

# Throughput-Optimal Broadcast in Wireless Networks with Point-to-Multipoint Transmissions

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# Outline

- 1 Introduction
- 2 Hardness of Wireless Broadcasting
- 3 Throughput-Optimal Broadcast Policy
- 4 Numerical Results
- 5 Conclusion

# Motivation

- We consider the problem of optimally broadcasting packets in a multi-hop wireless adhoc network

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- We consider the problem of optimally broadcasting packets in a multi-hop wireless adhoc network
- A primary measure of efficiency is *throughput-optimality*, i.e., policies that achieve the *entire capacity region*
- Vast literature for the **Unicast** problem (**Backpressure** policy), not so much for other flow problems
  - **Packet duplications** are harder to deal with (**no** flow conservations)

# Introduction

- We study the **Generalized Flow Problem** and design **throughput-optimal** policies.
- A **Fundamental** problem with wide ranging applications: Internet routing, in-network function computations, live multi-media streaming, military communications etc.
- Topics of this talk:
  - ① **Broadcast**: Specialized **dynamic** algorithms that solve the throughput-optimal broadcasting problem
    - It admits an inherently **decentralized** solution

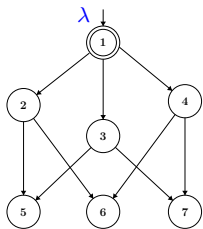
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- Topics of this talk:
  - 1 **Broadcast**: Specialized **dynamic** algorithms that solve the throughput-optimal broadcasting problem
    - It admits an inherently **decentralized** solution
  - 2 **Generalized Flow**: A general algorithmic paradigm that efficiently solves **all flow problems** (unicast+broadcast+multicast+anycast).

# System Model

- The Wireless Network is represented by a graph  $\mathcal{G}(V, E)$ , where each node has an **omnidirectional** antenna.
- Packets arrive at a source node  $r$  i.i.d. at every slot at rate  $\lambda$ .
- Due to the **local broadcast** nature of the wireless medium, packets transmitted by a node  $i$  is heard at all of its out-neighbor  $j \in \partial^+(i)$ .
- As a result, if two or more in-neighbors of a node transmits at a slot, it results in a collision.
- This talk considers **collision-free** schedules only. The set of all feasible collision-free node activations is given by  $\mathcal{M}$ .

