

# Optimization and Stochastic Control of MANETs

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# Cross-Layer Mechanisms for Wireless Networks

- **Scheduling, Routing, Network Coding:**
  - Queue-length based backpressure policies
- **Rate Control:**
  - Utility maximization framework for representing different QoS requirements
  - Convex optimization duality and subgradient methods used to develop distributed algorithms for rate allocation.
- Cross layer mechanisms have been designed by combining the two frameworks through the association:
  - **Differential queue length** on each link provides information about the **Lagrange Multiplier** (or the shadow cost) of that link

# Challenges and Ongoing Work

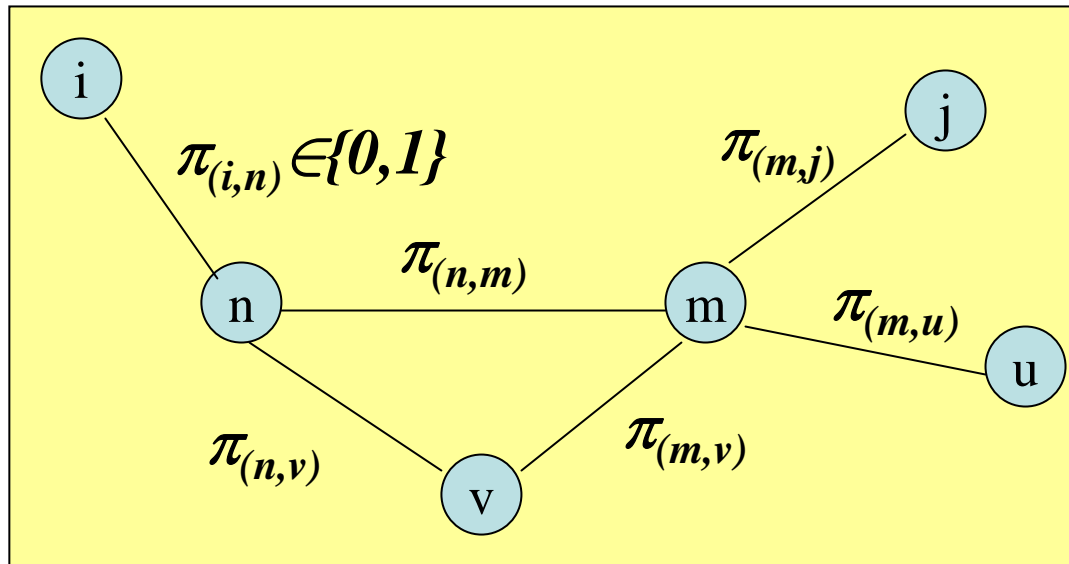
- **Backpressure Policies for Wireless Networks:**
  - Traditionally have relied on a centralized scheduler
  - Delay performance not well-understood.
- **Subgradient Methods:**
  - Convergence can be established using different stepsize rules.
  - Very few applicable results on the rate of convergence in the dual and primal space
  - In practice, we care about constructing near-optimal (near feasible) primal solutions
    - Not studied systematically for convex programs.
  - Nonconvex utility maximization (representing inelastic preferences)

# This Talk

- Polynomial Complexity Distributed Algorithms for Coding-Scheduling - Rate Control Decisions in Wireless Networks  
[Eryilmaz, Ozdaglar, Modiano 07]

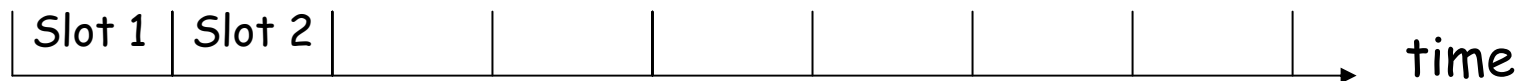
# Wireless Network Model

- The network is represented by a graph:  $\mathcal{G} = (\mathcal{N}, \mathcal{L})$ .



