

# An Overview of the Use of Distributed Mechanisms in Network Coding

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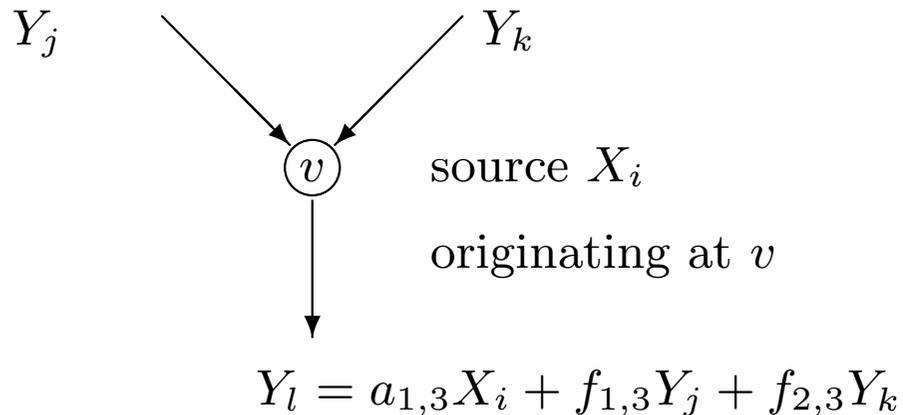
Massachusetts Institute of Technology

Deb, Effros, Ho, Karger, Koetter, Lun, Médard, Ratnakar

## Distributed Methods for Multicast Network Coding

- Can we build codes in a distributed manner?
- Can we have a distributed deployment of network coding that is cost efficient?
- How can we disseminate information in the absence of source information?

# Randomized Network Coding for Multicast



Determining feasibility: min-cut max-flow bound satisfied for each receiver [ACLY00]

For a feasible  $d$ -receiver multicast for independent or linearly correlated sources [HKMKE03, HMSEK03, WCJ03]

- choose code coefficients  $a_{i,j}$ ,  $f_{l,j}$  for  $\eta$  links independently and uniformly over  $F_q$
- success probability is at least  $(1 - d/q)^\eta$  for  $q > d$ .

## Randomized Network Coding

- Randomized network coding can use any subgraph which satisfies min-cut max-flow bound for each receiver
- Receiver nodes can decode if they receive as many independent linear combinations as the number of source processes
- Differs from traditional networking approaches which first do source/diversity coding followed by routing of coded information
- Closely related to random codes for compression

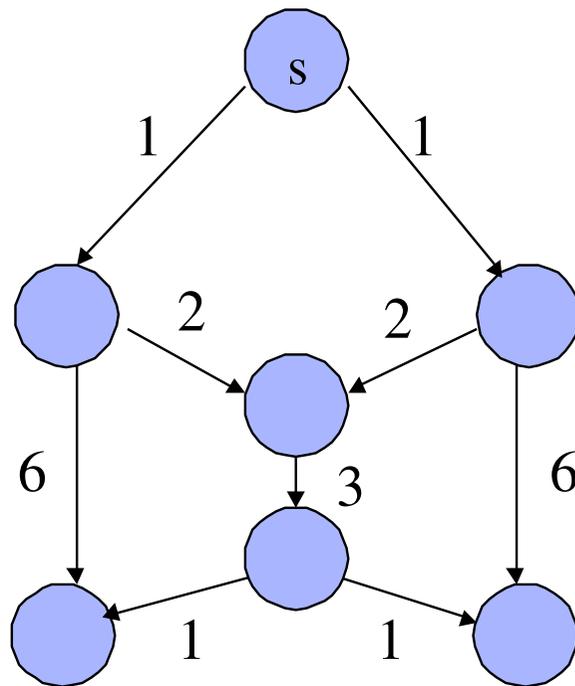
## Random Coding and Robustness

- For multicast recovery, the random code in the **interior** of the network need not be changed [KM01, HMK03]
- Robustness to corruption - what happens when a compromised node can transmit nefarious signals? Randomized distributed network coding can be used to achieve Byzantine modification detection using a simple polynomial hash functions included in transmitted packets. The modifications are detected with high probability [HLKMEK04]

