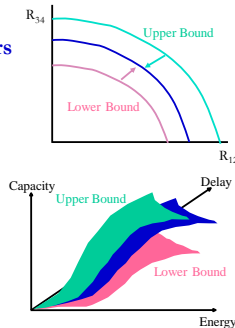


EE360: Lecture 10 Outline Capacity of Ad Hoc Nets

- **Announcements**
 - Revised proposals due tomorrow
 - HW 1 posted, due Feb. 24 at 5pm
- Definition of ad hoc network capacity
- Capacity regions
- Scaling laws and extensions
- Achievable rate regions
- Capacity under cooperation
- Interference alignment
- Cross layer design

Network Capacity: *What is it?*

- **n(n-1)-dimensional region**
 - Rates between all node pairs
 - Upper/lower bounds
 - Lower bounds achievable
 - Upper bounds hard
- Other possible axes
 - Energy and delay

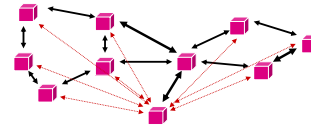


Some capacity questions

- How to parameterize the region
 - Power/bandwidth
 - Channel models and CSI
 - Outage probability
 - Security/robustness
- Defining capacity in terms of asymptotically small error and infinite delay has been highly enabling
 - Has also been limiting
 - Cause of unconsummated union in networks and IT
 - What is the alternative?



Ad-Hoc Network Capacity








- Fundamental limits on the maximum possible rates between all possible node pairs with vanishing probability of error
- Independent of transmission and reception strategies (modulation, coding, routing, etc.)
- Dependent on propagation, node capabilities (e.g. MIMO), transmit power, noise, etc

Fundamental Network Capacity *The Shangri-La of Information Theory*

- Much progress in finding the capacity limits of wireless single and multiuser channels
- Limited understanding about the capacity limits of wireless networks, even for simple models
- System assumptions such as constrained energy and delay may require new capacity definitions
- Is this elusive goal the right thing to pursue?

Shangri-La is synonymous with any earthly paradise; a permanently happy land, isolated from the outside world

Network Capacity Results

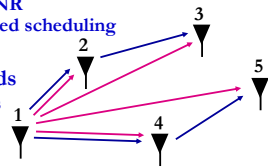
- Multiple access channel (MAC)  Gallager
- Broadcast channel  Cover & Bergmans
- Relay channel upper/lower bounds  Cover & El Gamal
- Strong interference channel  Sato, Han & Kobayashi
- Scaling laws  Gupta & Kumar
- Achievable rates for small networks

Capacity for Large Networks (Gupta/Kumar'00)

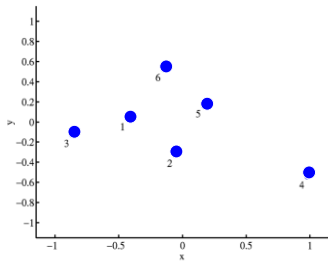
- Make some simplifications and ask for less
 - Each node has only a single destination
 - All nodes create traffic for their desired destination at a uniform rate λ
 - Capacity (throughput) is maximum λ that can be supported by the network (1 dimensional)
- Throughput of random networks
 - Network topology/packet destinations random.
 - Throughput λ is random: characterized by its distribution as a function of network size n.
- Find scaling laws for $C(n)=\lambda$ as $n \rightarrow \infty$.

Ad Hoc Network Achievable Rate Regions

- All achievable rate vectors between nodes
 - Lower bounds Shannon capacity
- An $n(n-1)$ dimensional convex polyhedron
 - Each dimension defines (net) rate from one node to each of the others
 - Time-division strategy
 - Link rates adapt to link SINR
 - Optimal MAC via centralized scheduling
 - Optimal routing
- Yields performance bounds
 - Evaluate existing protocols
 - Develop new protocols



Example: Six Node Network



Capacity region is 30-dimensional

Extensions

- Fixed network topologies (Gupta/Kumar'01)
 - Similar throughput bounds as random networks
- Mobility in the network (Grossglauser/Tse'01)
 - Mobiles pass message to neighboring nodes, eventually neighbor gets close to destination and forwards message
 - Per-node throughput constant, aggregate throughput of order n, delay of order n.
- Throughput/delay tradeoffs
 - Piecewise linear model for throughput-delay tradeoff (ElGamal et. al'04, Toumpis/Goldsmith'04)
 - Finite delay requires throughput penalty.
- Achievable rates with multiuser coding/decoding (GK'03)
 - Per-node throughput (bit-meters/sec) constant, aggregate infinite.
 - Rajiv will provide more details