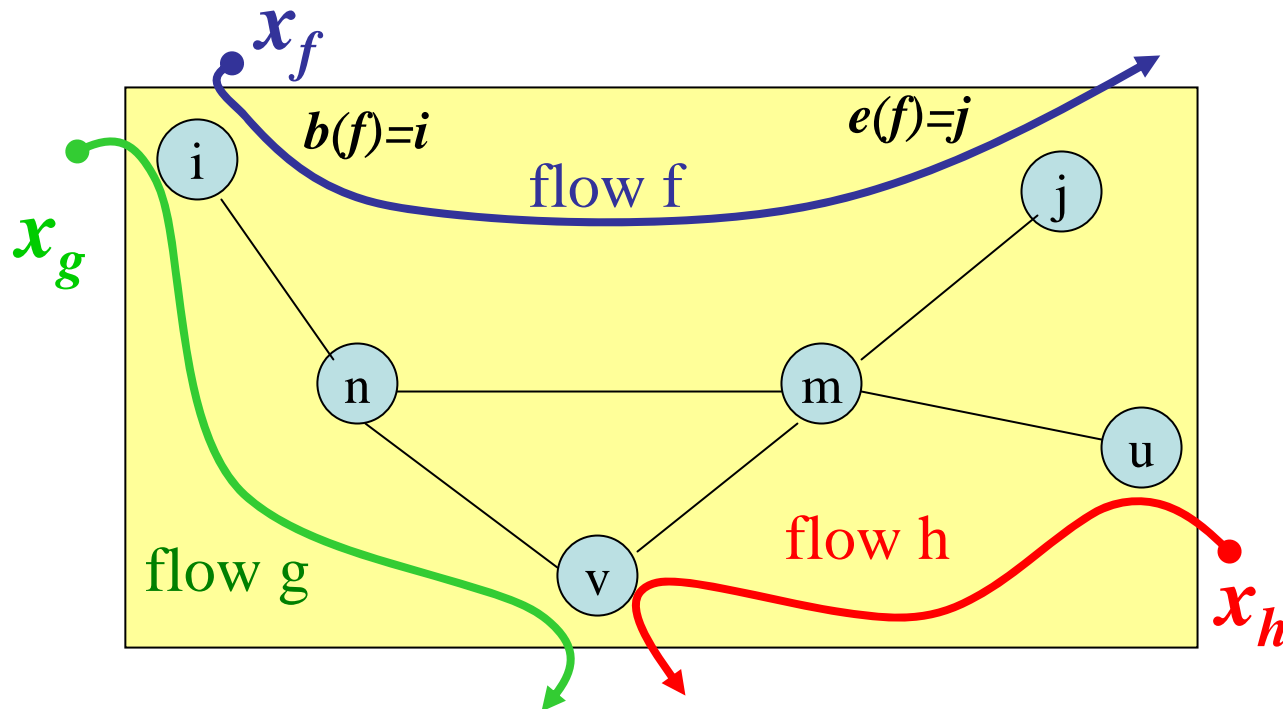
- $\Pi =$  Convex hull of the set of link rates that are allowable in a time slot, i.e., we have:

$$\pi[t] \in \Pi, \quad \forall t.$$



# Traffic Model

- $\mathcal{F}$  : The set of flows that share the network.
- Each Flow- $f$  is defined by a pair of nodes:  $(b(f), e(f))$ .



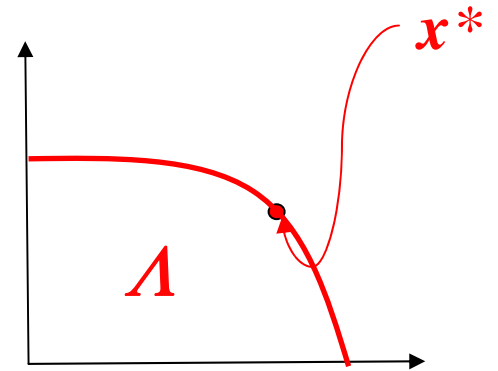
- Let  $x_f$  denote the mean rate of flow  $f \in \mathcal{F}$ .
- Link level interference constraints  $\Pi$  impose constraints on the maximum flow rate region  $\Lambda$ .

- Then,  $U_f(x_f)$  is a strictly concave function that measures the utility of Flow- $f$  as a function of  $x_f$ .

# Problem Statement

- We aim to design a *distributed scheduling-routing-flow control* mechanism that
  - **guarantees stability** of the queues,
  - and allocates the mean flow rates,  $\{x_f\}$ , so that they **attain the unique optimum** of the following problem:

$$\begin{array}{ll} \max_{\mathbf{x}} & \sum_{f \in \mathcal{F}} U_f(x_f) \\ \text{s.t.} & \mathbf{x} \in \Lambda \end{array}$$



- We use  $\mathbf{x}^*$  to denote the optimizer of the above problem, and also call it the *fair allocation*.

# *Background and Related Work [partial]*

- *Tassiulas and Ephremides ('92, '93)*
  - Introduced the queue-length based scheduling-routing mechanism
  - Central Controller necessary for Wireless Networks
  - Only *throughput-optimality*, no rate control
- *Tassiulas ('98)*
  - Introduced a Randomized Algorithm for switches
  - No rate control & assumed central information
- *Eryilmaz, Srikant, Lin et al., Neely et al., Stolyar ('05)*
  - Solved the utility maximization problem with rate control
  - Required centralized controller
- *Lin et al., Wu et al., Chaporkar et al. ('05)*
  - Considered distributed algorithms that sacrifice a portion of the capacity
- *Modiano, Shah, Zussman ('06)*
  - Proposed Gossip algorithms for distributed implementation
  - Focused on the switch scenario & no rate control



# *Joint Scheduling-Routing-Flow Control Policy*

- $q_{n,d}[t]$  = queue at node  $n$  with packets destined to node  $d$  in slot  $t$ .

