

# WiSwarm at the Edge: Wireless Networking for Collaborative Teams of UAVs

Igor Kadota

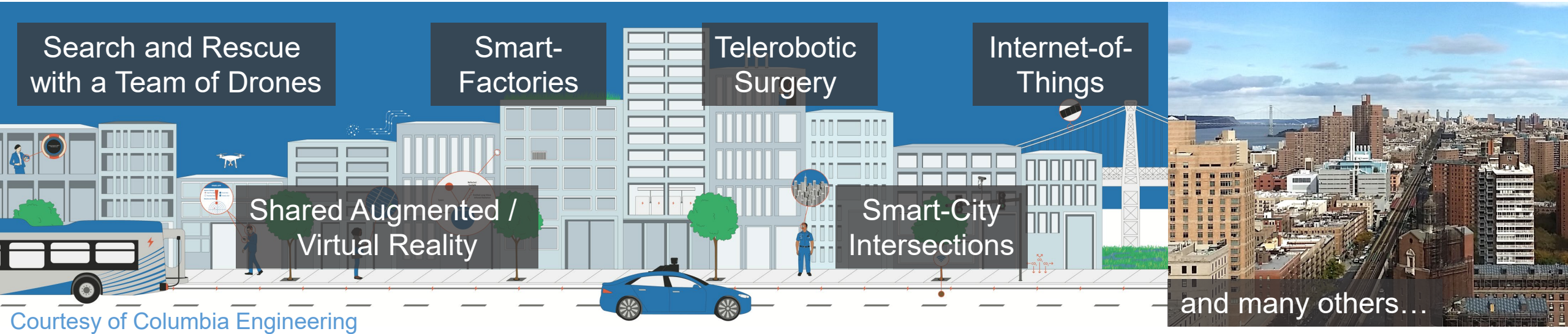
Assistant Professor

Department of Electrical and Computer Engineering

WiCI Seminars, Oct. 2025

Northwestern | McCORMICK SCHOOL OF  
**ENGINEERING**

# Motivation: Future Applications

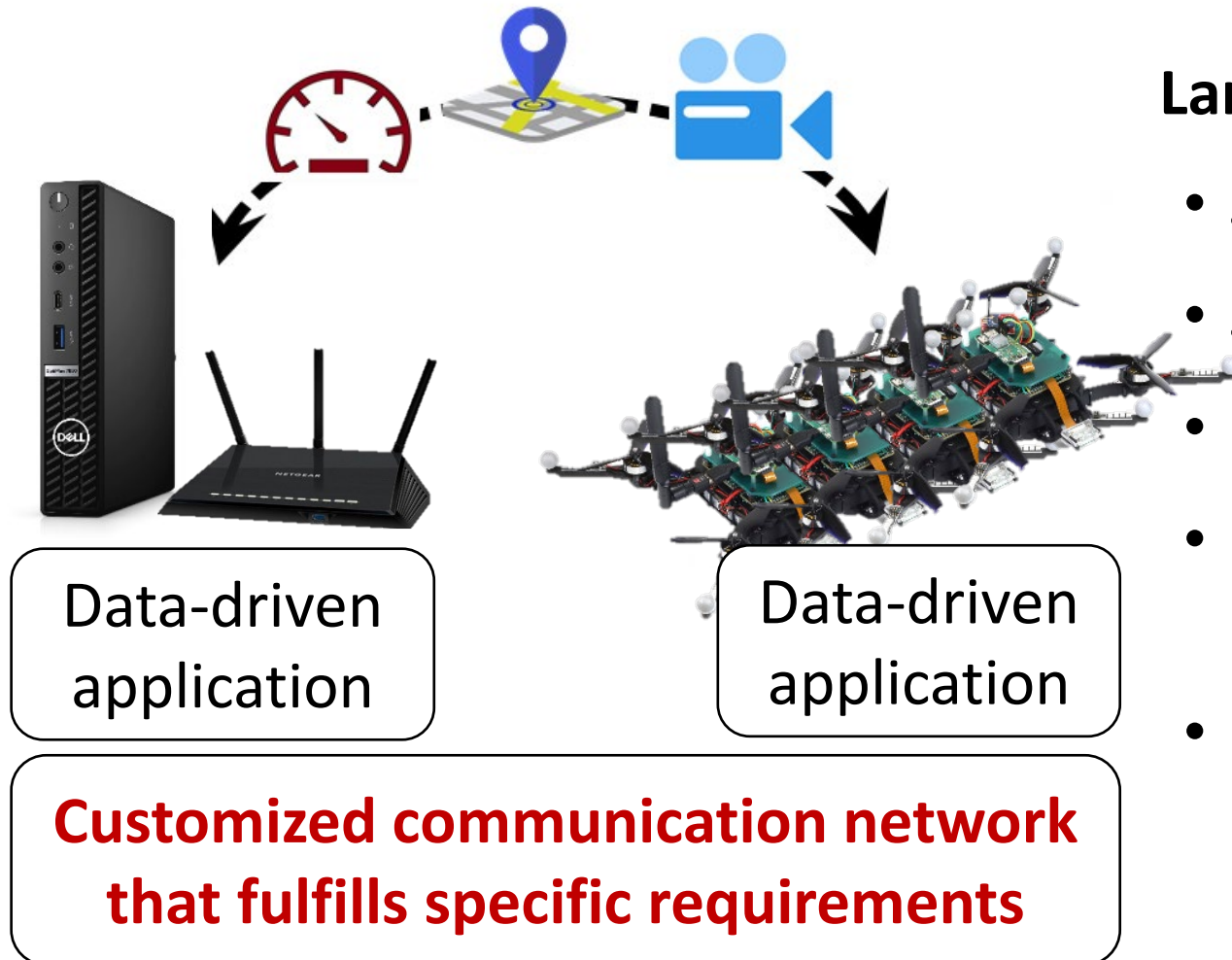


Rural applications (discussed at **AraFest'25**):

- Sprayer teleoperation and monitoring
- Animal herding via UAVs
- Autonomous precision planting
- UAV/UGV fleet control



# Motivation: Future of Networking



## Large-scale time-sensitive applications:

- Smart Cities (Autonomous vehicles)
- Smart-Factories (Machines)
- Multi-robot formations (Robots)
- **Exploration of dynamic environments (Drones)**
- IoT platforms for agriculture (Sensors)

⋮

# In this talk: Time-Sensitive Application



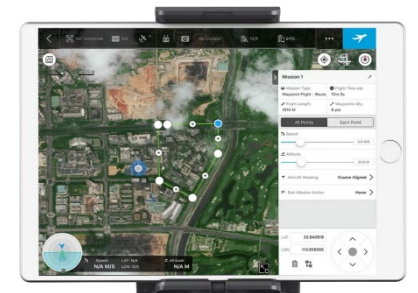
**Large-Scale Network of  
Sensing-Drones**

Wireless  
Access Point  
(AP)



**DYNAMIC  
ENVIRONMENT**

**Edge Server:**  
Monitoring  
and Control

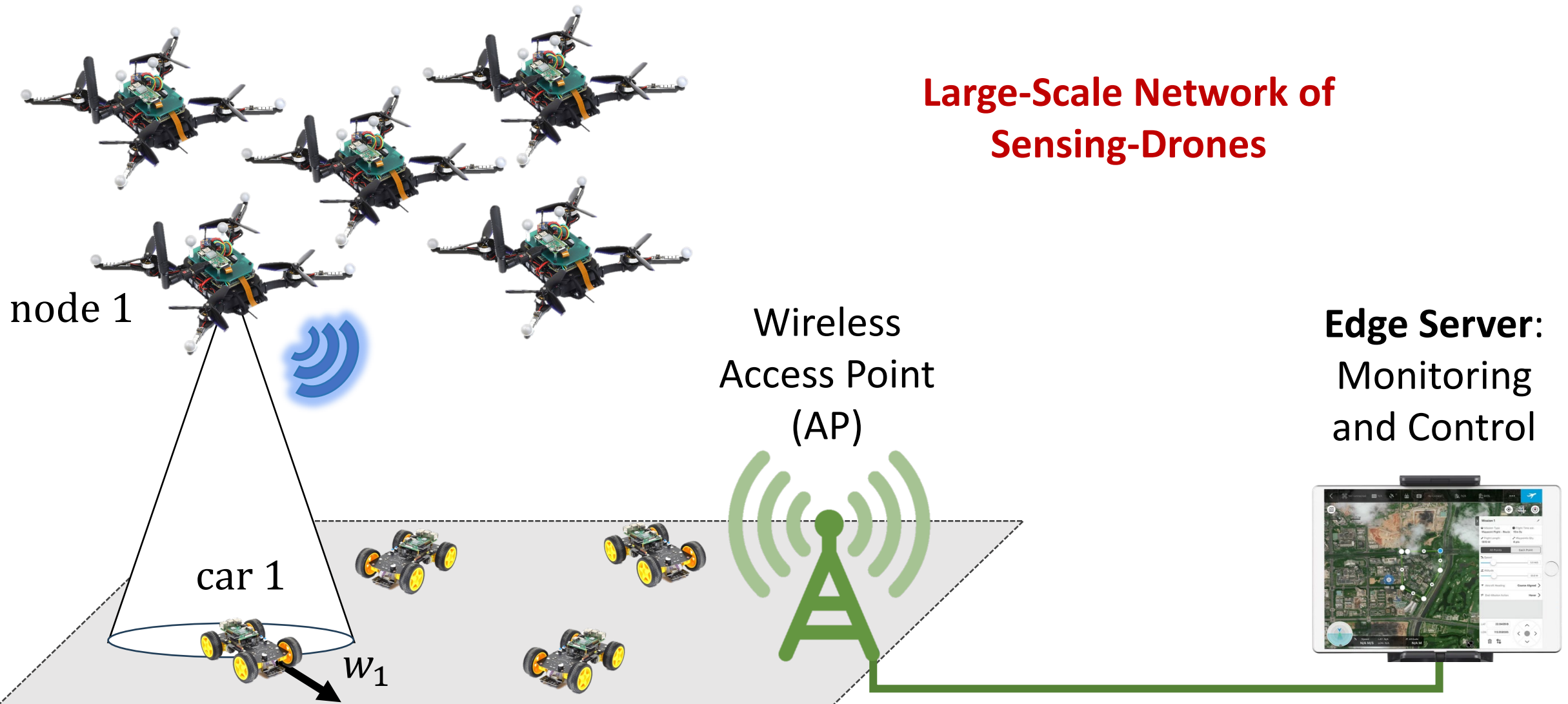




WiSwarm was introduced in  
IEEE INFOCOM 2023

Collaboration with  
V. Tripathi, E. Tal,  
M. Rahman, A. Warren,  
S. Karaman, E. Modiano

# In this talk: Time-Sensitive Application





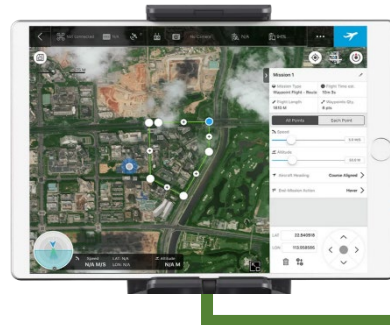
# In this talk: Time-Sensitive Application



**Large-Scale  
Network of  
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Wireless  
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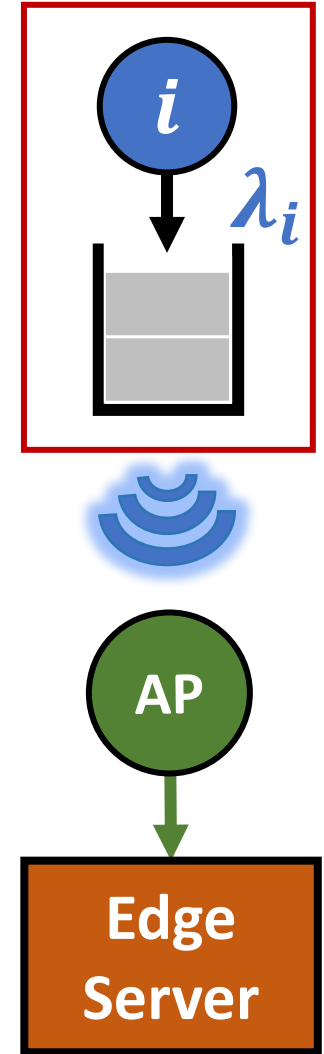
**Edge Server:**  
Monitoring  
and Control



Time-Sensitive  
Information

Queueing

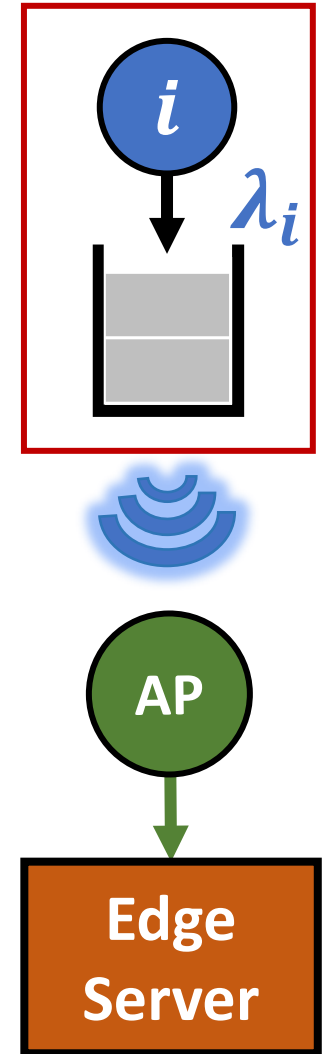
Wireless  
Channel  
Access



# Outline

- **Theory:**

- Introduction to Age-of-Information using a Simple System →





# Outline

- **Theory:**

- Introduction to Age-of-Information using a Simple System
- Formulation of the AoI Minimization Problem in a Wireless Network

## Minimizing the Age of Information in Broadcast Wireless Networks

Igor Kadota, Elif Uysal-Biyikoglu, Rahul Singh and Eytan Modiano

**Abstract**—We consider a wireless broadcast network with a base station sending time-sensitive information to a number of clients. The Age of Information (AoI), namely the amount of time that elapsed since the most recently delivered packet was generated, captures the freshness of the information. We formulate a discrete-time decision problem to find a scheduling policy that minimizes the expected weighted sum AoI of the clients in the network. To the best of our knowledge, this is the

providing a minimum throughput. Fig. 1 illustrates the case of two sequences of packet deliveries that have the same throughput but different delivery regularity. In general, a minimum throughput requirement can be fulfilled even if long periods with no delivery occur, as long as those are balanced by periods of consecutive deliveries.

Minimizing the AoI is particularly challenging in wireless

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- Theoretical Results: Lower Bound and Scheduling Policies

## Optimizing Age of Information in Wireless Networks with Throughput Constraints

Igor Kadota, Abhishek Sinha and Eytan Modiano  
Laboratory for Information & Decision Systems, MIT

*Abstract*—Age of Information (AoI) is a performance metric that captures the freshness of the information from the  $m$ th packet delivered by sensor  $i$  is the most recent. Then, the

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- **Generalizations:** Packet Generation + Queueing + Scheduling Policies

## Minimizing the Age of Information in Wireless Networks with Stochastic Arrivals

Igor Kadota  
Massachusetts Institute of Technology  
kadota@mit.edu

Eytan Modiano  
Massachusetts Institute of Technology  
modiano@mit.edu

### ABSTRACT

We consider a wireless network with a base station serving multiple traffic streams to different destinations. Packets from each stream

networks, other performance requirements are increasingly relevant. In particular, the Age of Information (AoI) is a performance metric that was recently proposed in [26, 27] and has been receiving

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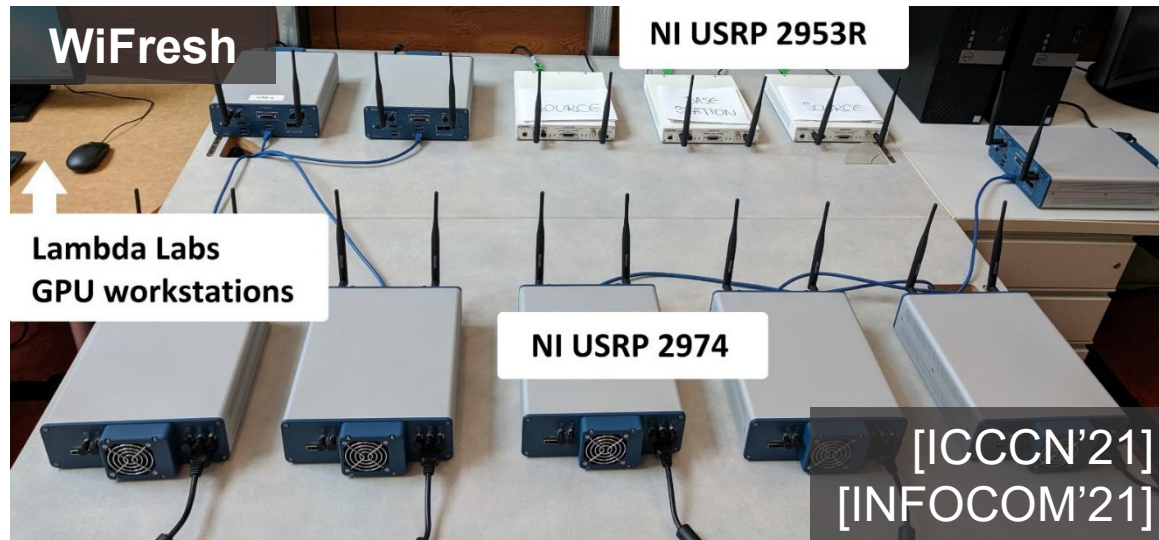
Yin Sun, I.K., Rajat Talak, and Eytan Modiano

*Age of Information: A New Metric for Information Freshness*

Morgan and Claypool Publishers, 2019.

