

Ad-hoc Affectance-selective Families for Layer Dissemination

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SEA 2017

Vehicle, asset, person & pet monitoring & controlling

Agriculture automation

Energy consumption

Security & surveillance

Building management

Embedded Mobile

Internet of things

Everyday things get connected for smarter tomorrow

M2M & wireless sensor network

Everyday things

Smart homes & cities

Telemedicine & healthcare

M2M & wireless sensor network



Ad-hoc Wireless Networks example

Introduction

A Sensor Network



Intel Berkeley Research Lab

Capabilities

- processing
- sensing
- communication

Limitations

- range
- memory
- life cycle



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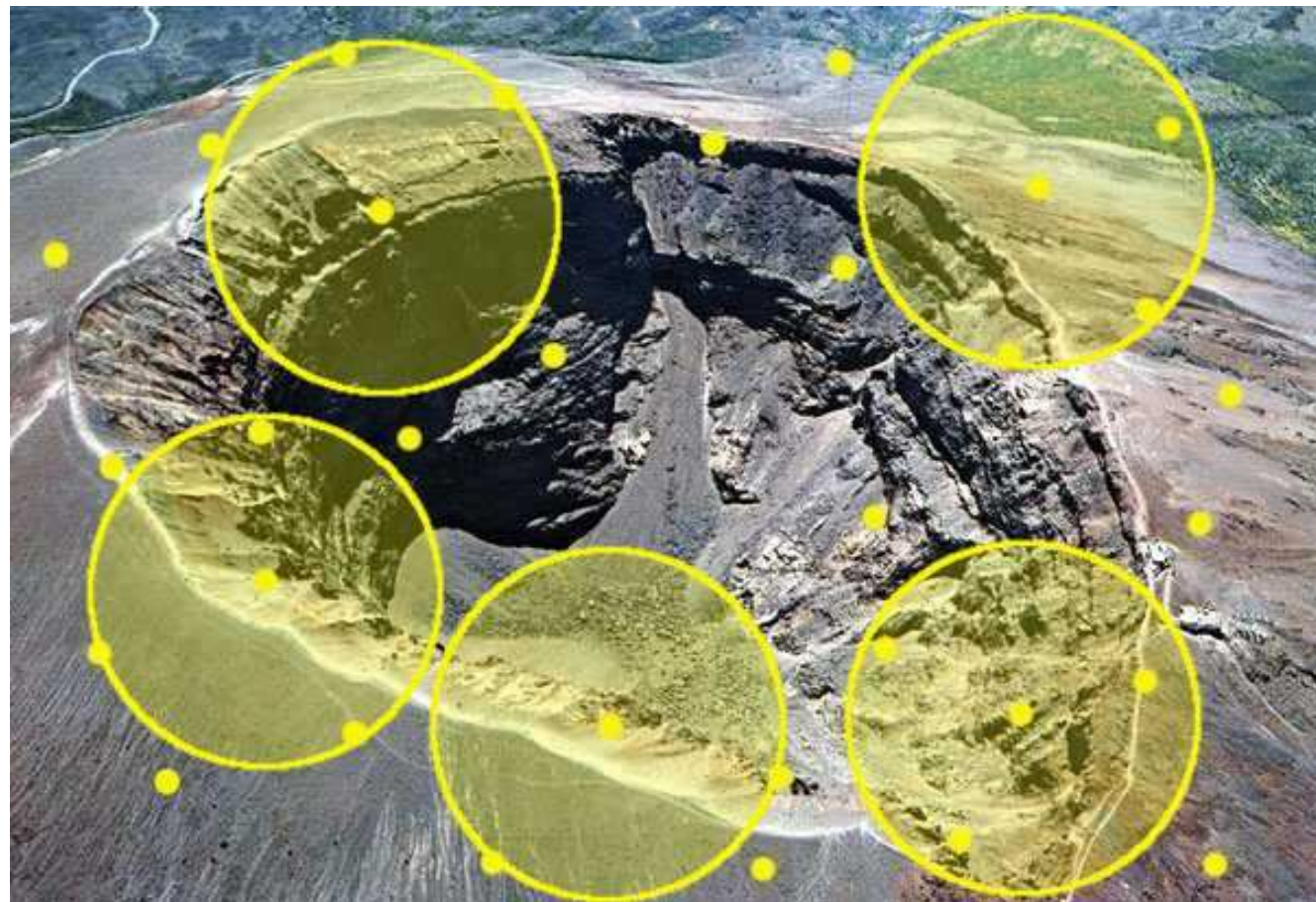
Intel Berkeley Research Lab

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Ad-hoc Wireless Networks example

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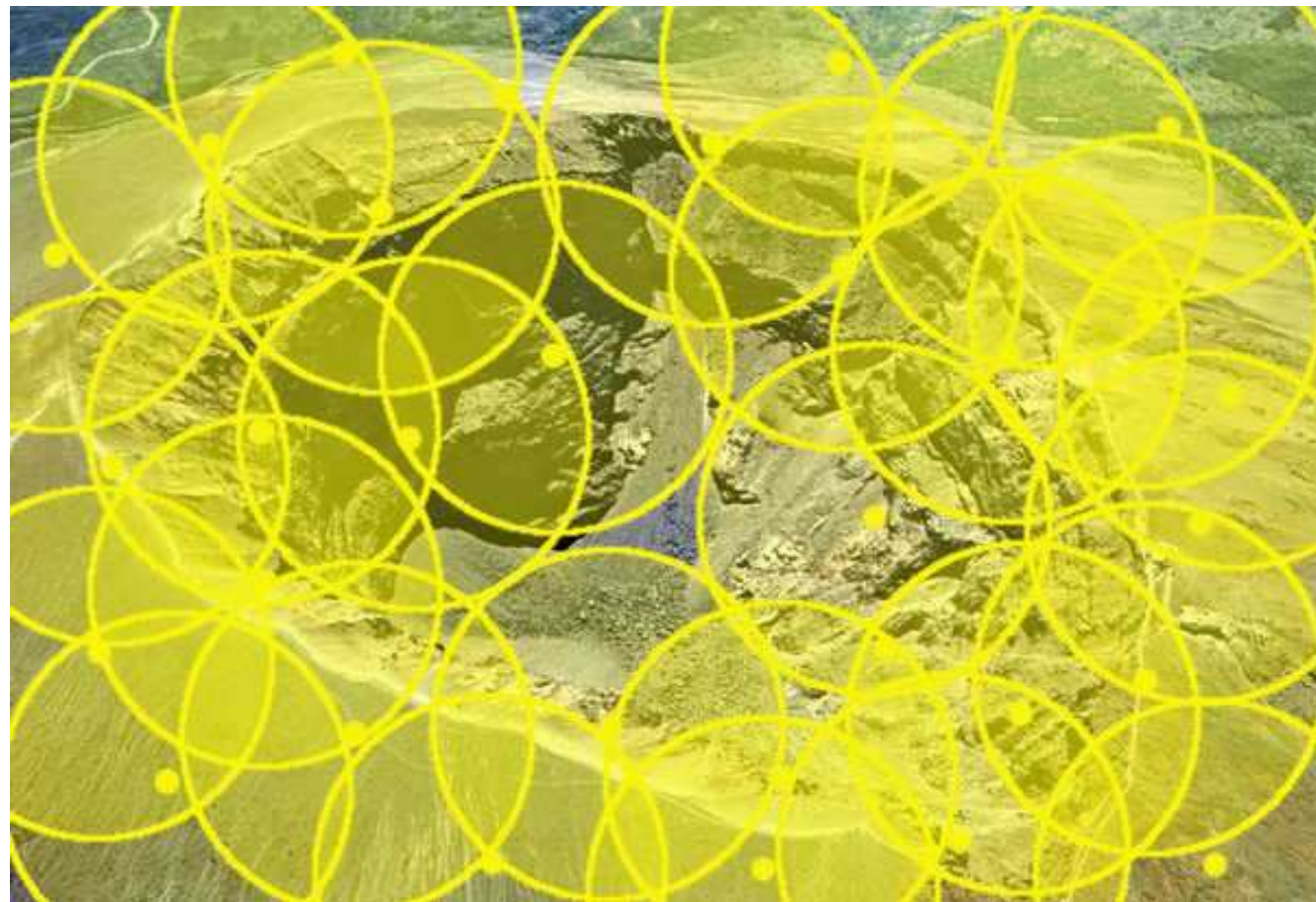
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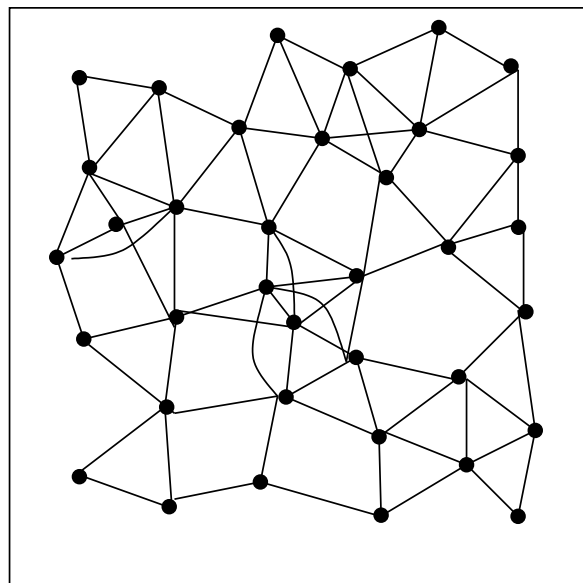
Limitations

