

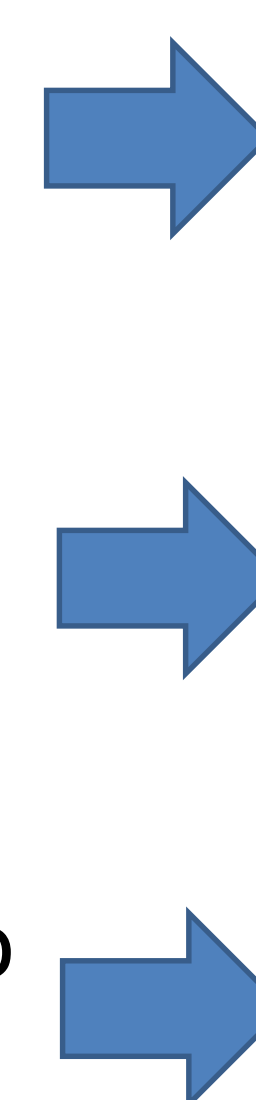
Sampling Rate Distortion: Global Inference from Partial Measurements



V. Praneeth Boda and P. Narayan

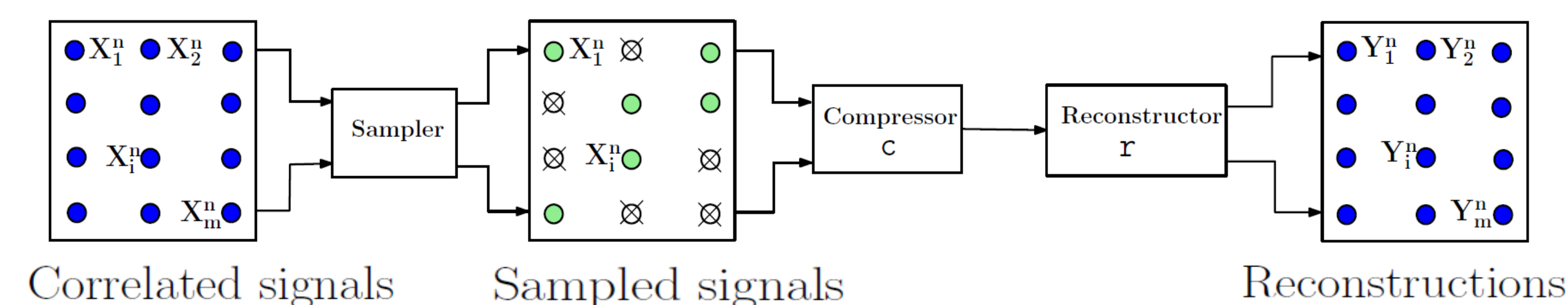
Problems?

- How do we **place** sensors for measurements? Dynamic thermal management in microprocessors.
- IoT: What can a central hub **learn** using compressed measurements from smart sensors?
- How do we **find** “central” nodes in large networks to capture the average behavior?



- Sampling rate distortion: **Statistics known**
 - Selection and **compression** of signals. Algorithms for reconstruction.
- Universal sampling rate distortion: **Statistics unknown**
 - Selection, **estimation** and **compression**.
- Multi-armed bandits: **Statistics unknown, finite-time horizon**
 - Selection, **estimation** and **identification** under **budget**.

- Correlated signals: $\{(X_{1t}, \dots, X_{Mt}) = X_{\mathcal{M}t}\}_{t=1}^{\infty}$
- k -random sampler: $P_{S^n|X_{\mathcal{M}}^n} = \prod_{t=1}^n P_{S_t|X_{\mathcal{M}}^t, S^{t-1}}$



- Sampling Rate Distortion function:

$$R(\Delta) \triangleq \inf_{k\text{-RS}, (c, r)} \frac{1}{n} \log \|c\|$$

$$\mathbb{E}[\text{Distortion}(\text{Signals}, r(c(\text{Sampled signals})))] \leq \Delta$$



Global inference when only subsets of signals can be accessed

Sampling Rate Distortion function

- For a **non-adaptive** random sampler,

$$P_{S_t|X_{\mathcal{M}}^t, S^{t-1}} = P_{S_t}, t = 1, \dots$$

$$R_{IRS}(\Delta) = \min_{P_S, P_{Y_{\mathcal{M}}|S, X_S}} \mathbb{E}[d(X_{\mathcal{M}}, Y_{\mathcal{M}})] \leq \Delta I(X_S \wedge Y_{\mathcal{M}}|S).$$

