



MIDDLE EAST TECHNICAL UNIVERSITY

Optimal Sampling and Scheduling for Remote Fusion Estimation of Correlated Wiener Processes



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Outline



Motivation



Problem Formulation



Main Results



Takeaways

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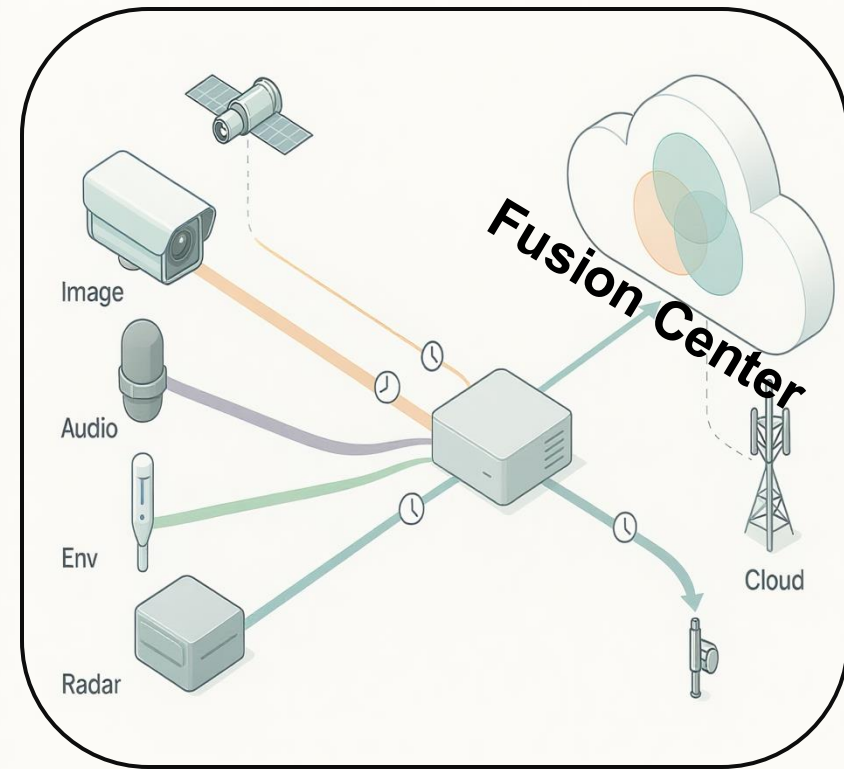
Motivation

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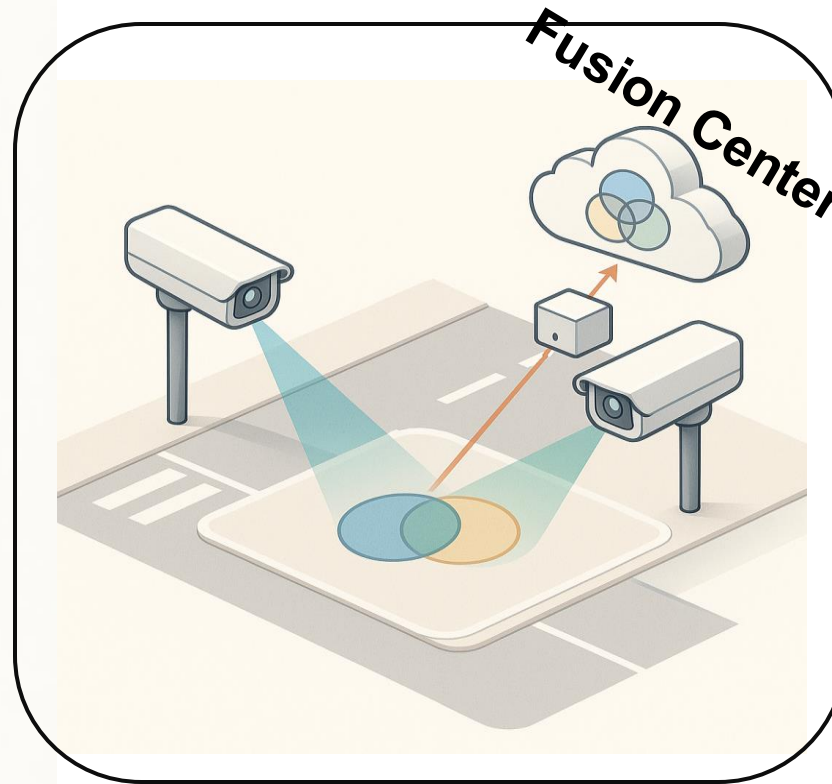