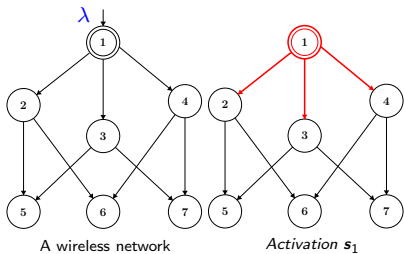
# An Illustrative Example



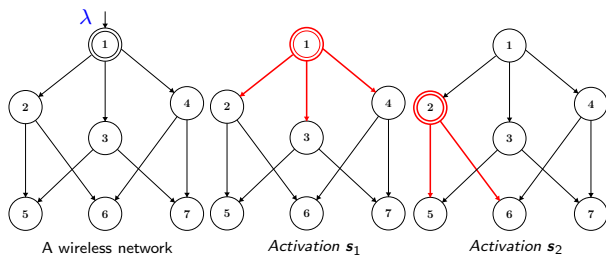
A wireless network



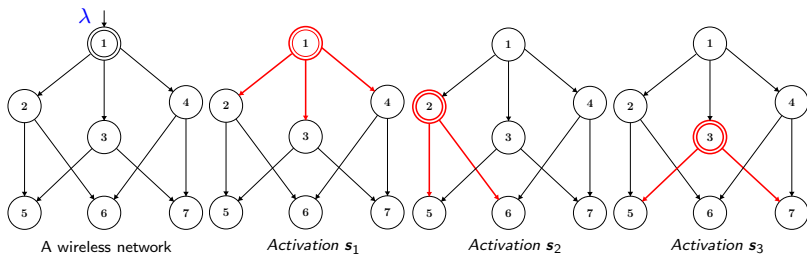
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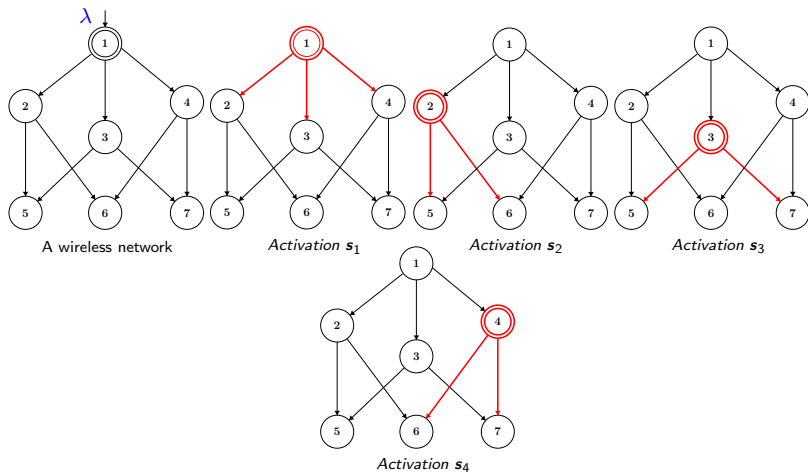
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A wireless network and its feasible link activations under the **primary** interference constraints.

$$\mathcal{M} = \{s_1, s_2, s_3, s_4\}$$

# WIRELESS BROADCAST: Problem Formulation

A feasible broadcast policy  $\pi \in \Pi$  executes the following two actions at every slot  $t$  :

- **Node Activation**  $\pi(\mathcal{A})$ : Activates a subset of nodes  $\mathbf{s}(t) \in \mathcal{M}$  subject to the underlying interference constraints.
- **Packet Scheduling**  $\pi(\mathcal{S})$ : The activated nodes *locally broadcasts* a set of packets subject to the capacity/power constraints of the nodes.

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- Let  $R^\pi(T)$  denote the number of packets received in **common by all nodes** under the action of a broadcast policy  $\pi$ .
- The objective is to design a policy  $\pi$  such that for all  $\lambda < \lambda^*$

$$\liminf_{T \rightarrow \infty} \frac{R^\pi(T)}{T} = \lambda, \quad \text{w.p. 1,}$$

where  $\lambda^*$  is the broadcast capacity of the network.

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Our first result in this **point-to-multipoint** broadcast setting is the following:

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- This is surprising, because we showed earlier [[Sinha et al, 2015, 2016](#)] that the problem is efficiently solvable in case of wireless DAGs with [point-to-point](#) links.
- The hardness comes from the requirement of optimally distributing the packets, which is intimately related to [Boolean Constraint Satisfaction](#), described next.

## Hardness Reduction: Proof Sketch

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**Problem**: Does there exist a satisfying assignment such that each clause contain at least one false literal? (Y/N)

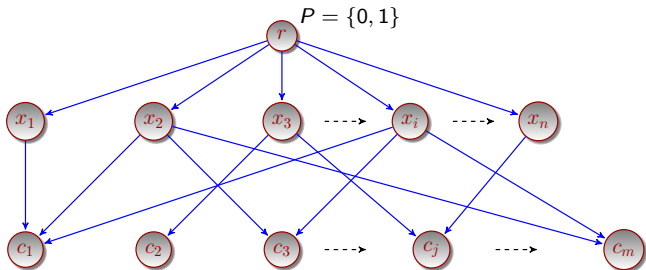
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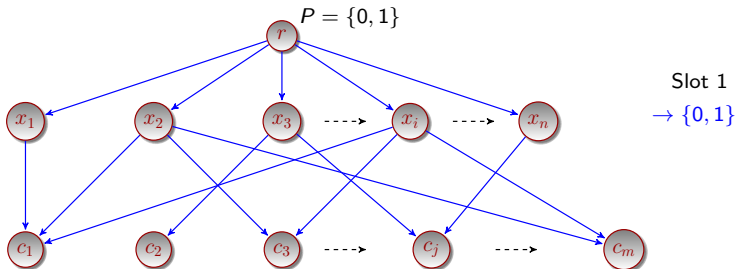
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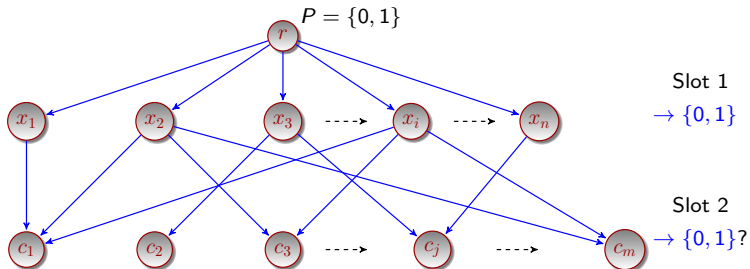
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# Routing of Packets: Connected Dominating Sets (CDS)