# Outline

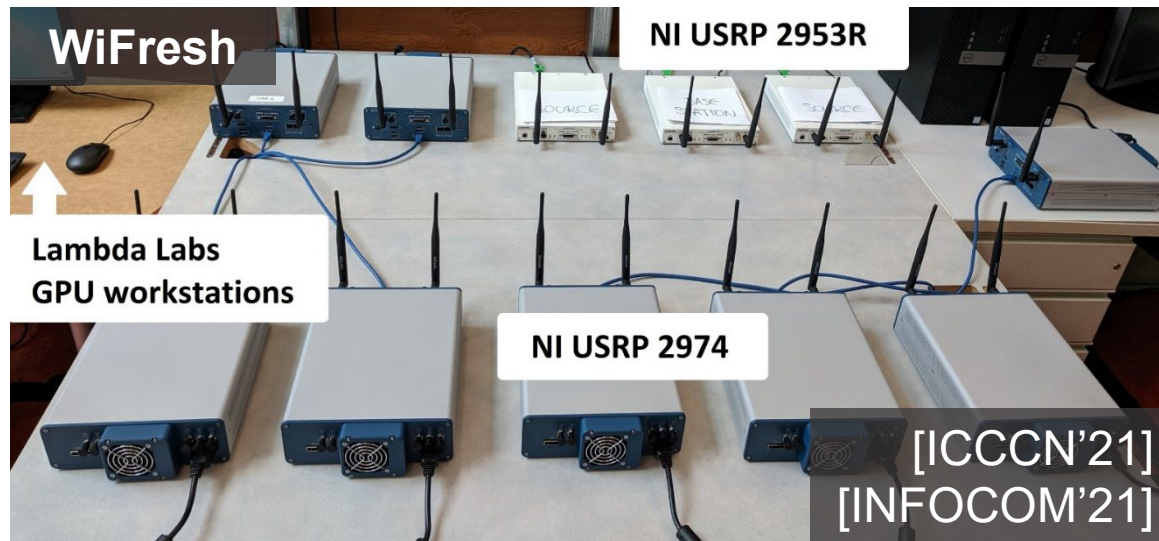
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- **System Implementation in a SDR Testbed and Flight Tests with Drones**





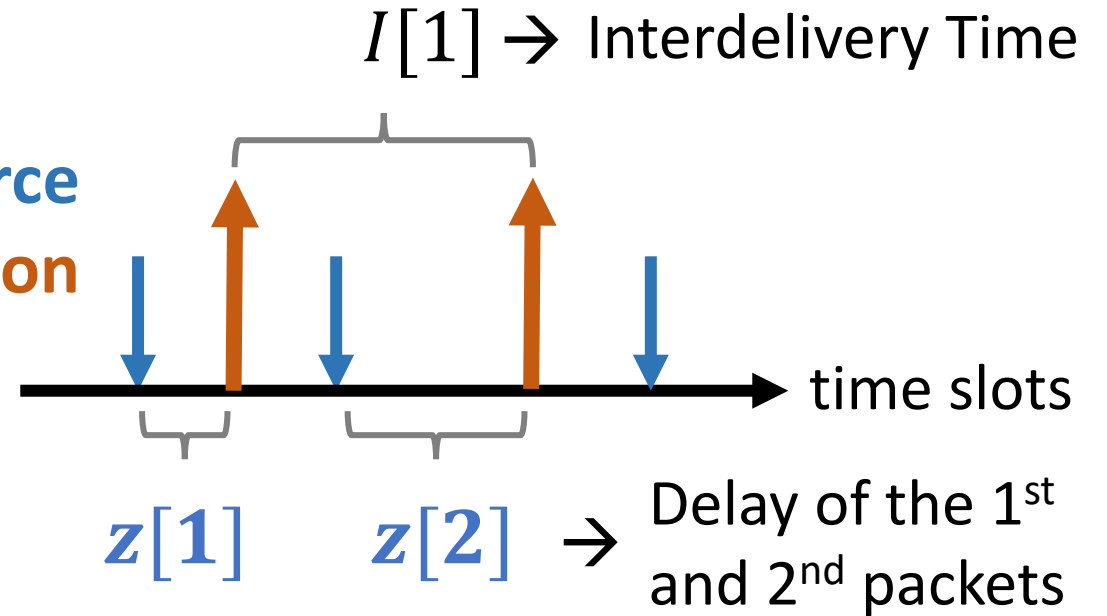
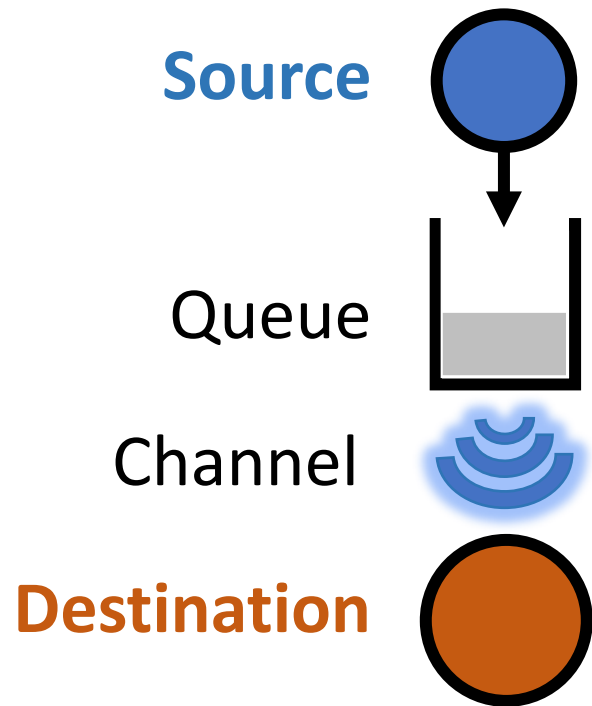
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# Background: Age-of-Information

Packet generation at the source  
Delivery of packets to the destination

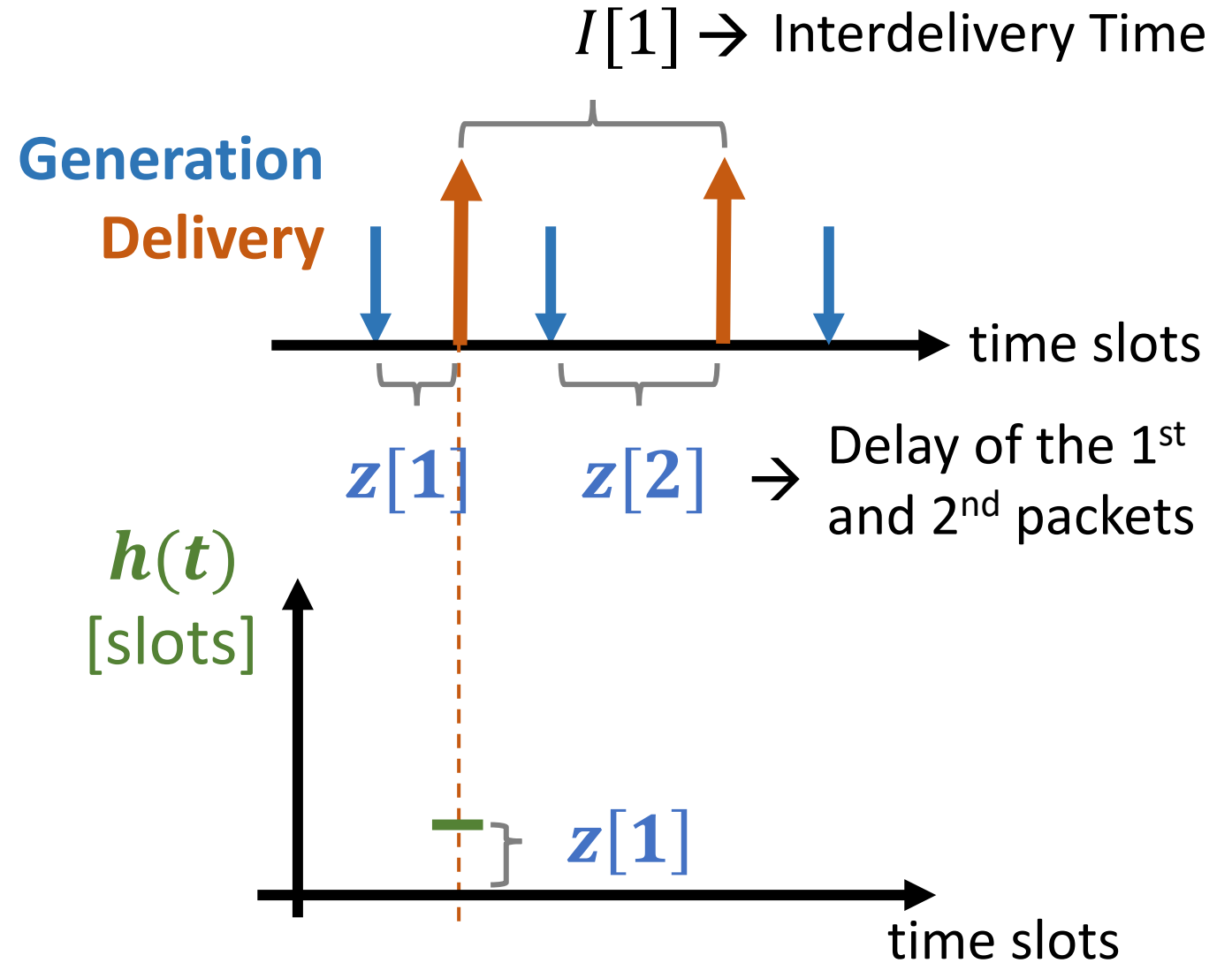
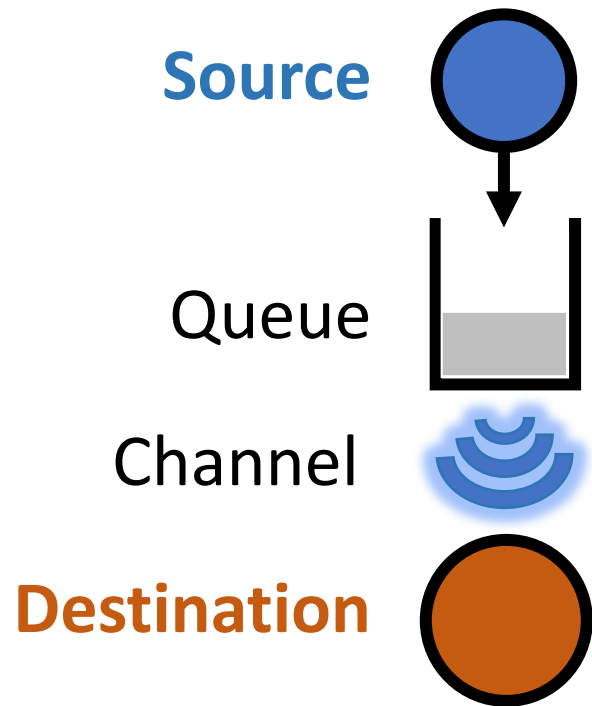


How **old** is the information  
at the **destination**?



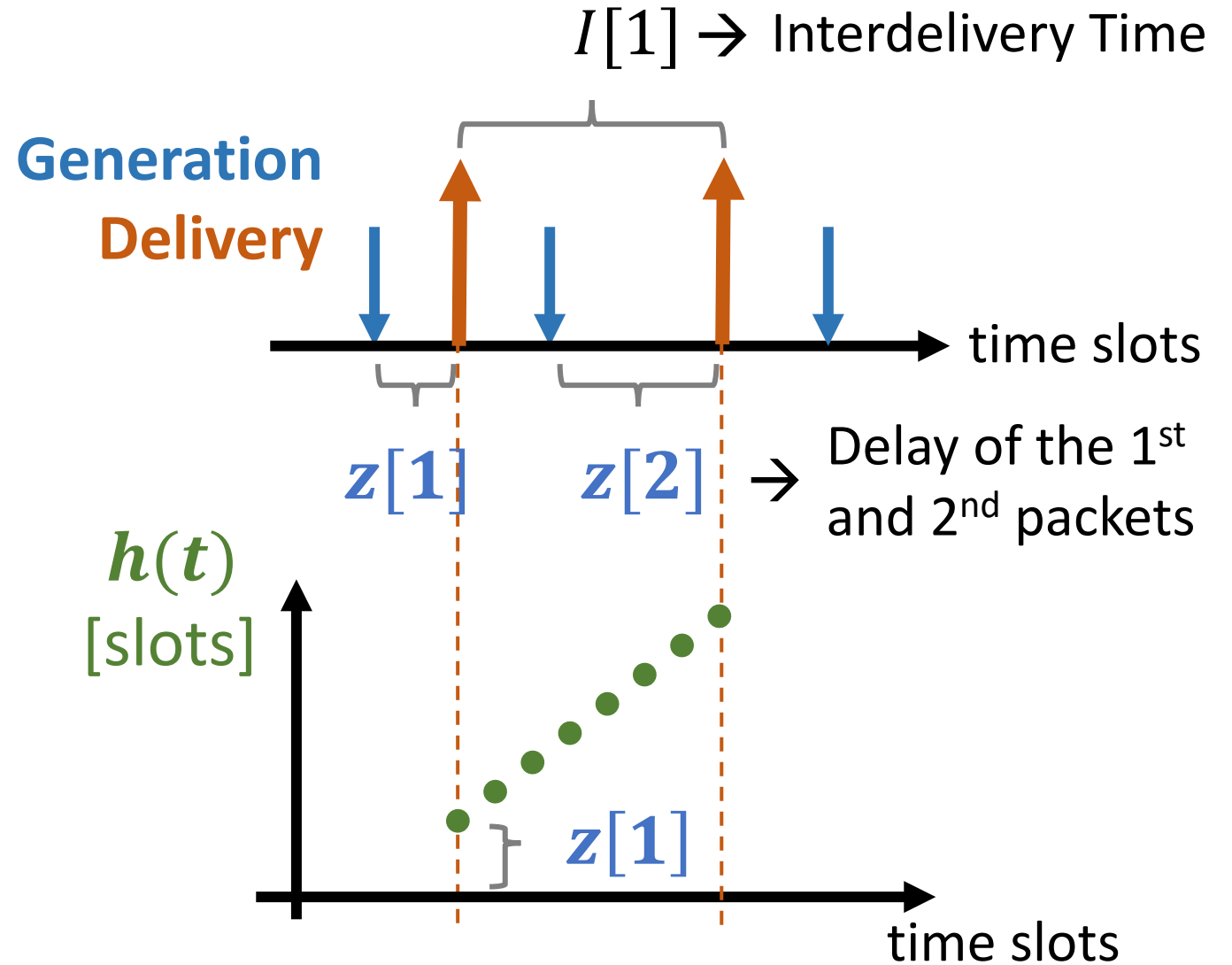
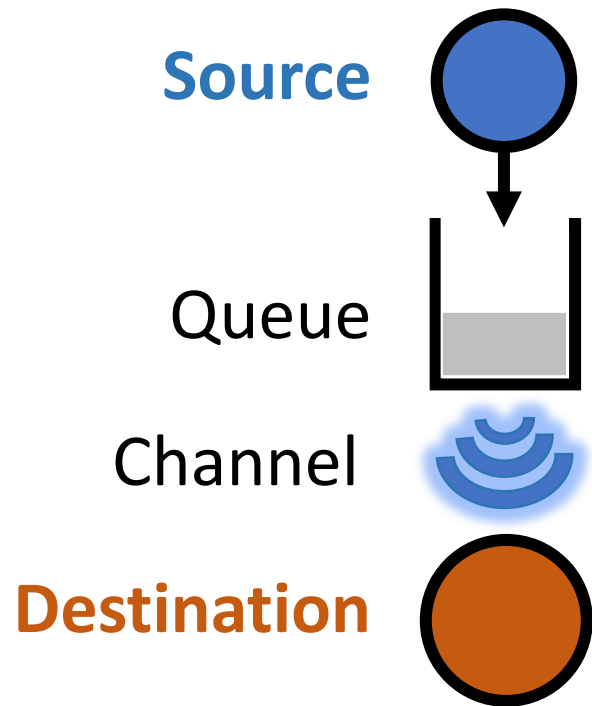
# Background: Age-of-Information

**Aol:** time elapsed since the generation of the freshest received packet



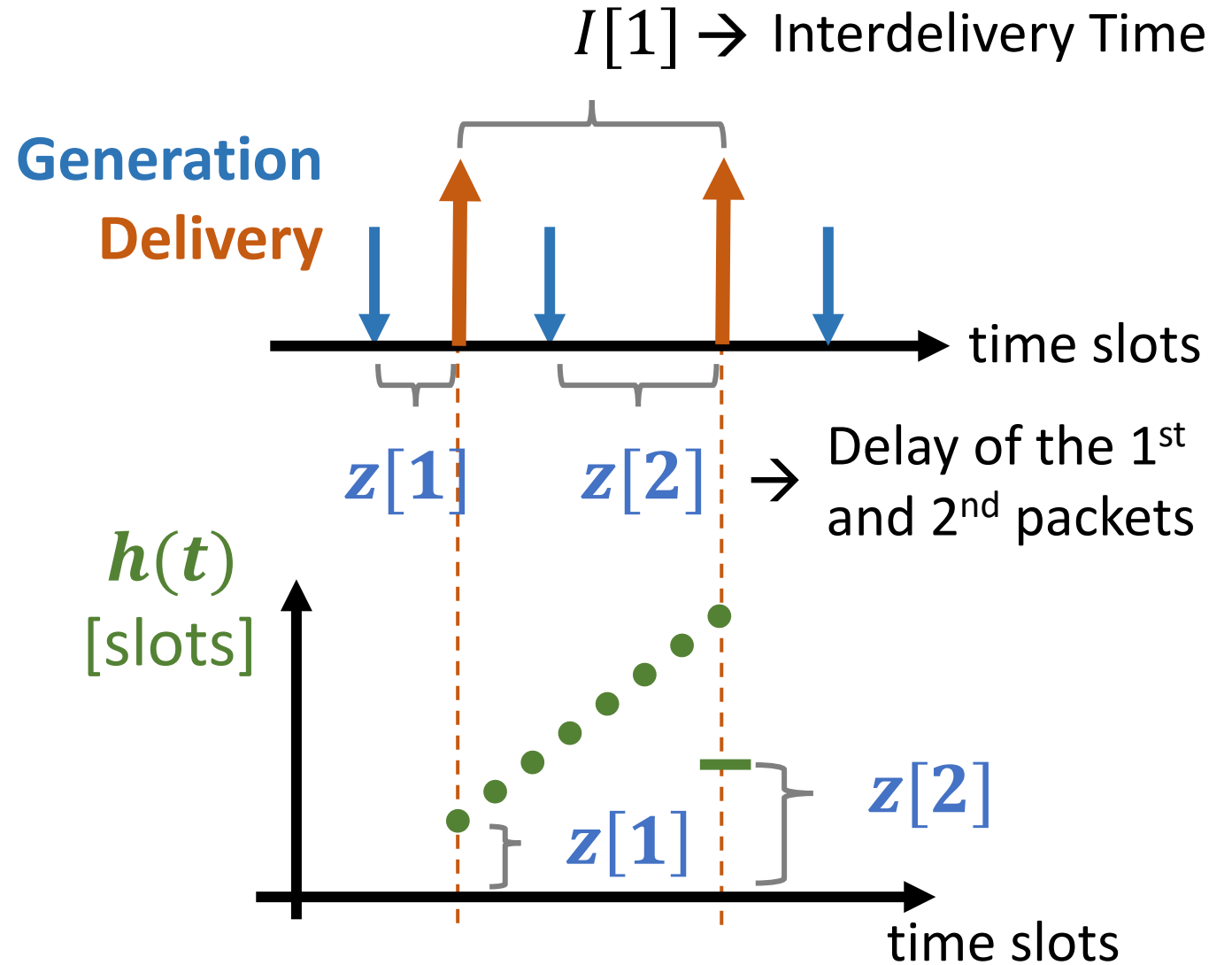
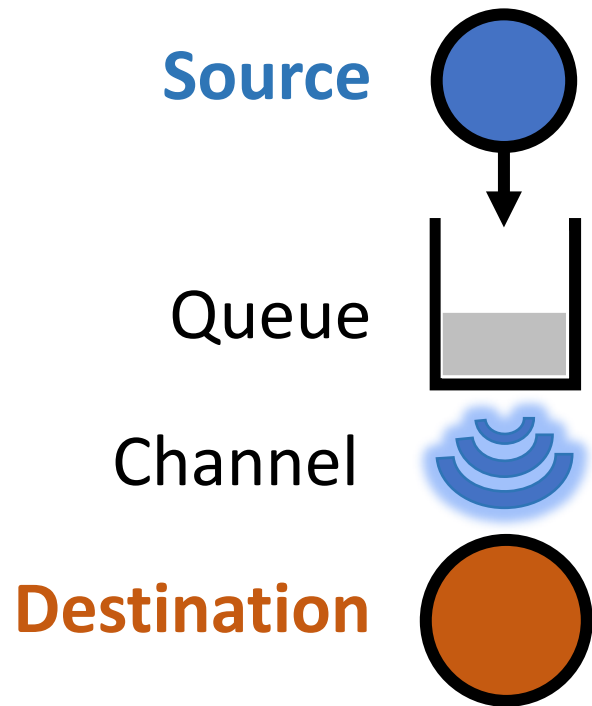
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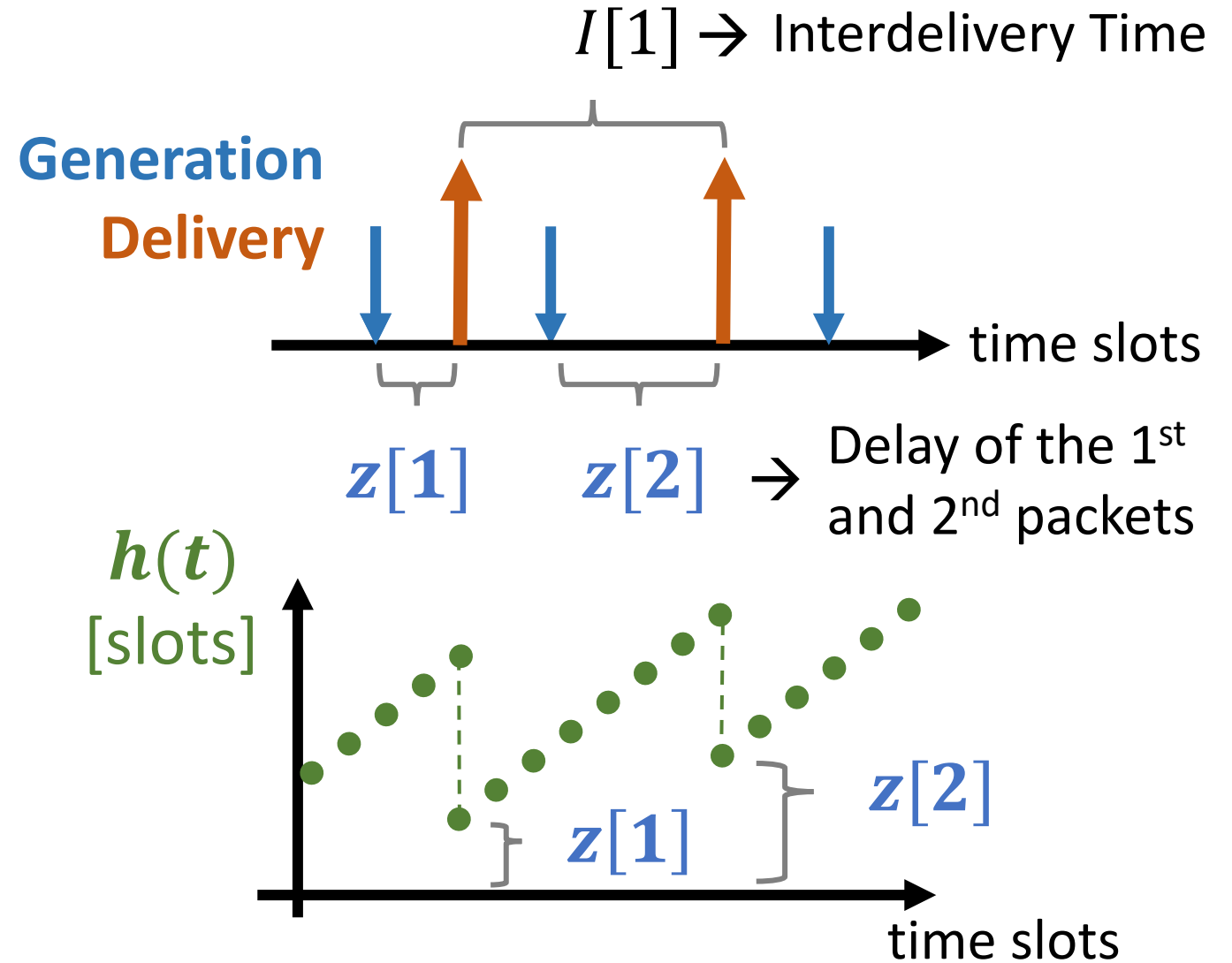
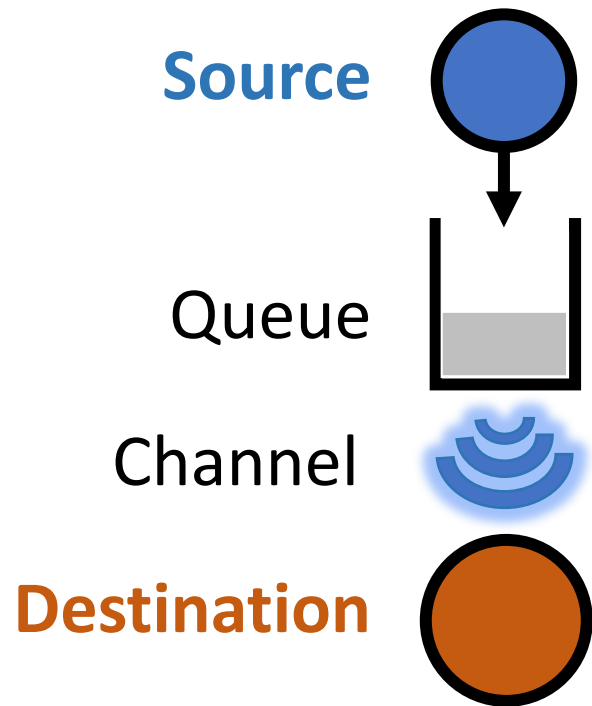
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# Background: Age-of-Information

