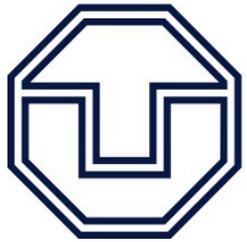
$$D_2(f) = (F_i(f) - F_c(f) - \sigma^2)^+.$$

Conclusion

- The theoretical framework allows us to quantify the utility-privacy tradeoff in smart meter data.
- The privacy guarantee comes from the bound on information leakage while the utility guarantee come from the upper bound on the MSE
- The distortion model can be viewed as a filter on the load signal that suppresses those appliance (intermittent) signatures that reveal the most private information

Future Works

- Apply the model on measured data to validate whether the filter eliminates or decreases the signatures of intermittent devices to the desired degree.
- Another interesting avenue to explore would be to apply and demonstrate the power of these concepts in a practical context



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