## CDS: Defintion

A connected dominating set in a directed graph  $\mathcal{G}(V, E)$  and root  $r$  is a set of vertices  $S \subseteq V$  such that:

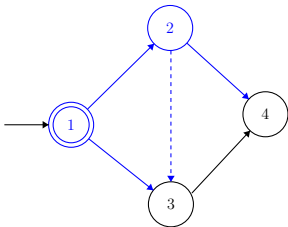
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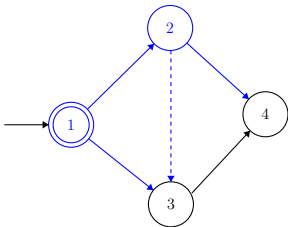
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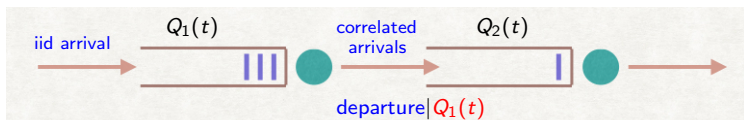
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## Key observation: Route of a Packet

Every packet must be transmitted **sequentially by a CDS** in order to be broadcasted.

# Design of UMW: Motivation and Insight

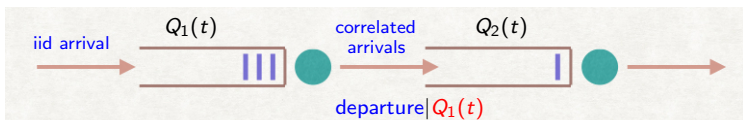
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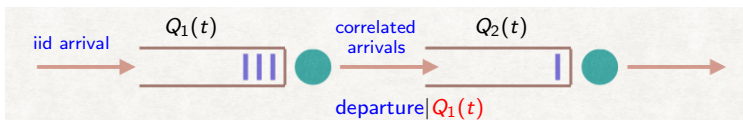
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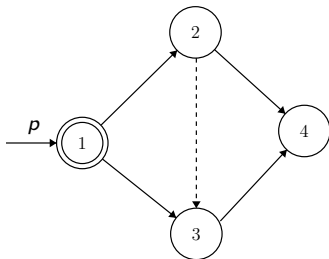
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**Ans:** The Precedence Constraints!

## Precedence Constraint: Example

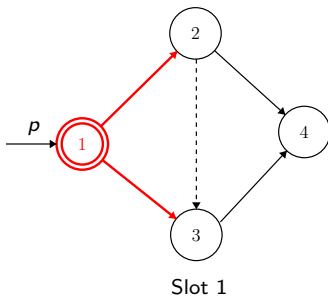
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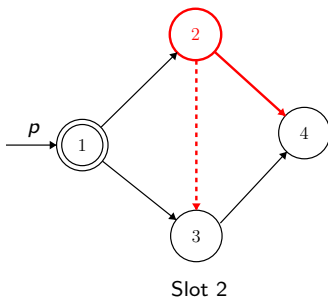
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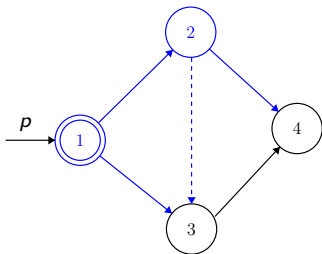
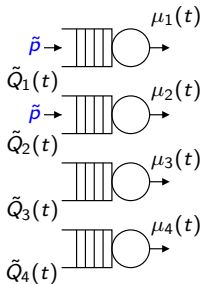
Consider an incoming packet  $p$  with the **specified** broadcasting route  $CDS_p = \{1, 2\}$ .



**Observation:** Due to the precedence, the packet  $p$  is transmitted by Node 2 **after** it has been transmitted by Node 1

## Precedence Relaxation: Example

We maintain a virtual system where the packet  $p$  is injected to the virtual queues  $\tilde{Q}_1, \tilde{Q}_2$  **immediately upon arrival**.