Achievable Rates

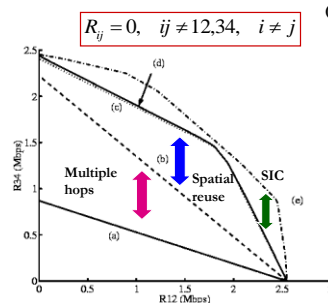
Achievable rate vectors achieved by time division \rightarrow Capacity region is convex hull of all rate matrices

- A matrix R belongs to the capacity region if there are rate matrices $R_1, R_2, R_3, \dots, R_n$ such that

$$R = \sum_{i=1}^n \alpha_i R_i; \quad \sum_{i=1}^n \alpha_i \leq 1; \alpha_i > 0$$

- Linear programming problem:
 - Need clever techniques to reduce complexity
 - Power control, fading, etc., easily incorporated
 - Region boundary achieved with optimal routing

Capacity Region Slice (6 Node Network)

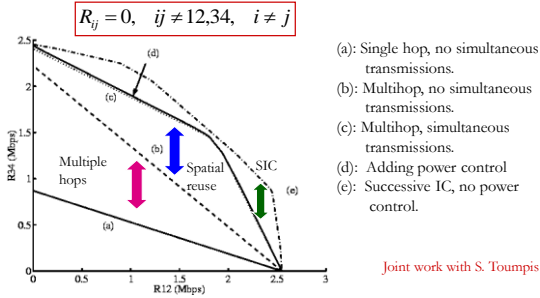


- (a): Single hop, no simultaneous transmissions.
- (b): Multihop, no simultaneous transmissions.
- (c): Multihop, simultaneous transmissions.
- (d): Adding power control
- (e): Successive interference cancellation, no power control.

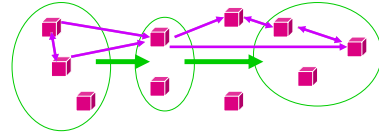
Extensions:

- Capacity vs. network size
- Capacity vs. topology
- Fading and mobility
- Multihop cellular

Achievable Region Slice (6 Node Network)

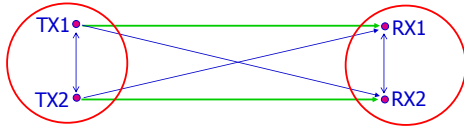


Cooperation in Wireless Networks



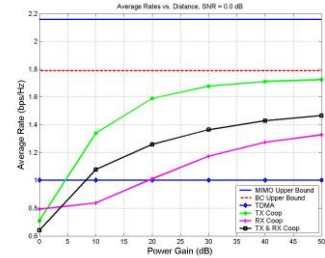
- Routing is a simple form of cooperation
- Many more complex ways to cooperate:
 - Virtual MIMO, generalized relaying, interference forwarding, and one-shot/iterative conferencing
- Many theoretical and practice issues:
 - Overhead, forming groups, dynamics, synch, ...

Virtual MIMO



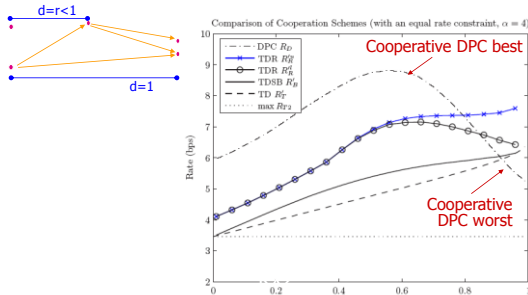
- TX1 sends to RX1, TX2 sends to RX2
- TX1 and TX2 cooperation leads to a MIMO BC
- RX1 and RX2 cooperation leads to a MIMO MAC
- TX and RX cooperation leads to a MIMO channel
- Power and bandwidth spent for cooperation

Capacity Gain with Cooperation (2x2)



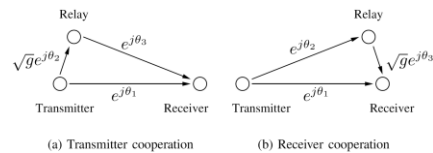
- TX cooperation needs large cooperative channel gain to approach broadcast channel bound
- MIMO bound unapproachable

