#### Quantifying and Achieving the Capacity of Wireless 1-Hop Network Coding — A Code-Alignment-Based Approach

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- Energy and power savings, [Cui et al. 08], [Goseling et al. 09], etc.
- Security (cryptography) [Bhattad *et al.* 06], [Ngai *et al.* 09], etc.
- Error correction [Ahlswede *et al.* 09], [Silva *et al.* 08], etc.
- Network tomography [Sattari *et al.* 09], [Gjoka *et al.* 08], etc.
- Speed up computation of the min-cuts and min-cut values [Wu *et al.* 06]
   [Wang *et al.* 09].
- Storage [Wu 09], P2P [M. Wang *et al.* 07], etc.
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#### **But** ... What is Network Coding?

#### Many different interpretations of NC:

- Link-by-Link Forward Error Control [Ghaderi et al. 07],
- Fountain codes [Luby 02],
- Network-wide multiple-description codes,
- Cooperative wireless networks [Médard 09],
- Generalization of the store-&-forward policy [Yeung 06],



#### Yet Another Definition . . .

- A conservative Packet Erasure Channel (PEC) abstraction of packet transmission that is independent from the PHY layer schemes:
  - Input:  $X \in GF(2^b)$  for some sufficiently large *b*.
  - Output:  $Y \in \{X\} \cup \{*\}$  where "\*" is the erasure symbol.
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• 
$$P(Y = * | X = x) = P(Y = *).$$
  
(3) Broadcast PEC (2)  
(4) Broadcast PEC (2)  
(5) PEC (2)  
(6) PEC (2)  
(7) PEC (2)  
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(7)

Network coding is ... the network information theory study (especially the achievability part) of a PEC-based network.



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  - Discard the corrupted packets completely, instead of using hybrid ARQ schemes.
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Wireless Networks (modeled as Packet Erasure Channels) +
 Throughput Analysis + Digital Network Coding
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- Intersession NC (INC): Coding over multiple unicast sessions.
- The COPE protocol 2-hop relay networks [Katti *et al.* 06] 4 transmissions w/o coding vs. 3 transmissions w. coding
  - r sends [X + Y];  $d_1$  decodes X by subtraction.
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- INC is a hard problem. Ex: Pure random lin. NC does not work.
- Shannon capacities of 1-hop INC remain unknown. It becomes non-trivial for random broadcast PECs and M > 2 sessions.

## **Main Theoretical Results**

The benefits of COPE follows
from the message side
information (MSI).  $\underbrace{\$_1 X \quad Y \, \$_2}_{r}$ Generation Generation

| # of sessions | COPE-like Protocols<br>(Broadcast PECs<br>w. MSI)       | Gaussian broadcast<br>channels w. MSI |
|---------------|---|---------------------------------------|
| M=2           | Full capacity region                                    | Full capacity region<br>[Wu 07]       |
| M=3           | Full capacity region                                    | ?                                     |
| General M     | Outer and inner<br>bounds that are<br>numerically close | ?                                     |



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The benefits of ER follows
 from the channel output
 feedback (COF).

$$\underbrace{\underbrace{X_1}}_{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{2}}}}}}} \underbrace{X_1}_{X_2} \underbrace{X_2 + Y_1}_{\sqrt{\sqrt{\sqrt{2}}}} X_1, X_2}_{\sqrt{\sqrt{2}}} \underbrace{X_1, X_2}_{\sqrt{\sqrt{2}}} \underbrace{X_1, X_2}_{\sqrt{2}} \underbrace{X_1, X_2} \underbrace{X_$$

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|---------------|---|---------------------------------------|
| M=2           | Full capacity region<br>[Georgiadis et al.<br>09]   | Outer and inner<br>bounds [Ozarow 84] |
| M=3           | Full capacity region  | ?                                     |
| General M     | <ul> <li>(1) Capacity for fair<br/>systems;</li> <li>(2) Outer and inner<br/>bounds that meet<br/>numerically.</li> </ul> | ?                                     |



# Part I: Quantifying and achieving the capacity of COPE-like protocols

# Part II: Quantifying and achieving the capacity of ER-like protocols



■ Memoryless broadcast PECs: Ex: A 1-to-2 PEC is governed by the success probabilities  $p_{s\to 12}$ ,  $p_{s\to 12^c}$ ,  $p_{s\to 1^c2}$ ,  $p_{s\to 1^c2^c}$ .



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- Sequentially,  $s_1$  to  $s_M$ , and r each can send n packets.
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PEC parameters for M = 2: Joint Prob.:

$$p_{s_1 \to 2r}, p_{s_1 \to 2r^c}, p_{s_1 \to 2^c r}, p_{s_1 \to 2^c r^c};$$

$$p_{s_2 \to 1r}, p_{s_2 \to 1r^c}, p_{s_2 \to 1^c r}, p_{s_2 \to 1^c r^c};$$

$$p_{r \to 12}, p_{r \to 12^c}, p_{r \to 1^c 2}, p_{r \to 1^c 2^c}.$$

Marginal Prob.:

 $p_{r;1} \stackrel{\Delta}{=} p_{r \to 12} + p_{r \to 12^c}$ 

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• Each round:  $s_1$  to  $s_M$  first and then r. Totally  $(M+1) \cdot n$  pkts.