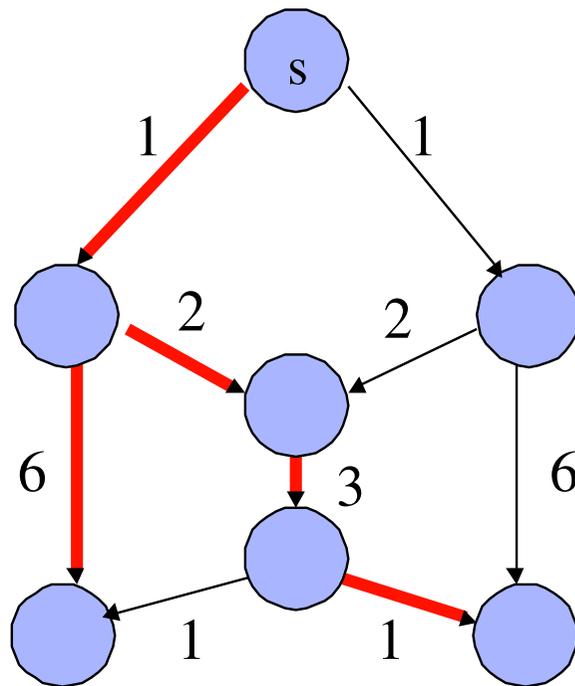
## Network Coding for Cost

- Random coding in effect decouples routing decisions and code selection decisions in the multicast case
- Are there any true benefits to obtaining a decentralized solution for coding if the choice of subgraphs must itself be centralized?
- Could a decentralized approach, of the Bellman-Ford type, be applied, even though we are not dealing with point-to-point routes?

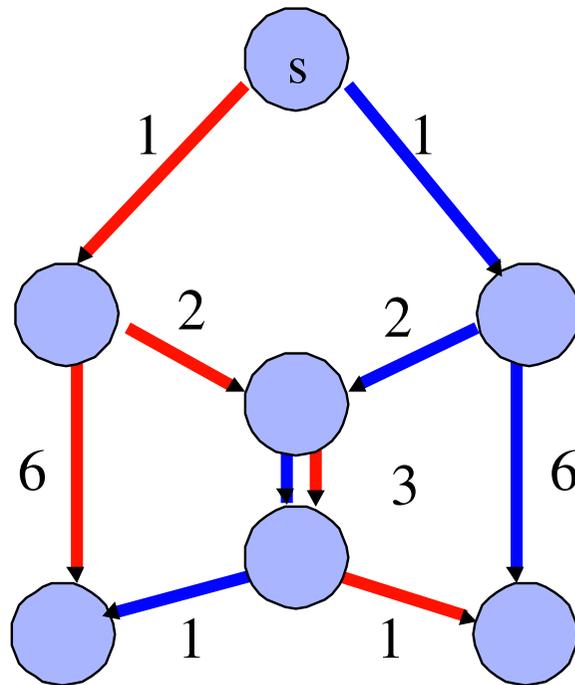
# Network Coding for Cost



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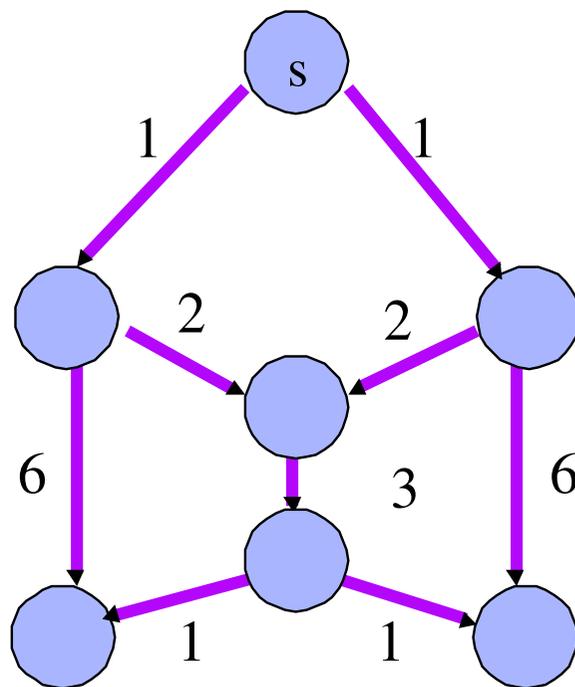