## **1. Generic Randomized Scheduling-Routing Policy:**

- Define link weights for each link  $(n,m)$ :  $w_{(n,m)}[t]$

## Examples to link weights

- Unicast sessions without coding :
  - $q_{n,d}[t]$  = queue at node  $n$  with packets destined to node  $d$  in slot  $t$ .
  - $w_{(n,m)}[t] = \max_d |q_{n,d}[t] - q_{m,d}[t]|$
- Multicast sessions with intra-session coding [*Ho et al. ('05)*]
  - $q_{n,d^D}[t]$  = queue at node  $n$  with packets destined to  $d \in D$  in slot  $t$ .
  - $w_{(n,m)}[t] = \max_D \sum_{d \in D} (q_{n,d}^D[t] - q_{m,d}^D[t])^+$
- Unicast sessions with inter-session coding [*Eryilmaz and Lun ('06)*]
  - $w_{(n,m)}[t] = \max \left\{ \max_{D \in \{D_1, D_2\}} \sum_{d \in D} (q_{n,d}^D[t] - q_{m,d}^D[t])^+, \sigma_{(n,m)}^{D_1, D_2} \right\}$

# Joint Scheduling-Routing-Flow Control Policy

## 1. Generic Randomized Scheduling-Routing Policy:

- Define link weights for each link  $(n,m)$ :  $w_{(n,m)}[t]$

- Let 
$$\pi^*[t] \in \arg \max_{\pi \in \Pi} \sum_{(n,m) \in \mathcal{L}} \pi_{(n,m)} w_{(n,m)}[t]$$

- Assuming that there exists a randomized policy ( $\mathbf{R}$ ) satisfying

$$P(\pi^{(R)}[t] = \pi^*[t] \mid \mathbf{q}[t]) \geq \delta > 0 \quad \forall \mathbf{q}[t]. \quad (1)$$

- Repeat:

$\pi^{(R)}[t] \leftarrow$ Pick a random allocation satisfying (1);	<b>PICK</b>
$\pi[t+1] = \begin{cases} \pi[t] & \text{if } (\mathbf{w}[t] \cdot \pi[t]) \geq (\mathbf{w}[t] \cdot \pi^{(R)}[t]) \\ \pi^{(R)}[t] & \text{otherwise} \end{cases}$ ;	<b>COMPARE</b>
$t \leftarrow t + 1;$	

# *Lagrangian Duality for Joint Scheduling-Routing-Flow Control*

## **2. Dual Flow-Control Policy:**

- At each slot  $t$ , Flow- $f$  updates its arrival rate as

$$x_f[t] = U_f'^{-1} \left( \frac{q_{b(f),e(f)}[t]}{K} \right),$$

where  $K$  is a positive design parameter.

- The scheduling-routing policy is randomized but not yet distributed.
- The flow-control policy is decentralized and uses only local queue-length information.

# Optimality of the Joint Algorithm

- **Theorem 1:** Given any randomized algorithm that satisfies (1), there exists positive constants,  $C_1$  and  $C_2$ , for which

$$\sum_{n \in \mathcal{N}} \sum_{d \in \mathcal{N}} \overline{q_{n,d}} \leq C_1 K,$$