#### M=2





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- Each round:  $s_1$  to  $s_M$  first and then r. Totally  $(M+1) \cdot n$  pkts.
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- From the relay's perspective, it becomes a broadcast PEC problem with side information (SI).



No feedback is allowed during the transmission of the last *n* packets by relay *r*.





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For the sake of illustration, we first focus on linear codes.  $\mathbf{X}^{[2]} \mathbf{X}^{[2^c]} \mathbf{Y}^{[1]} \mathbf{Y}^{[1^c]}$ 







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• The cap. outer bound, M = 2 users: main info + minimal interference  $\leq$  the overall available slots  $d_1$ 's perspective:  $nR_{1;2} + nR_{1;2^c} + nR_{2;1^c} \leq np_{r;1}$  $d_2$ 's perspective:  $nR_{2;1} + nR_{2;1^c} + \frac{p_{r;2}}{p_{r;1}}nR_{1;2^c} \leq np_{r;2}$ .



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The achievability: A 2-Stage coding scheme.



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Given that  $R_1 = R_{1;2} + R_{1;2^c}$ , maximizing  $R_2$  is equivalent to allocating the smallest  $R_1$  to  $R_{1;2^c}$ . I.e., the stronger overhearing the better.



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• Therefore:

$$R_{1} \leq p_{r;1} - (R_{2} - p_{s_{2};1})^{+}$$
$$R_{2} \leq p_{r;2} - \frac{p_{r;2}}{p_{r;1}}(R_{1} - p_{s_{1};2})^{+}$$



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• After combining the  $s_i \rightarrow r$  coding:

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 $X_1 \cdots X_{nR_1}$ 

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$$\frac{q_{2}}{\hat{y_{1}} \cdots \hat{y_{nR_{2}}}} = \hat{y_{nR_{2}}} + \hat{y_{1}} + \hat{y_{nR_{2}}} + \hat{y_{nR_$$

• Can be combined with opportunistic routing (jump over 2 hops):



 $Y_1 \cdots Y_{nR_2}$ 

• After combining the  $s_i \rightarrow r$  coding:

$$R_{1} \leq p_{r;1} - (R_{2} - p_{s_{2};1})^{+}$$
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 $\begin{array}{cccc} X_1 \cdots X_{nR_1} & Y_1 \cdots Y_{nR_2} \\ t_{S_1} & & S_2 \\ \end{array} \\ t_{S_1} & & t_{S_2} \\ \end{array}$ 

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To ensure that s<sub>i</sub> can convey all the info. to r, we must also have R<sub>1</sub> ≤ p<sub>s1;r</sub> and R<sub>2</sub> ≤ p<sub>s2;r</sub>.
Final Results:

$$R_{1} \leq \min\left(p_{s_{1};r}, p_{r;1} - (R_{2} - p_{s_{2};1})^{+}\right)$$

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$$P_{PEC}$$

$$\frac{d_{2}}{\hat{y_{1}} \cdots \hat{y_{nR_{2}}}}$$

- Can be combined with opportunistic routing (jump over 2 hops):
- Can be combined with cross-layer optimization: Each round of 3n packets  $\Rightarrow$  variable scheduling  $t_{s_1}, t_{s_2}, t_r$ .

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# Is The Hybrid Scheme Optimal?



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Any network code.

 $\mathbf{Z}^{[1\overline{2}]}$ 

630

 $X^{[\overline{2}3]} Y^{[13]}$ 

L







Any network code.

240

 $\mathbf{Z}^{[1\overline{2}]}$ 

630

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 $\dim = 900$ 

L



 $\mathbf{X}^{[\overline{2}3]}$ 

 $\dim = 900$ 

 $A \stackrel{\Delta}{=}$ 

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 $\mathbf{Z}^{[1\overline{2}]}$ 

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Decodability at  $d_1 \Rightarrow \text{Ran}$ Decodability at  $d_3 \Rightarrow \text{Rank}(A(1/3)) \ge 630$ . Concavity of information transmission.

1200

1000

800

600

400

200

Rank

info.

interference

rank(A(p))



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Inner bound: Hybrid schemes with stage-based approaches + code alignment.







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#### A 3-User Cap. Illustration



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# **The Throughput Improvements**





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2-hop random networks, Rayleigh fading, proportional fairness.





# Part I: Quantifying and achieving the capacity of COPE-like protocols

# Part II: Quantifying and achieving the capacity of ER-like protocols



# 1-Hop Cellular (AP) Networks

- 1-hop access point networks. M dest.
- M can be large, say  $\approx 20$ .
  (For 2-hop relay networks  $M \leq 6$ ).
- Each session has  $nR_i$  packets.
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- For M = 2, w. feedback, the capacity is [Georgiadis *et al.* 09].



$$\begin{cases} \frac{R_1}{p_{1\cup 2}} + \frac{R_2}{p_2} \le 1\\ \frac{R_1}{p_1} + \frac{R_2}{p_{1\cup 2}} \le 1 \end{cases}$$



Outer bound [Ozarow *et al.* 84]: Introduce auxiliary pipes to convert it into *physically degraded channels*, for which feedback does not increase the capacity [El Gamal 78].