- range
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- life cycle



Dissemination Problems in Wireless Networks

Radio Network = abstraction of a radio communication network



A geometric graph.

k nodes

hold a piece of information to disseminate.

- $k = 1 \rightarrow \text{Broadcast}$ [BGI'92,KM'98]
- $k = n \rightarrow \text{Gossiping}$ [CGLP'01,LP'02]
- k arbitrary $\rightarrow k\text{-selection}$ [K'05]
- ...
- *Multiple-message broadcast*
- *Dynamic multiple-message broadcast*
- *etc.*

Models for Wireless Networks

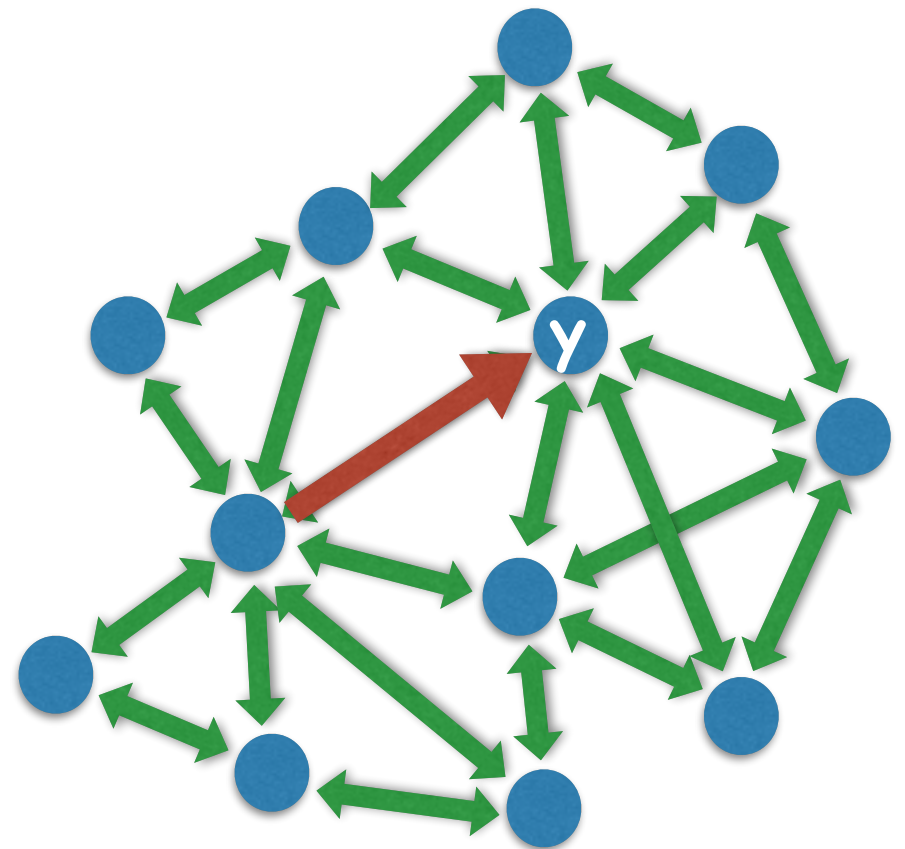
- **Topology Models :**
 - Undirected Graph
 - Unit Disk Graph
 - Time-varying Graph
- **Node Capabilities Models :**
 - Computational Resources
 - Communication Capabilities
 - Weak Sensor Model
- **Interference Models :**
 - Radio Network (RN)
 - Signal to Interference plus Noise Ratio (SINR)
 - Affectance (AFF)

Interference Models

RN Model [1]:

- **Collision/success:**

Node y receives if and only if exactly one neighbor of y transmits at a given time, and y is not transmitting.



[1] Chlamtac and Kutten. *Trans. on Computers*. IEEE, 1987.

Interference Models

SINR Model [1]:

- **Collision/success:**

A signal that overcomes interference from others plus background noise is received.

$$\frac{p((x, y))}{d(x, y)^\alpha} \geq \beta \left(\sum_{(u, v) \in \mathcal{R}(t) \setminus (x, y)} \frac{p((u, v))}{d(u, y)^\alpha} + N \right)$$

Defs. :

