

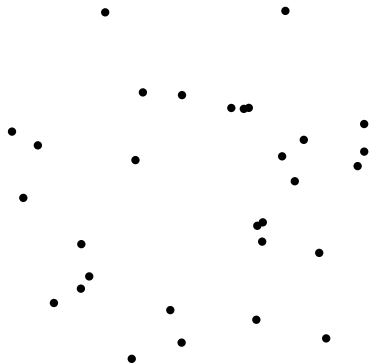
Ad Hoc Wireless Networks: Information-Theoretic Optimal Operating Regimes

Ayfer Özgür (EPFL), Ramesh Johari (Stanford), David Tse (Berkeley),
Olivier Lévêque (EPFL)

April 4, Spaswin 08

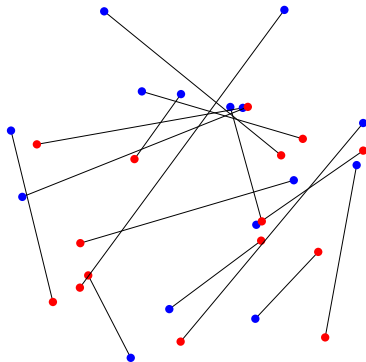
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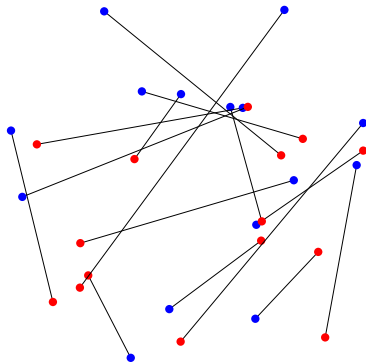
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- a set of random connections (logical links) that need to be established, each carrying data at rate $R(n)$
- What is the scaling of the aggregate throughput $T(n) = nR(n)$ with increasing number of nodes n ?



Communication Model

$$Y_i = \sum_k H_{ik} X_k + Z_i$$

where:

- Z_i is additive Gaussian noise with variance σ^2
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- **Related Question:**

What is the “right” scaling to consider for $A(n)$?
 $A(n) = 1$ (dense model)? $A(n) = n$ (extended model)?

Key Observation

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- How can one then talk about scaling laws for wireless networks?
- One needs to choose a model which remains valid in the range of interest (say, e.g., $n \leq 10'000$).

Another Issue about Scaling Laws: The Example of MIMO systems

- Foschini-Gans / Telatar, 95: the capacity of a MIMO system scales linearly with the number of antennas at Tx and Rx: $C(n) = O(n)$.

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- So if $A \sim n$ (constant density), then $C(n) = O(\sqrt{n})$?
- **The question is**: for which values of n does this limitation kick in?
- **And the answer is**: for a highly dense multiple antenna system only! (before this, the assumption of i.i.d. channel coefficients is reasonable)

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Numerical Example for a Wireless Network

- network area: $A = 1 \text{ km}^2 = 10^6 \text{ m}^2$

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- \Rightarrow area limitation? $C \leq \sqrt{10^8} = 10'000 = n$: **not a limitation!**

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- no final decision for the scaling of $A(n)$ yet

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Dense Networks ($A(n) = 1$):

The optimal aggregate throughput scales as $T(n) = n$.

More precisely:

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Extended Networks ($A(n) = n$):

The optimal aggregate throughput scales as

$$T(n) = \begin{cases} n^{2-\alpha/2} & 2 \leq \alpha < 3 \\ n^{1/2} & \alpha \geq 3 \end{cases}$$

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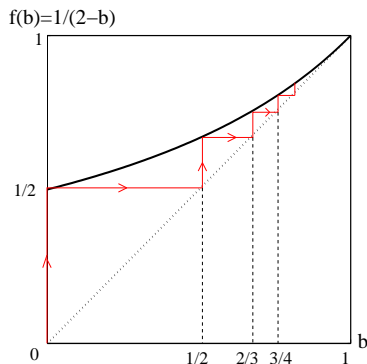
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- Key ideas:
 - ▶ **Distributed MIMO**: pool the nodes of the network so as to form virtual multiple antenna arrays.
 - ▶ **Hierarchical Cooperation**: set up cooperation progressively between the nodes, via a hierarchical structure.