Capacity Gain vs Network Topology



Optimal cooperation coupled with access and routing

Relative Benefits of TX and RX Cooperation

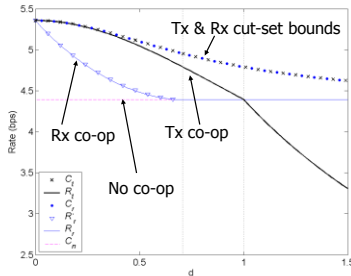


- Two possible CSI models:
 - Each node has full CSI (synchronization between Tx and relay).
 - Receiver phase CSI only (no TX-relay synchronization).
- Two possible power allocation models:
 - Optimal power allocation: Tx has power constraint aP , and relay $(1-a)P$; $0 \leq a \leq 1$ needs to be optimized.
 - Equal power allocation ($a = 1/2$).

Joint work with C. Ng

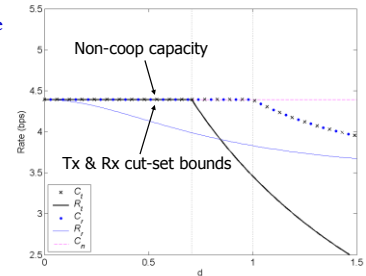
Example 1: Optimal power allocation with full CSI

- Cut-set bounds are equal.
- Tx co-op rate is close to the bounds.
- Transmitter cooperation is preferable.



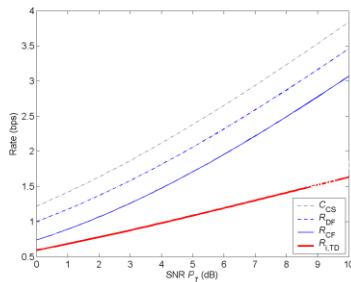
Example 2: Equal power allocation with RX phase CSI

- Non-cooperative capacity meets the cut-set bounds of Tx and Rx co-op.
- Cooperation offers no capacity gain.



Capacity: Non-orthogonal Relay Channel

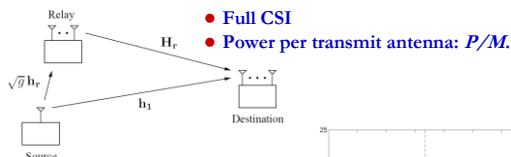
- Compare rates to a full-duplex relay channel.
- Realize conference links via time-division.
- Orthogonal scheme suffers a considerable performance loss, which is aggravated as SNR increases.



Transmitter vs. Receiver Cooperation

- Capacity gain only realized with the right cooperation strategy
- With full CSI, Tx co-op is superior.
- With optimal power allocation and receiver phase CSI, Rx co-op is superior.
- With equal power allocation and Rx phase CSI, cooperation offers no capacity gain.
- Similar observations in Rayleigh fading channels.

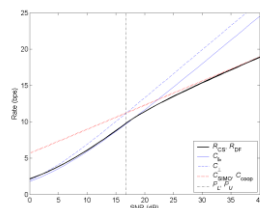
Multiple-Antenna Relay Channel



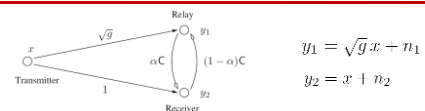
- Full CSI
- Power per transmit antenna: P/M .

- Single-antenna source and relay
- Two-antenna destination
 - $\text{SNR} < P_r$: MIMO Gain
 - $\text{SNR} > P_r$: No multiplexing gain; can't exceed SISO channel capacity (Host-Madsen'05)

Joint work with C. Ng and N. Laneman



Conferencing Relay Channel

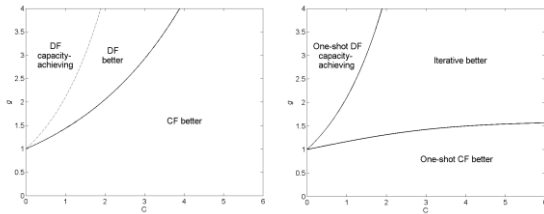


$$y_1 = \sqrt{g}x + n_1$$

$$y_2 = x + n_2$$

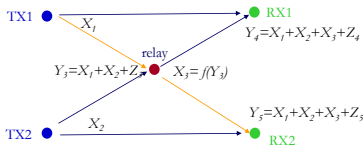
- Willems introduced conferencing for MAC (1983)
 - Transmitters conference before sending message
- We consider a relay channel with conferencing between the relay and destination
- The conferencing link has total capacity C which can be allocated between the two directions