$\alpha > 0$: *path-loss exponent.*

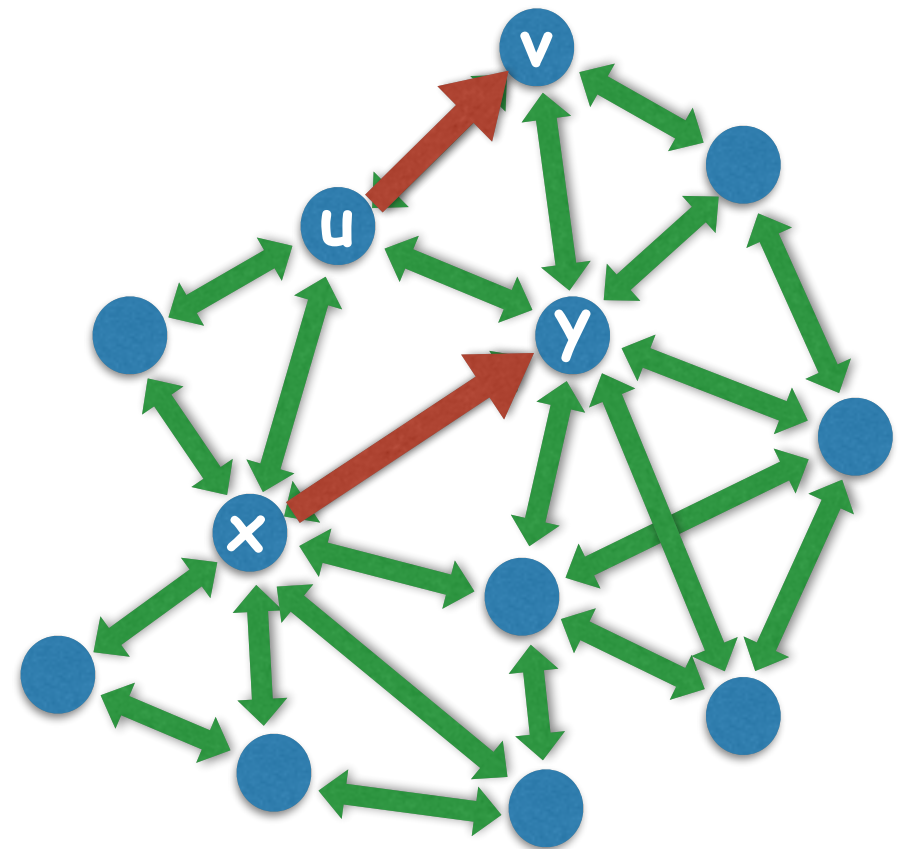
$\beta \geq 0$: *gain.*

$p((i, j))$: *transmission power on link (i, j) .*

$d(i, j)$: *Euclidean distance between i and j .*

$\mathcal{R}(t)$: *set of links transmitting at time t .*

N : *background noise.*



[1] Moscibroda and Wattenhofer. Infocom 2006.

Interference Models

AFF Model [1,2,3]:

$$a(u, (x, y))$$

matrix quantifying interference from node u on communication through link (x, y) .

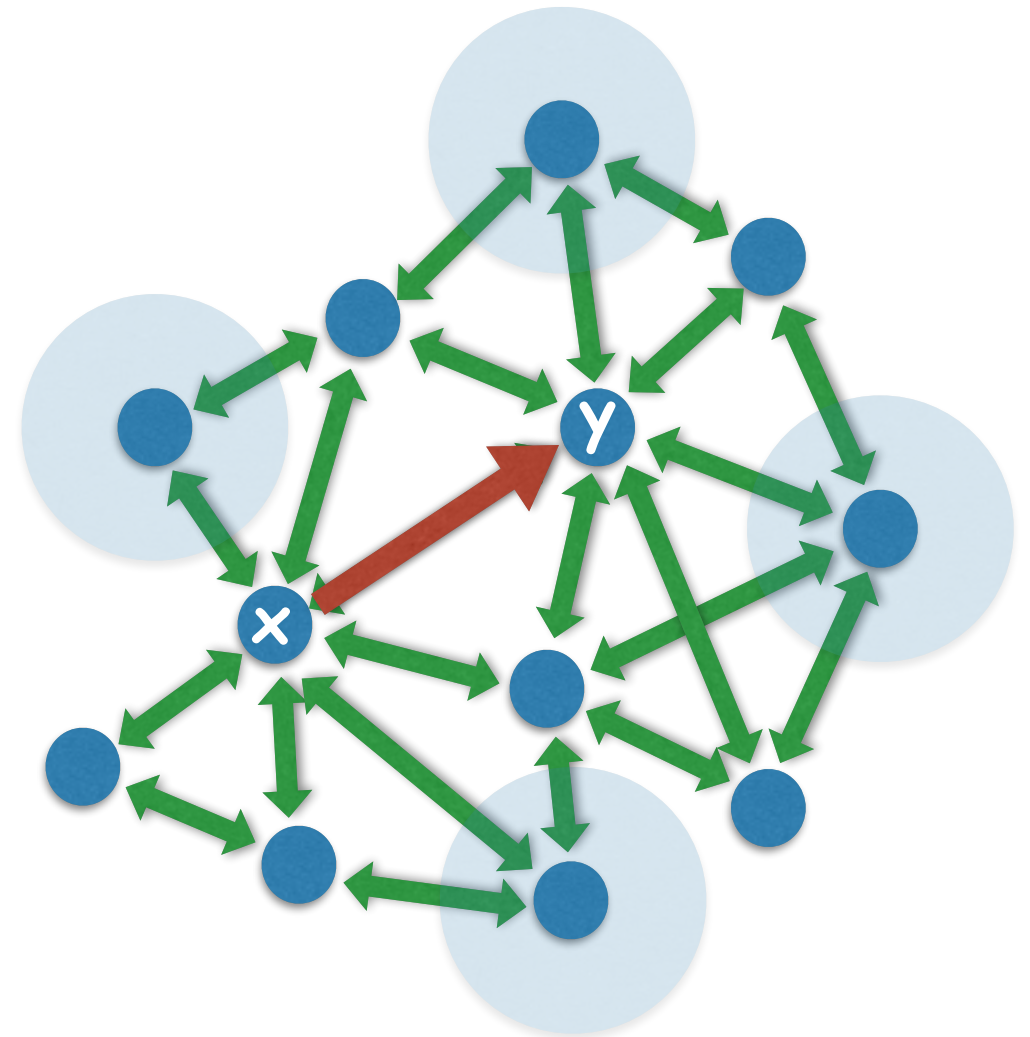
- **Collision/success:**

For any link (x, y) , a transmission from x is received by y in t if and only if

- » x transmits in t and

- » $\sum_{u \in V(t)} a(u, (x, y)) < 1$,

$V(t) \subseteq V$: set of nodes transmitting in t .



[1] Halldórsson and Wattenhofer. ICALP 2009.

[3] Fanghänel, Kesselheim and Vöcking. ICALP 2009.

[3] Kesselheim and Vöcking. DISC 2010.

Dynamic Multi-Broadcast [ACM-FOMC 2014]

Introduction

Dynamic Multiple-Message Broadcast (MMB) [1]:

- *problem:*
packets arrive at some nodes **continuously**, to be delivered to **all** nodes
- *metric:*
competitive throughput of deterministic distributed MMB algorithms
- *analysis:*
in the **Affectance model**:
 - Affectance subsumes many interference models, e.g. RN and SINR models
 - conceptual idea: parameterize interference from transmitting **nodes into links**
 - introduced [2,3,4] for link scheduling as link-to-link affectance