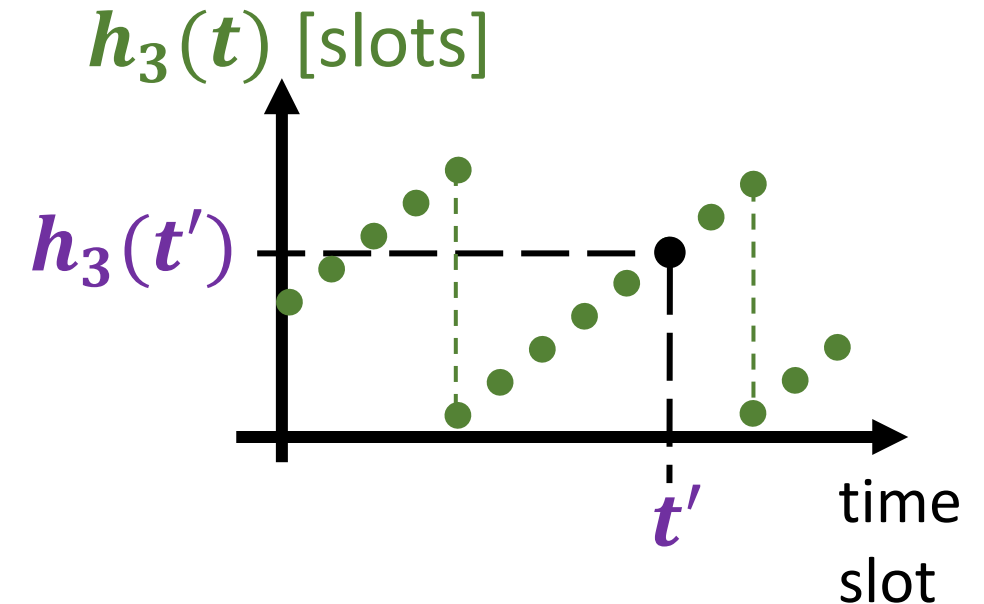
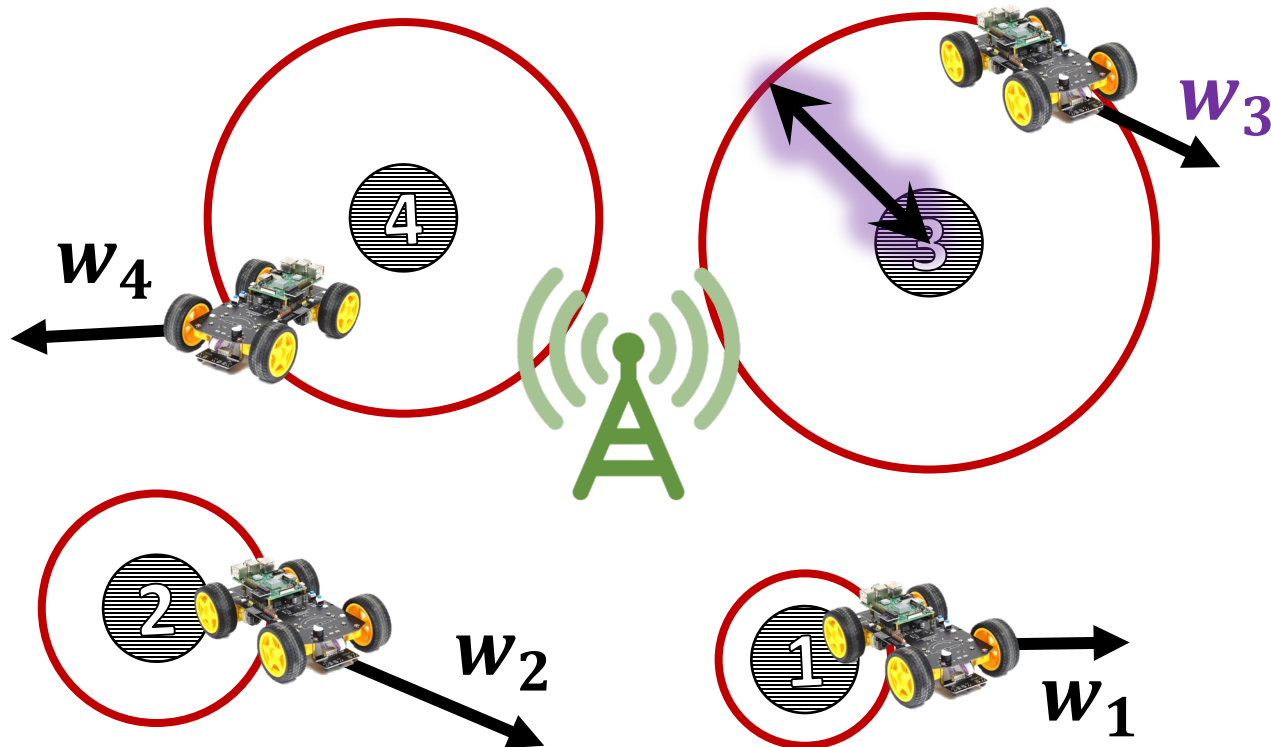
**Aol:** time elapsed since the generation of the freshest received packet



# Background: Aol and Position Uncertainty

From the perspective of the edge server:

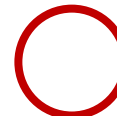
**Position uncertainty** of car 3 in slot  $t'$   
is given by the **radius**  $w_3 h_3(t')$



current position of car



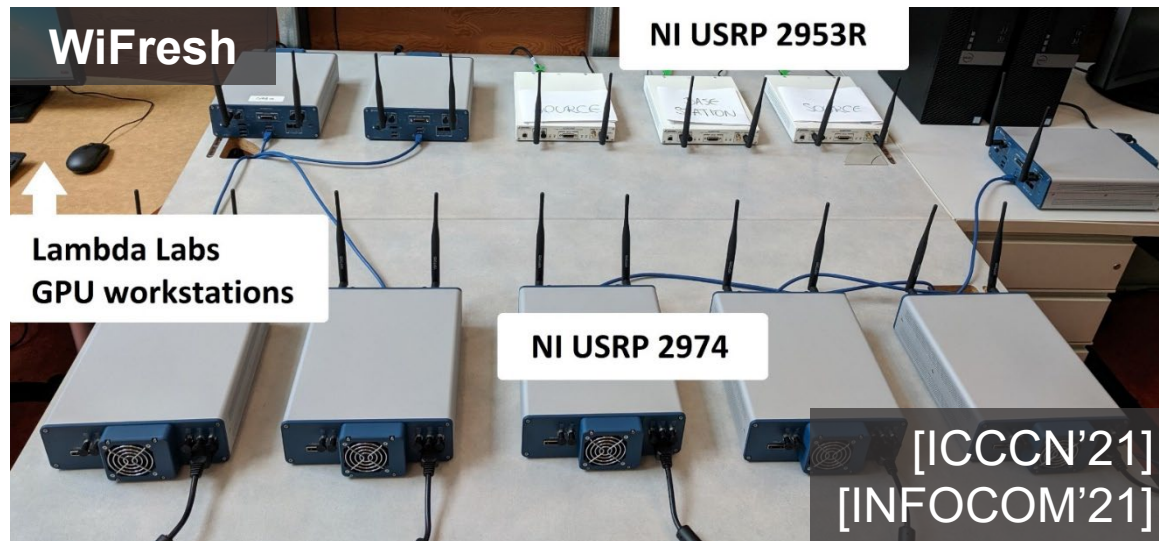
last known position  
(from last packet received)



position uncertainty

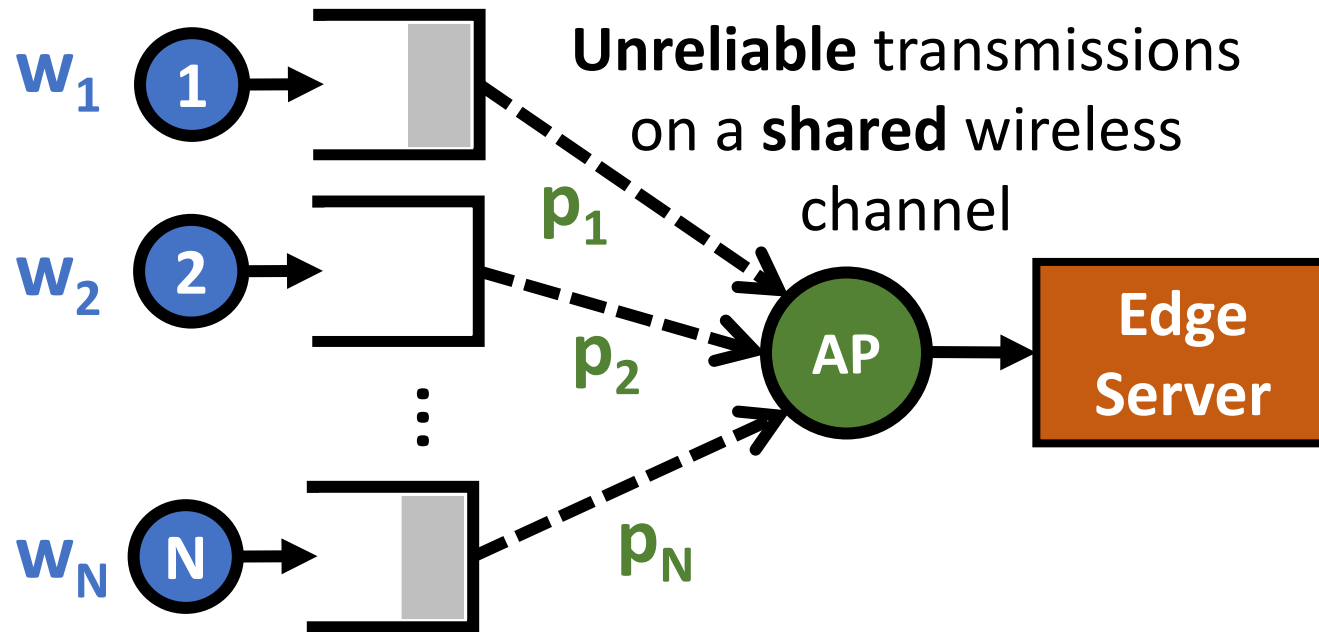
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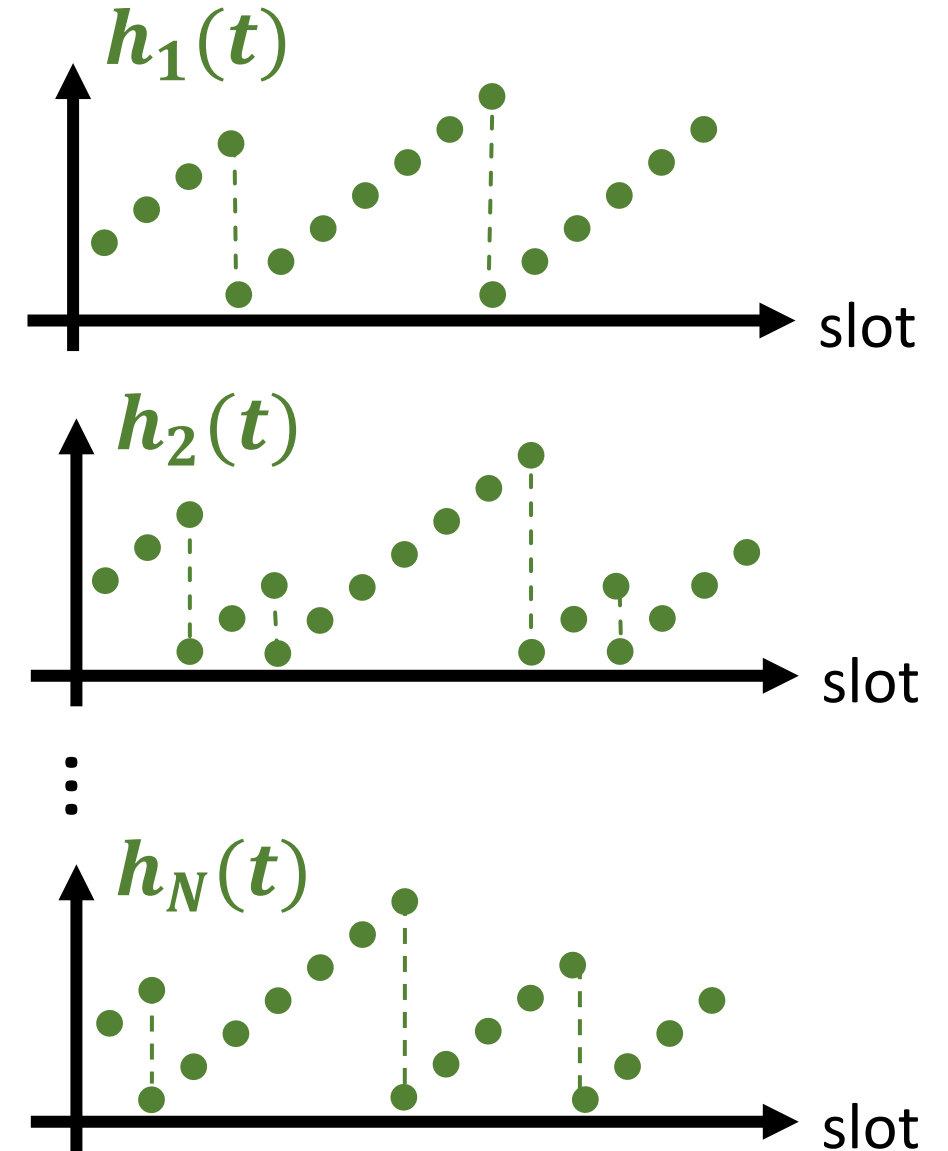
# Network Model

## Packet generation and queueing



Weight  $w_i > 0$  represents **priority** of source  $i$

Probability  $p_i \in (0,1]$  represents **quality of the link**





# Network Model

Packet g

$w_1$

1

$w_2$

2

$w_N$

N

**For simplicity, in this talk, we assume that:**

- Sources generate fresh packets at will

**As a result, the Aol evolves as follows:**

$$h(t+1) = \begin{cases} 1, & \text{if packet is delivered in slot } t \\ h(t) + 1, & \text{otherwise} \end{cases}$$