Motivation: Goal-Oriented Communications

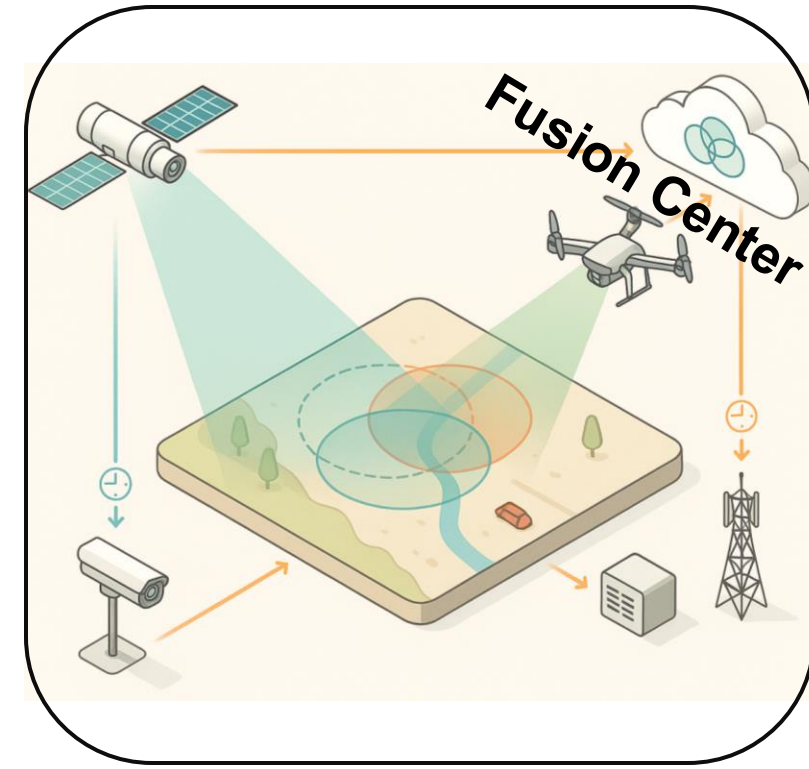
- Scenario: Multiple Correlated Sources



Correlated **Multi-Modal Data**
Comming From
Heterogeneous Sensors

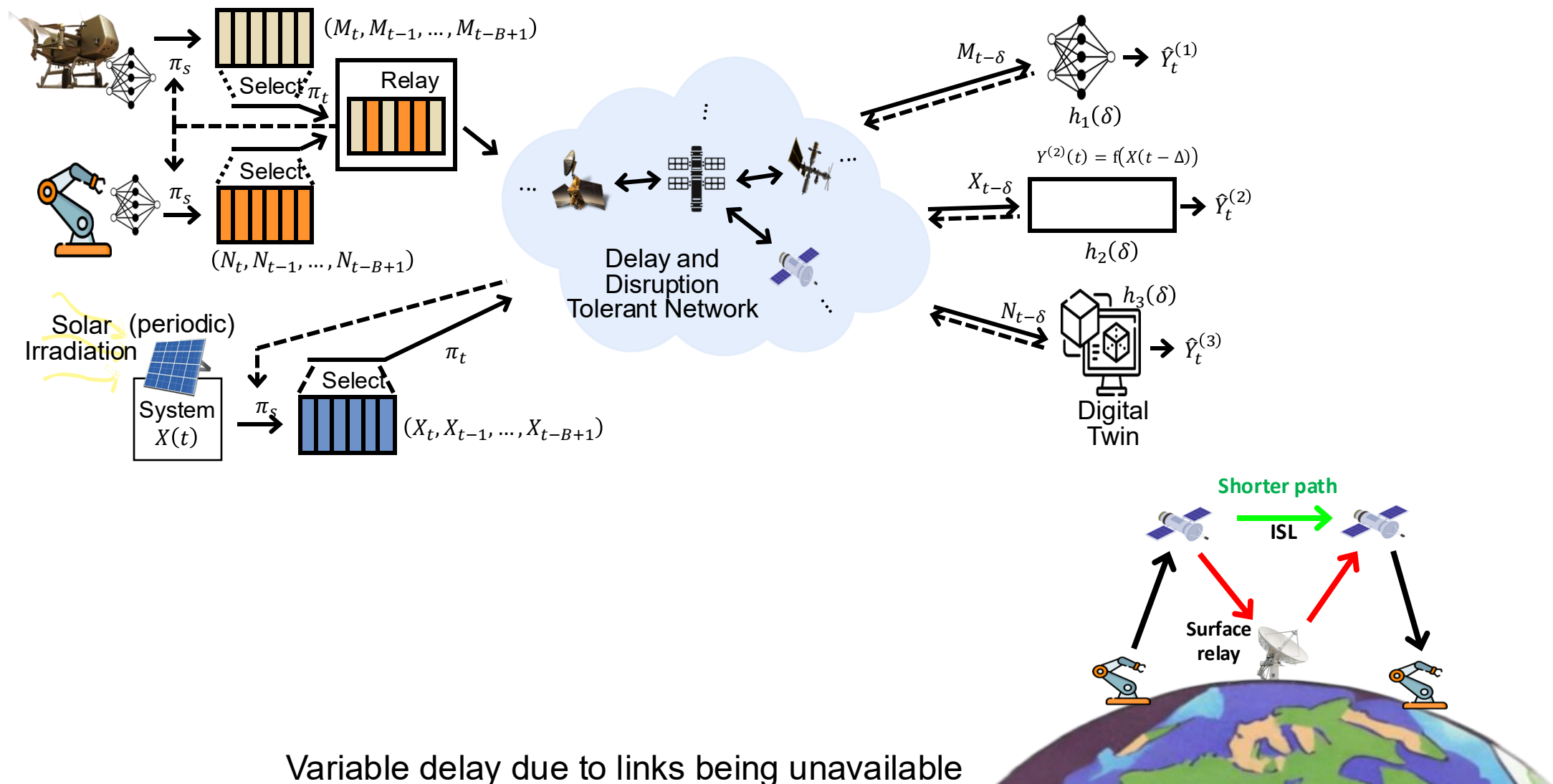


Distributed Sensors Observing
an Overlapping Area



Cross-Platform
Situational Awareness
(Satellite-UAV-Ground)

Motivation: Goal-Oriented Communication



Motivation: Goal-Oriented Communication



Open Challenge: How to sample and transmit, to optimize **goal-oriented application-layer performance** in the presence of **network delay**

Prior insights (Single Source):

- Wiener process estimation^[2]: **threshold policy for sampling f(change, age)**
- Single-Source OU process estimation^[3]: **MSE monotonic in age**
- General inference: **non-linear, non-monotonic performance degradation in Aol**

Open problem: Multiple sources (correlated, delayed):

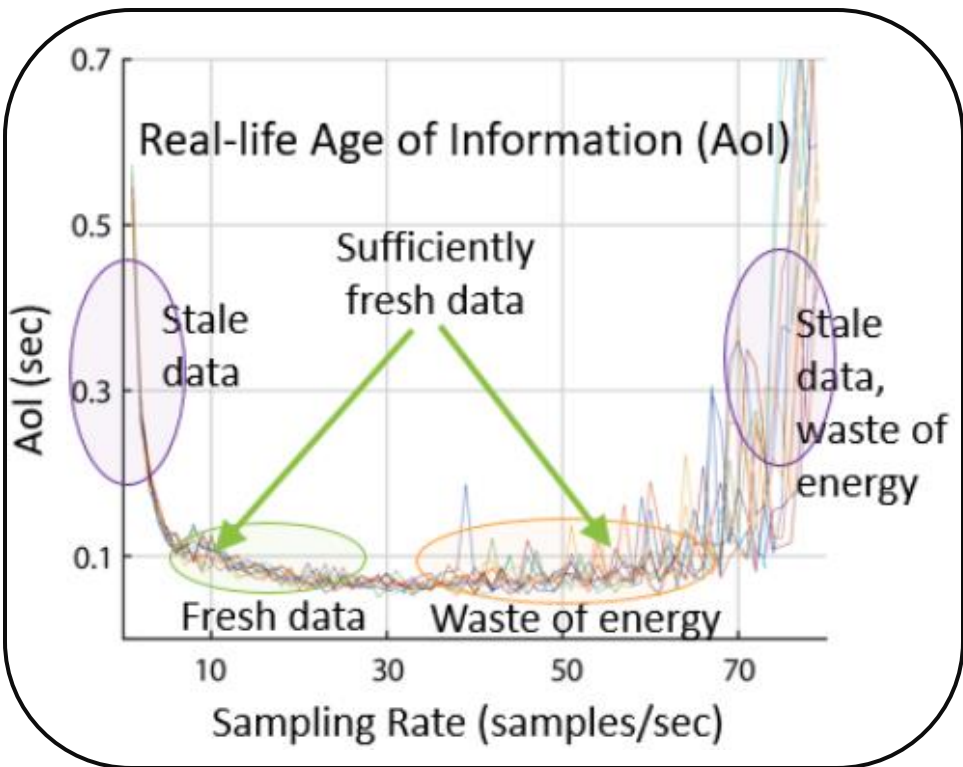
- Most studies assume **independent** multiple sources
- Heuristic metrics for correlated sources: **Age of Correlated Information**, etc.
- **Our formulation**: optimal pull-based sampling for remote **joint estimation**

[2] Y Sun, Y Polyanskiy, E Uysal. Sampling of the Wiener Process for Remote Estimation over a Channel with Random Delay. IEEE Transactions on Information Theory, 2019

[3] TZ Ornee, Y Sun. Sampling and remote estimation for the Ornstein-Uhlenbeck process through queues: Age of information and beyond. ACM/IEEE Transaction on Networking, 2021

Motivation: Age and Beyond

● Insights from Age of Information (AoI) research



- AoI captures the freshness of information:
Sparse Update → **Stale**;
Frequent Update → **Waste of Energy + Stale**.
- AoI as a key metric whenever **fresher data is more valuable**.
- Age-optimal policy neither transmits at the highest possible rate nor requires the smallest end-to-end delay ^[1]
(Waiting before transmitting may be efficient).

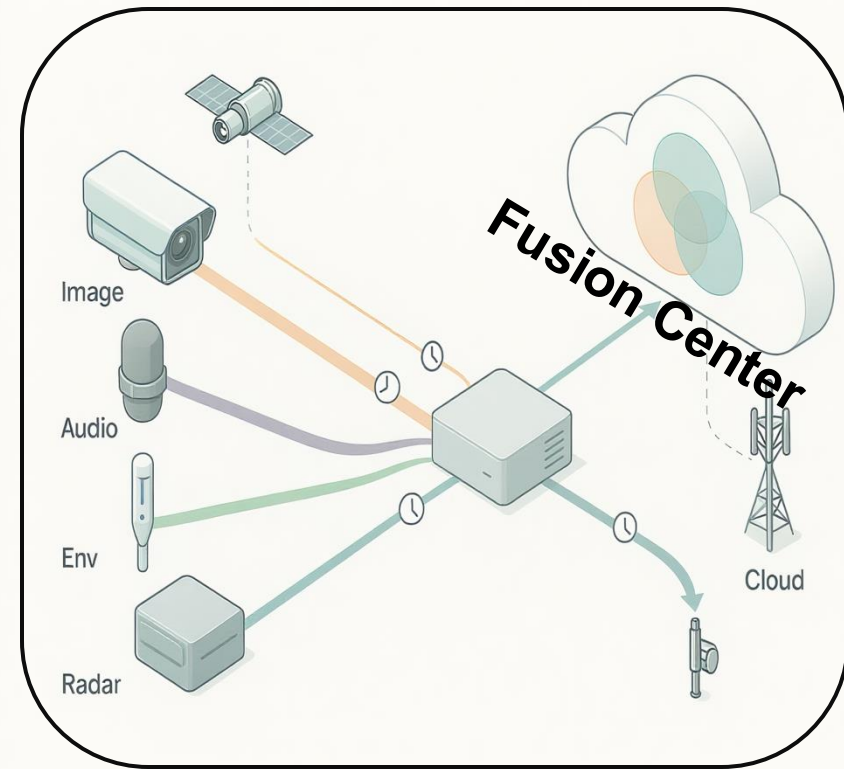


Open Challenge: **Impact of AoI** on **goal-oriented application-layer performance** remains not fully understood.