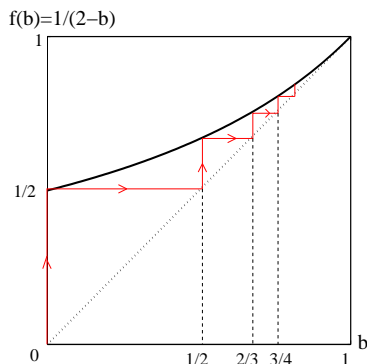
Dense Networks: Hierarchical Cooperation

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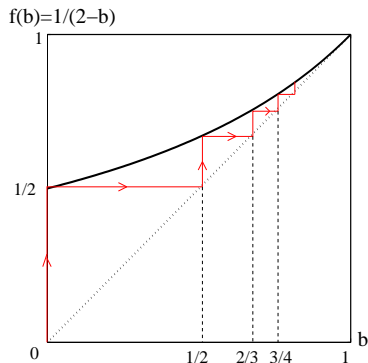
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$T(n) = n^{h/(h+1)}$ after h iterations and near network-wide distributed MIMO is achieved (i.e. $n^{h/(h+1)}$ simultaneous **long range** transmissions!)

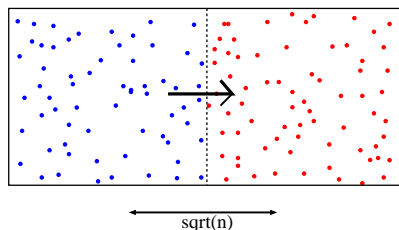


Extended Networks: Proof Idea of the Upper Bound

- Another key limiting factor in extended networks: **power**

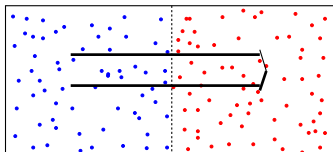
Extended Networks: Proof Idea of the Upper Bound

- Another key limiting factor in extended networks: **power**
- Cut-set bound argument: the aggregate throughput is limited by the total power received on one side of the cut.



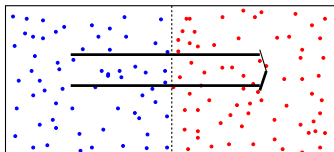
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- When $\alpha < 3$, the total received power is $n^{2-\alpha/2}$, dominated by transfer between **all** the users: hierarchical cooperation is optimal.

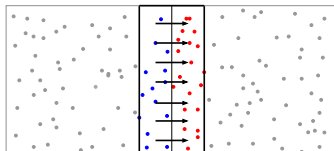


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- When $\alpha > 3$, the total received power is \sqrt{n} , dominated by transfer between the **boundary** users: multi-hop is optimal.



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Beyond “Dense vs Extended Networks”

- Intermediate regime: $A(n) = n^\gamma$
- For $\alpha > 3$ and $0 < \gamma < 1$, it turns out that **neither hierarchical cooperation nor multihop is the optimal strategy** (in terms of throughput scaling).
- But a combination of these achieves the optimal throughput scaling:
 “cooperate locally, multi-hop globally”
(the optimal size of the cooperative clusters being determined by the parameters).

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- These parameters are of order:

$$\text{SNR}_s = n^\beta \quad \text{and} \quad \text{SNR}_l = n^{1-\alpha/2+\beta}$$

where $\beta = (1 - \gamma)\alpha/2$.

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- $\alpha > 3$, $\text{SNR}_I \ll 0$ dB and $\text{SNR}_S \gg 0$ dB: a combination of hierarchical cooperation and multi-hop is optimal

Summary: Optimal Operating Regimes