## Network Coding for Cost



Cost of trees = 26

## Network Coding for Cost



Cost of code = 23

## Network Coding for Cost

- Without coding, the problem of multicast is the Steiner tree problem over dags, possibly with decompositions into several trees
- An immediately attractive approach would be to overlay trees to create codes, attempting to increase overlaps and counting only once several uses of a link - code is built automatically
- Complexity is high and does not make use of distributed random code construction, which works well in practice
- A linear (or convex) program statement of the problem (polynomial-time) can be solved in a **distributed manner** [LMHK04, LRKMLA05]

## A LP-based Solution

$$\begin{aligned}
 & \text{minimize} && \sum_{(i,j) \in A} a_{ij} z_{ij} \\
 & \text{subject to} && z_{ij} \geq x_{ij}^{(t)}, \quad \forall (i,j) \in A, t \in T, \\
 & && \sum_{\{j|(i,j) \in A\}} x_{ij}^{(t)} - \sum_{\{j|(j,i) \in A\}} x_{ji}^{(t)} = \\
 & \left\{ \begin{array}{l} R \text{ if } i = s, \\ \\ -R \text{ if } i = t, \quad \forall i \in N, t \in T, \\ \\ 0 \text{ otherwise,} \end{array} \right. && (1) \\
 & && c_{ij} \geq x_{ij}^{(t)} \geq 0, \quad \forall (i,j) \in A, t \in T.
 \end{aligned}$$

## A LP-based Solution

- The vector  $z$  is part of a feasible solution for the LP problem if and only if there exists a network code that sets up a multicast connection in the graph  $G$  at rate arbitrarily close to  $R$  from source  $s$  to terminals in the set  $T$  and that puts a flow arbitrarily close to  $z_{ij}$  on each link  $(i, j)$
- Proof follows almost immediately from min-cut max-flow necessary and sufficient conditions
- Polynomial-time
- Steiner-tree problem can be seen to be this problem with extra integrality constraints

## A Distributed Approach

Consider the dual problem

$$\begin{aligned} & \text{maximize} && \sum_{t \in T} q^{(t)}(p^{(t)}) \\ & \text{subject to} && \sum_{t \in T} p_{ij}^{(t)} = a_{ij} \quad \forall (i, j) \in A, \\ & && p_{ij}^{(t)} \geq 0 \quad \forall (i, j) \in A, t \in T, \end{aligned} \tag{2}$$

where

$$q^{(t)}(p^{(t)}) = \min_{x^{(t)} \in F^{(t)}} \sum_{(i,j) \in A} p_{ij}^{(t)} x_{ij}^{(t)}, \tag{3}$$

and  $F^{(t)}$  is the bounded polyhedron of points  $x^{(t)}$  satisfying the conservation of flow constraints and capacity constraints

## Subgradient Approach

- Consider a subgradient approach
- Start with an iterate  $p[0]$  in the feasible set
- Solve subproblem (3) for each  $t$  in  $T$  to obtain  $x[n]$

$$p_{ij}[n+1] := \arg \min_{v \in P_{ij}} \sum_{t \in T} (v^{(t)} - (p_{ij}^{(t)}[n] + \theta[n]x_{ij}^{(t)}[n]))^2 \quad (4)$$

for each  $(i, j) \in A$ , where  $P_{ij}$  is the  $|T|$ -dimensional simplex

$$P_{ij} = \left\{ v \mid \sum_{t \in T} v^{(t)} = a_{ij}, v \geq 0 \right\} \quad (5)$$

and  $\theta[n] > 0$  is an appropriate step size

- $p_{ij}[n+1]$  is set to be the Euclidean projection of  $p_{ij}[n] + \theta[n]x_{ij}[n]$  onto  $P_{ij}$