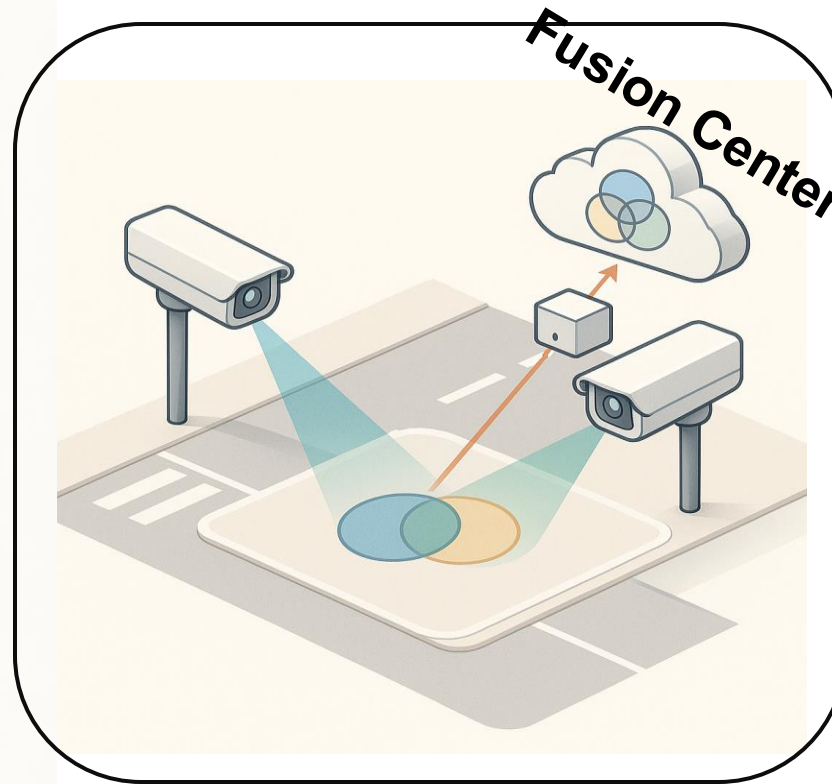
[1] Y Sun, E Uysal, RD Yates, CE Koksal, NB Shroff. Update or Wait: How to Keep Your Data Fresh. IEEE Transaction on Information Theory, 2017.

Motivation: Goal-Oriented Communications

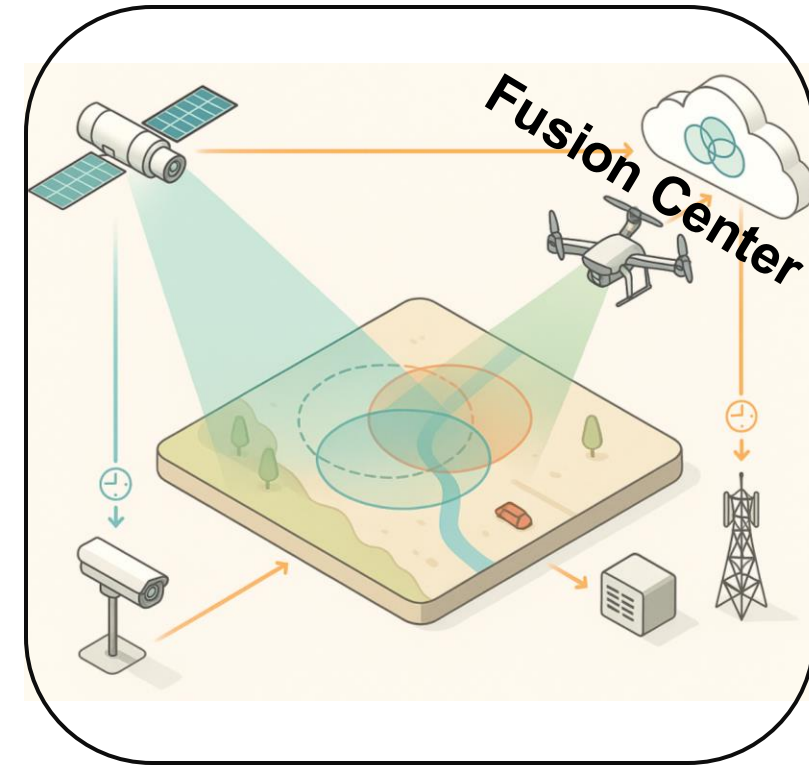
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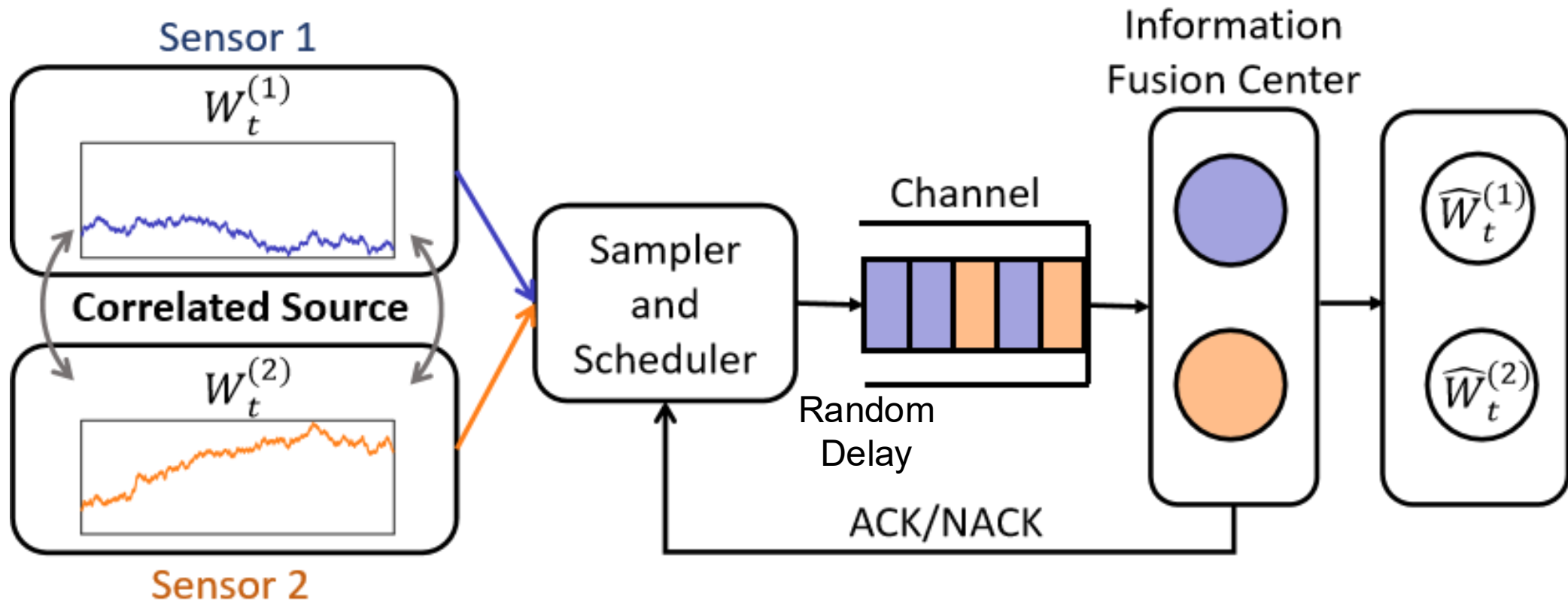
Problem Formulation

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Motivation: MSE-Oriented Remote Information Fusion