[1] (non-dynamic MMB) Khabbazzian-Kowalski PODC 2011

[2] Halldórsson-Wattenhofer, ICALP 2009

[3] Kesselheim, PODC 2012

[4] Kesselheim-Vöcking, DISC 2010

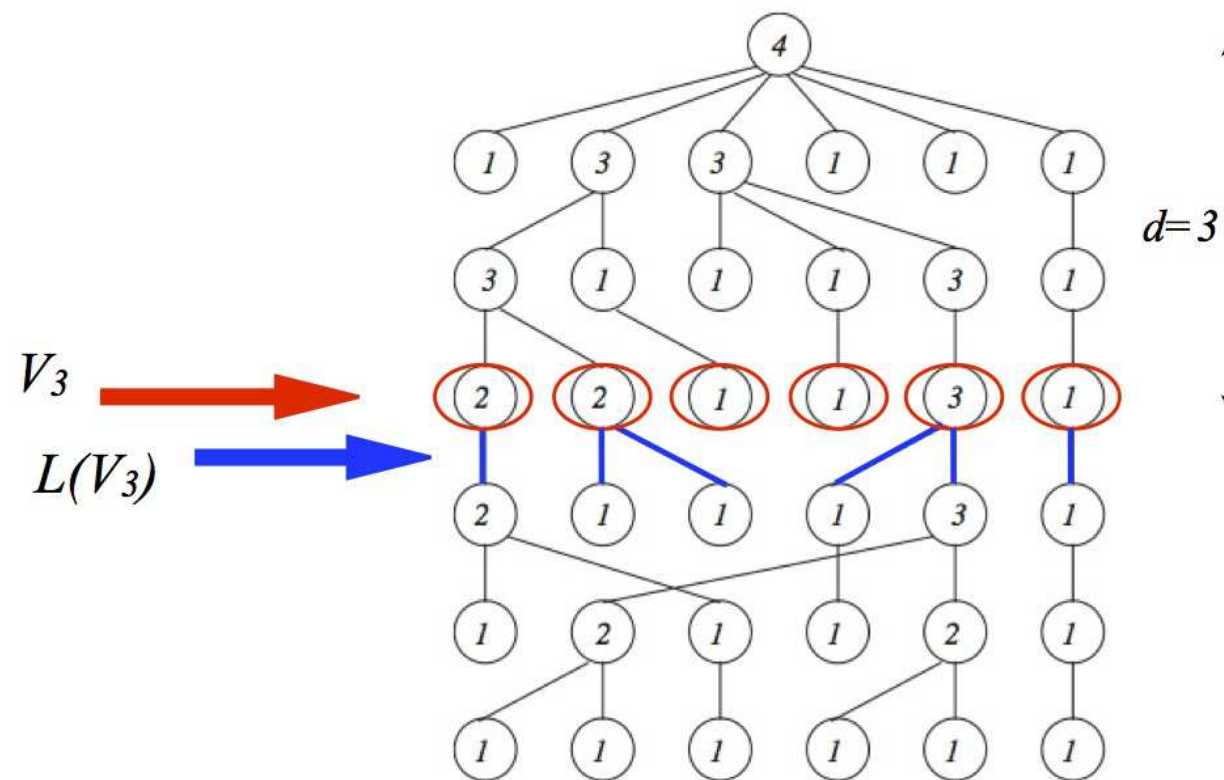
Dynamic Multi-Broadcast [ACM-FOMC 2014]

Affectance Characterization

Maximum average tree-layer affectance

Quantifies the difficulty to disseminate from one layer to the next one.

$$K(T, s) = \max_d \max_{V' \subseteq V_d(T)} \frac{a_{V'}(L(V'))}{|L(V')|}$$



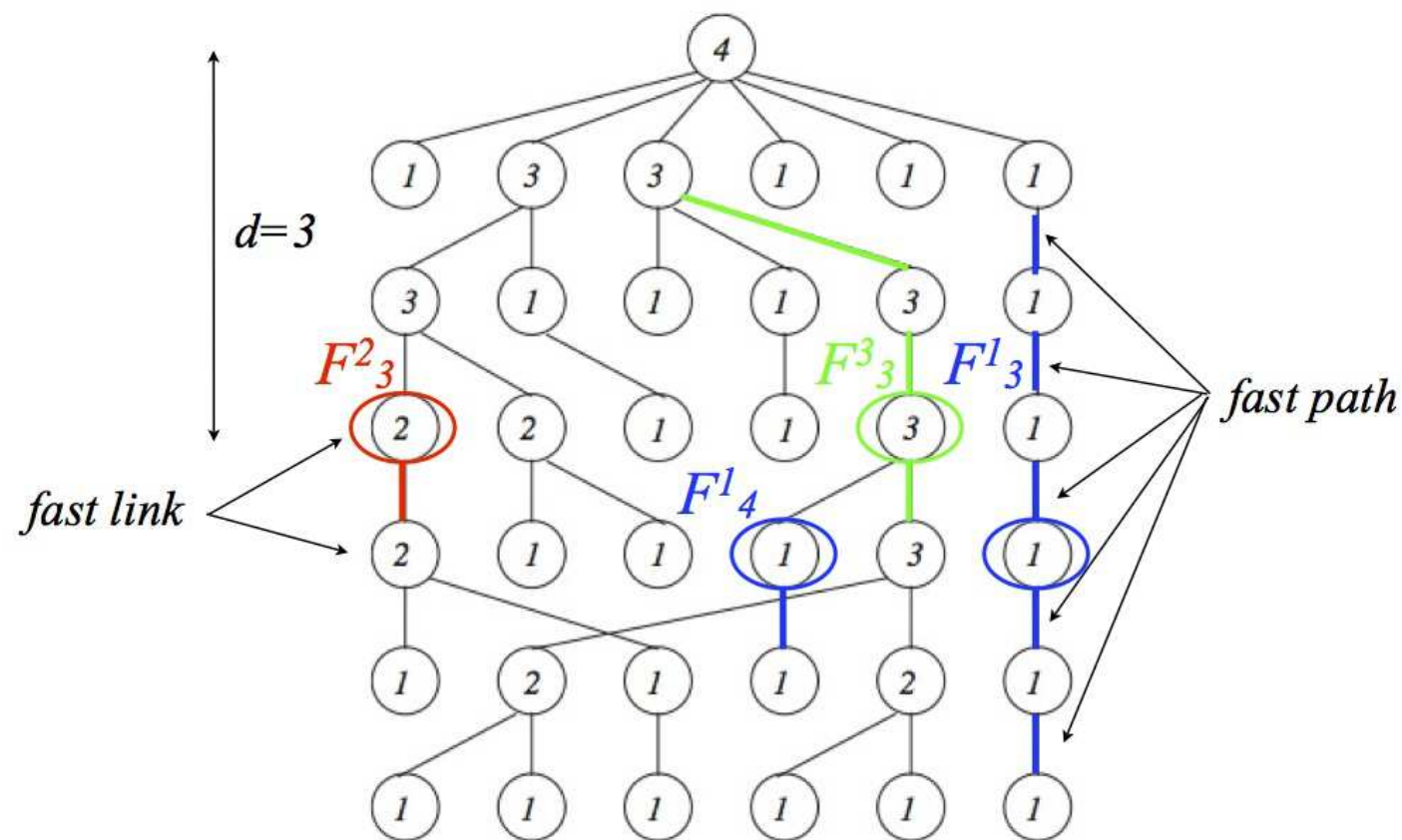
Dynamic Multi-Broadcast [ACM-FOMC 2014]

Affectance Characterization

Maximum fast-paths affectance

Quantifies the difficulty for dissemination on a path due to other paths.

$$M(T, s) = \max_{d,r} \max_{\ell \in F_d^r(T)} a_{F_d^r(T)}(\ell)$$



Dynamic Multi-Broadcast [ACM-FOMC 2014]