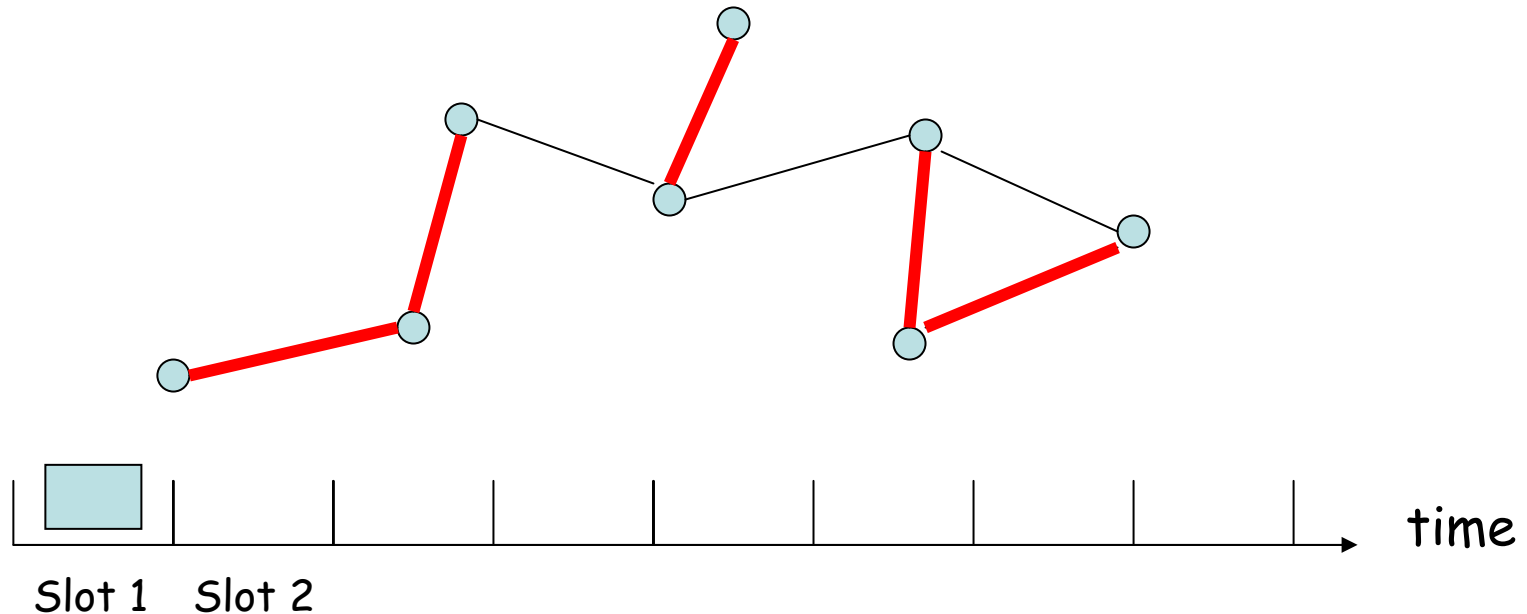
$$\sum_{f \in \mathcal{F}} U_f(\bar{x}_f) \geq \sum_{f \in \mathcal{F}} U_f(x_f^*) - \frac{C_2}{K},$$

where  $\bar{x}_f := \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} E[x_f[t]]$ , and similarly for  $\overline{q_{n,d}}$ .

- Trade-off between average delay and fairness
- The result applies to a large class of interference models

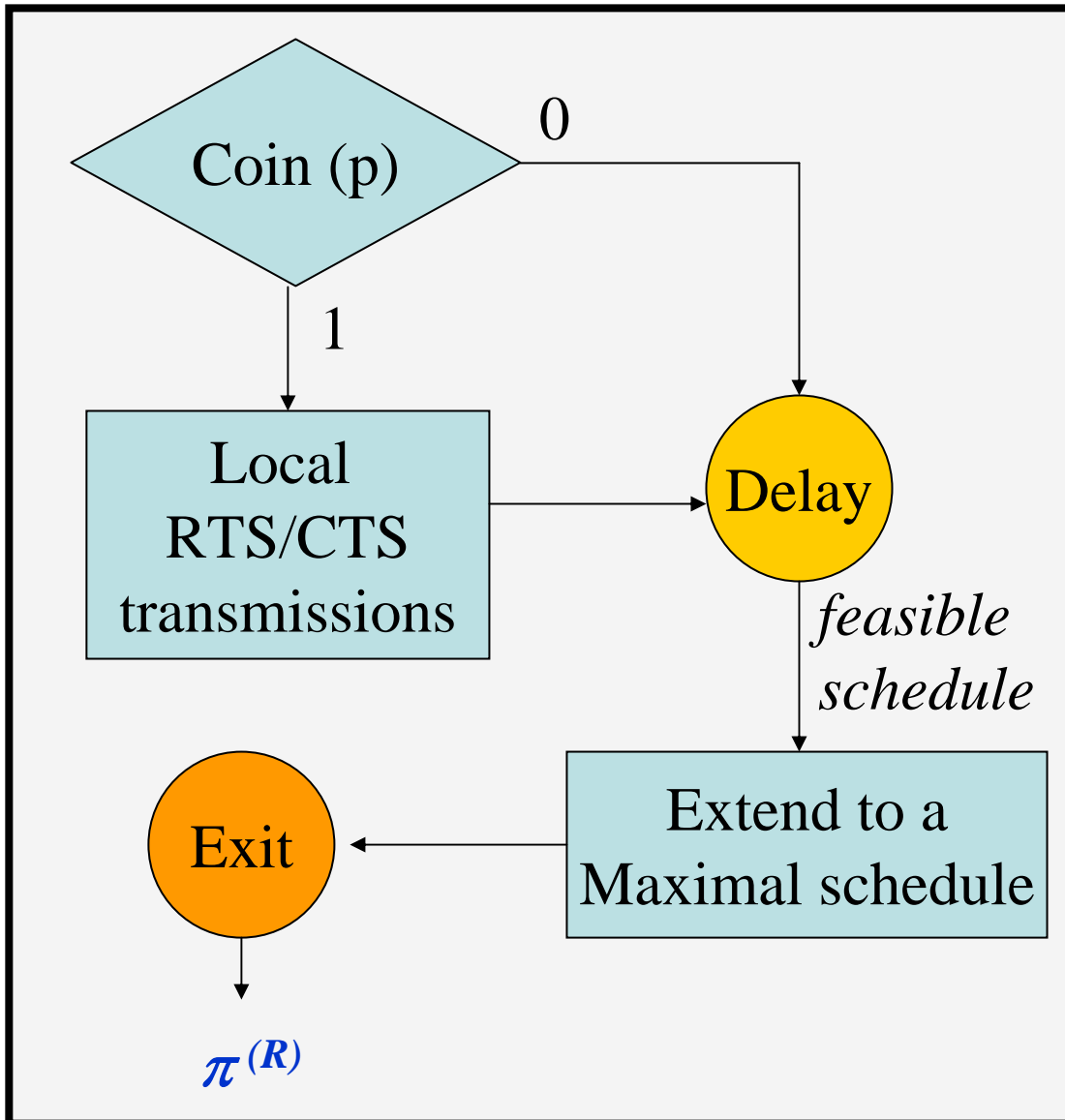
# Distributed Algorithm for 2<sup>nd</sup> Order Interference