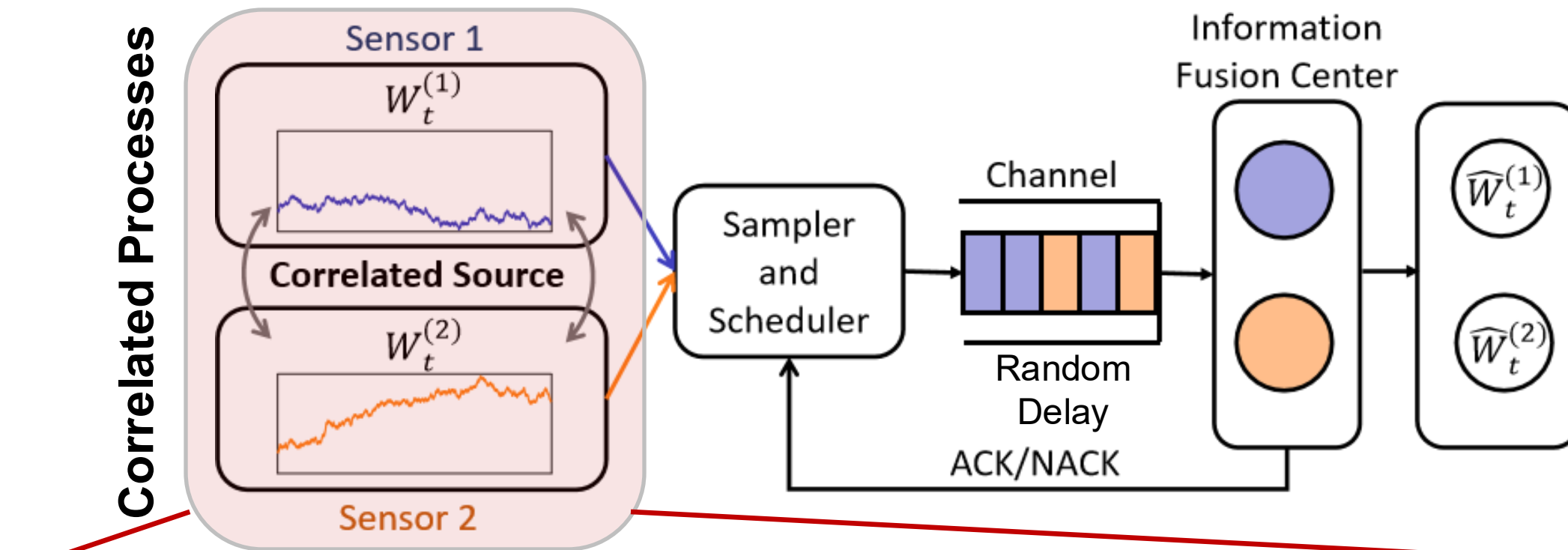
- Information Fusion and Estimation



Challenge: Naïve aggregation of **delayed data from correlated sources** may lead to redundant or misleading information.

Question: How to sample to optimize the staleness level.

System Model: Multiple Correlated Sources



Multiple Sources: Correlated Wiener Processes

Source Dynamics:

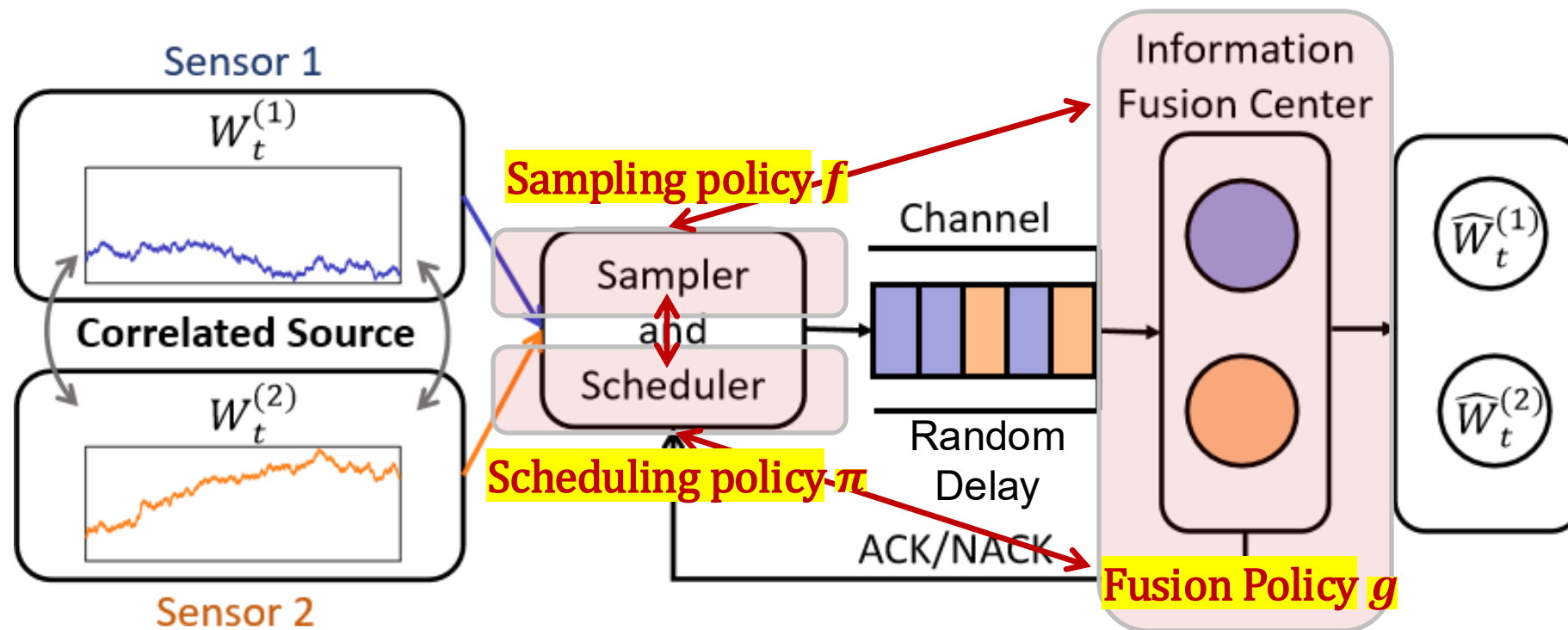
The two observed sources $W_t^{(1)}$ and $W_t^{(2)}$ are **Correlated Wiener process**:

$$\begin{aligned} dW_t^{(1)} &= dB_t^{(1)} \\ dW_t^{(2)} &= \rho dB_t^{(1)} + \sqrt{1 - \rho^2} dB_t^{(2)} \end{aligned}$$

Mutually Independent
Wiener Process

ρ correlation coefficient; if $\rho = 1$, $W_t^{(2)} = W_t^{(1)}$; if $\rho = 0$, $W_t^{(2)}$ is independent of $W_t^{(1)}$

System Model: Objective



Co-Design Sampler, Scheduler, and Remote Information Fusion

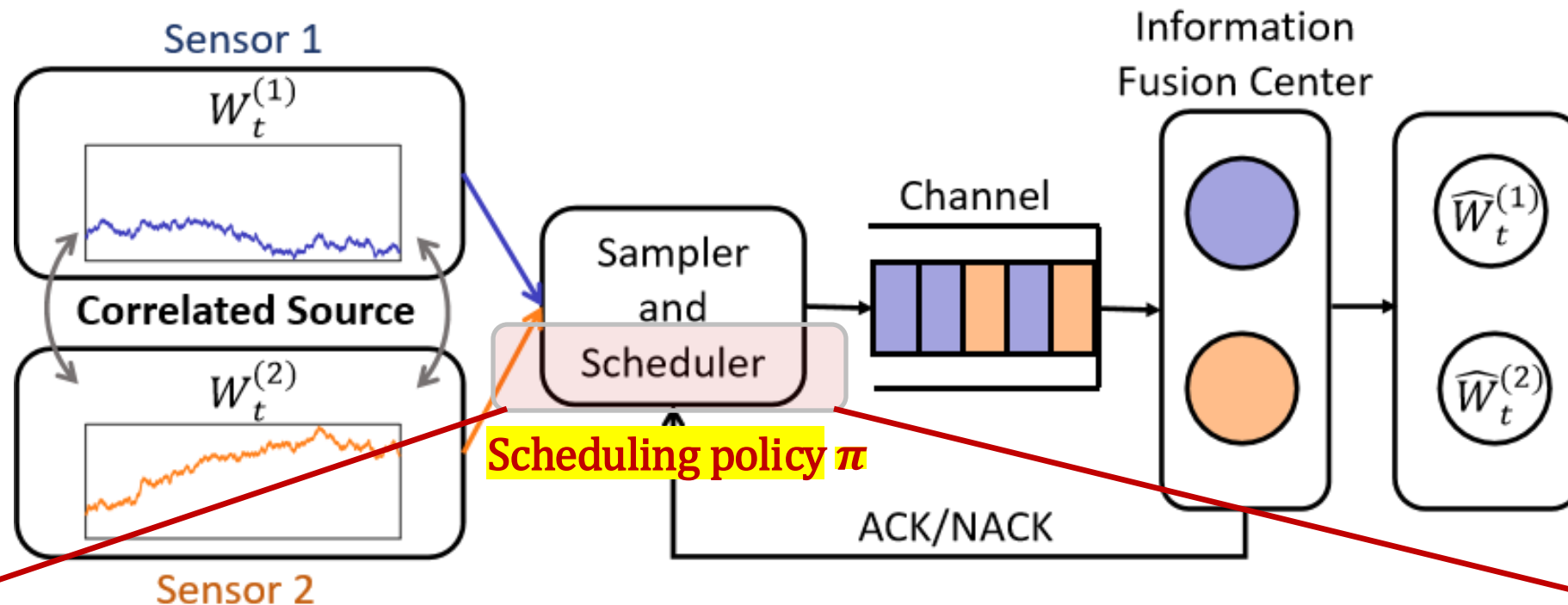
Objective:

- Find out the optimal sampling, scheduling, remote estimation to minimize the long-term average MSE

$$\inf_{g, \pi, f} \limsup_{T \rightarrow \infty} \frac{1}{T} \mathbb{E} \left[\int_0^T (W_t^{(1)} - \widehat{W}_t^{(1)})^2 + (W_t^{(2)} - \widehat{W}_t^{(2)})^2 dt \right]$$

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System Model: Scheduling Policy Definition



Scheduler Design (Which Source to Transmit?)

Scheduling Policy π

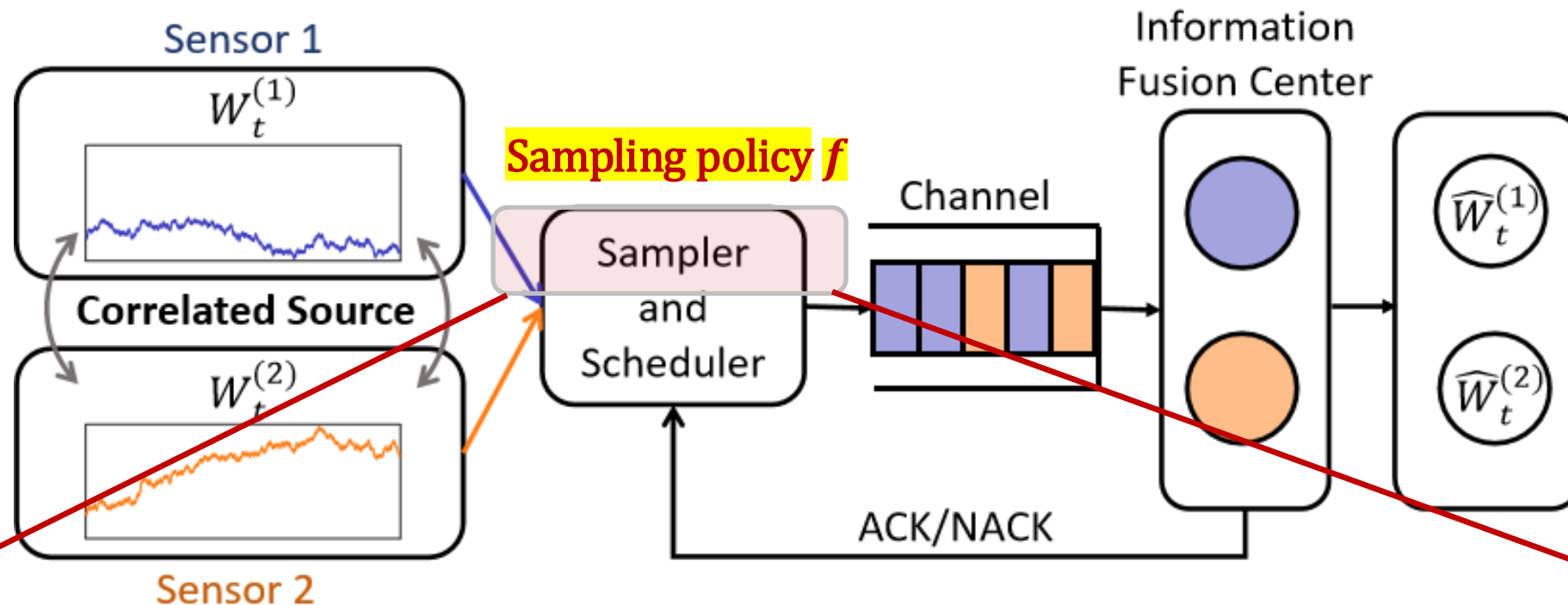
A scheduler is a **policy function** that determines **which source is selected for transmission**.

$$a_i = \pi(\mathcal{I}_{D_{i-1}}), \quad \forall i \in \mathbb{N}$$

Here $a_i \in \{1, 2\}$ is the scheduling action; $\mathcal{I}_{D_{i-1}}$ is the history before the time D_{i-1}

$$\mathcal{I}_t = \{a_j, S_j, W_{S_j}^{(a_j)}, D_j | D_j \leq t\}$$

System Model: Sampling Policy Definition



Sampler Design (When to Transmit?)

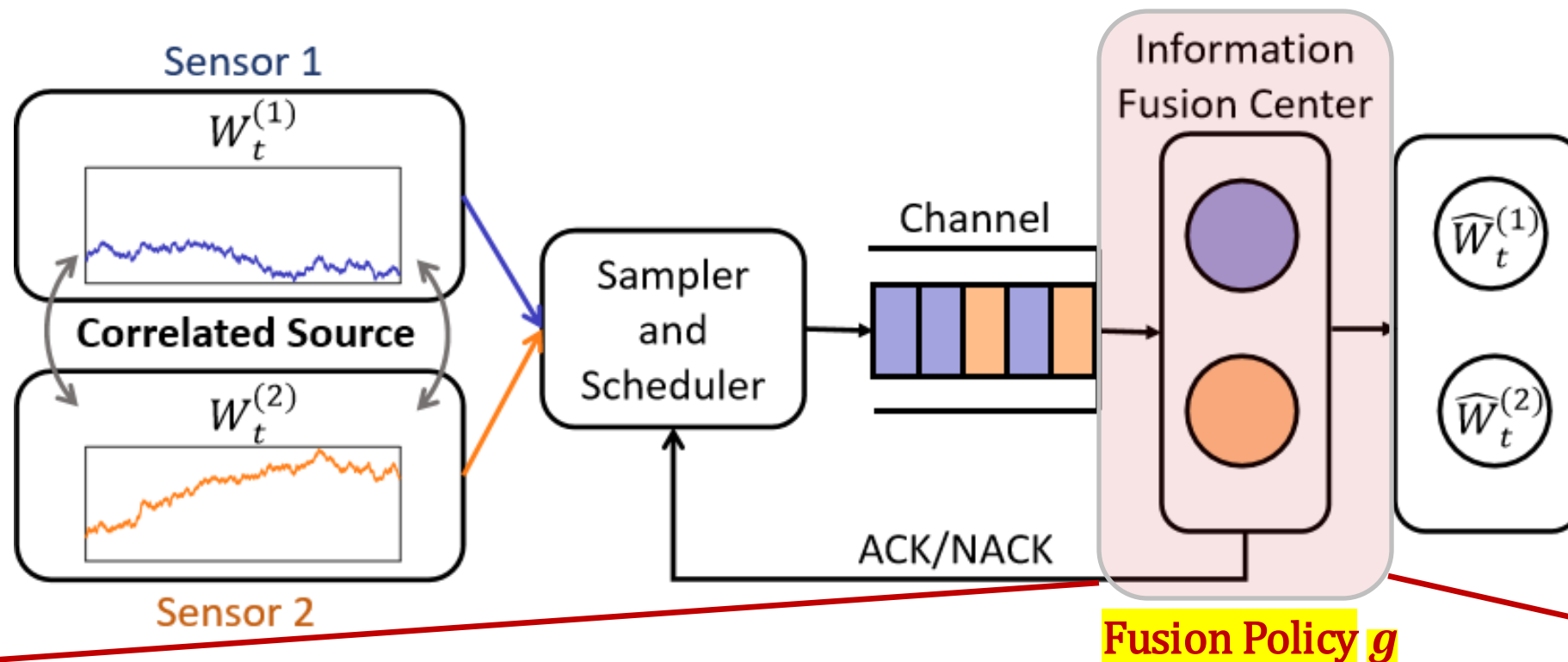
Sampling Policy f

A sampler is a **policy function** that determines **when each sensor is sampled and transmitted**.