Iterative vs. One-shot Conferencing



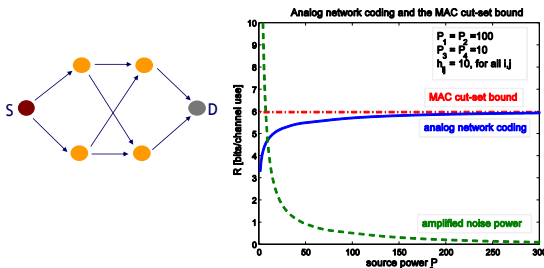
- Weak relay channel: the iterative scheme is disadvantageous.
- Strong relay channel: iterative outperforms one-shot conferencing for large C.

Generalized Relaying



- Can forward message and/or interference
 - Relay can forward all or part of the messages
 - Much room for innovation
 - Relay can forward **interference**
 - To help subtract it out

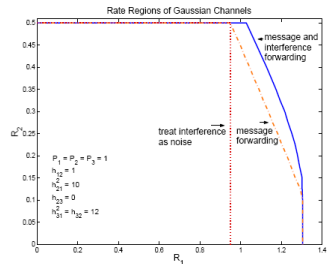
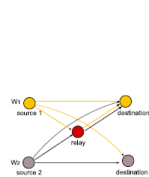
In fact, it can achieve capacity



Lessons Learned

- Orthogonalization has considerable capacity loss
 - Applicable for clusters, since cooperation band can be reused spatially.
- DF vs. CF
 - DF: nearly optimal when transmitter and relay are close
 - CF: nearly optimal when transmitter and relay far
 - CF: not sensitive to compression scheme, but poor spectral efficiency as transmitter and relay do not joint-encode.
- The role of SNR
 - High SNR: rate requirement on cooperation messages increases.
 - MIMO-gain region: cooperative system performs as well as MIMO system with isotropic inputs.

Beneficial to forward both interference and message



Interference Alignment

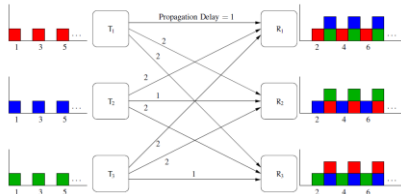
- Addresses the number of interference-free signaling dimensions in an interference channel
- Based on our orthogonal analysis earlier, it would appear that resources need to be divided evenly, so only $2BT/N$ dimensions available
- Jafar and Cadambe showed that by aligning interference, $2BT/2$ dimensions are available



- Everyone gets half the cake!

Basic Premise

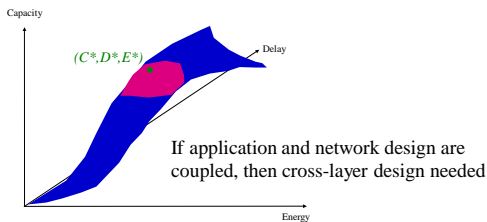
- For any number of TXs and RXs, each TX can transmit half the time and be received without any interference
 - Assume different delay for each transmitter-receiver pair
 - Delay odd when message from TX i desired by RX j even otherwise.
 - Each TX transmits during odd time slots and is silent at other times.
 - All interference is aligned in even time slots.



Is a capacity region all we need to design networks?

Yes, if the application and network design can be decoupled

Application metric: $f(C, D, E)$; $(C^*, D^*, E^*) = \arg \max f(C, D, E)$



Consummating Unions

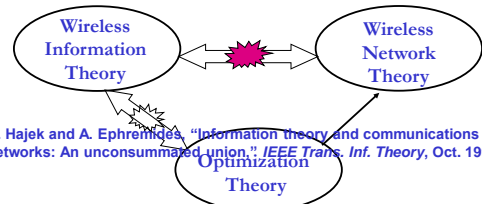


- When capacity is not the only metric, a new theory is needed to deal with nonasymptopia (i.e. delay, random traffic) and application requirements
 - Shannon theory generally breaks down when delay, error, or user/traffic dynamics must be considered
- Fundamental limits are needed outside asymptotic regimes
- Optimization, game theory, and other techniques provide the missing link

Extensions

- Multipath channels
- Fading channels
- MIMO channels
- Cellular systems
- Imperfect channel knowledge
- ...