Weight  $w_i > 0$  represents **priority** of source  $i$

Probability  $p_i \in (0,1]$  represents **quality of the link**

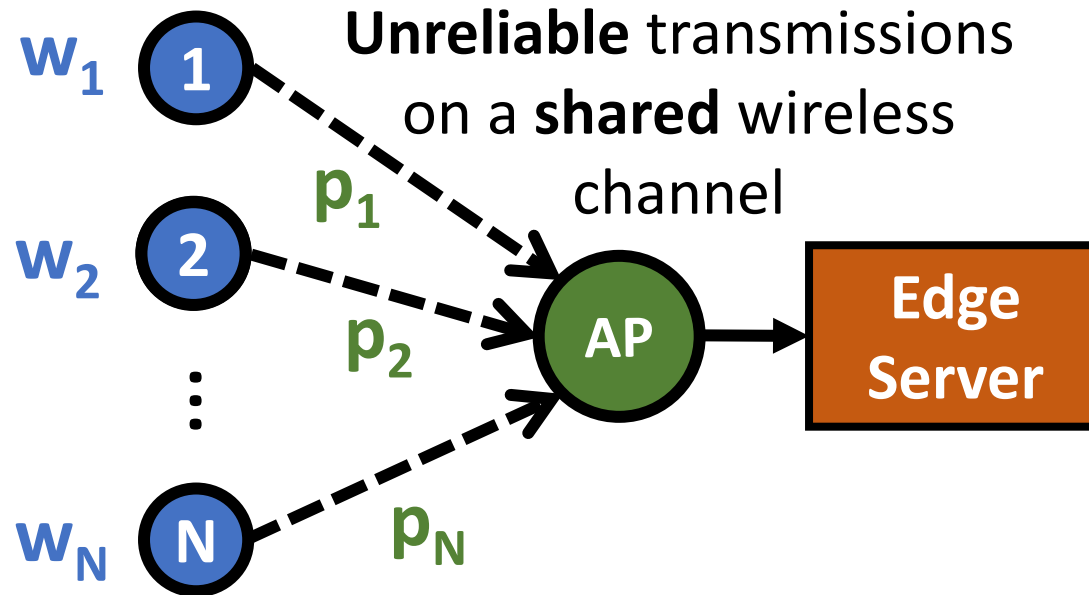
slot

slot

slot

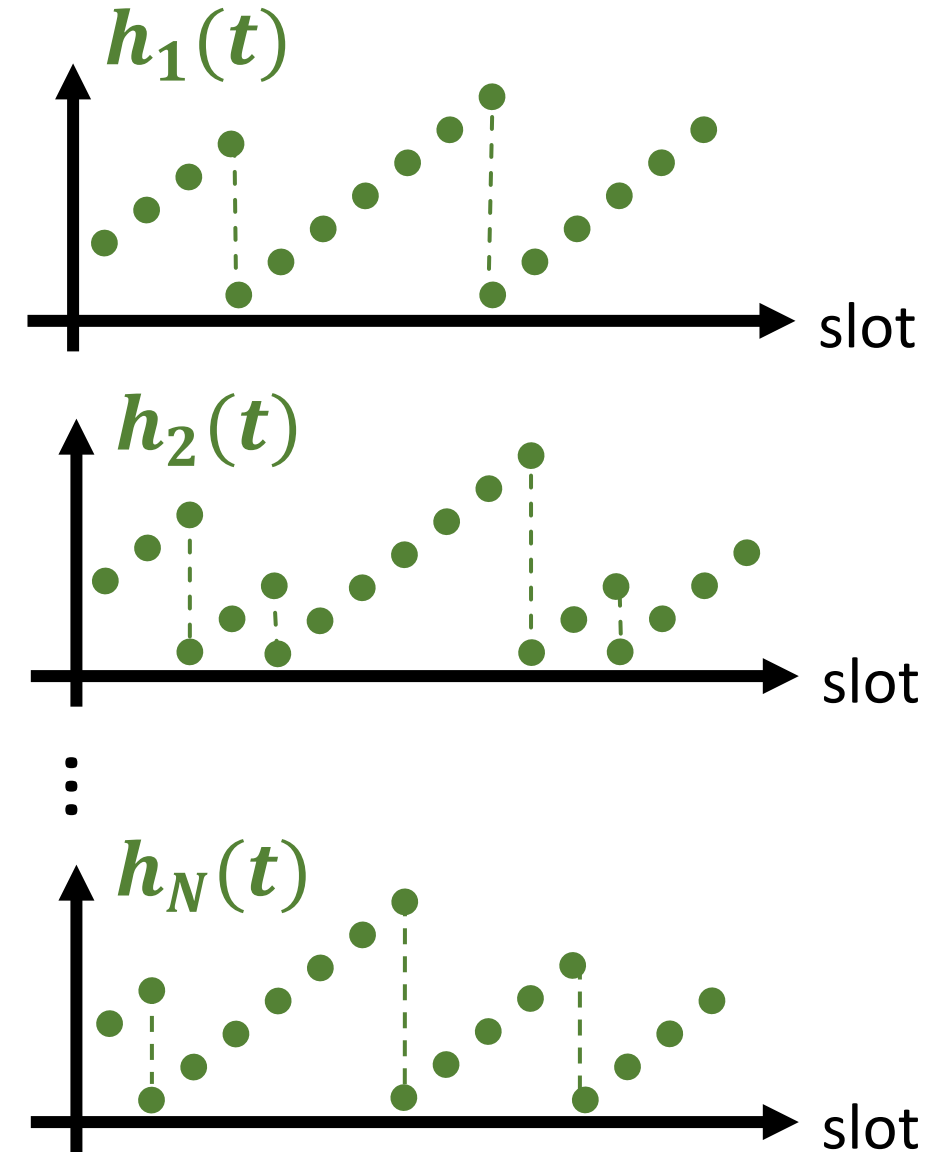
# Network Model

For simplicity: **Packet generation at will**



Weight  $w_i > 0$  represents **priority** of source  $i$

Probability  $p_i \in (0,1]$  represents **quality of the link**



# Network Model: Objective Function

**Goal:** find a **transmission scheduling policy**  $\pi^* \in \Pi$  that minimizes the Expected Weighted Sum Age-of-Information (**EWSAoI**):

$$\mathbb{E}[J^*] = \min_{\pi \in \Pi} \left\{ \lim_{T \rightarrow \infty} \mathbb{E}[J_T^\pi] \right\}, \text{ where } \mathbb{E}[J_T^\pi] = \frac{1}{TN} \sum_{t=1}^T \sum_{i=1}^N w_i \mathbb{E}[h_i^\pi(t)]$$

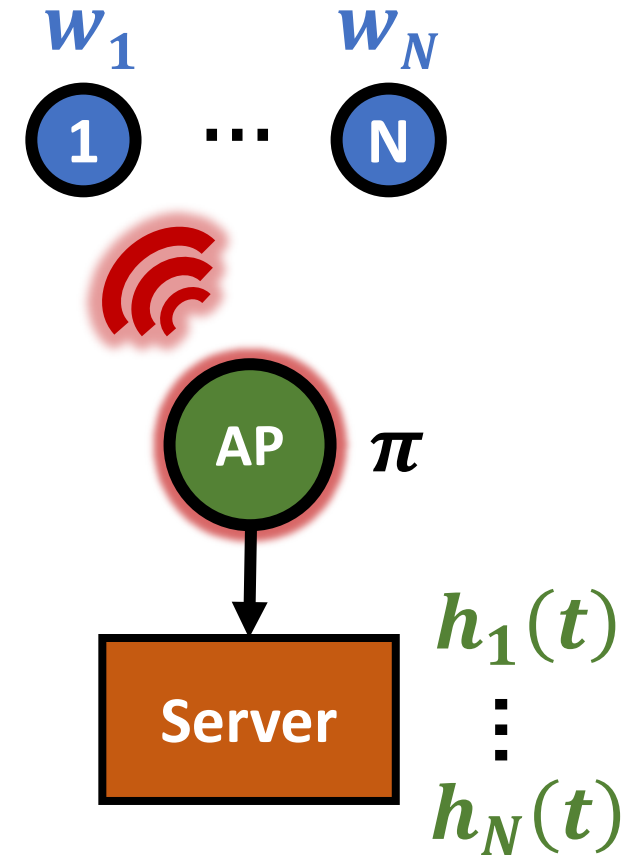
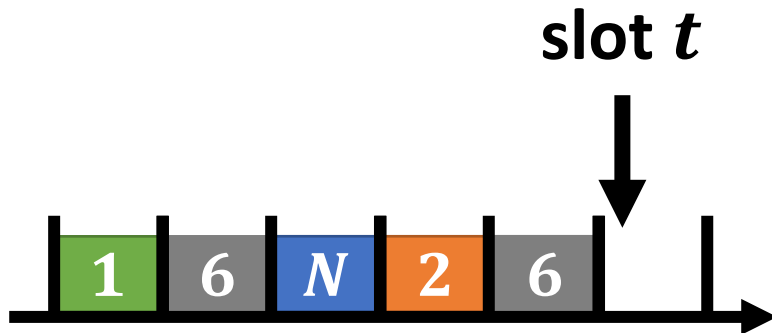
Class of non-anticipative policies  $\Pi$ . Arbitrary policy  $\pi \in \Pi$ .

I. K., A. Sinha, and E. Modiano, "Scheduling Algorithms for Optimizing Age of Information in Wireless Networks with Throughput Constraints," IEEE/ACM ToN, 2019. [conference version received the best paper award at IEEE INFOCOM 2018]  
I. K. and E. Modiano, "Minimizing the Age of Information in Wireless Networks with Stochastic Arrivals," IEEE TMC 2021. [conference version was a best paper award finalist at ACM MobiHoc 2019]

# Network Model: Scheduling Policy $\pi$

**Scheduling decision** during slot  $t$ :

- 1) AP selects** a single source  $i \in \{1, 2, \dots, N\}$   
[ $u_i(t) = 1$  and  $u_j(t) = 0, \forall j \neq i$ ]

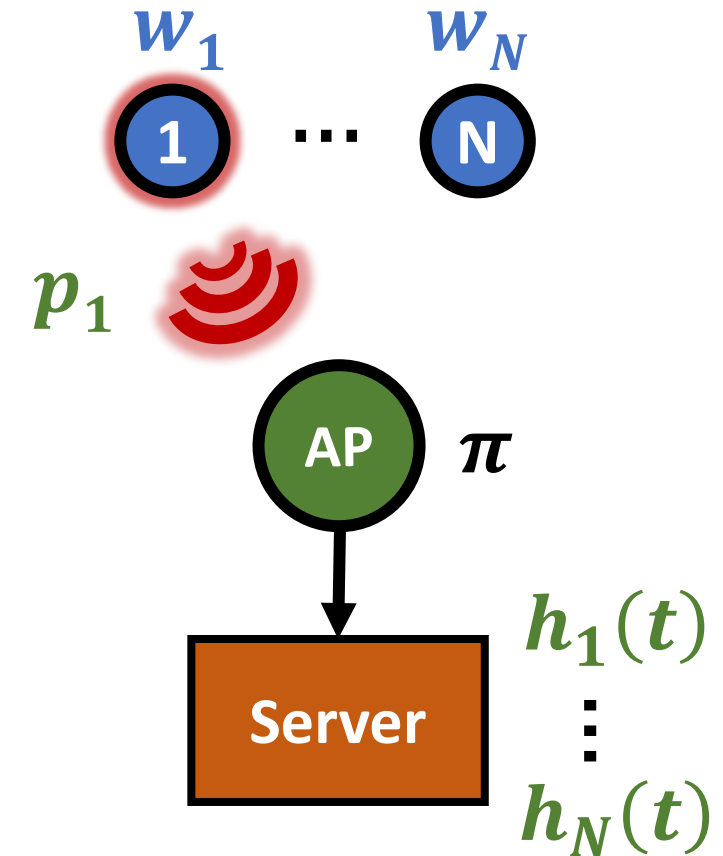


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# Network Model: Scheduling Policy $\pi$

**Scheduling decision** during slot  $t$ :

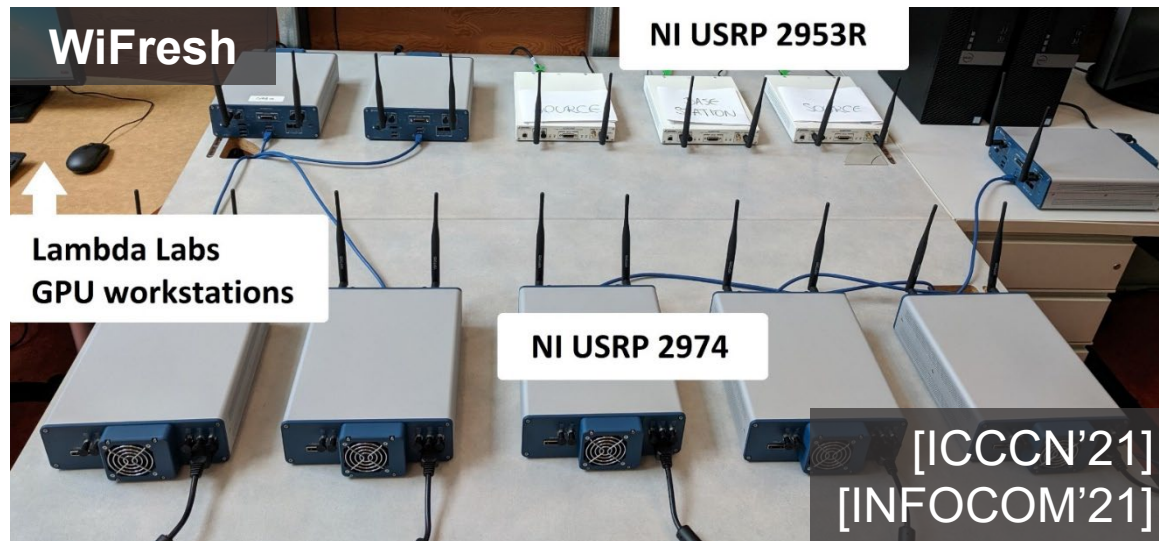
- 1) **AP selects** a single source  $i \in \{1, 2, \dots, N\}$   
[ $u_i(t) = 1$  and  $u_j(t) = 0, \forall j \neq i$ ]
- 2) **Selected source** generates and transmits a packet to the AP
- 3) Packet is **delivered** to the AP w.p.  $p_i$   
[ $d_i(t) = 1$  and  $d_j(t) = 0, \forall j \neq i$ ] and  
transmission error occurs w.p.  $1 - p_i$



Class of non-anticipative policies  $\Pi$ . Arbitrary policy  $\pi \in \Pi$ .

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**Goal:** find a **transmission scheduling policy**  $\pi^*$  that selects, in each slot  $t$ , the optimal source to poll in order to minimize the EWSAol.

**Dynamic Programming** to compute  
the Aol-optimal policy

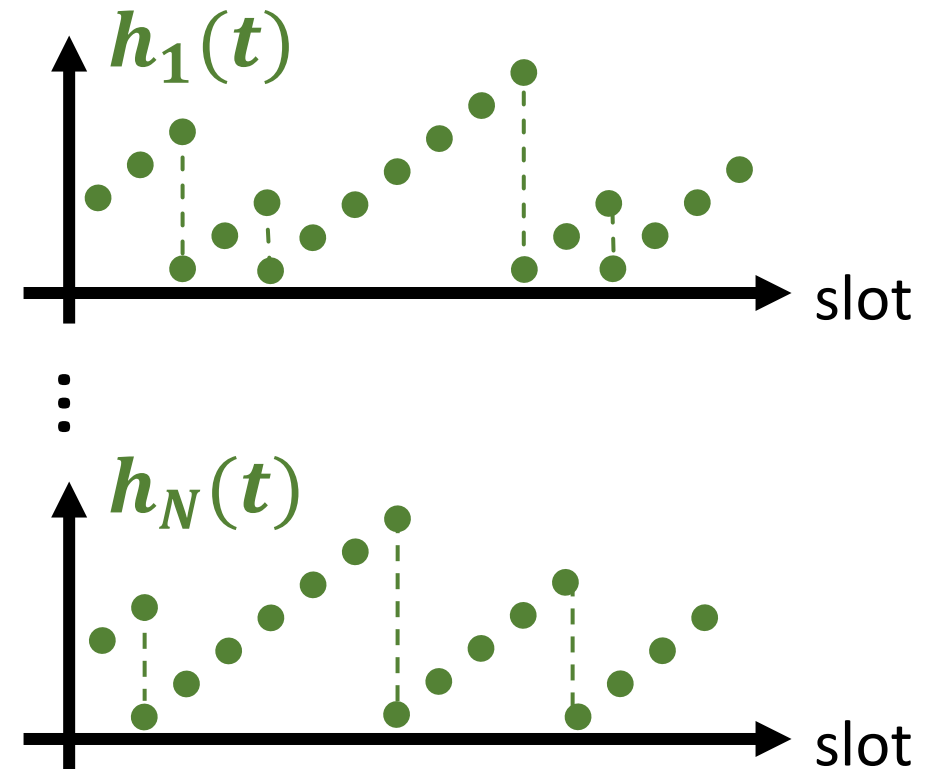


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**Greedy Policy** is Aol-optimal for networks with  $p_i = p$  and  $w_i = w$

Proof: Stochastic Coupling to compare Greedy with any other admissible policy



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**Lower Bound:** for any given network with  $(N, w_i, p_i)$ :  $\lim_{T \rightarrow \infty} \mathbb{E}[J_T^\pi] \geq L_B$

where :

$$L_B = \frac{1}{2N} \left( \sum_{i=1}^N \sqrt{\frac{w_i}{p_i}} \right)^2 + \frac{1}{N} \sum_{i=1}^N w_i$$

Proof: Sample path + Fatou's Lemma + Generalized Mean Inequality

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**Lower Bound:** for any given network with  $(N, w_i, p_i)$ :  $\lim_{T \rightarrow \infty} \mathbb{E}[J_T^\pi] \geq L_B$

**Then:** low-complexity policy  $\pi \in \Pi$  with performance guarantee.

**Performance Guarantee for  $\pi$ :** for any given network with  $(N, w_i, p_i)$

$$L_B \leq \mathbb{E}[J^*] \leq \mathbb{E}[J^\pi] \leq \beta^\pi L_B$$

Means that  $\pi$  is  $\beta^\pi$ -optimal

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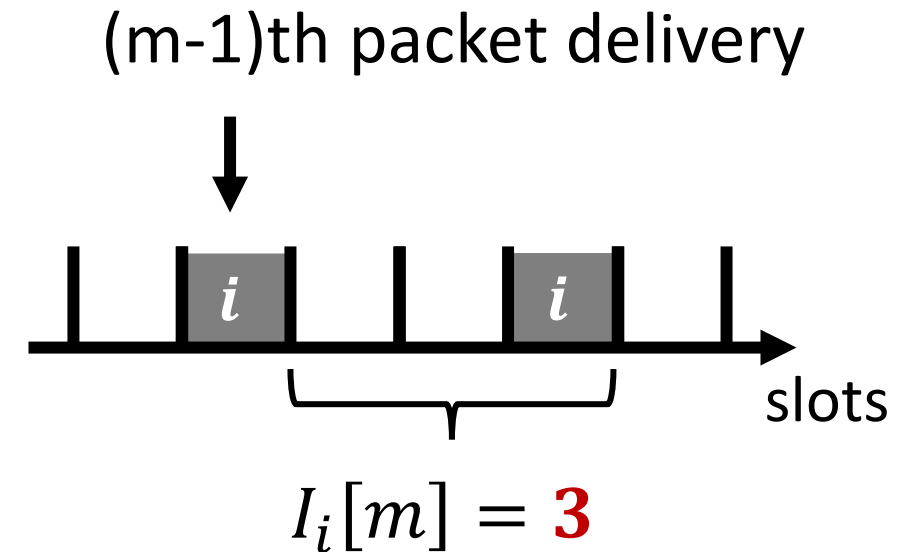
# Aol from a different perspective

**Lemma:**

$$\lim_{T \rightarrow \infty} J_T^\pi \triangleq \lim_{T \rightarrow \infty} \frac{1}{TN} \sum_{t=1}^T \sum_{i=1}^N \mathbf{w}_i \mathbf{h}_i(t) = \frac{1}{N} \sum_{i=1}^N \frac{\mathbf{w}_i}{2} \left[ \frac{\overline{\mathbb{M}}[I_i^2]}{\overline{\mathbb{M}}[I_i]} + 1 \right], \text{wp1}$$

where  $I_i[m]$  is the inter-delivery time of node  $i$  and  $\overline{\mathbb{M}}[I_i]$  is the **sample mean** of  $I_i[m]$ .

$$\lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T \mathbf{h}_i(t) = \frac{1}{2} \left[ \frac{\overline{\mathbb{M}}[I_i^2]}{\overline{\mathbb{M}}[I_i]} + 1 \right], \text{wp1}$$