Introduction

Contributions:

- introduce new model characteristics:
(based on comm network, affectance function, and a chosen BFS tree)
 - maximum average tree-layer affectance K
 - maximum fast-paths affectance M
- show how these characteristics influence broadcast time complexity:
if one uses a specific BFS tree (GBST [1]) that minimizes $M(K + M)$
single broadcast can be done in time $D + O(M(K + M) \log^3 n)$
- extend this to dynamic packet arrival model and the MMB problem:
new MMB algorithm reaching throughput of $\Omega(1/(\alpha K \log n))$
- ... also simulations for RN

**Dissemination
bottleneck is from
layer to layer!!**

[1] Gašieniec-Peleg-Xin, DC 2007

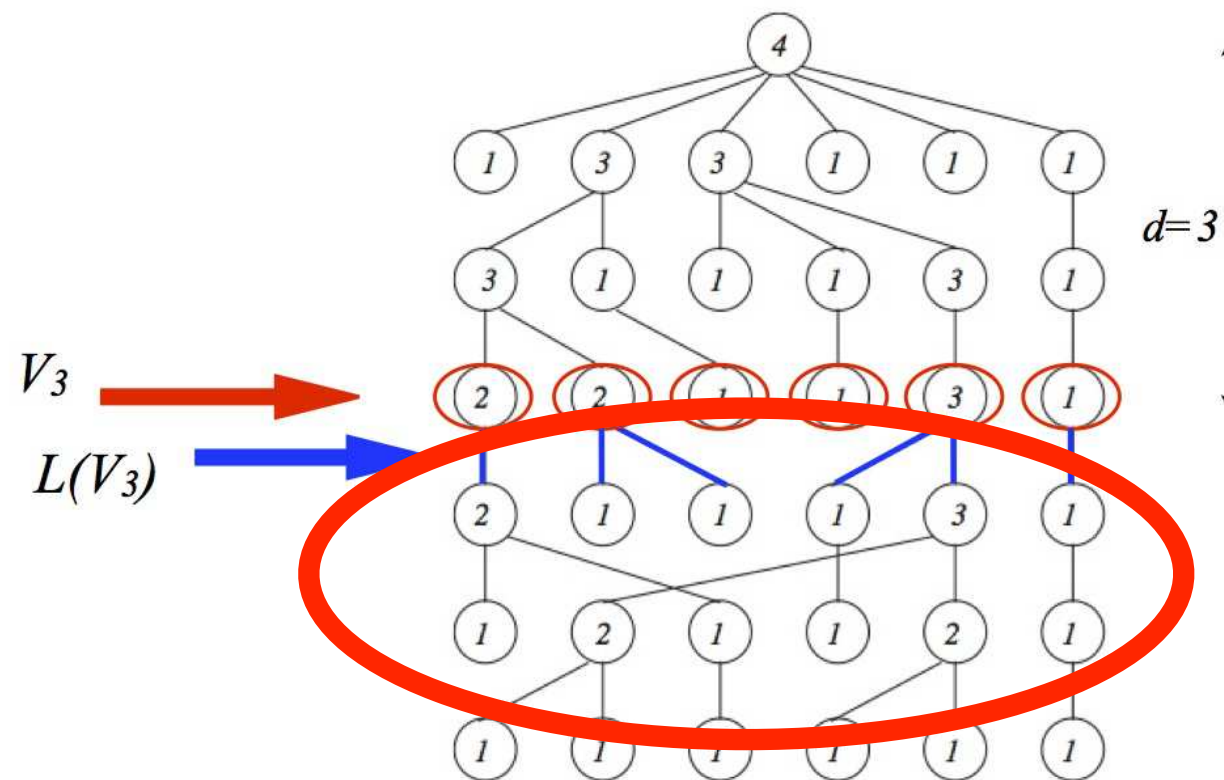
Dynamic Multi-Broadcast [ACM-FOMC 2014]

Affectance Characterization

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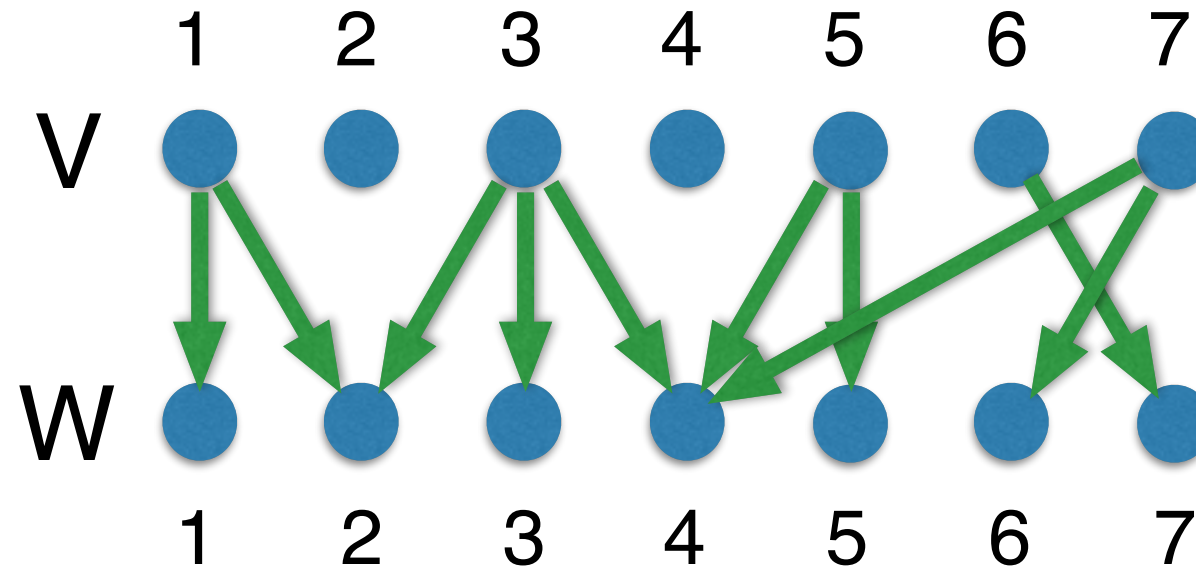
Quantifies the difficulty to disseminate from one layer to the next one.

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Each layer is a bipartite graph.