## Step Size Selection

- $u := p_{ij}[n] + \theta[n]x_{ij}[n]$
- We index the elements of  $T$  such that  $u^{(t_1)} \geq u^{(t_2)} \geq \dots \geq u^{(t_{|T|})}$
- Take  $k^*$  to be the smallest  $k$  such that

$$\frac{1}{k} \left( a_{ij} - \sum_{r=1}^{t_k} u^{(r)} \right) \leq -u^{(t_{k+1})} \quad (6)$$

or set  $k^* = |T|$  if no such  $k$  exists

- Projection is achieved by

$$p_{ij}^{(t)}[n+1] = \begin{cases} u^{(t)} + \frac{1}{k^*} \left( a_{ij} - \sum_{r=1}^{t_{k^*}} u^{(r)} \right) & \text{if } t \in \{t_1, \dots, t_{k^*}\}, \\ 0 & \text{otherwise.} \end{cases} \quad (7)$$

## Distributed Approach - Bringing it Together

- Problem of recovering primal from approximation of dual
- Use approach of [SC96] for obtaining primal from subgradient approximation to dual
- The conditions can be coalesced into a single algorithm to iterate in a distributed fashion towards the correct cost
- There is inherent robustness to change of costs, as in classical distributed Bellman-Ford approach to routing

## Application - Wireless Networks

- Omnidirectional antennas - when transmitting from node  $i$  to node  $j$ , we get transmission to all nodes whose distance from  $i$  is less than that from  $i$  to  $j$  “for free”
- We consider energy efficiency
- We do not consider interference (bursty set-up, for instance)
- We impose an ordering  $\preceq$  on the set of outgoing links from node  $i$ , such that  $(i, j) \preceq (i, k)$  if and only if  $a_{ij} \leq a_{ik}$
- Typically, the set of outgoing links from  $i$  will be the set of all nodes within a certain, fixed radius of  $i$  and the cost  $a_{ij}$  of the link between nodes  $i$  and  $j$  will be proportional to their distance raised to some power  $\alpha$ , where  $\alpha \geq 2$

## LP for Wireless Network

- Owing to the omnidirectionality of the antennas, flow can be pushed from  $i$  to  $j$  by pushing it to any node  $k$  such that  $(i, k) \in A$  and  $(i, k) \succeq (i, j)$
- Thus, the maximum flow  $x_{ij}^{(t)}$  that can be pushed for a given  $t$  in  $T$  is

$$z_{ij} + \sum_{\{k | (i,k) \in A, (i,k) \succeq (i,j)\} \setminus \{j\}} (z_{ik} - x_{ik}^{(t)}) \quad (8)$$

- Hence

$$\sum_{\{k | (i,k) \in A, (i,k) \succeq (i,j)\}} (z_{ik} - x_{ik}^{(t)}) \geq 0 \quad (9)$$

for all  $t \in T$ .

## Usefulness of LP

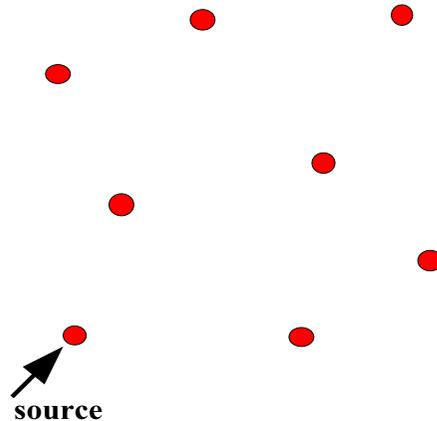
- We can extend this approach to other types of cost functions, for instant typical cost functions used to represent cost of congestion
- Can use to obtain equivalence of distances in networks, extend minimum first derivative length approaches or other convex cost considerations
- Many open issues: asynchronicity, speed of convergence, compatibility with times associated with routing-based solutions, pricing of resources and economic incentives

## Distributed Sources

Robustness to source location - what happens when the sources of traffic may not be readily identified?