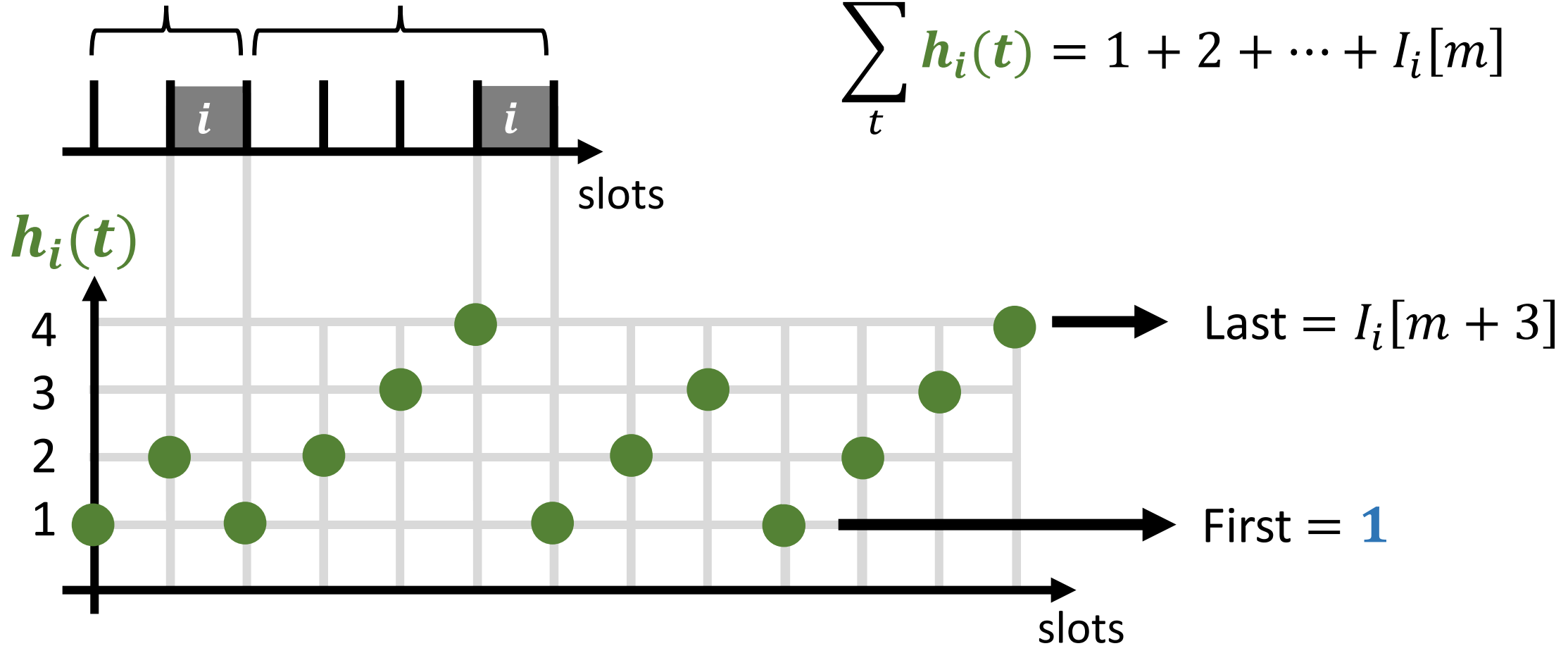


# Intuition of the proof

$$I_i[m] = \mathbf{2} \quad I_i[m+1] = \mathbf{4}$$

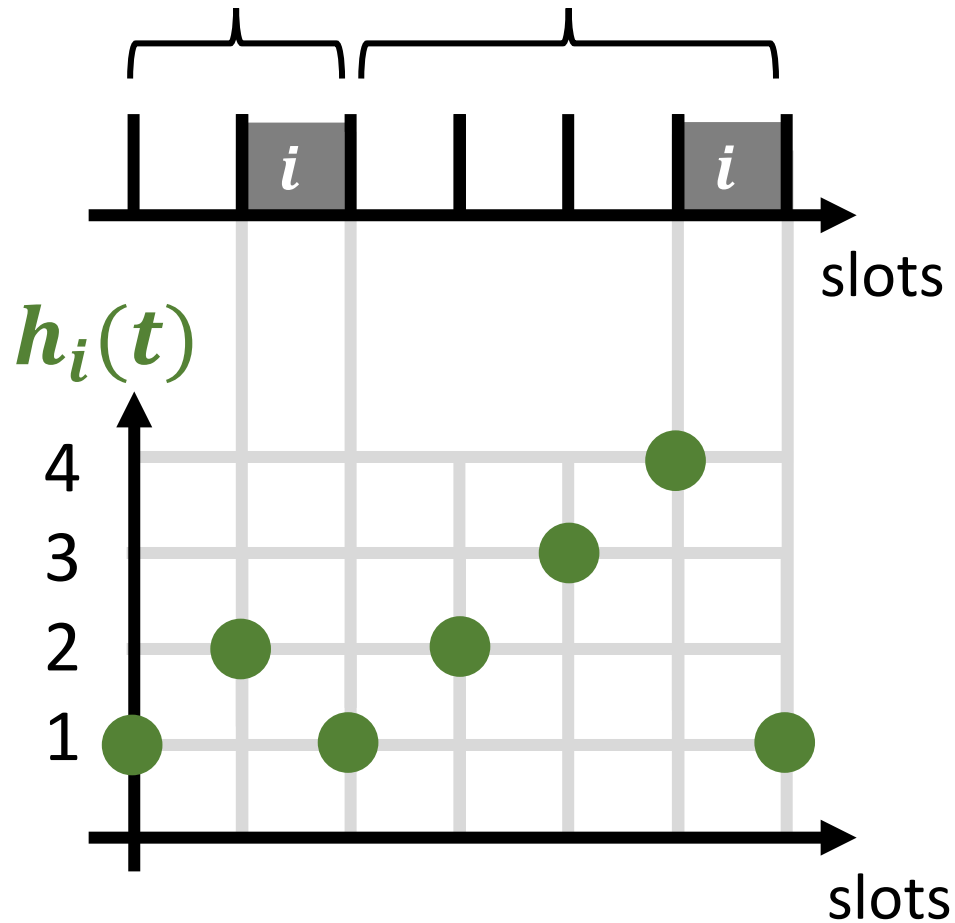
Between any two consecutive deliveries:

$$\sum_t h_i(t) = 1 + 2 + \dots + I_i[m]$$



# Intuition of the proof

$$I_i[m] = \mathbf{2} \quad I_i[m+1] = \mathbf{4}$$



Between any two consecutive deliveries:

$$\sum_t \mathbf{h_i(t)} = \frac{I_i[m]^2 + I_i[m]}{2}$$

Hence, the **time-average Aol** is:

$$\frac{1}{T} \sum_{t=1}^T \mathbf{h_i(t)} \approx \frac{1}{\overline{\mathbb{M}}[I_i]} \left[ \frac{\overline{\mathbb{M}}[I_i^2] + \overline{\mathbb{M}}[I_i]}{2} \right]$$

$$\frac{1}{T} \sum_{t=1}^T \mathbf{h_i(t)} \approx \frac{1}{2} \left[ \frac{\overline{\mathbb{M}}[I_i^2]}{\overline{\mathbb{M}}[I_i]} + 1 \right]$$



# Lower Bound

**Lemma:**

$$\lim_{T \rightarrow \infty} J_T^\pi \triangleq \lim_{T \rightarrow \infty} \frac{1}{TN} \sum_{t=1}^T \sum_{i=1}^N \mathbf{w}_i \mathbf{h}_i(t) = \frac{1}{N} \sum_{i=1}^N \frac{\mathbf{w}_i}{2} \left[ \frac{\bar{\mathbb{M}}[I_i^2]}{\bar{\mathbb{M}}[I_i]} + 1 \right], \text{wp1}$$

**Theorem:**

$$\lim_{T \rightarrow \infty} \mathbb{E}[J_T^\pi] \geq \mathbf{L}_B, \text{ where } \mathbf{L}_B = \frac{1}{2N} \min_{\pi \in \Pi} \left\{ \sum_{i=1}^N \mathbf{w}_i \mathbb{E}[\bar{\mathbb{M}}[I_i^\pi]] \right\} + \frac{1}{2N} \sum_{i=1}^N \mathbf{w}_i$$

Proof outline: applying Fatou's Lemma to the non-negative sequence  $J_T^\pi$  and then the generalized mean inequality  $\bar{\mathbb{M}}[I_i^2] \geq (\bar{\mathbb{M}}[I_i])^2$ .

# Lower Bound

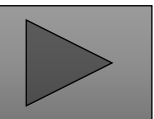
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**Theorem:**

$$\lim_{T \rightarrow \infty} \mathbb{E}[J_T^\pi] \geq \mathbf{L}_B, \text{ where } \mathbf{L}_B = \frac{1}{2N} \left( \sum_{i=1}^N \sqrt{\frac{\mathbf{w}_i}{\mathbf{p}_i}} \right)^2 + \frac{1}{2N} \sum_{i=1}^N \mathbf{w}_i$$

Proof outline: applying Fatou's Lemma to the non-negative sequence  $J_T^\pi$  and then the generalized mean inequality  $\overline{\mathbf{M}}[I_i^2] \geq (\overline{\mathbf{M}}[I_i])^2$ .



# Performance Guarantees

- **Performance Guarantee for  $\pi$ :** for any given network with  $(N, \mathbf{w}_i, \mathbf{p}_i)$

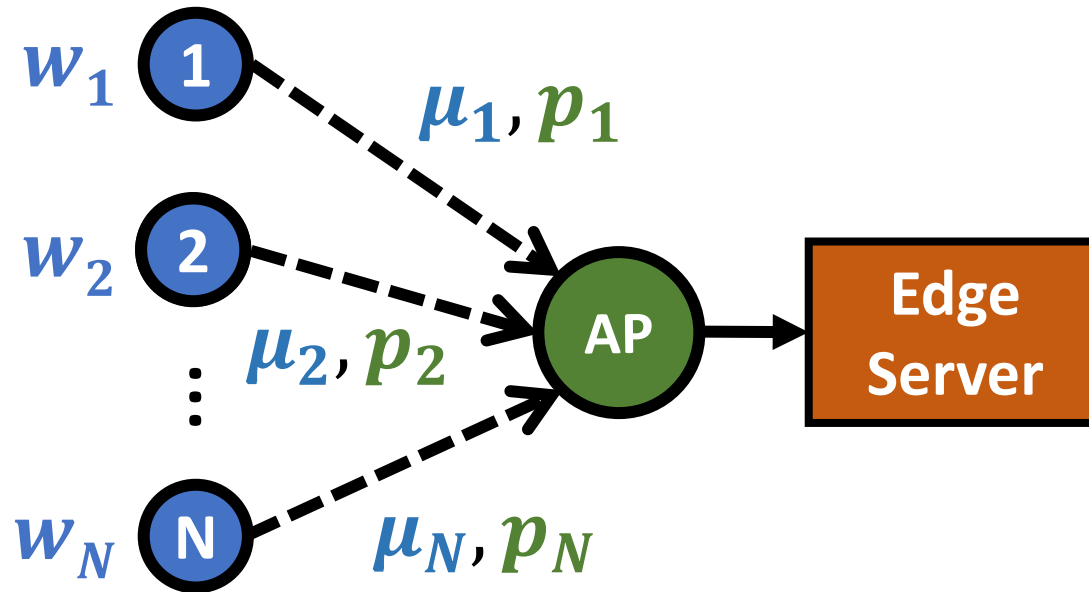
$$\mathbf{L}_B \leq \mathbb{E}[J^*] \leq \mathbb{E}[J^\pi] \leq \beta^\pi \mathbf{L}_B$$

- Three **scheduling policies** with performance guarantees:

Scheduling Policy	Proof Technique	Optimality Ratio
Optimal Stationary Randomized Policy	Renewal Theory	2-optimal
Age-Based Max-Weight Policy	Lyapunov Optimization	2-optimal
Whittle's Index Policy	RMAB Framework	8-optimal

# Stationary Randomized Policies

- **Randomized Policy  $\mathbf{R}$ :** in slot  $t$ , select source  $i$  with probability  $\mu_i$ .
- $d_i^{\mathbf{R}}(t) \sim \text{Ber}(\mu_i p_i)$  and  $I_i^{\mathbf{R}}[m] \sim \text{Geo}(\mu_i p_i)$  iid for node  $i$

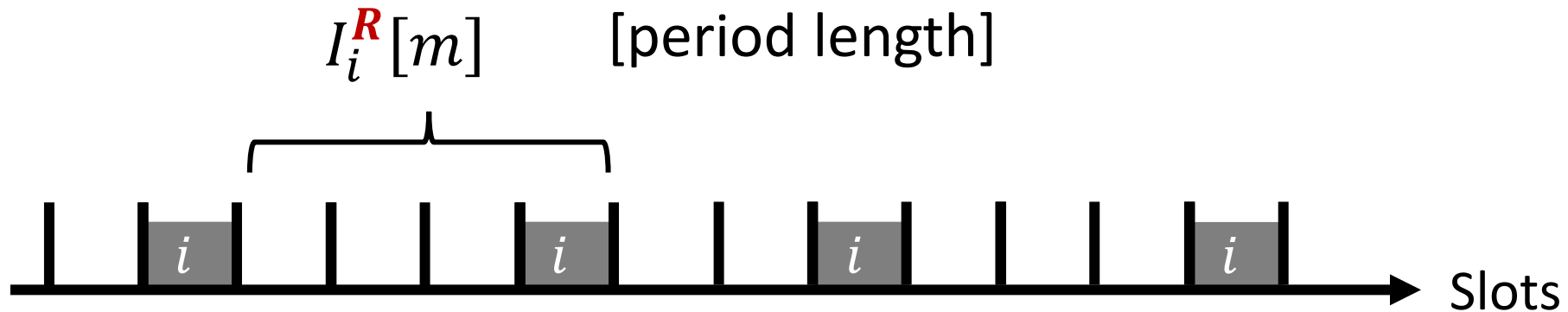


I. K., A. Sinha, and E. Modiano, "Scheduling Algorithms for Optimizing Age of Information in Wireless Networks with Throughput Constraints," IEEE/ACM ToN, 2019. [conference version received the best paper award at IEEE INFOCOM 2018]

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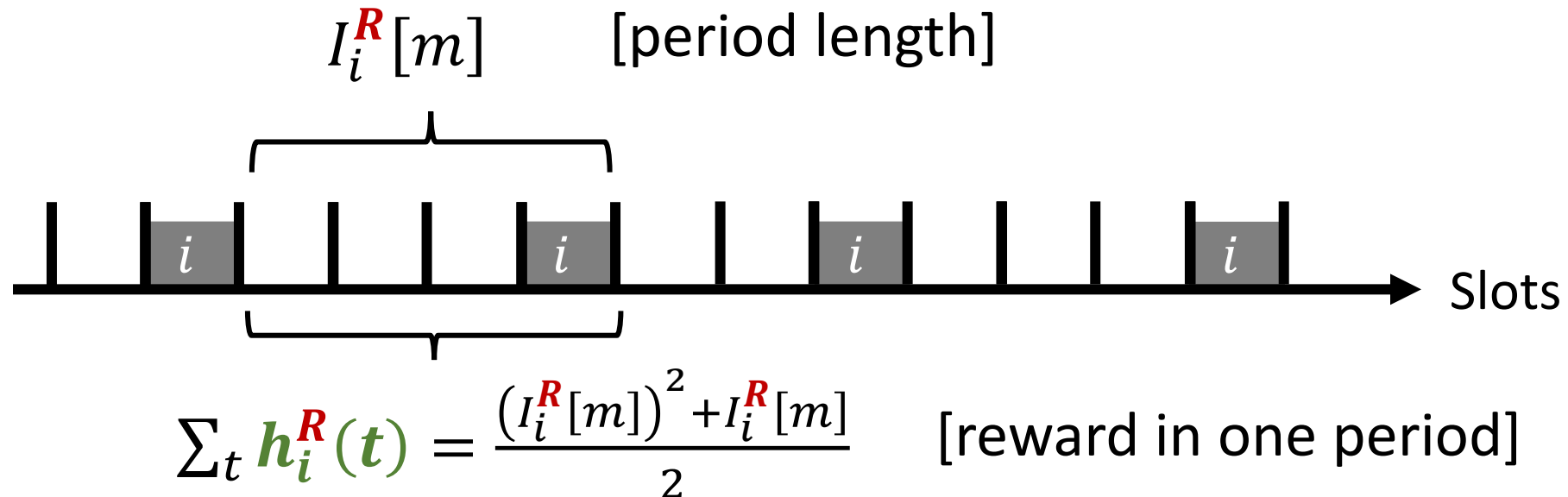
# Stationary Randomized Policies

- **Randomized Policy  $R$ :** in slot  $t$ , select source  $i$  with probability  $\mu_i$ .
- $d_i^R(t) \sim \text{Ber}(\mu_i p_i)$  and  $I_i^R[m] \sim \text{Geo}(\mu_i p_i)$  iid for node  $i$
- Sequence of packet deliveries from node  $i$  is a **renewal process**



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- Sum of  $h_i^R(t)$  is a **renewal-reward process**:



# Stationary Randomized Policies

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- Sequence of packet deliveries from node  $i$  is a **renewal process**
- Sum of  $h_i^{\mathbf{R}}(t)$  is a **renewal-reward process**:
  - Period length  $\mathbb{E}[I_i^{\mathbf{R}}] = (\mu_i p_i)^{-1}$  and reward is  $\mathbb{E} \left[ (I_i^{\mathbf{R}})^2 + I_i^{\mathbf{R}} \right] / 2 = (\mu_i p_i)^{-2}$

Hence, by the elementary renewal theorem:

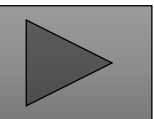
$$\lim_{T \rightarrow \infty} \mathbb{E}[J_T^{\mathbf{R}}] = \frac{1}{N} \sum_{i=1}^N w_i \frac{\mathbb{E}[\text{reward}_i]}{\mathbb{E}[\text{period}_i]} = \frac{1}{N} \sum_{i=1}^N \frac{w_i}{\mu_i p_i}$$

# Stationary Randomized Policies

- **Randomized Policy  $R$ :** in slot  $t$ , select source  $i$  with probability  $\mu_i$ .
- **Optimal Randomized policy  $R^*$ :**  $(\mu_i^*)_{i=1}^N = \operatorname{argmin}_{\mu_i, \forall i} \left\{ \frac{1}{N} \sum_{i=1}^N \frac{w_i}{\mu_i p_i} \right\}$

The optimal solution is  $\mu_i^* = \frac{\sqrt{w_i/p_i}}{\sum_{j=1}^N \sqrt{w_j/p_j}} \propto \sqrt{w_i/p_i}$

- **EWSAol:**  $\lim_{T \rightarrow \infty} \mathbb{E}[J_T^{R^*}] = \frac{1}{N} \sum_{i=1}^N \frac{w_i}{\mu_i^* p_i} = \frac{1}{N} \left( \sum_{i=1}^N \sqrt{\frac{w_i}{p_i}} \right)^2$





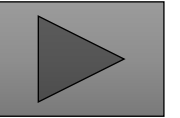
# Stationary Randomized Policies

**Thm:** for any given network configuration, **optimal randomized  $R^*$  is 2-optimal**

- **EWSAol:**  $\lim_{T \rightarrow \infty} \mathbb{E}[J_T^{R^*}] = \frac{1}{N} \sum_{i=1}^N \frac{w_i}{\mu_i^* p_i} = \frac{1}{N} \left( \sum_{i=1}^N \sqrt{\frac{w_i}{p_i}} \right)^2$
- **Lower Bound:**  $\lim_{T \rightarrow \infty} \mathbb{E}[J_T^\pi] \geq L_B$ , where  $L_B = \frac{1}{2N} \left( \sum_{i=1}^N \sqrt{\frac{w_i}{p_i}} \right)^2 + \frac{1}{N} \sum_{i=1}^N w_i$
- **Performance guarantee:**  $L_B \leq \lim_{T \rightarrow \infty} \mathbb{E}[J_T^{R^*}] \leq 2L_B$  [2-optimal]