- For an **adaptive** random sampler,

$$P_{S_t|X_{\mathcal{M}}^t, S^{t-1}} = P_{S_t|X_{\mathcal{M}t}}, t = 1, \dots$$

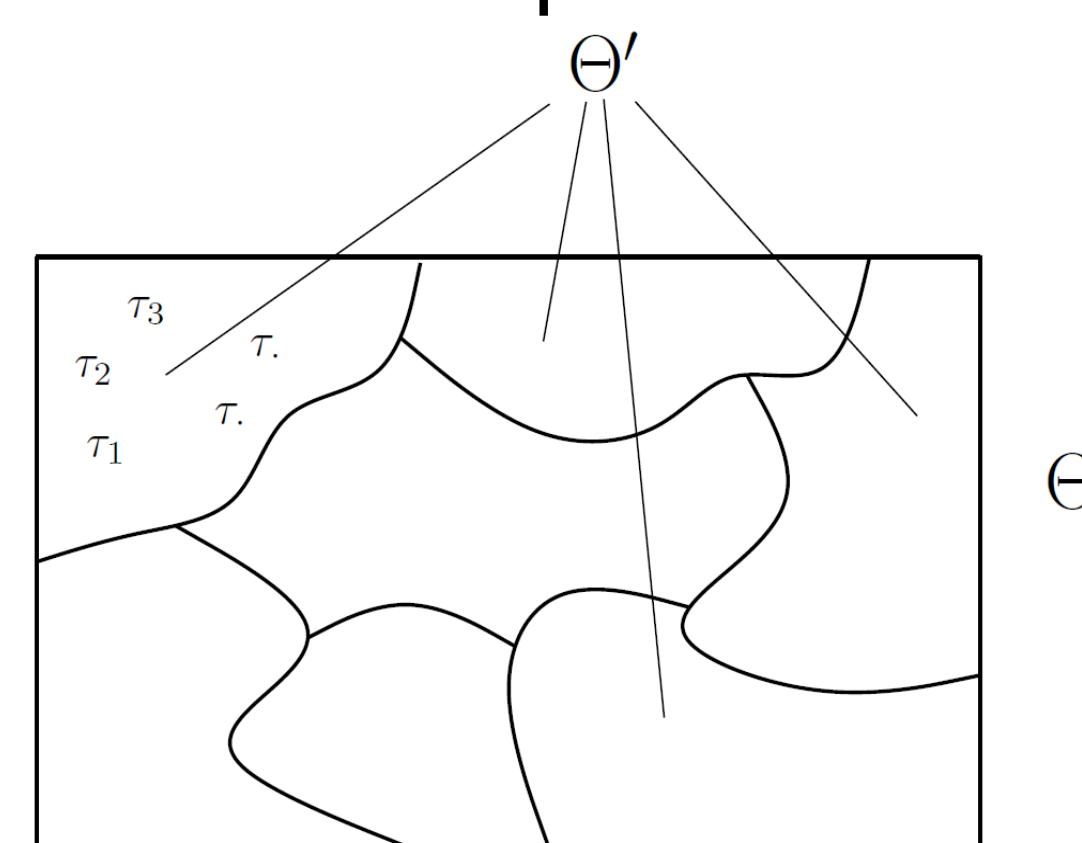
- Conditional **deterministic sampling** is optimal!
- Reduce computational complexity.
- Memory of previously sampled signals doesn't help.
- Does knowledge of sampling at reconstructor help?

What if statistics are unknown?

- Distribution of $\{(X_{1t}, \dots, X_{Mt}) = X_{\mathcal{M}t}\}_{t=1}^{\infty}$ known only to

$$P_{X_{\mathcal{M}}} \in \mathcal{P} = \{P_{\tau}, \tau \in \Theta\}$$

- Compressor:
 - Forms an “estimate” of statistics from partial observations
 - Cannot learn all correlations**
 - Uses estimate to compression and reconstruction.



Ambiguity atoms over class of distributions

Identify “central” nodes in networks

(Joint work with Prashanth L. A.)

- Capture **average** behavior in communication/social networks.
- Statistics unknown; data acquisition expensive.
- Observations only from subsets of nodes/people.
- Sequential decision algorithms** require less data.