A Wireless Network  $\mathcal{G}$ 

Virtual Queues

# Virtual Queues: Operation

Formally,

- ① Associate a virtual queue  $\tilde{Q}_v(t)$  with each **node**  $v$  of the graph.
- ② Upon packet **arrival**:
  - Determine a **CDS**  $T_p^*(t)$  for the packet  $p$
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**Question:** How to design the optimal controls:  $T_p^*(t)$  and  $\mu^*(t)$  ?

## Dynamics of the Virtual Queues $\tilde{Q}(t)$

The virtual queue lengths can be mathematically identified with an  $n$ -dimensional vector taking values in  $\mathbb{Z}_+^n$ .

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► Denote the (controlled) arrival to the VQ  $\tilde{Q}_i$  by  $\tilde{A}_i(t)$ . Then, the virtual queues evolve as:

$$\tilde{Q}_i(t+1) = (\tilde{Q}_i(t) + \tilde{A}_i(t) - \mu_i(t))^+, \quad (\text{Lindley recursion}) \quad (1)$$

► Note that, the arrivals to the virtual queues ( $\tilde{A}_i(t), i \in V$ ) are **explicit control variables** at the source.

► Unlike the original system, given the controls, the virtual queues are **independent** of each other. This makes their exact analysis tractable.



# Stabilizing Controls for $\tilde{Q}(t)$ : Drift Analysis

- A natural first-step is to design  $\pi^{\text{UMW}} \equiv (\mathbf{A}(t), \boldsymbol{\mu}(t))_{t \geq 0}$ , such that, it stabilizes the virtual system  $\{\tilde{Q}(t)\}_{t \geq 0}$ .
- The policy consists of the routing decisions : **routing  $\mathbf{A}^\pi(t)$** , and **scheduling  $\boldsymbol{\mu}^\pi(t)$** .
- **Intuition:** This control is *likely to stabilize* the physical queues as well
  - However, note that the dynamics of the physical queues depend explicitly on the packet scheduling policy (e.g., FIFO, LIFO etc.). We will come to this issue later.

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  - However, note that the dynamics of the physical queues depend explicitly on the packet scheduling policy (e.g., FIFO, LIFO etc.). We will come to this issue later.
- To stabilize the virtual queues, we choose the control that **minimizes the drift of the Quadratic Lyapunov (potential) function of the Virtual Queues**.

## Derivation of the Control-Policy

- Define a Quadratic Lyapunov (potential) function

$$L(\tilde{\mathbf{Q}}(t)) \stackrel{\text{def}}{=} \sum_{i \in V} \tilde{Q}_i^2(t)$$

- The one-slot drift of  $L(\tilde{\mathbf{Q}}(t))$  under any admissible policy  $\pi$  may be computed to be

$$\begin{aligned} \Delta^\pi(t) &\stackrel{\text{def}}{=} L(\tilde{\mathbf{Q}}(t+1)) - L(\tilde{\mathbf{Q}}(t)) \\ &\leq B + 2 \left( \underbrace{\sum_{i \in V} \tilde{Q}_i(t) A(t) \mathbb{1}(i \in T^\pi(t))}_{(a)} - \underbrace{\sum_{i \in V} \tilde{Q}_i(t) \mu_i^\pi(t)}_{(b)} \right) \quad (2) \end{aligned}$$

Where  $T^\pi(t) \in \mathcal{T}$  and  $\mu^\pi(t) \in \mathcal{M}$  are routing and activation control variables chosen for slot  $t$ .

- The drift upper-bound (2) has a nice **separable** form and may be minimized over the **routing** and **activation** controls individually.

## Optimal Routing Policy $T_p^*(t)$

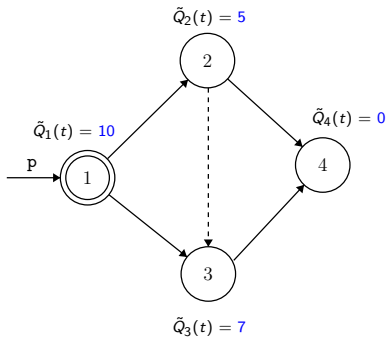
Let  $\mathcal{T}$  denote the set of all CDS in  $\mathcal{G}$ . Minimizing the routing term (a), we get the following optimal routing policy.