# Age-Based Max-Weight policy

- Lyapunov Function:  $L(t) = \frac{1}{N} \sum_{i=1}^N \gamma_i h_i(t)$ , where  $\gamma_i > 0$  is a constant
- Lyapunov Drift:  $\Delta(t) = \mathbb{E}\{ L(t+1) - L(t) \mid h_i(t) \}$



# Age-Based Max-Weight policy

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- Lyapunov Drift:  $\Delta(t) = \mathbb{E}\{ L(t+1) - L(t) \mid h_i(t) \}$
- Substituting  $h_i(t+1) = h_i(t)[1 - d_i(t)] + 1$  into the drift

$$\Delta(t) = -\frac{1}{N} \sum_{i=1}^N \gamma_i h_i(t) p_i \mathbb{E}[u_i(t) \mid h_i(t)] + \frac{1}{N} \sum_{i=1}^N \gamma_i$$

- **MW policy:** in slot  $t$ , schedule node ( $u_i(t) = 1$ ) with highest  $\gamma_i h_i(t) p_i$

# Age-Based Max-Weight policy

**Thm:** for any given network configuration, MW with  $\gamma_i = w_i / \mu_i^* p_i$  is **2-optimal**

**MW policy:** in slot  $t$ , select node with highest  $w_i p_i h_i^2(t)$

# Age-Based Max-Weight policy

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**MW policy:** in slot  $t$ , select node with highest  $\mathbf{w}_i \mathbf{p}_i h_i^2(t)$

- Proof outline:

MW minimizes drift while optimal randomized  $\mathbf{R}^*$  does not. Thus:

$$\Delta^{\mathbf{MW}}(t) \leq \Delta^{\mathbf{R}^*}(t)$$

Manipulating the expression and substituting  $\gamma_i = \mathbf{w}_i / \mu_i^* \mathbf{p}_i$ , gives:

$$\mathbf{L}_B \leq \lim_{T \rightarrow \infty} \mathbb{E}[J_T^{\mathbf{MW}}] \leq \lim_{T \rightarrow \infty} \mathbb{E}[J_T^{\mathbf{R}^*}] \leq 2\mathbf{L}_B$$

Hence, **Age-Based Max-Weight is 2-optimal.**

# Summary of Results

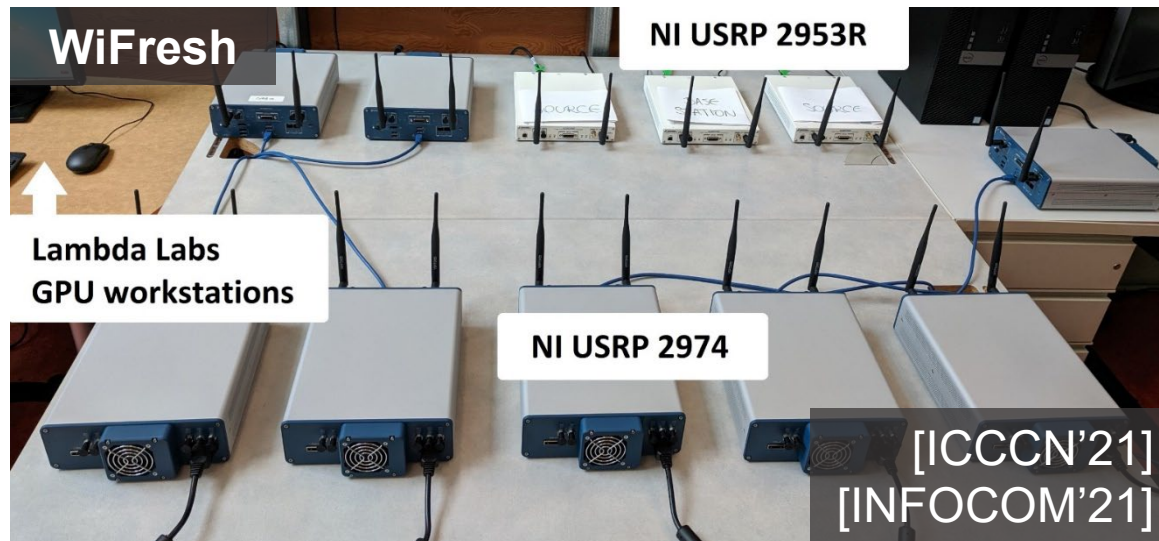
- **Performance Guarantee for  $\pi$ :** for any given network with  $(N, \mathbf{w}_i, \mathbf{p}_i)$

$$L_B \leq \mathbb{E}[J^*] \leq \mathbb{E}[J^\pi] \leq \beta^\pi L_B$$

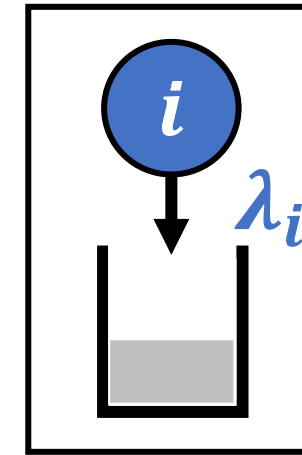
Scheduling Policy	Decision in slot $t$	Performance Guarantee	Simulation Result
Greedy	highest $h_i(t)$	optimal when symmetric	optimal when symmetric
Randomized	node $i$ w.p. $\propto \sqrt{\mathbf{w}_i/\mathbf{p}_i}$	2-optimal	$\sim$ 2-optimal
Age-Based Max-Weight	highest $\mathbf{w}_i \mathbf{p}_i h_i^2(t)$	2-optimal	close to optimal

# Outline

- **Theory:**
  - Introduction to Age-of-Information using a Simple System
  - Formulation of the AoI Minimization Problem in a Wireless Network
  - Theoretical Results: Lower Bound and Scheduling Policies
- **System Implementation in a SDR Testbed and Flight Tests with Drones**



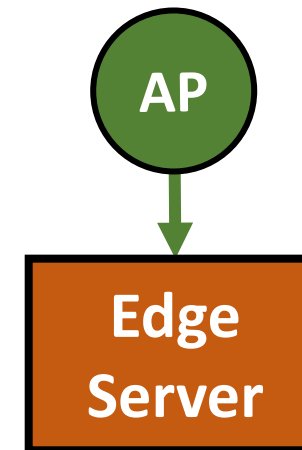
# WiFi: Network of Sensing-Drones



Generates  
time-sensitive  
information

**FCFS queue**

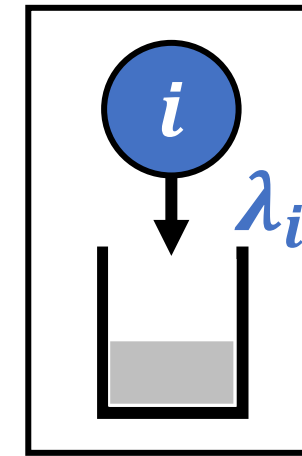
Wireless  
Access Point  
uses **Random  
Access**



Keeps track  
of the  
information



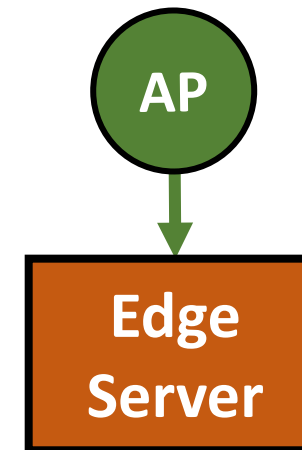
# WiSwarm: Network of Sensing-Drones



Generates  
time-sensitive  
information

**LCFS queue**

Wireless  
Access Point  
**schedules  
transmissions**



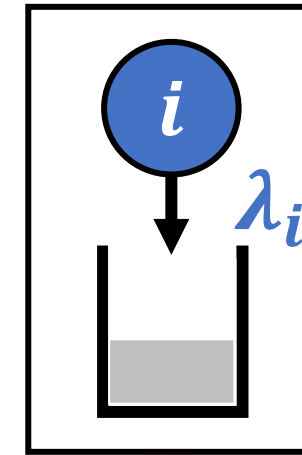
Keeps track  
of the  
information

# WiSwarm: Network of Sensing-Drones

AP runs **Max-Weight** policy,  
sends **poll** packet, and  
broadcasts **trajectory updates**



**MW policy** selects, in each decision time  $t$ ,  
the source  $i^*(t)$  with highest value of  
$$\mathcal{I}(i, t) = w_i p_i h_i^2(t)$$



Generates  
time-sensitive  
information

**LCFS queue**



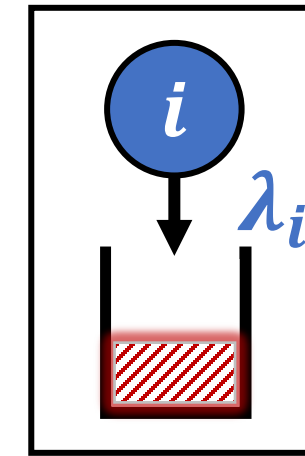
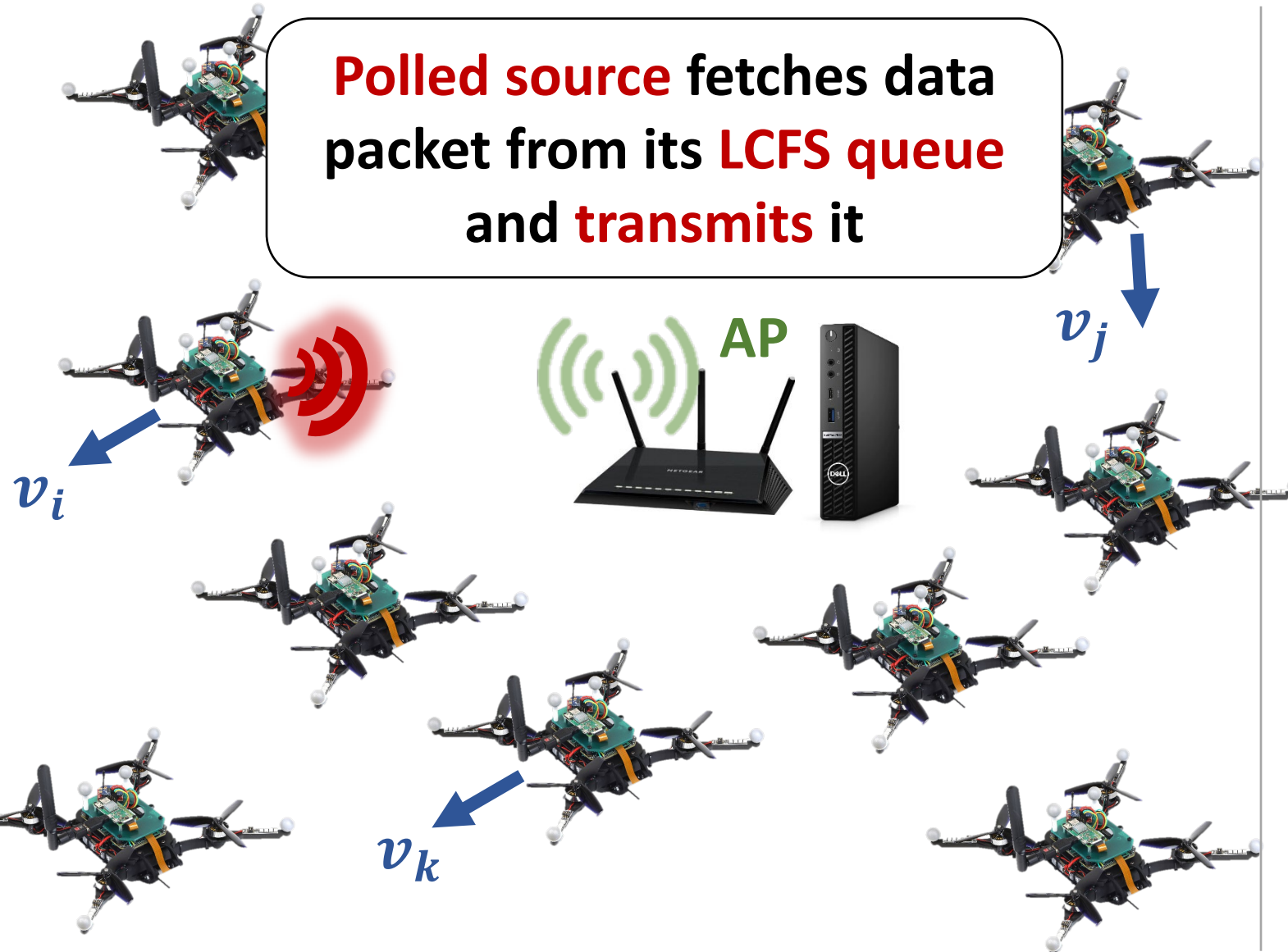
Wireless  
Access Point  
**schedules**  
**transmissions**

Edge  
Server

Keeps track  
of the  
information

# WiSwarm: Network of Sensing-Drones

**Polled source** fetches data packet from its **LCFS queue** and **transmits it**



Generates time-sensitive information

**LCFS queue**



Wireless Access Point **schedules transmissions**



**Edge Server**

Keeps track of the information

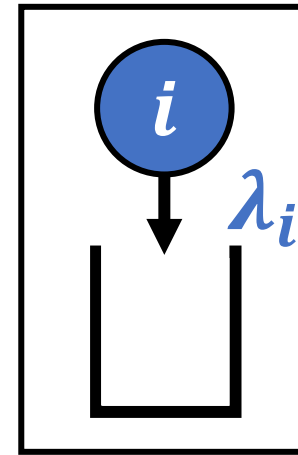
# WiSwarm: Network of Sensing-Drones

AP **updates its belief** about the network state ( $\hat{h}_i(t)$ ,  $\hat{p}_i(t)$ ) and about the **environment**



**MW policy** selects, in each decision time  $t$ , the source  $i^*(t)$  with highest value of

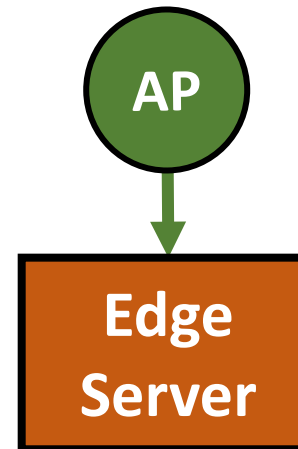
$$J(i, t) = w_i(t) \hat{p}_i(t) \hat{h}_i^2(t)$$



Generates time-sensitive information

**LCFS queue**

Wireless Access Point **schedules transmissions**



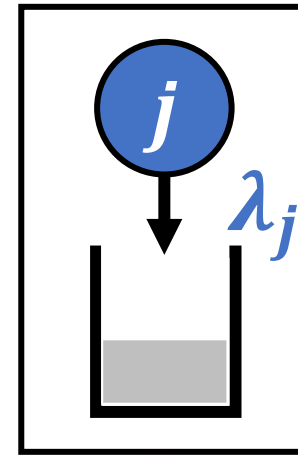
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Generates  
time-sensitive  
information

**LCFS queue**



Wireless  
Access Point  
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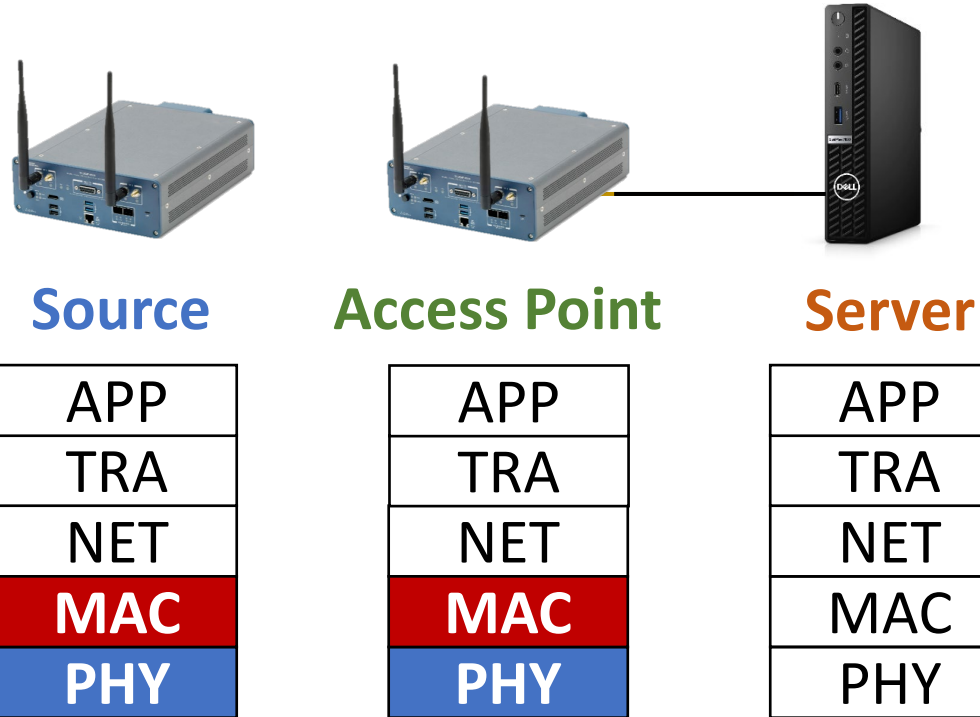
Edge  
Server

Keeps track  
of the  
information

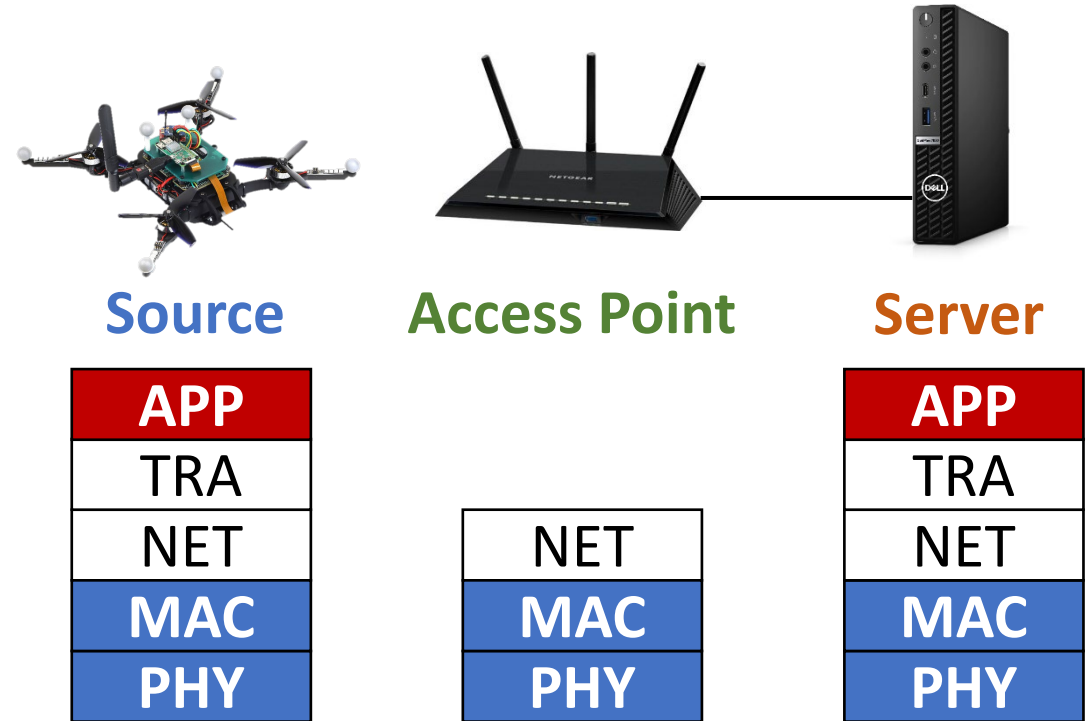


# Implementations (It takes a village...)

WiFresh in a SDR testbed



WiSwarm in a drone testbed



- I. K. and E. Modiano, "Age of Information in Random Access Networks with Stochastic Arrivals," IEEE INFOCOM, 2021.
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WiFresh in a SDR testbed