Layer Dissemination [SEA 2017]



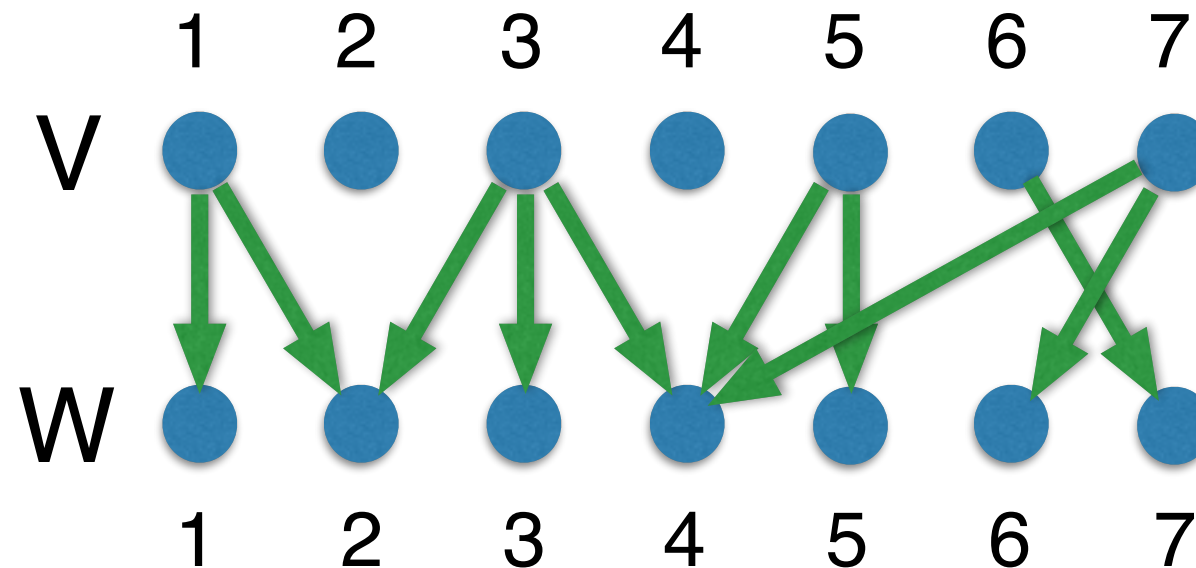
Bipartite network with

- V : set of transmitters
- W : set of receivers
- F_w : set of transmitters connected to $w \in W$

Layer Dissemination problem:

- Each $w \in W$ must receive at least one successful transmission from some $v \in F_w$, despite interference.

Layer Dissemination



INPUT: Affectance matrix $A=[a(u,(v,w))]$ and a family $F = \{F_w \mid w \in W\}$ of subsets of transmitters connected to each receiver:

$F_1=\{1\}$

$F_2=\{1,3\}$

$F_3=\{3\}$

$F_4=\{3,5,7\}$

...

OUTPUT: Family $S = \{S_t \mid t=1,2,3,\dots\}$ of subsets of transmitters transmitting in each time slot:

$S_1=\{1,5,7\}$

$S_2=\{2,3,4,6\}$

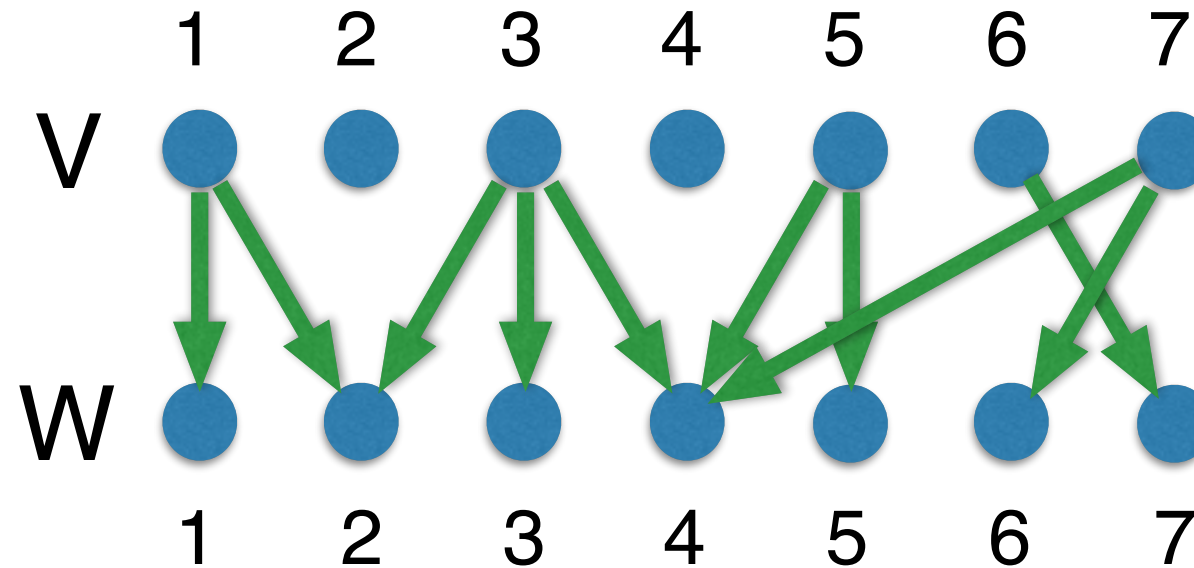
$S_3=\{1,4,7\}$

$S_4=\{2,5\}$

...

Affectance-selective Families

Transmissions
schedule



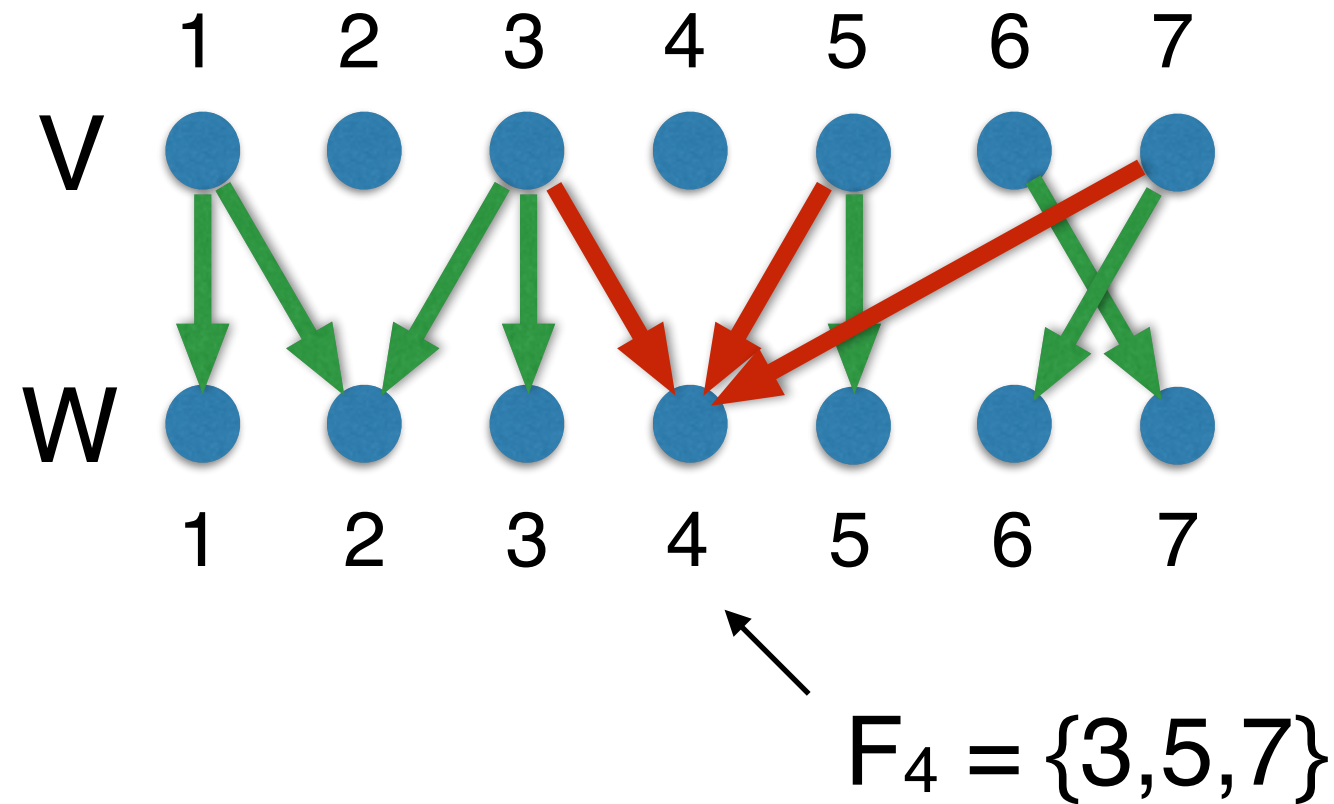
A family $\mathcal{S} = \{S_1, S_2, \dots, S_t\}$ of subsets of $[n]$ is *affectance-selective* on the family $\mathcal{F} = \{F_1, F_2, \dots, F_n\}$ of subsets of $[n]$ if and only if, for each $w \in [n]$, there exists $j \in [t]$ such that:

- $|F_w \cap S_j| \geq 1$, and
- for some $v \in (F_w \cap S_j)$ it is $\sum_{u \in S_j} a(u, (v, w)) < 1$.

Subsets of
transmitters

We say that the family \mathcal{S} has *length* t , and that each w is *affectance-selected*.

Layer Dissemination



Bound family
size on network
characteristic

Maximum Average Affectance:

$$\overline{A} = \max_{w \in [n]} \max_{F \subseteq F_w} \sum_{v \in F} \sum_{u \in [n]} a(u, (v, w)) / |F|.$$

Layer Dissemination

Existence of Aff-selective families:

► **Theorem 1.** For any $n > 0$, consider a family $\mathcal{F} = \{F_1, F_2, \dots, F_n\}$ of subsets of integers in $[n]$ and any affectance matrix A defined on \mathcal{F} . For each $w \in [n]$, let $\bar{A}_w = \max_{F \subseteq F_w} \sum_{v \in F} \sum_{u \in [n]} a(u, (v, w)) / |F|$ be the maximum average affectance on w . If there exists a constant $c > 1$ such that $\bar{A}_w \leq c|F_w|$ for all $w \in [n]$, then, there exists a family $\mathcal{S} = \{S_1, S_2, \dots, S_s\}$ that is affectance-selective on \mathcal{F} , and its size s satisfies

$$s \in O(1 + \log n \log \bar{A}),$$

where $\bar{A} = \max_{w \in [n]} \bar{A}_w$ is the maximum average affectance.

Logarithmic in the
network
characterization.

Layer Dissemination

Sketch of proof:

- Assume each v transmits with some probability p .
- Probabilistic method: show that the probability of a given w not being selected is < 1 .
- Prove Markov-type inequality: to bound such probability by the expected average affectance on w .
- Prob is < 1 if p is within a constant-factor b range \rightarrow try all.
- Union bound: after enough number of rounds, the probability of any w not being selected is still less than 1 \rightarrow add some multiplicity m .

We redefine \mathcal{S} as the family $\{S_{i,j}\}$ of subsets of $[n]$ where the set $S_{i,j}$ is obtained including each $v \in [n]$ in $S_{i,j}$ independently with probability $p = 1/b^i$, for each $i = 0, 1, 2, \dots, \max\{\lceil \log_b(2\bar{A}) \rceil, 0\}$ and each $j = 1, 2, \dots, m$.

$$\Pr(\exists w \in [n] : Z_w = 0) \leq nd^m \quad \longrightarrow \quad s \in O(1 + \log n \log \bar{A})$$

Layer Dissemination

Proof of Thm 1 yields a **randomized** protocol of same length:

```
1  $b \leftarrow 1 + 1/(2c)$ 
2  $m \leftarrow \lceil 2 \log_{1/d} n \rceil$ 
3 for  $i = 0, 1, 2, \dots, \max\{\lceil \log_b(2\bar{A}) \rceil, 0\}$  do
4   |   for  $m$  times do
5   |   |   transmit with probability  $1/b^i$ 
```

Algorithm 1: Randomized Layer Dissemination protocol for each node $v \in$

Theorem 2 Consider a layer of a Radio Network with affectance matrix A and topology $G = (V, W, E)$, where $|V| = |W| = n$, where for each receiver $w \in W$ there is at least one transmitter $v \in V$ such that $(v, w) \in E$. Then, if there exists a constant $c > 1$ such that $\bar{A}_w \leq c|F_w|$ for all $w \in W$, where $\bar{A}_w = \max_{F \subseteq F_w} \sum_{v \in F} \sum_{u \in V} a(u, (v, w)) / |F|$ is the maximum average affectance on w , Algorithm 1 solves the Layer Dissemination problem with high probability¹, and the running time is in $O(1 + \log n \log \bar{A})$, where $\bar{A} = \max_{w \in W} \bar{A}_w$ is the maximum average affectance.

Layer Dissemination

De-randomization yields a **deterministic** protocol of same length:

```
// Initialization
1  $p \leftarrow 0$ 
2  $b \leftarrow 1 + 1/(2c)$ 
3  $m \leftarrow \max\{\lceil \log_b(2\bar{A}) \rceil, 0\}$ 
4  $W'_0 \leftarrow \{w \in W : \bar{A}_w \leq 1/2\}$ 
5 for  $r = 1, \dots, m$  do  $W'_r \leftarrow \{w \in W : b^{r-1}/2 < \bar{A}_w \leq b^r/2\}$ 

// Protocol
6 for each time slot while  $\exists r = 0, 1, \dots, m : W'_r \neq \emptyset$  do
7   if  $p \leq 1/(2b\bar{A})$  then
8      $p \leftarrow 1$ 
9      $r \leftarrow 0$ 
10  set  $V'[1 \dots n]$  array of booleans //  $V'[i] \equiv i$  transmits
11  for  $i = 1, 2, \dots, n$  do
12     $\mathbb{E}_{true} \leftarrow \mathbb{E}_{V'[i+1 \dots n]} (\# \text{ selected in } W'_r | V'[i] = true)$ 
13     $\mathbb{E}_{false} \leftarrow \mathbb{E}_{V'[i+1 \dots n]} (\# \text{ selected in } W'_r | V'[i] = false)$ 
14     $V'[i] \leftarrow \mathbb{E}_{true} > \mathbb{E}_{false}$ 
15  if  $V'[v]$  then transmit
16   $W'_r \leftarrow W'_r \setminus \{w | w \text{ was selected}\}$ 
17   $p \leftarrow p/b$ 
18   $r \leftarrow r + 1$ 
```

Algorithm 2: Deterministic Layer Dissemination protocol for each node $v \in$

... but computing those expectations is exponential,
due to computing probs of low affectance.



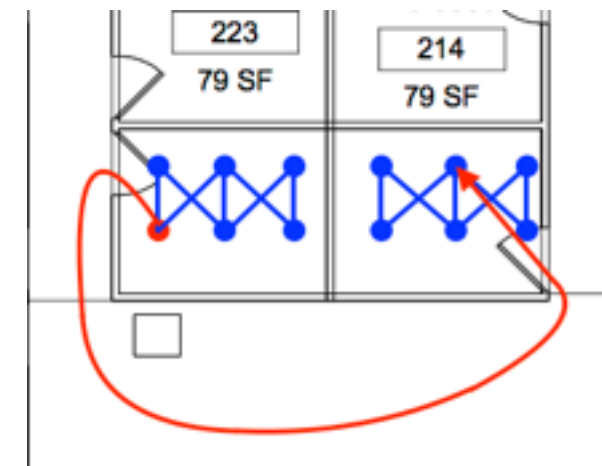
So?