Limitations in theory of ad hoc networks today

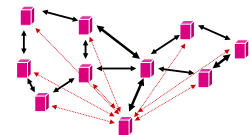


B. Hajek and A. Ephremides, "Information theory and communications networks: An unconsummated union," *IEEE Trans. Inf. Theory*, Oct. 1998.

- Shannon capacity pessimistic for wireless channels and intractable for large networks
- Large body of wireless (and wired) network theory that is ad-hoc, lacks a basis in fundamentals, and lacks an objective success criteria.
- Little cross-disciplinary work spanning these fields
- Optimization techniques applied to given network models, which rarely take into account fundamental network capacity or dynamics

Crosslayer Design in Ad-Hoc Wireless Networks

- Application
- Network
- Access
- Link
- Hardware

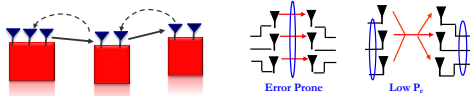


Substantial gains in throughput, efficiency, and end-to-end performance from cross-layer design

Why a crosslayer design?

- The technical challenges of future mobile networks cannot be met with a layered design approach.
- QoS cannot be provided unless it is supported across all layers of the network.
 - The application must adapt to the underlying channel and network characteristics.
 - The network and link must adapt to the application requirements
- Interactions across network layers must be understood and exploited.

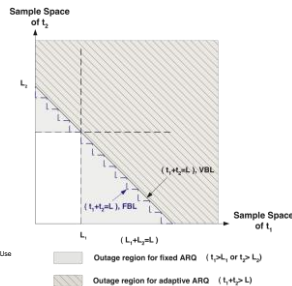
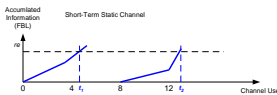
Diversity-Multiplexing-Delay Tradeoffs for MIMO Multihop Networks with ARQ



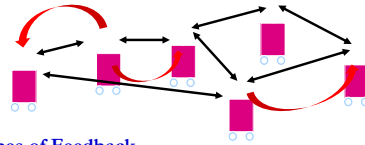
- MIMO used to increase data rate or robustness
- Multihop relays used for coverage extension
- ARQ protocol:
 - Can be viewed as 1 bit feedback, or time diversity,
 - Retransmission causes delay (can design ARQ to control delay)
- Diversity multiplexing (delay) tradeoff - DMT/DMDT
 - Tradeoff between robustness, throughput, and delay

Asymptotic DMDT Optimality

- Theorem: VBL ARQ achieves optimal DMDT in MIMO multihop relay networks in long-term and short-term static channels.
- Proved by cut-set bound
- An intuitive explanation by stopping times: VBL ARQ has the smaller outage regions among multihop ARQ protocols



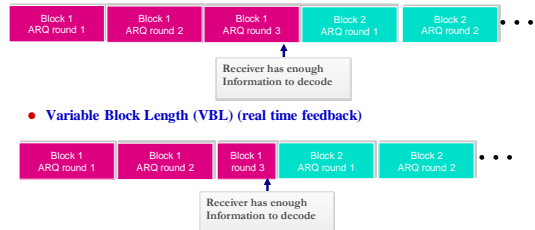
How to use Feedback in Wireless Networks



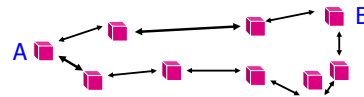
- Types of Feedback
 - Output feedback
 - CSI
 - Acknowledgements
 - Network/traffic information
 - Something else
- What is the metric to be improved by feedback
 - Capacity
 - Delay
 - Other

Multihop ARQ Protocols

- Fixed ARQ: fixed window size
 - Maximum allowed ARQ round for i th hop L_i satisfies $\sum_{i=1}^N L_i \leq L$
- Adaptive ARQ: adaptive window size
 - Fixed Block Length (FBL) (block-based feedback, easy synchronization)