Average Aol of a network with five sources is **< 10 msec**

WiSwarm in a drone testbed

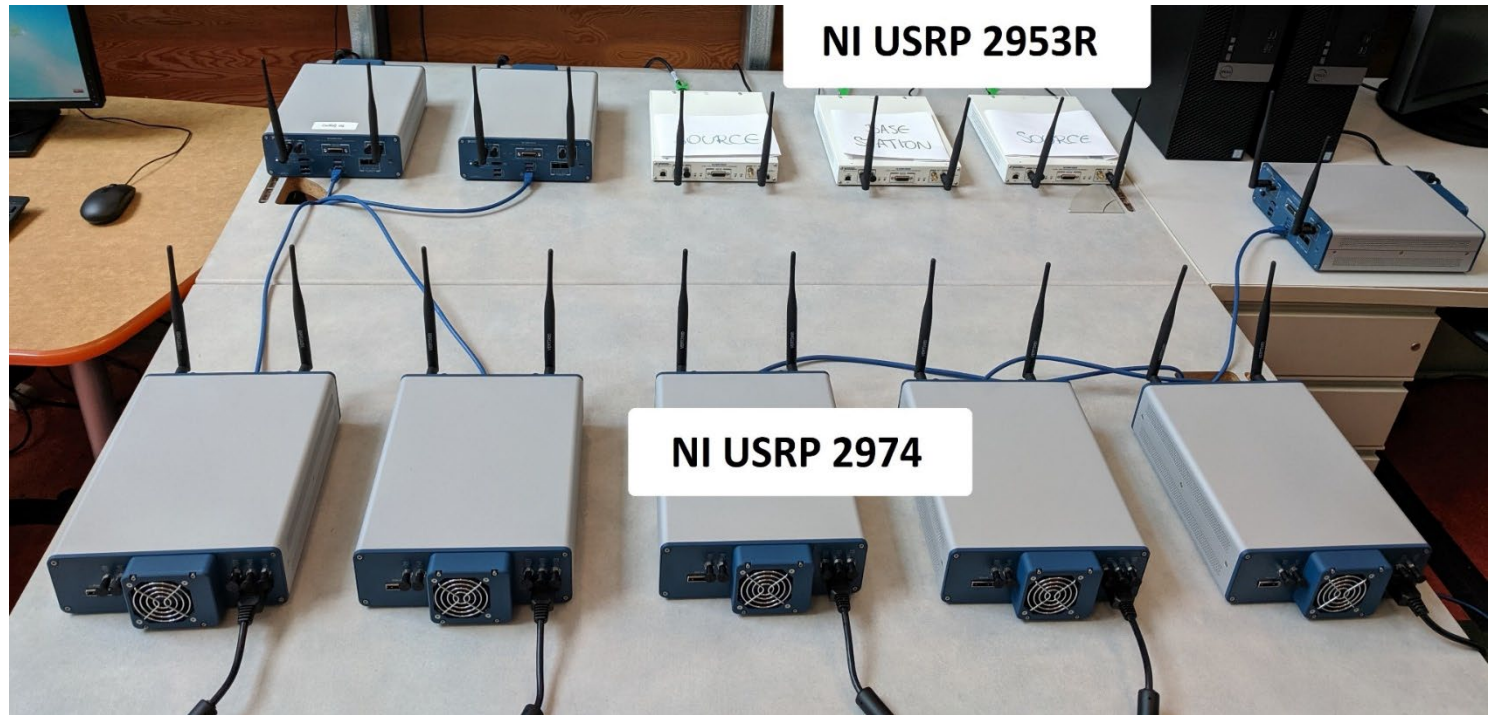


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# Implementation Challenges

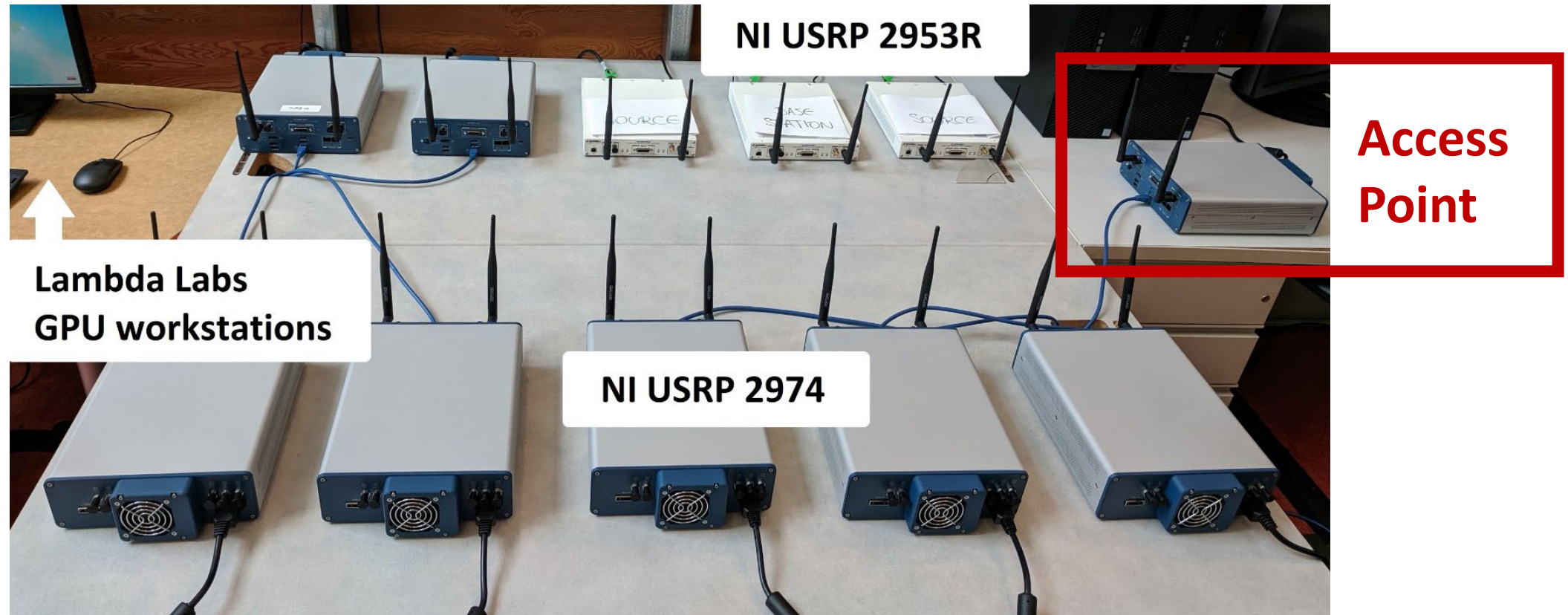
- **Complexity:** Microsecond time-scale scheduling decisions achieved by implementing WiFresh at the MAC layer using FPGAs and hardware-level programming.
- **From Theory to Practice:**
  - Clock Synchronization
  - Estimation of Parameters
  - Packet Fragmentation
- **Barrier to adoption:**
  - Alternative implementation (WiSwarm) that can be easily integrated into applications already run over WiFi





# WiFresh: Measurements

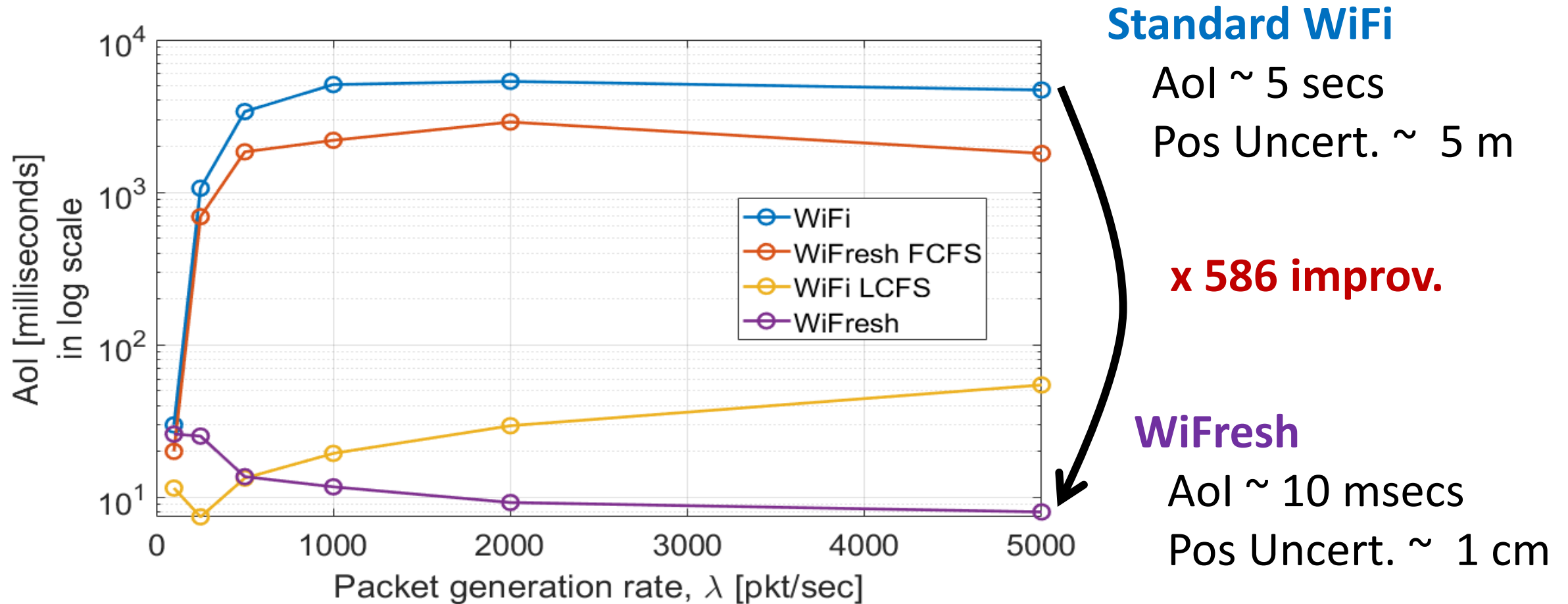
- Ten SDRs generating packets of 150 bytes with frequency  $\lambda$  packets per sec.



# WiFresh: Measurements

- Ten SDRs generating packets of 150 bytes with frequency  $\lambda$  packets per sec.
- We measure the EWSAol of the following communication systems:
  - 1) **WiFresh**;
  - 2) **Standard WiFi**
  - 3) **WiFresh FCFS**; and
  - 4) **WiFi LCFS**.
- Each experiment runs for a total of 10 minutes.

# WiFresh: Measurements



# WiFresh



# Standard WiFi

Overview

Instructions

Demonstrates 802.11 Rx and Tx

1. Cable device depending on operation mode selected.
2. Make sure "RIO Device" names the RIO alias of the device.
3. Start the VI and enable the station.

## Station

Primary Channel Center Frequency

2.437500 GHz

Primary Channel Selector

1

Power Level

0.0 dBm

RF ports are applicable for USRP only

TX RF Port

TX1/RX 0

RX RF Port

RX2/RX 1

Subcarrier Format

20 MHz (IEEE 802.11 ac)

MCS

5

64-QAM (2/3)

AGC

Enable

Manual RX gain

17.5 dB

Applied RX Gain

17.5 dB

valid Device MAC Address

48:4F:4B:75:6D:60

valid Destination MAC Address

48:4F:4B:75:6D:61

## 802.11 Application Framework

See the diagram for more information.

RIO Device

RIO0

Reference Clock

Internal

Station Number

0

Device Ready

Stop

Enable Station

On

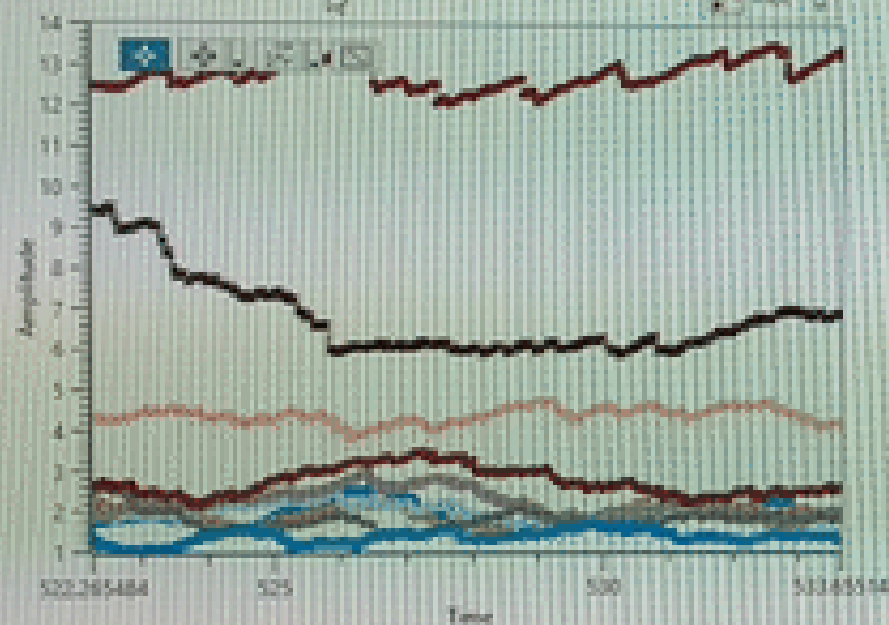
On

Station Active

Target FIFO Overflow

MAC RF & Phy WinFlow Aol Statistics Status Aol Instructions Debug 1 Debug 2 Debug 3 Events

Aol Plot



Number of Full Packets Sent

0 0 0 0 0 0 0 0 0 0

Number of Packets Received

34612 231437 79542 190041 206483 165408 201696 55100 189131 268723

Windowed Estimated Probability

1 1 1 1 1 1 1 1 1 1

Long-term Estimated Probability

1.99213 1.99219 1.99219 1.99219 1.99219 1.99219 1.99219 1.99219 1.99219 1.99219

Time Reference at the FPGA

1042902586

WinFlow Timestamp sent to FPGA

52347740

WinFlow Timestamp from FPGA

52347740

Firstest Received Timestamp

52150788 52164042 52147134 52164344 52166550 52163576 52165409 52114013 52157889 52171113