$$S_i = f(\mathcal{I}_{D_{i-1}}), \quad \forall i \in \mathbb{N}$$

Here $S_i \in \mathbb{R}^+$ is the sampling time of the i -th transmitted packet;

System Model: Fusion Policy Definition



Remote Information Fusion

Fusion policy g

A **function** that **determines the real-time estimate** $(\widehat{W}_t^{(1)}, \widehat{W}_t^{(2)})$:

$$(\widehat{W}_t^{(1)}, \widehat{W}_t^{(2)}) = g(\mathcal{I}_t), \quad \forall t \in \mathbb{R}^+$$



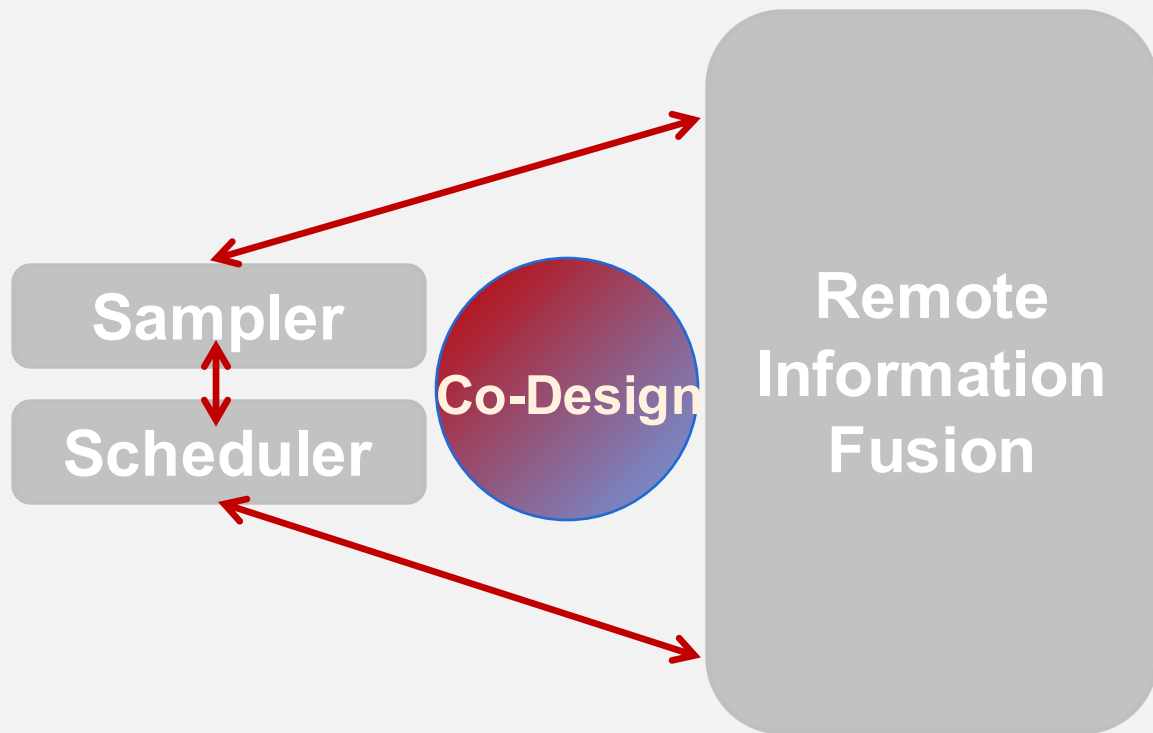
Main Results

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|| Main Results: Separation Principle

Coupled Policies



- **Challenge**

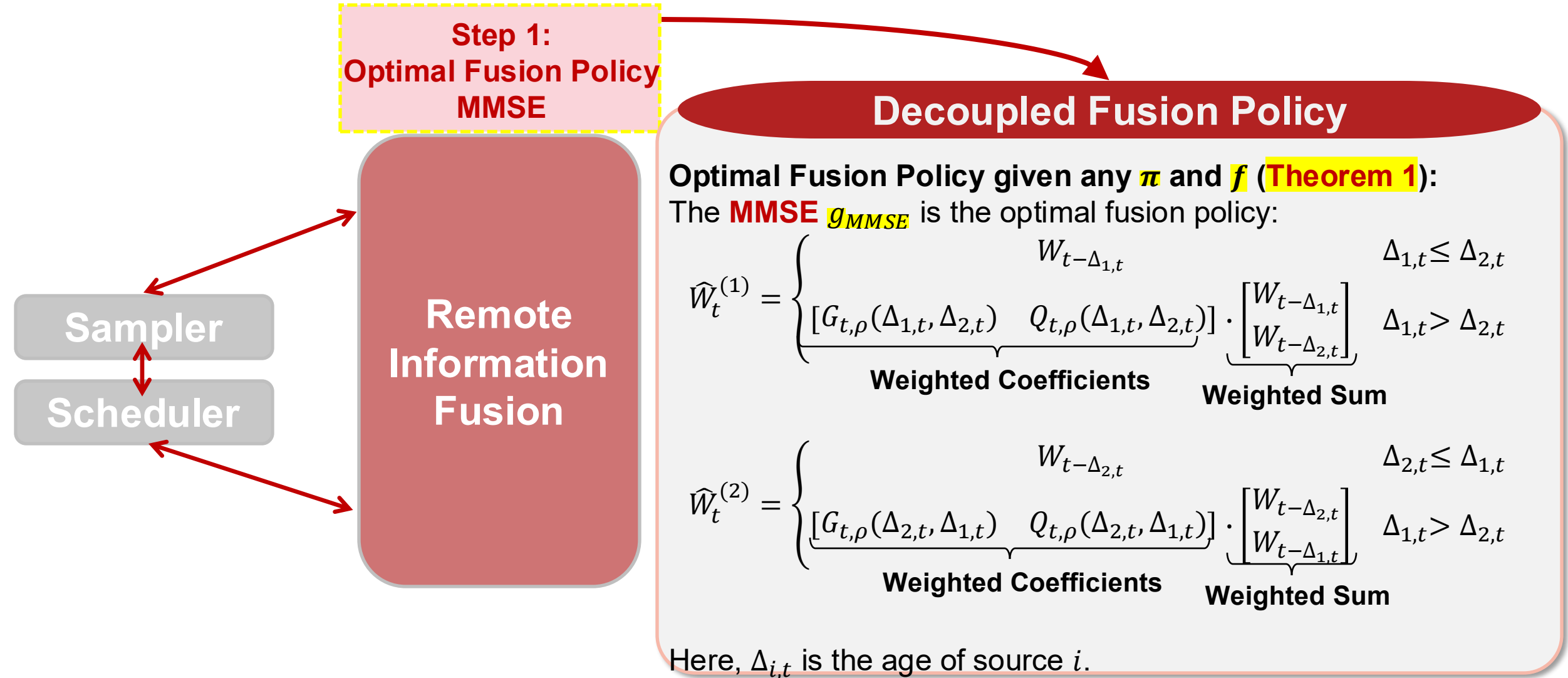
- These three policies are mutually coupled.
- Challenging for explicit optimal policies.