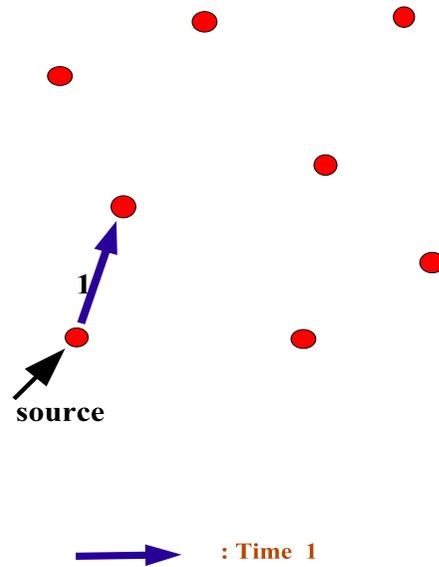
We may apply random network coding for message dissemination in networks, speeding the dissemination of  $\Theta(N)$  messages from  $O(N \log(N))$  to  $O(N)$  [DM04]

## The Original Gossip Problem



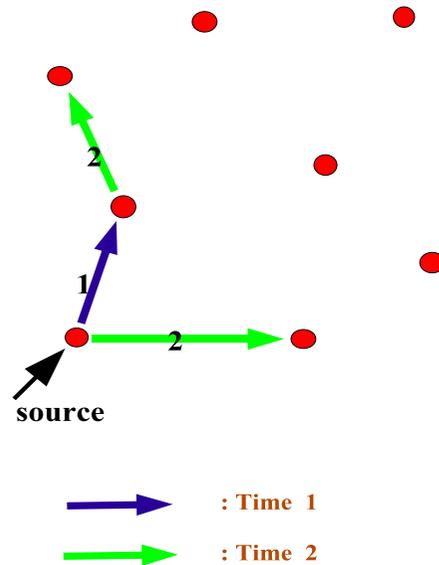
- A group of  $N$  people/nodes. Somebody has a message/rumor
- How much time does it take for the rumor to disseminate completely?

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## Our Problem: The Multiple Message Case

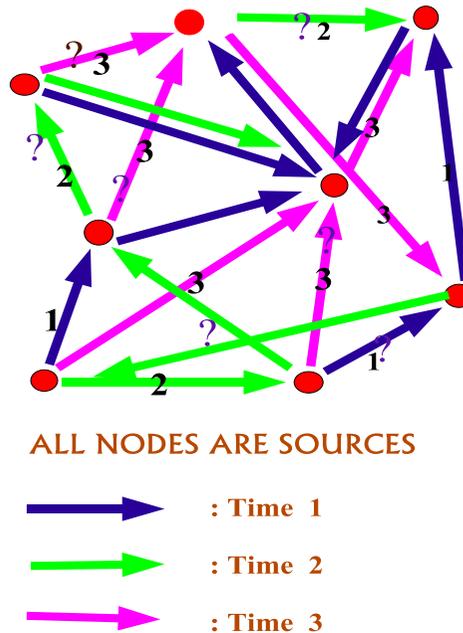
- Suppose there are  $k = \Theta(N)$  messages to start with
- Every node has only one of the  $k$  messages

Sensor network

File downloading from distributed storage locations in the network

- Goal: To disseminate all the messages among all the nodes
- Can we do better by disseminating the messages simultaneously? How?

## Main Issue: The Multiple Message Case



- Only one message can be transmitted per communication
- Communication Protocol: What to transmit? Nodes do not know the requirements of the communication partner

## Multiple Message Dissemination: A Closer Look

- If there is an omniscient central controller that decides who transmits what, complete dissemination occurs in  $\Omega(N)$  rounds
- Takes  $O(\ln(N))$  time if entire data-base exchange is allowed (almost the single message framework)

Assumes unlimited bandwidth between nodes

- Can we disseminate the messages in  $O(N)$  transmissions in a decentralized manner?

Nodes only have local knowledge

A sequential approach takes  $\Theta(N \ln(N))$  time

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## A Distributed Random Coding Approach

- $O(N)$  rounds with approach, provided we allow a slight overhead with every transmission. The nodes collect linear combination of the messages, building up **degrees of freedom**
- We allow random algebraic mixing of the messages rather than treating them as mere transportation elements
- An uncoded approach does poorly because:  
Once all the nodes have  $O(N)$  messages, a new message is likely to be an old one (coupon collector problem)  
Probability of getting a new message (from virtually any node) goes down with the number of messages collected

## Conclusions

- Distributed methods appear to have a natural place in network coding for multicast applications
- Network coding may in effect render several problems for multicasting more easily implemented in a distributed fashion
- Other naturally distributed settings lend themselves well to network coding approaches:
  - distributed storage
  - networks with varying costs
  - networks with erasures