Parameters for Randomized Policy

1 1 1 1 1 1 1 1 1 1

USRP timestamp

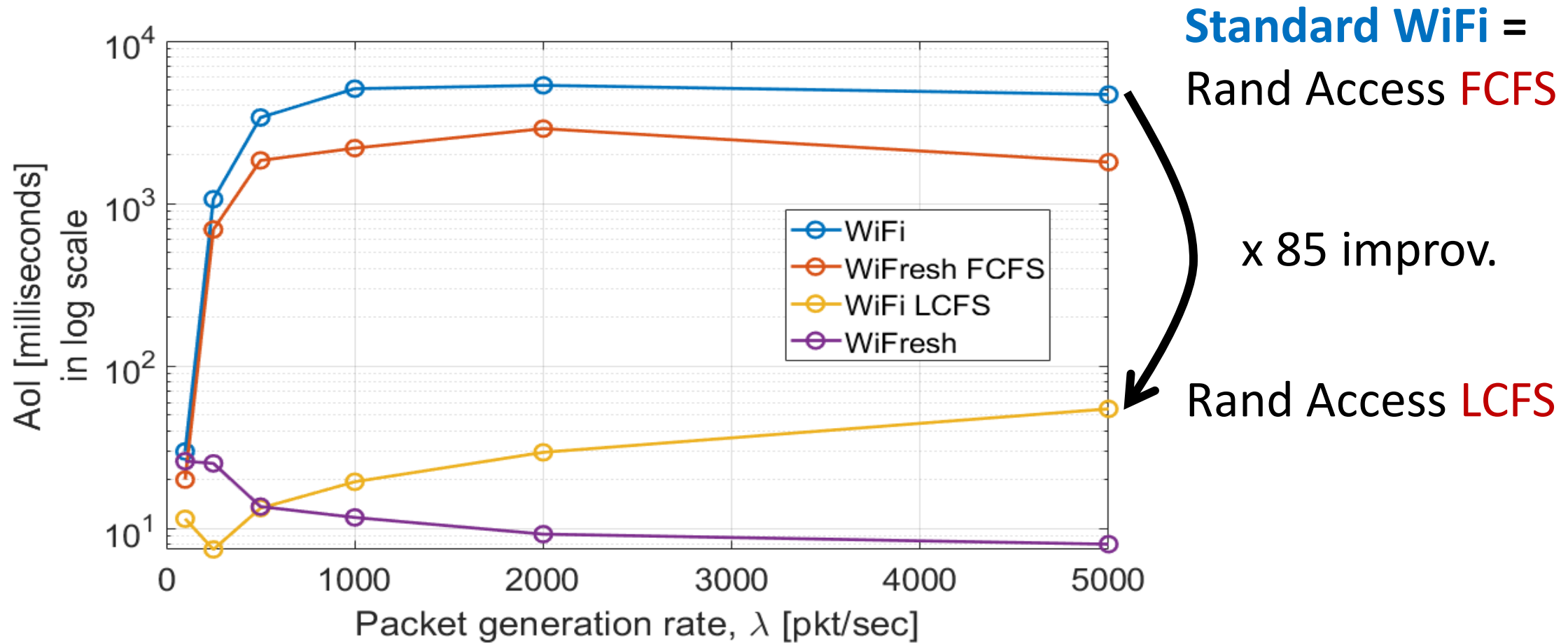
whole seconds

1565141254

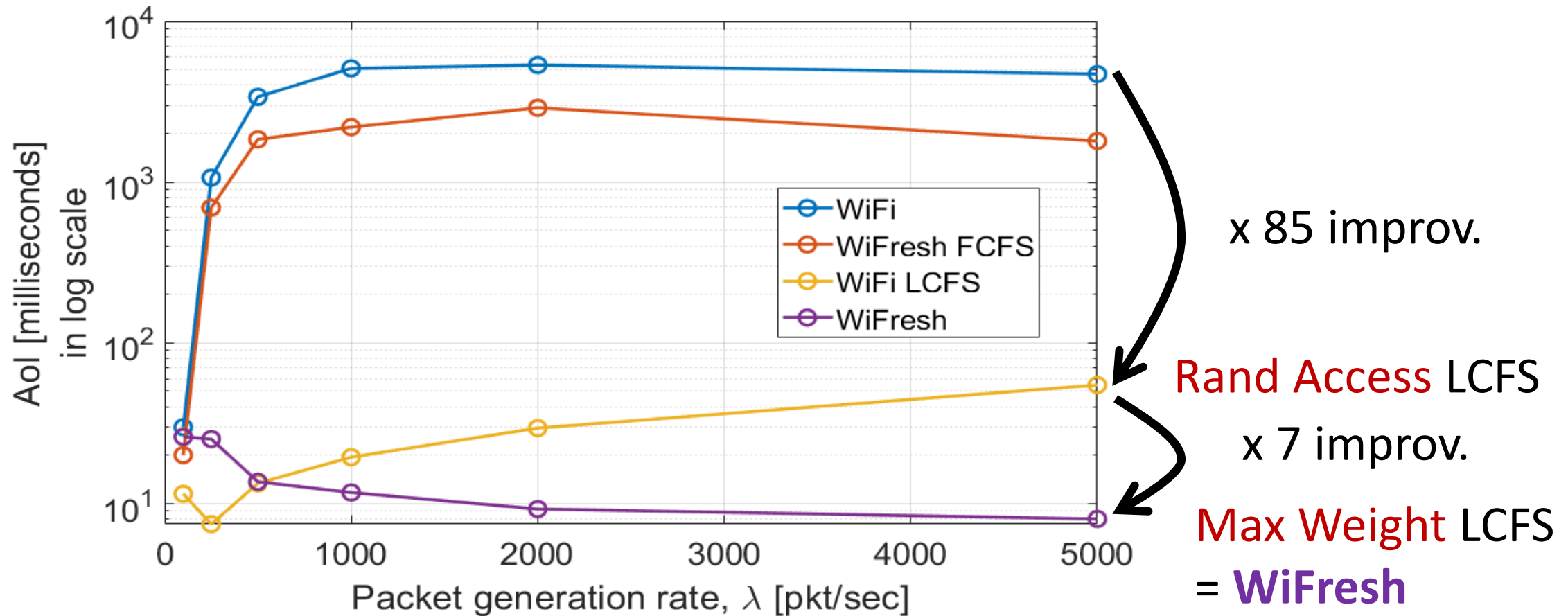
fractional seconds

0.77199

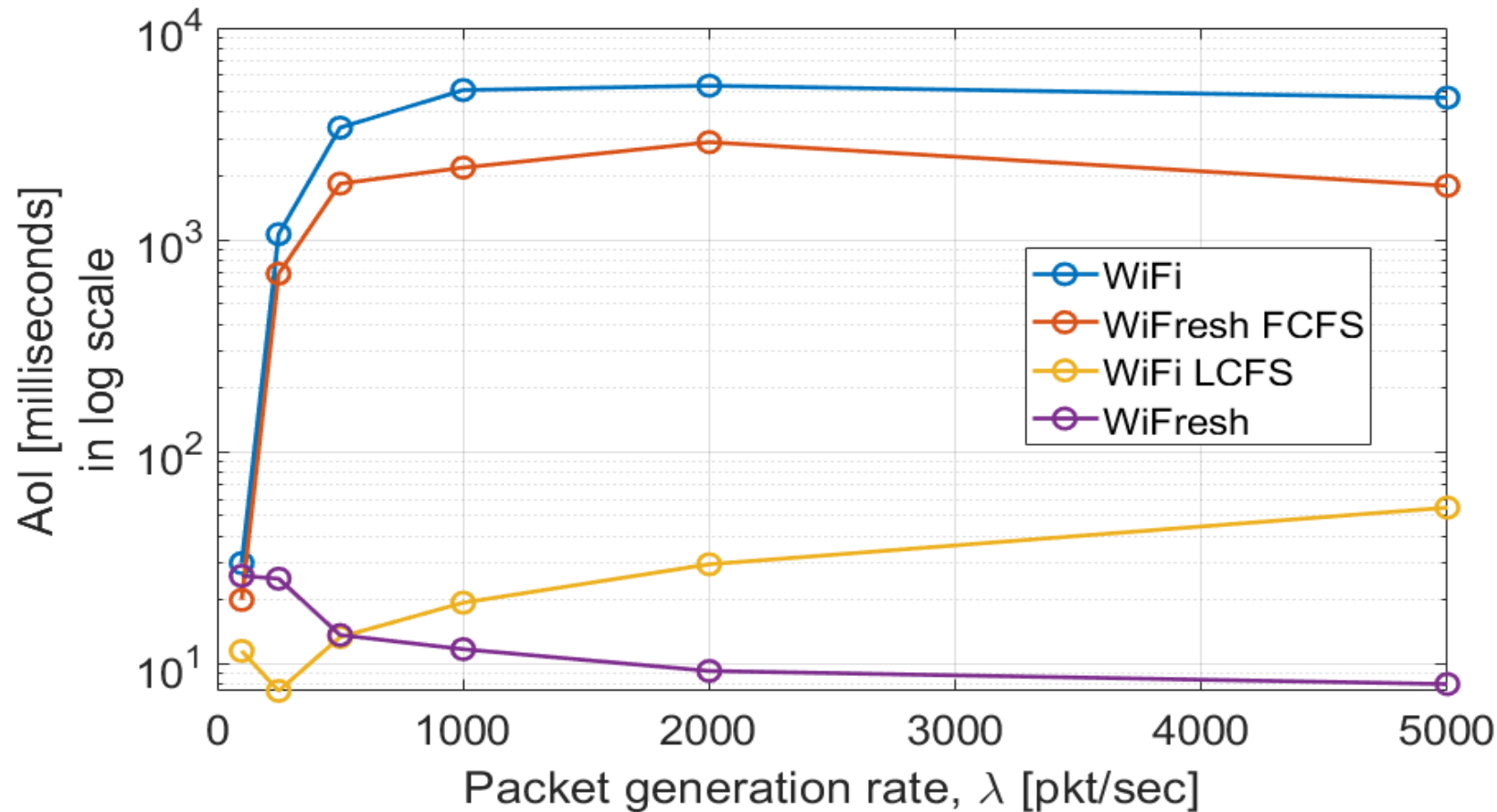
# WiFresh: Measurements



# WiFresh: Measurements



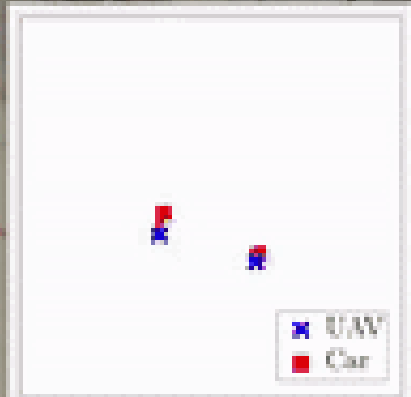
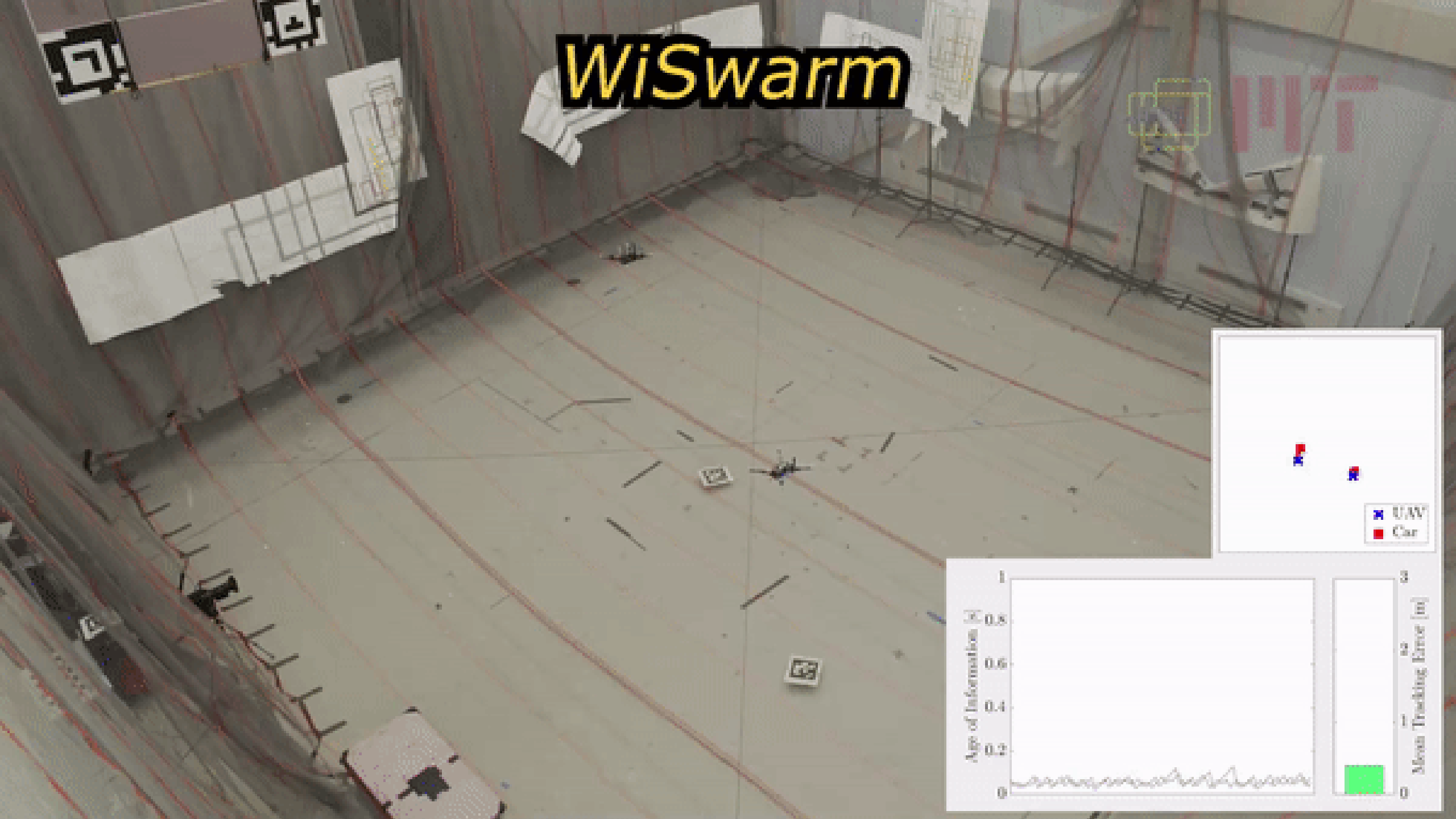
# WiFresh: Measurements



**WiFresh** is the **only** in which a higher  $\lambda$  always leads to fresher info. at the destination

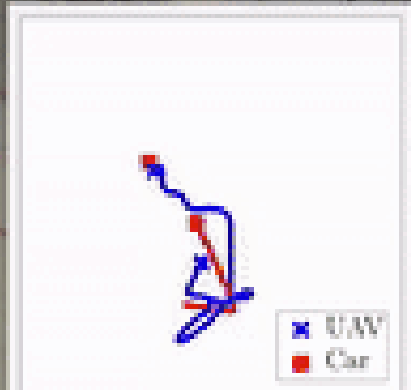


# WiSwarm

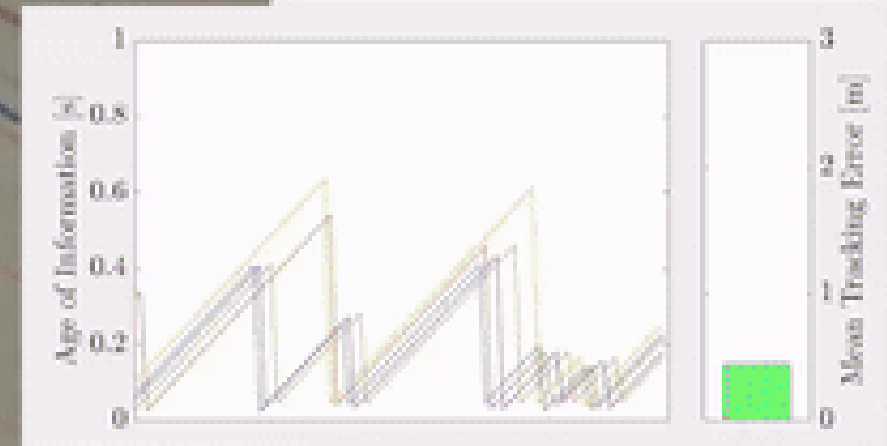


WiFi

Conventional WiFi enables tracking with at most two UAVs.

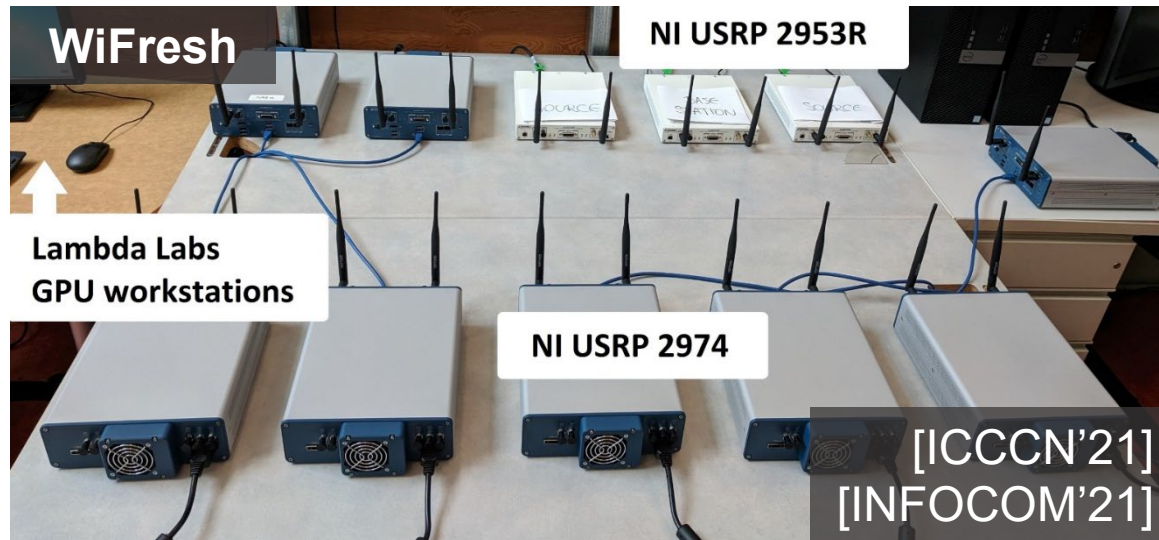


# WiSwarm



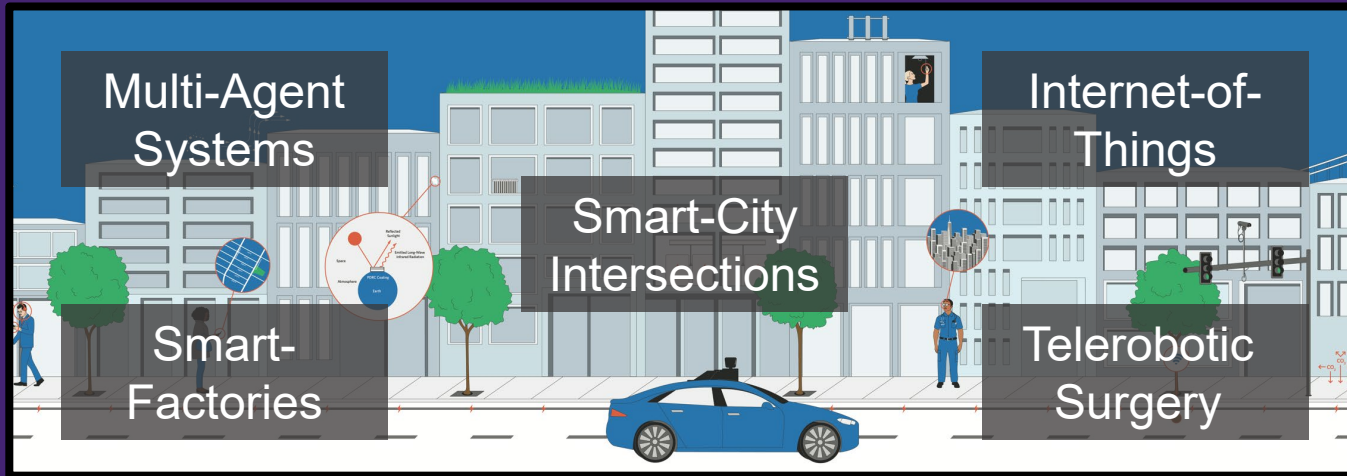
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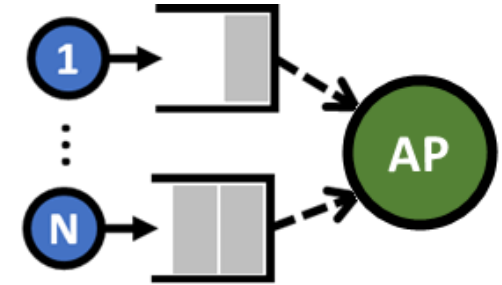


# Research @Northwestern

## Future Applications

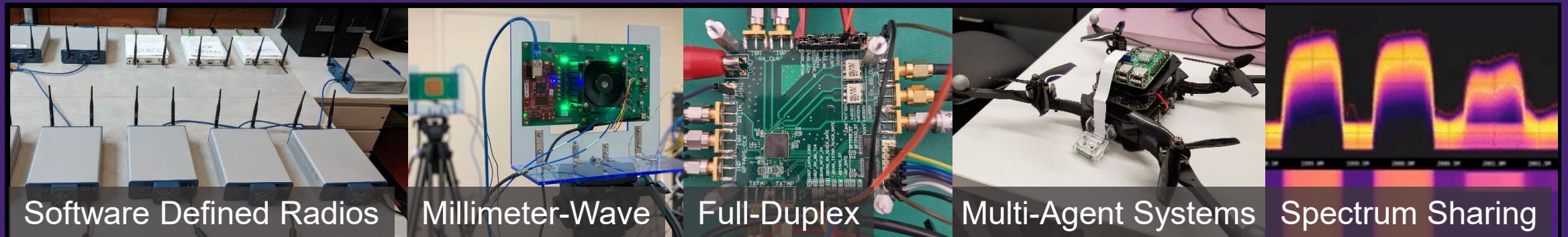


## Rigorous Theory



**Multi-Armed Bandits**  
**Markov Decision Process**  
**Lyapunov Optimization**  
**Dynamic Programming**

## Next-Gen Technology



# Supplemental Slides

# Lower Bound



# Lower Bound

**Lemma:**

$$\lim_{T \rightarrow \infty} J_T^\pi \triangleq \lim_{T \rightarrow \infty} \frac{1}{TN} \sum_{t=1}^T \sum_{i=1}^N \mathbf{w}_i \mathbf{h}_i(t) \geq \frac{1}{2N} \sum_{i=1}^N \mathbf{w}_i \bar{\mathbf{M}}[\mathbf{I}_i] + \frac{1}{2N} \sum_{i=1}^N \mathbf{w}_i, \text{ wp1}$$

Proof outline:

$$\begin{aligned} \sum_{i=1}^N \mathbf{w}_i \bar{\mathbf{M}}[\mathbf{I}_i] &= \lim_{T \rightarrow \infty} \sum_{i=1}^N \mathbf{w}_i \frac{\sum_{m=1}^{D_i(T)} I_i^\pi[m]}{D_i(T)} \approx \lim_{T \rightarrow \infty} T \sum_{i=1}^N \frac{\mathbf{w}_i}{D_i(T)} \geq \\ &\geq \lim_{T \rightarrow \infty} \sum_{i=1}^N \Upsilon_i(T) \sum_{i=1}^N \frac{\mathbf{w}_i}{D_i(T)} \geq \lim_{T \rightarrow \infty} \left( \sum_{i=1}^N \sqrt{\frac{\Upsilon_i(T) \mathbf{w}_i}{D_i(T)}} \right)^2 = \left( \sum_{i=1}^N \sqrt{\frac{\mathbf{w}_i}{p_i}} \right)^2 \end{aligned}$$





# Optimal Randomized Policy

# Stationary Randomized Policies

- For fixed and positive  $w_i$  and  $p_i$ . Consider the optimization problem below:

$$\min_{\mu_i, \forall i} \left\{ \frac{1}{N} \sum_{i=1}^N \frac{w_i}{\mu_i p_i} \right\} \text{ s. t. } \sum_{i=1}^N \mu_i \leq 1$$

- The Cauchy-Schwarz Inequality (below) holds with **equality** when:

$$\left( \sum_{i=1}^N \sqrt{\frac{w_i}{p_i}} \right)^2 \leq \left( \sum_{i=1}^N \mu_i \right) \left( \sum_{i=1}^N \frac{w_i}{\mu_i p_i} \right)$$

$$\mu_i^* = \frac{\sqrt{w_i/p_i}}{\sum_{j=1}^N \sqrt{w_j/p_j}}, \forall i$$