- We say a link is *active* if there is a transmission over it.
- A transmission over link  $(n,m)$  is successful iff no neighbors of  $n$  and  $m$  have an active link incident to them.



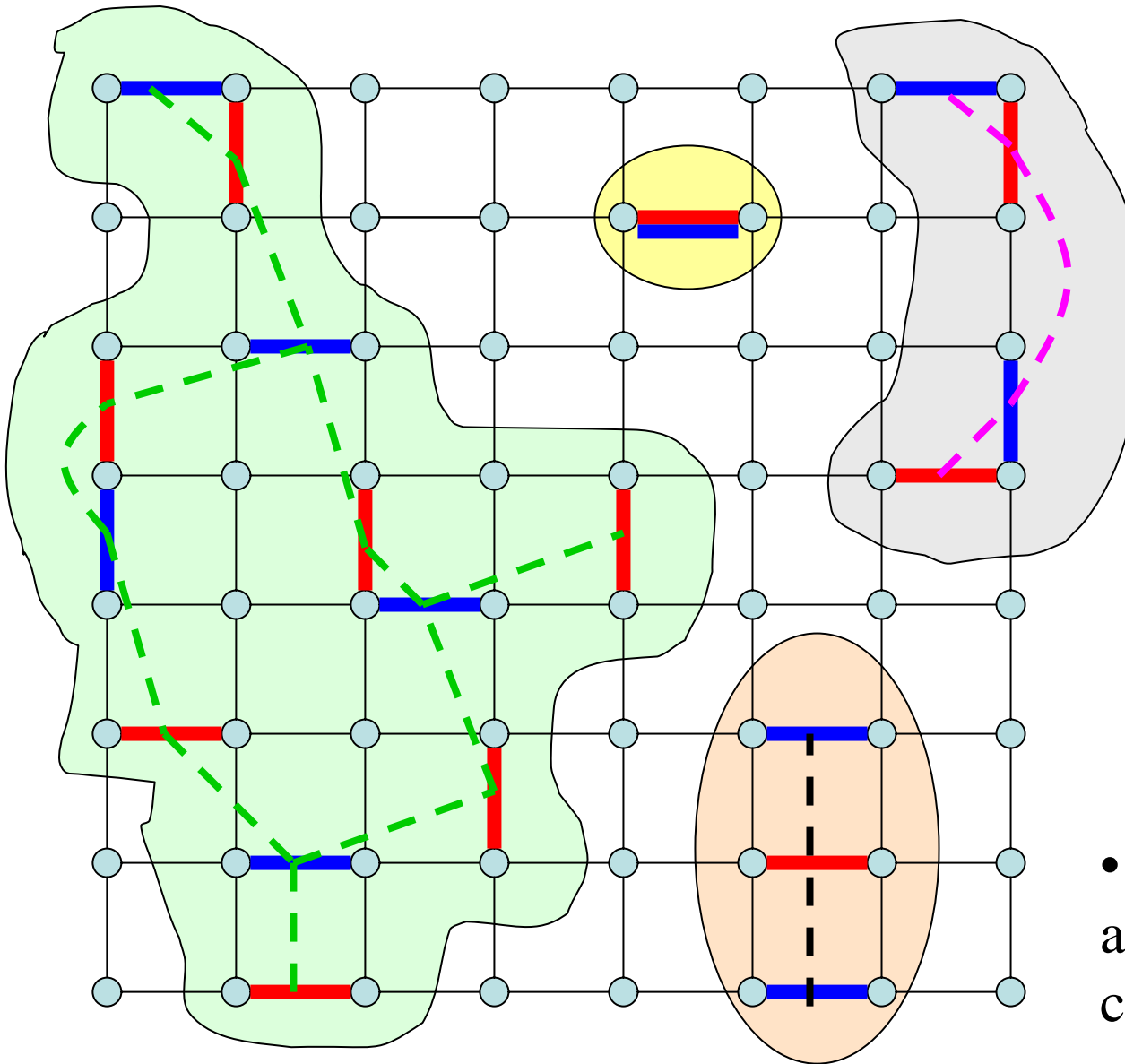
- Remark: For this model, finding the maximum weight schedule is an **NP-hard** problem.