Our work achieves explicit solutions

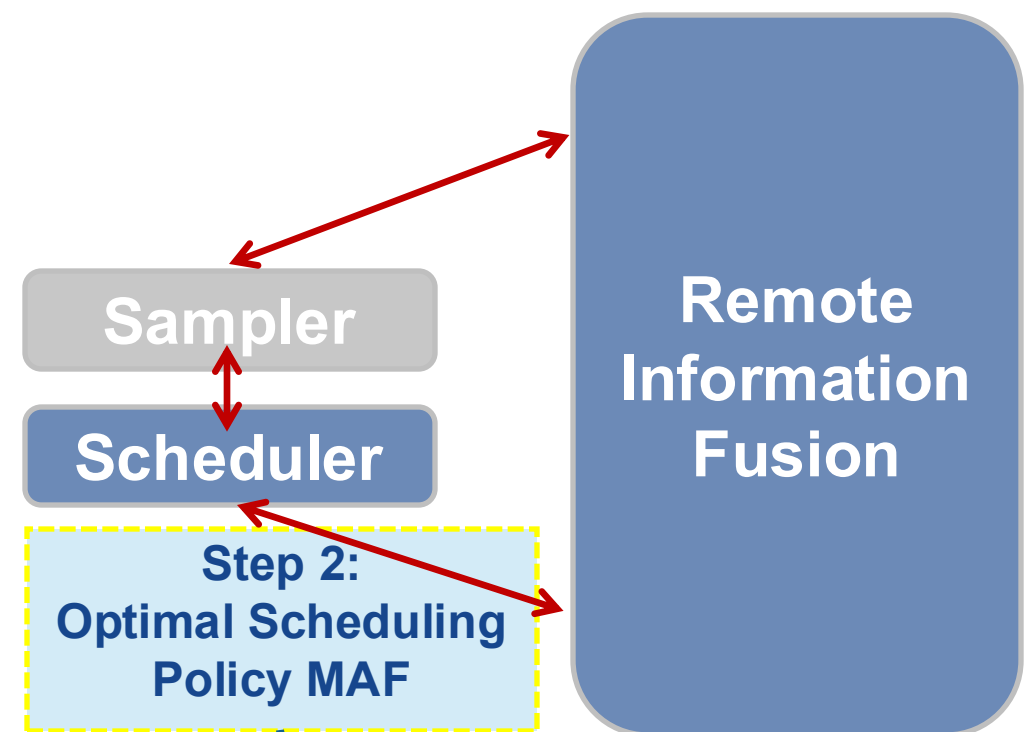
Separation Principles

- **Optimal Remote Information Fusion**
 - Given any sampling and scheduling policy pair, the **MMSE fusion policy is optimal**.
- **Optimal Scheduling Policy**
 - Given any sampling policy and MMSE remote fusion, the optimal scheduling policy is the **Maximum Age First (MAF) scheduling policy**.
- **Optimal Sampling Policy**
 - Given the remote MMSE fusion policy and MAF scheduling policy, the optimal scheduling policy is the **threshold-based Age-optimal sampling policy**.

Main Results: Optimal Remote Information Fusion



Main Results: Optimal Scheduling



Decoupled Scheduling Policy

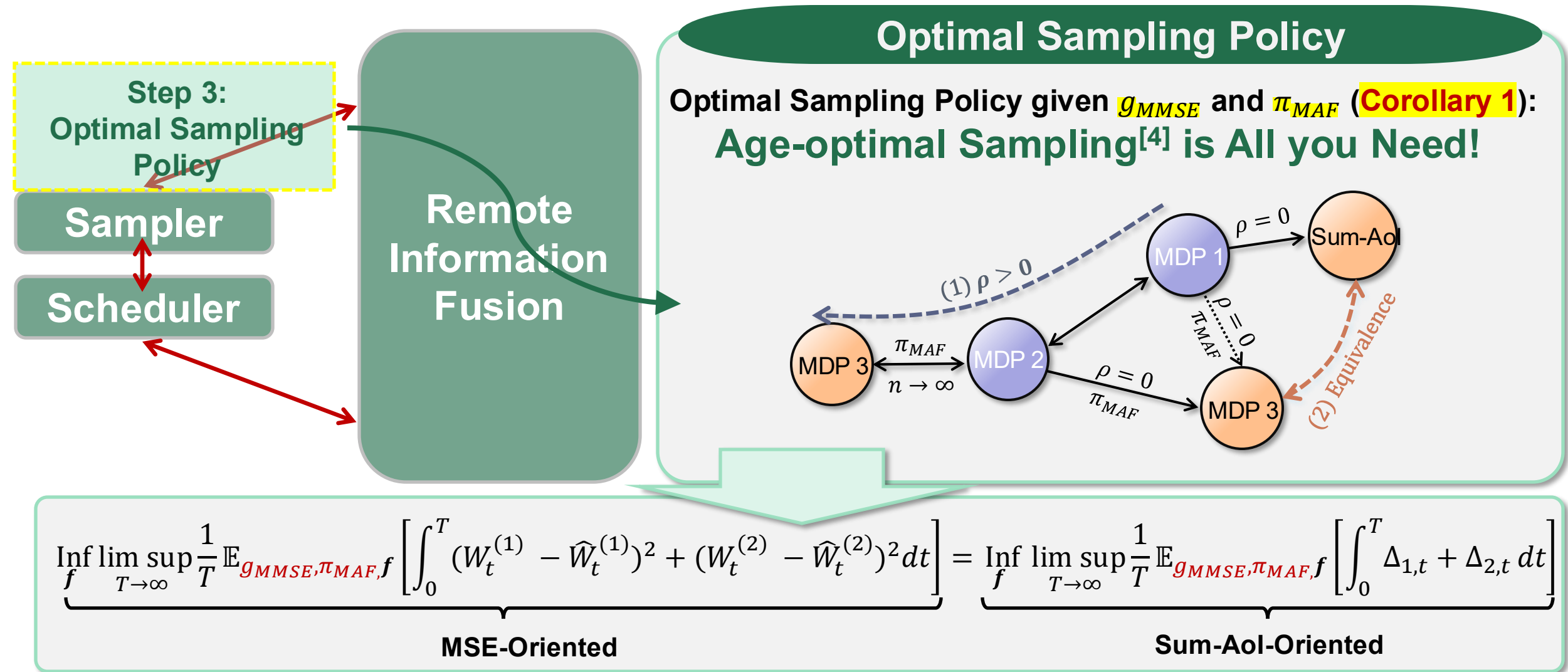
Optimal Scheduling Policy given g_{MMSE} and f (Theorem 2):
The Maximum Age First (MAF) π_{MAF} scheduling is optimal

$$\hat{W}_t^{(1)} = \begin{cases} W_{t-\Delta_{1,t}} & \Delta_{1,t} \leq \Delta_{2,t} \\ [G_{t,\rho}(\Delta_{1,t}, \Delta_{2,t}) \quad Q_{t,\rho}(\Delta_{1,t}, \Delta_{2,t})] \cdot \underbrace{\begin{bmatrix} W_{t-\Delta_{1,t}} \\ W_{t-\Delta_{2,t}} \end{bmatrix}}_{\text{Weighted Sum}} & \Delta_{1,t} > \Delta_{2,t} \end{cases}$$

$$\hat{W}_t^{(2)} = \begin{cases} W_{t-\Delta_{2,t}} & \Delta_{2,t} \leq \Delta_{1,t} \\ [G_{t,\rho}(\Delta_{2,t}, \Delta_{1,t}) \quad Q_{t,\rho}(\Delta_{2,t}, \Delta_{1,t})] \cdot \underbrace{\begin{bmatrix} W_{t-\Delta_{2,t}} \\ W_{t-\Delta_{1,t}} \end{bmatrix}}_{\text{Weighted Sum}} & \Delta_{1,t} > \Delta_{2,t} \end{cases}$$

Here, $\Delta_{i,t}$ is the age of source i .