Optimal Routing :  $T_p^*(t)$

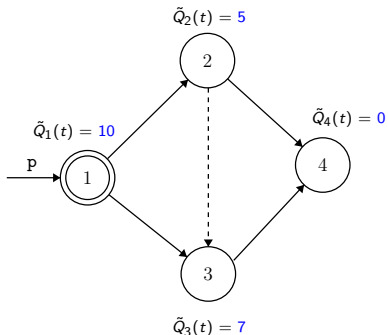
$$T_p^*(t) \in \arg \min_{T \in \mathcal{T}} \sum_{i \in V} \tilde{Q}_i(t) \mathbb{1}(i \in T)$$

In other words, the drift minimizing routing policy is to route the incoming packet along the **Minimum Weight CDS**, where each node  $i$  is weighted by the corresponding virtual queue  $\tilde{Q}_i(t)$ .

# Example of Optimal Routing



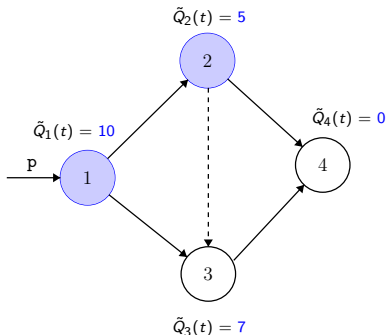
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Chosen route=MCDS=  $\{1, 2\}$

## Optimal Node Scheduling Policy $\mu^*(t)$

Let  $\mathcal{M}$  denote the set of all non-interfering activations in  $\mathcal{G}$ . Minimizing the scheduling term (b) in the drift expression, we get the following optimal scheduling policy.

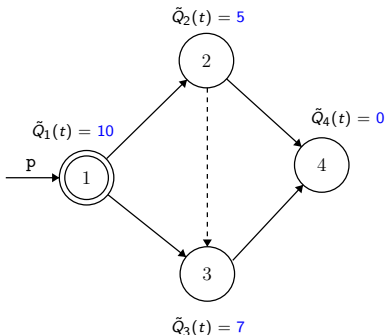
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In other words, the drift minimizing node scheduling policy is to schedule the **Max-Weight activation**, where each node  $i$  is weighted by the corresponding virtual queue length  $\tilde{Q}_i(t)$ .

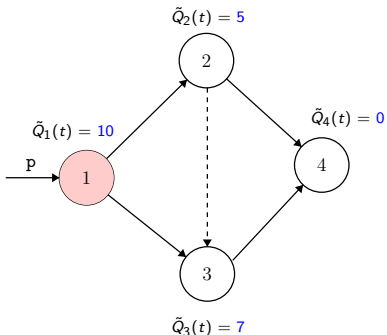


# Example of Optimal Scheduling



Due to interference, can activate only **one node** per slot

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Optimal Schedule = Activate the Node 1

# Stability of the Virtual Queue

## Theorem 6: Strong Stability of $\tilde{Q}(t)$

Under the above routing and scheduling policy, for all arrival rate  $\lambda \leq \lambda^*$  the virtual queue process is Strongly stable and has a limiting M.G.F, i.e.,

$$\limsup_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T \sum_i \mathbb{E}(\tilde{Q}_i(t)) \leq B$$

and,

$$\limsup_{T \rightarrow \infty} \mathbb{E}(\exp(\theta^* \sum_i \tilde{Q}_i(t))) \leq C$$

for some finite  $B, C$  and strictly positive  $\theta^*$ .

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The above leads to the following :

### Lemma: Sample Path bound on Virtual Queues

Under the same condition, we have

$$\sum_i \tilde{Q}_i(t) = \mathcal{O}(\log t), \quad \text{a.s.}$$

# Optimal Control of the Physical Queues : Packet Scheduling

- How do we decide which packet to transmit over a link at any given time slot?
  - Why does it matter? Cannot we just use FCFS?
- Nearest to Origin (NTO) policy [Gamarnik, 1998]
- **Extended Nearest to Origin policy (ENTO)**: When multiple packets contend for an edge, schedule the one which has traversed the least number of edges
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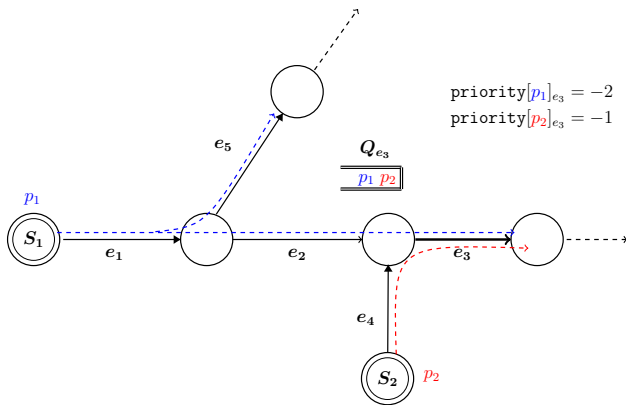
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## Theorem 7: Stability of the Physical Queues