# PICK Algorithm



- Every link has a unique ID number.
- The algorithm finds a (maximal) feasible schedule for which (1) is satisfied.
- At the end of the algorithm, each link knows the ID number of its neighboring nodes that are in  $\pi^{(R)}$  in the conflict graph.

# COMPARE Algorithm



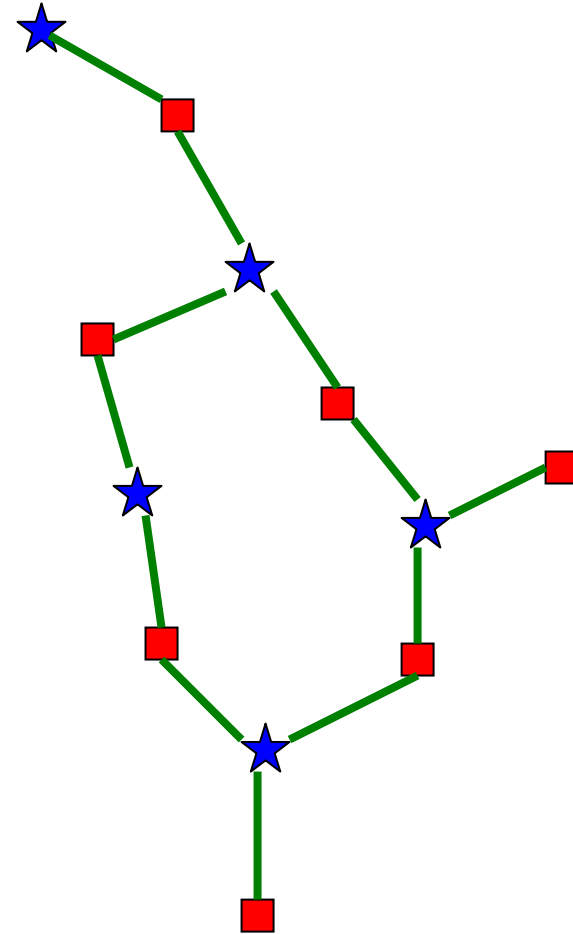
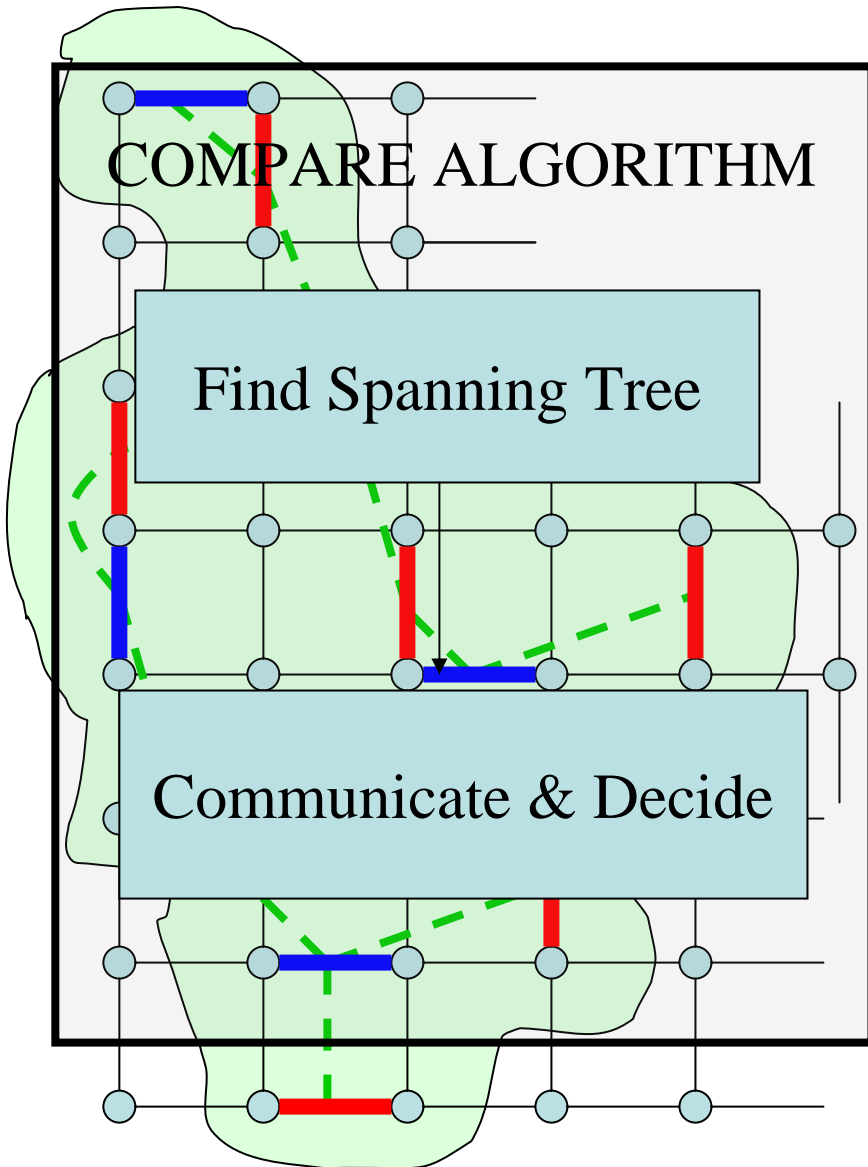
- Grid Topology
- Two schedules:

$$\pi^{\text{old}}, \pi^{(R)}$$

- Connecting interfering links
- Several disjoint connected components
- It suffices to consider a single disconnected component.



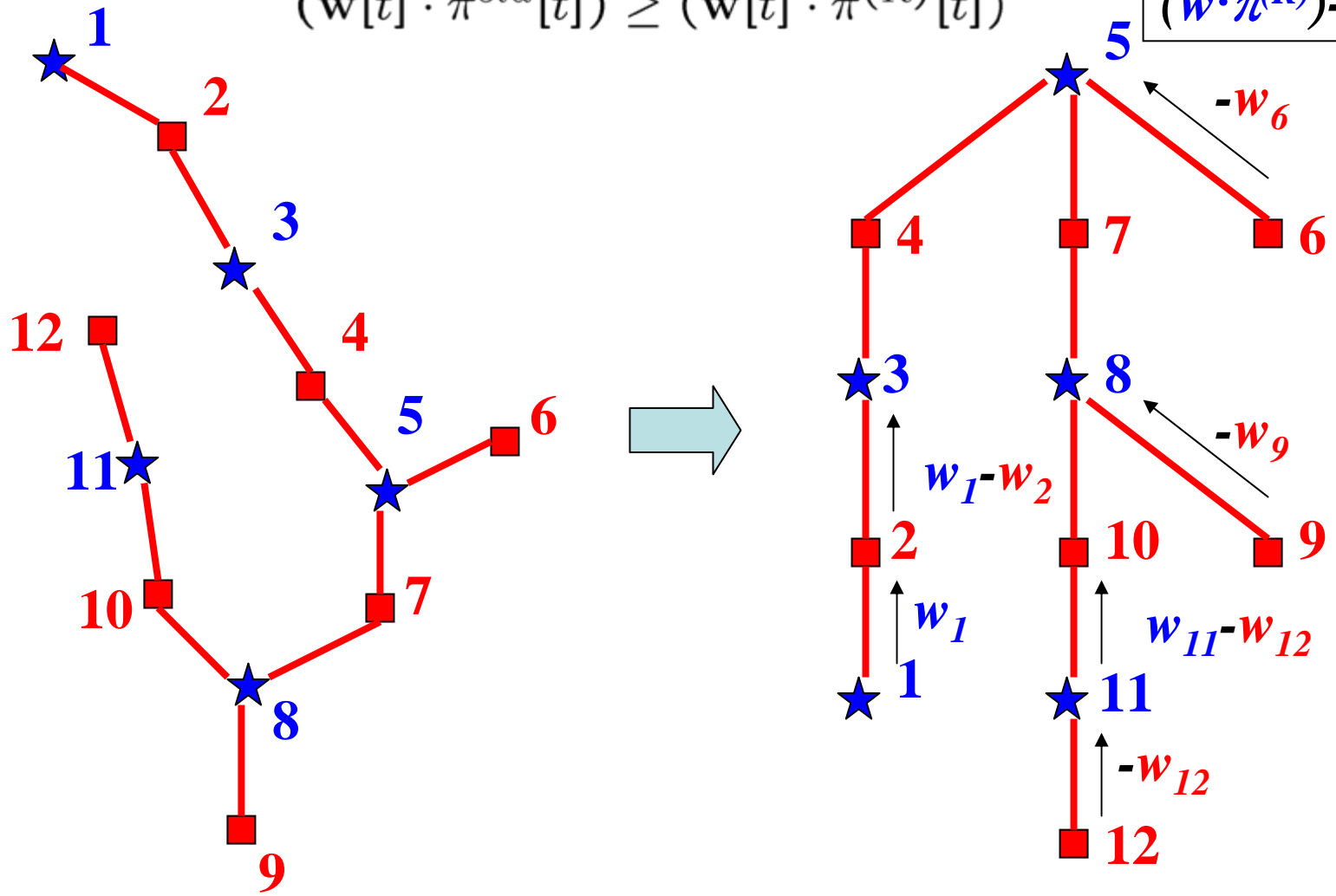
# Conflict graph



# COMMUNICATE & DECIDE Procedure

$$(w[t] \cdot \pi^{old}[t]) \geq (w[t] \cdot \pi^{(R)}[t])$$

$$(w \cdot \pi^{(R)}) - (w \cdot \pi^{old})$$

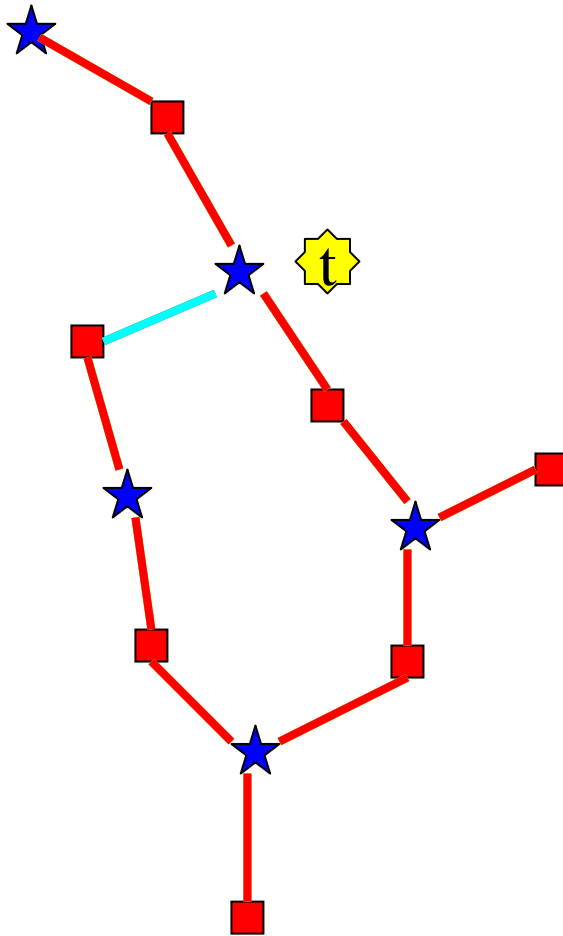


# Result

**Theorem 2:** The distributed implementations of PICK and COMPARE Algorithms designed for the second order interference model asymptotically (in  $K$ ) achieve *full utilization of network resources* with  $O(N^3)$  time and  $O(N^2)$  message exchanges per node, per stage.

- Thus, the randomized algorithm is guaranteed to achieve full utilization of the resources with polynomial complexity.
- This result is particularly interesting when we note that the maximum weight problem is an NP-hard problem for the second order interference model.

# *FIND SPANNING TREE Procedure*



- Token based procedure
- Assume there is a single token generated in the graph.
  - Every node accepts the token if it has not traversed it so far.
  - If the node has already forwarded the token, it accepts it only if it is returning from the neighbor to which the token was forwarded.
  - A token is returned to the parent of a node only if all the other neighbors are tried.

## *Proof idea [based on Tassiulas ('98)]*

- Show the negative mean drift of the Lyapunov function for

$\mathbf{y} = ( \mathbf{q} , \pi^{(R)} )$ :

$$V(\mathbf{y}) = \underbrace{\sum_{n \in \mathcal{N}} \sum_{d \in \mathcal{N}} q_{n,d}^2}_{\text{Measures the stability level of the queues.}} + \underbrace{\left( \sum_{l \in \mathcal{L}} w_l ((\pi_{\mathbf{w}}^*)_l - \pi_l^{(R)}) \right)^2}_{\text{Measures the accuracy of the random schedule to the optimal one.}}$$

Measures the stability level of the queues.

Measures the accuracy of the random schedule to the optimal one.