Are affectance-based protocols better? worse?

Let's try some experiments!



Simulations topology



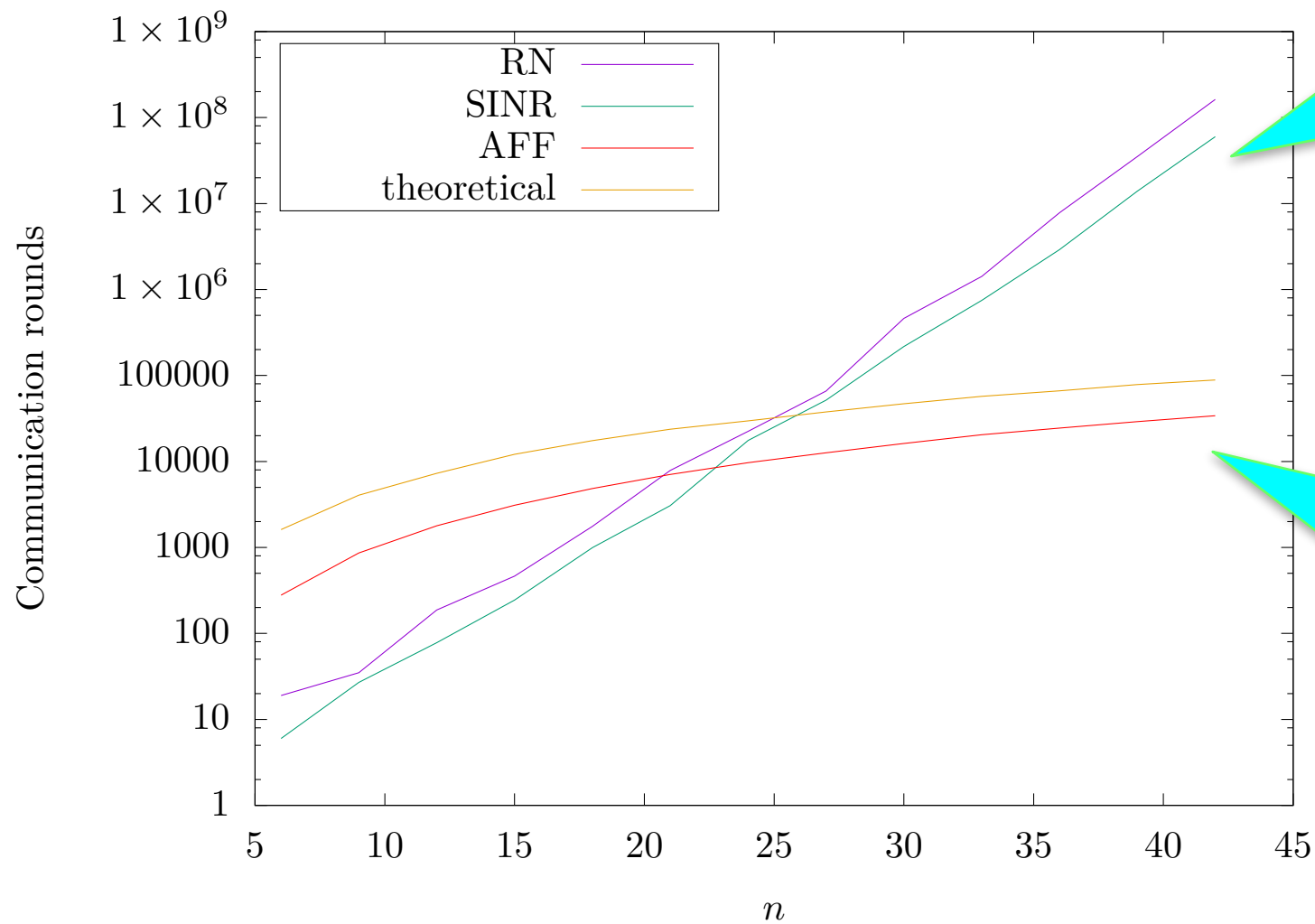
Layer Dissemination

Simulation protocols:

- **Transform Montecarlo into Las Vegas protocols:**
 - count how many rounds to complete dissemination.
- **Affectance model evaluation:**
 - Compare performance with an RN protocol and and SINR protocol.
 - Successful transmission according to Affectance model.
 - Compare also with theoretical performance.

Layer Dissemination

Simulations results



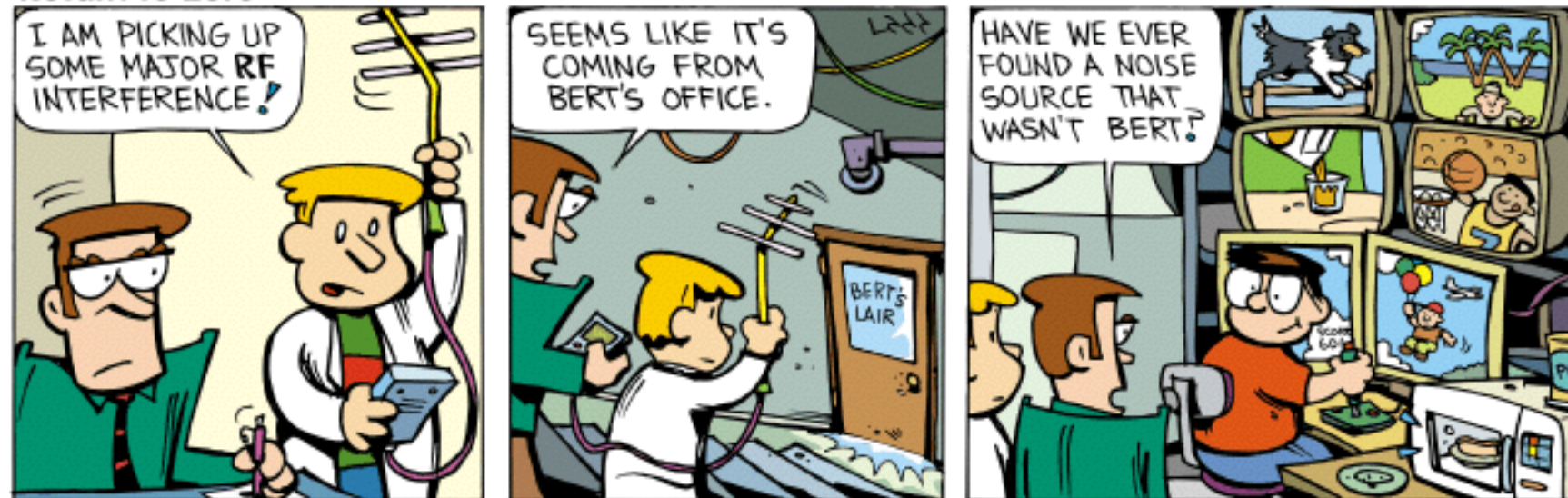
RN & SINR are exponential under affectance!

Better than theoretical bound!

Thank you!

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Return to Zero



EEWeb.com