■ Main Results: Optimal Sampling



[4] Ahmed M. Bedewy, Yin Sun, Sastry Kompella, and Ness B. Shroff. Optimal Sampling and Scheduling for Timely Status Updates in Multi-source Networks. *IEEE Transactions on Information Theory*, 2021



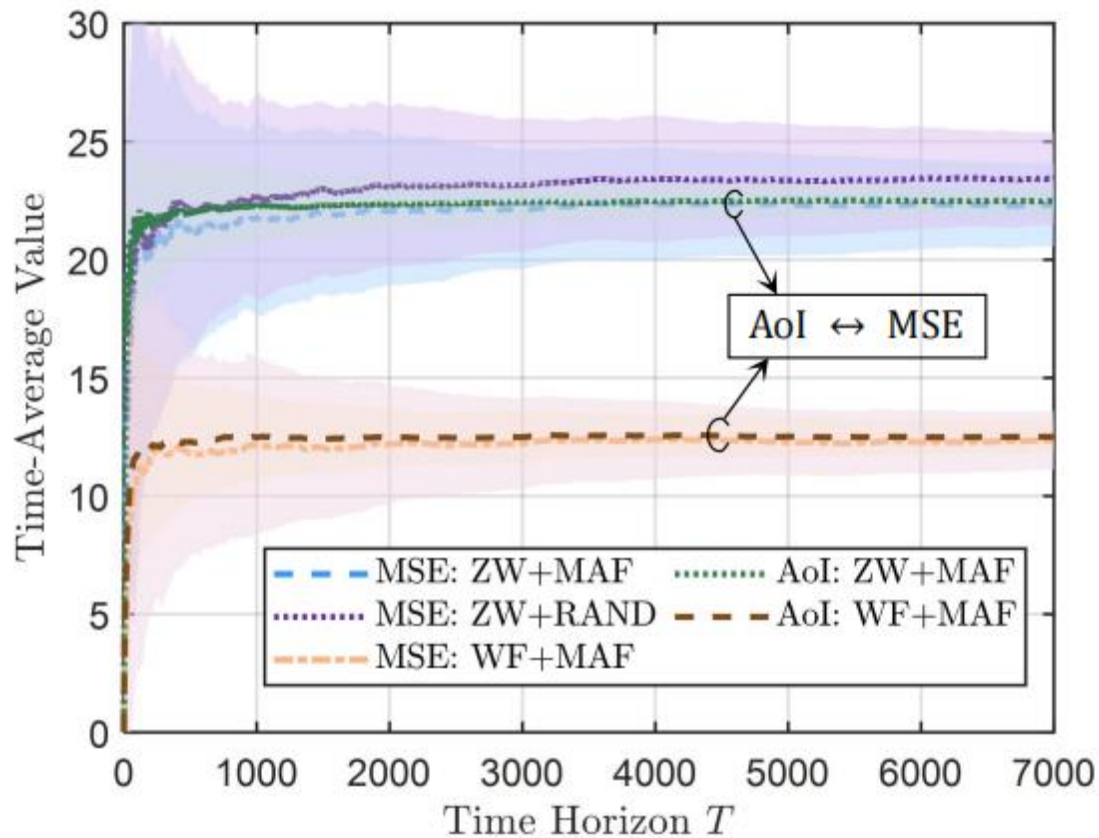
Numerical Results

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Simulations Results

Sum-AoI \leftrightarrow MSE



Insight:

- Even for correlated sources, minimizing sum-AoI is equivalent to minimizing MSE.
- As a result, AoI can serve as a surrogate for remote estimation and information fusion for Wiener Processes, no matter whether the sources are correlated.

Comparative Analysis: New Trends and Challenges

System Model in [5]

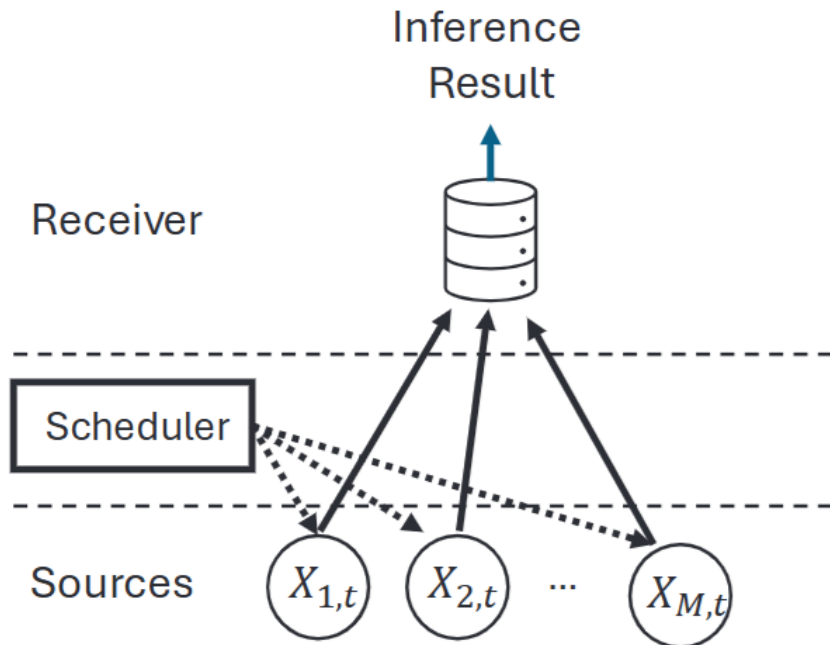


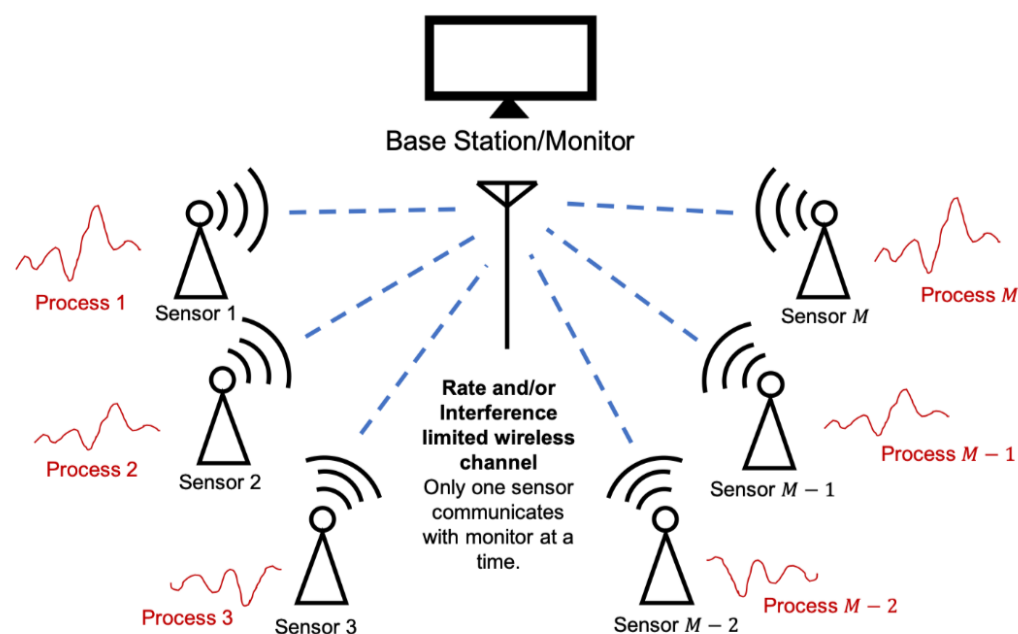
Fig. 1: A remote inference system with M correlated sources.

Comparsion	This Work	Arxiv Paper [5]
Objective	MSE (Not only function Age)	A General Joint Function of Ages (Age-based Inference Error)
Policies Components	Sampling & Scheduling & Fusion over Continuous Time	Scheduling over Discrete Time Slots
Channel	Random Delay	Zero Delay
Addressed Challenges	Explicit Optimal Co-Design	How to Decouple Ages for scheduling?

[5] Md Kamran Chowdhury Shisher, Vishrant Tripathi, Mung Chiang, and Christopher G. Brinton. Aol-based Scheduling of Correlated Sources for Timely Inference. Arxiv, **September 2, 2025**.

Comparative Analysis: New Trends and Challenges

System Model in [6]



Comparsion	This Work	TMC [6]
Objective	MSE	MSE
Policy Components	Sampling & Scheduling & Fusion over Continuous Time	Scheduling Only over Discrete Time Slot
Channel	Random Delay	Zero Delay
Addressed Challenges	Explicit Optimal Co-Design	How to Schedule Correlated Sources?
Sources	Non-stationary Markov	Non-stationary Hidden Markov

[6] R. V. Ramakanth, V. Tripathi and E. Modiano. Monitoring Correlated Sources: AoI-based Scheduling is Nearly Optimal, IEEE Transaction on Mobile Computing, **February, 2025**

Amin Li and Elif Uysal (METU)

Thank you

On behalf of METU CNG
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