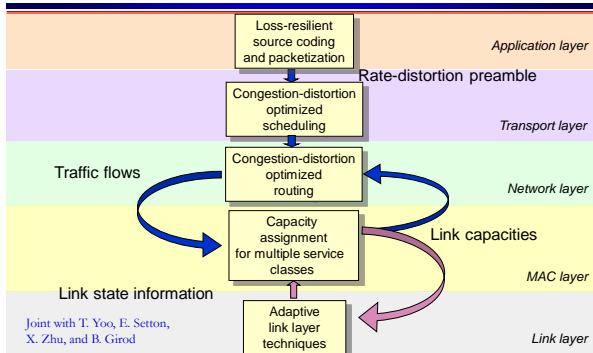


Delay/Throughput/Robustness across Multiple Layers

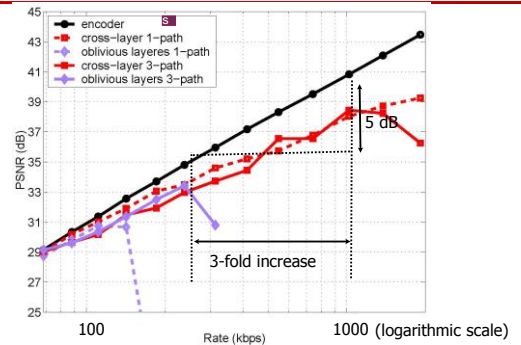


- Multiple routes through the network can be used for multiplexing or reduced delay/loss
- Application can use single-description or multiple description codes
- Can optimize optimal operating point for these tradeoffs to minimize distortion

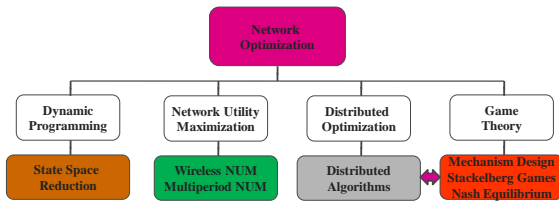
Cross-layer protocol design for real-time media



Video streaming performance



Approaches to Cross-Layer Resource Allocation*



*Much prior work is for wired/static networks

Network Utility Maximization

- Maximizes a network utility function

$$\max \sum_k U_k(\bar{r}_k)$$

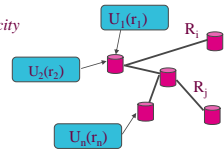
flow k

$$s.t. \quad Ar \leq R$$

routing *Fixed link capacity*

- Assumes

- Steady state
- Reliable links
- Fixed link capacities

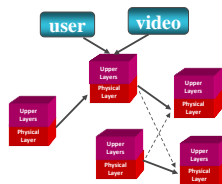


- Dynamics are only in the queues

Wireless NUM

- Extends NUM to random environments
- Network operation as stochastic optimization algorithm

$$\begin{aligned} \max \quad & E[\sum U(r_m(G))] \\ \text{st} \quad & E[r(G)] \leq E[R(S(G), G)] \\ & E[S(G)] \leq \bar{S} \end{aligned}$$

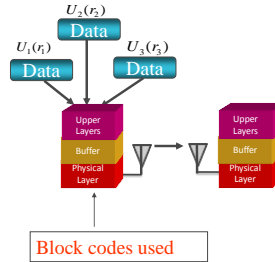


WNUM Policies

- Control network resources
- Inputs:
 - Random network channel information G^k
 - Network parameters
 - Other policies
- Outputs:
 - Control parameters
 - Optimized performance, that
 - Meet constraints
- Channel sample driven policies

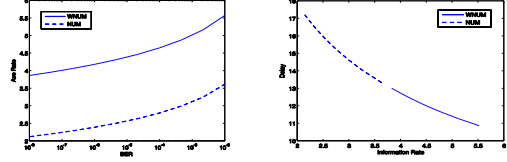
Example: NUM and Adaptive Modulation

- Policies
 - Information rate
 - Tx power
 - Tx Rate
 - Tx code rate
- Policy adapts to
 - Changing channel conditions
 - Packet backlog
 - Historical power usage



Rate-Delay-Reliability

• Policy Results



Game theory

- Coordinating user actions in a large ad-hoc network can be infeasible
- Distributed control difficult to derive and computationally complex
- Game theory provides a new paradigm
 - Users act to “win” game or reach an equilibrium
 - Users heterogeneous and non-cooperative
 - Local competition can yield optimal outcomes
 - Dynamics impact equilibrium and outcome
 - **Adaptation via game theory**

Summary

- Capacity of wireless ad hoc networks largely unknown, even for simple canonical models.
- Scaling laws, degrees of freedom (interference alignment) and other approximations promising
- Capacity not the only metric of interest
- Cross layer design requires new tools such as optimization and game theory

Presentation

- “Hierarchical Cooperation Achieves Optimal Capacity Scaling in Ad Hoc Networks” by Ayfer Ozgur, Olivier Leveque, and David N. C. Tse
- Presented by Alexandros Manolakos