The overall UMW policy is throughput-optimal.

Proof uses the previous **almost sure** arrival bound on a typical sample path with an **inductive argument** on the edges.

## ENTO: Example



Packet  $p_1$  has higher priority than  $p_2$  to cross  $e_3$  as it has traversed **less** number of edges

## Proof Ideas for Theorem 7: Stability of the Physical Queues

- ① **Observation:** Since **routes are fixed at source**, total number of arrival  $\tilde{A}_e(t_1, t_2)$  in interval  $[t_1, t_2]$  at virtual queue  $\tilde{Q}_i = A_i(t_1, t_2)$  total number of packets that **wish to be trabsmitted** by the node  $i$  in the physical network **sometime in future**.



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- ② Theorem 6 (Stability of the **Virtual Queues**) + **Skorokhod Map** representation + Almost Sure Bound  $\implies$

$$A_i(t_0, t) \leq S_i(t_0, t) + \mathcal{O}(\log(t)), \forall i \in V, t_0 \leq t, \text{ w.p. } 1$$

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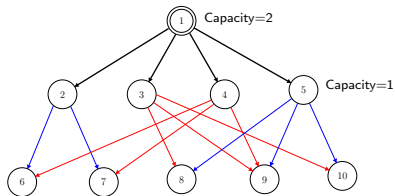
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- ③ With the **universal stability** property of the ENTO packet scheduling policy it is finally shown that the physical queues are rate stable.
    - Involves induction on the number of hops from the source.

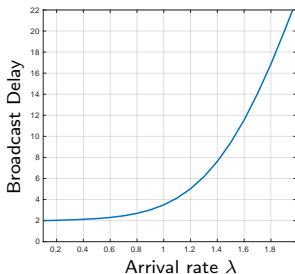
# Outline

- 1 Introduction
- 2 Hardness of Wireless Broadcasting
- 3 Throughput-Optimal Broadcast Policy
- 4 Numerical Results**
- 5 Conclusion

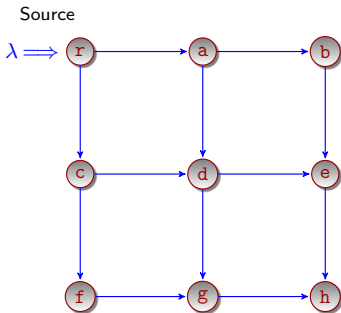
# Broadcasting: Network without Interference



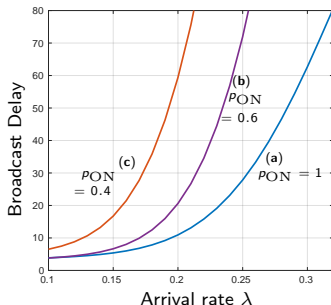
A wireless network with non-interfering channels. The broadcast capacity of the network is  $\lambda^* = 2$ .



## Broadcasting Simulation: Time-Varying Wireless Network



The Grid Network



Plot of the broadcast delay incurred by the UMW policy as a function of the arrival rate  $\lambda$  in the  $3 \times 3$  wireless grid network.

# Outline

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# Conclusion

- Our understanding of network control theory has progressed enormously over the past 25 years, starting with the seminal Backpressure policy of Tassiulas and Ephremides (1992).
- We have derived a throughput-optimal algorithm, UMW, for broadcasting in wireless networks with **point-to-multipoint** links.
- This important problem was proposed by Massoulié and Twigg, and has remained open for last ten years.
- The virtual network framework used to solve the problem is surprisingly general and may be applied to other open problems in this area (e.g., Sinha, Modiano, [INFOCOM '17](#)).
- Opens up exciting new directions for research with lots of interesting problems.