The cap. of the original CH with feedback

- $\prec$  The cap. of the new physically degraded CH with feedback
- = The cap. of the new physically degraded CH without feedback



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Inner bound: A 2-phase approach. (Creating its own side info.)





















# What if $M \ge 3$ ?

The CH. parameters become more involved.

- $M = 2: p_{12}, p_{12^c}, p_{1^c2}, p_{1^c2^c}$ .
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- Can we also quantify the Shannon capacity for  $M \ge 3$ ?
  - Generalization of the outer bound is straightforward.
  - Generalization of the inner bound is more difficult.











• For any permutation  $\pi : [M] \mapsto [M]$ ,

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  - *p*<sub>∪S</sub>: Prob. at least one *d<sub>i</sub>* ∈ *S* is successful.
    *S*<sup>π</sup><sub>k</sub> = {π(*j*) : ∀*j* = 1, · · · , *k*}.







• For each  $\pi$ , the capacity of the degraded channel is

$$\sum_{k=1}^{M} \frac{R_{\pi(k)}}{p_{\cup S_k^{\pi}}} \le 1.$$





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• A capacity outer bound is thus  $\forall \pi$ ,  $\sum_{k=1}^{M} \frac{R_{\pi(k)}}{p_{\cup S_{k}}} \leq 1$ .






How to achieve the outer bound:  $\forall \pi$ ,  $\sum_{k=1}^{M} \frac{R_{\pi(k)}}{p_{\cup S_k^{\pi}}} \leq 1$ 











How to achieve the outer bound:  $\forall \pi$ ,

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First try was by [Larsson et al. 06], an M-phase approach.

Phase 1 Creating New Coding Opp.





|                                      | Phase 2                |
|--------------------------------------|------------------------|
| rcv'd by 1                           | Exploiting Coding Opp. |
| rcv'd by $\overline{1}2\overline{3}$ |                        |
| $\frac{1}{10}$                       | Pha                    |
| rev a by 123                         | Explo                  |
| rcv'd by $\overline{1}23$            |                        |
|                                      |                        |
| rcv'd by 2                           |                        |
| rcv'd by $\overline{12}3$            |                        |
| rcv'd by $1\overline{23}$            |                        |
| rcv'd by $1\overline{2}3$            |                        |





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How to achieve the outer bound:  $\forall \pi$ ,  $\sum_{k=1}^{\infty}$ 

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#### New Cap. Inner Bound

- Again, we need code alignment in order to recoup the overheard coding opportunities during Phases 2 to *M*.
- That is, the overheard coding vectors [X + Y] has to remain aligned in the subsequent mixing stages.
- We propose a new Packet Evolution scheme.
- For each packet,
  - The overhearing status keeps evolving to create more coding opportunities.
  - The representative coding vector keeps evolving to ensure code alignment.





















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# Packet Evolution (Cont'd)

- When we have a transmission opportunity:
  - Use the overhearing status to decide which packets to be mixed
  - Instead of mixing the original packets, we mix the representative coding vectors.



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- When we receive a channel feedback:
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  - Update the representative coding vector to stay aligned in the code space. Code Alignment
- The overhearing status and the coding vector of each packet keep evolving.



- Capacity outer bound:  $\forall \pi, \sum_{k=1}^{M} \frac{R_{\pi(k)}}{p_{\cup S_{k}}} \leq 1.$
- By analyzing the throughput of the packet evolution scheme, we obtain new inner bounds for 1-to-*M* broadcast PECs with arbitrary  $p_{S([M]\setminus S)}$  parameters.



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Time Sharing

W. Feedback

0.6

0.4 R₁

0.2

0L 0

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- Spatially independent 1-to-*M* PECs with rate-fairness constraints (when  $R_1 \approx R_2 \approx \cdots \approx R_M$ ).
- For all our experiments, the outer/inner bounds always meet.





#### **Numerical Evaluation**


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- In practice, per-packet feedback is costly.
- We modify the packet evolution scheme and develop a Mixing-reAlignment-Mixing (MAM) scheme that requires only infrequent periodic feedback.



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- We have implemented practical MAM in Glomosim simulator. Group sessions into groups of M = 4 sessions and perform MAM within each group. Rayleigh fading model with 802.11 CSMA-CD. Packet loss rate: 0.5.





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 Wireless network coding — From practice (ex: COPE, ER, and MORE protocols) back to theory.



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- Message side information vs. channel output feedback:
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- From theory back to practice: Combining the information-theoretic and algorithmic studies.
  - Ex: How to guarantee termination in a noisy environment?
  - Ex: The linear independence guaranteed by  $GF(q), q \rightarrow \infty$ does not hold with prob. 1 for the practical choice  $GF(2^8)$ . How to guarantee decodability?

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#### Questions?

