From Entropy to Networks

The Future of Communication Engineering Through the Lens of Network Information Theory

Young-Han Kim

University of California, San Diego

Department of Information Engineering Chinese University of Hong Kong July 19, 2013

Joint work with Fatemeh Arbabjolfaei (UCSD), Bernd Bandemer (UCSD), Sae-Young Chung (KAIST), Abbas El Gamal (Stanford), Sung Hoon Lim (Samsung), Eren Şaşoglu (Berkeley), Lele Wang (UCSD)





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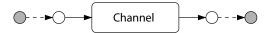


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- Tasks: Communicate the source reliably over the channel



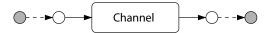
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Information flow questions



Question 1 What is the limit on communication (how much/how fast)?

Information flow questions



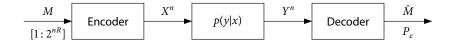
Question 1

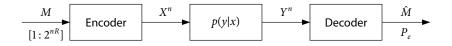
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Question 2

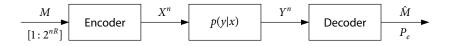
What are the coding schemes/techniques that achieve this limit?



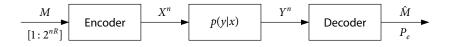




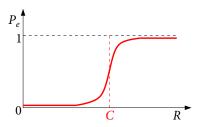
• Tradeoff between *R*, *P*_e, and *n*

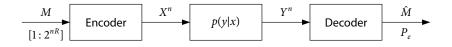


- Tradeoff between *R*, *P*_e, and *n*
- Capacity C: maximum R such that $P_e \rightarrow 0$ as $n \rightarrow \infty$

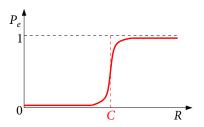


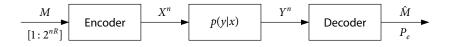
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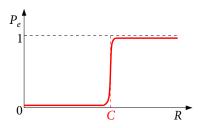


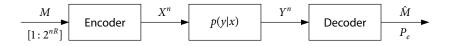
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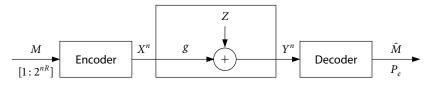




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Shannon (1948)

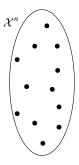
 $C = \max_{p(x)} I(X; Y)$

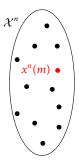


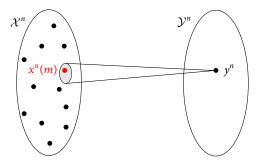
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Shannon (1948)

$$C = \frac{1}{2}\log(1+S) = \mathsf{C}(S)$$







- Average performance of randomly generate codes is good
 - Probabilistic method: there exists a good code

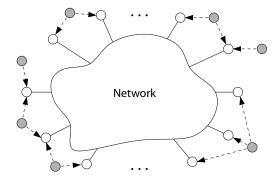
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- IEEE CTW 2010 panel discussion: is communication theory dead?

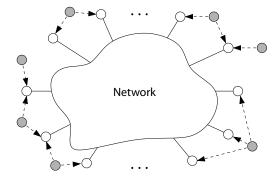


Networked information processing system



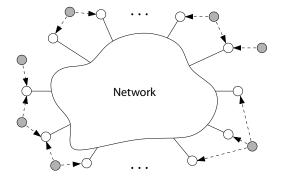
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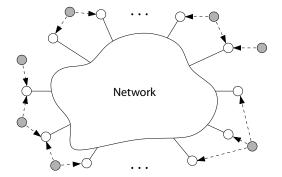
- System: Internet, peer-to-peer network, sensor network, ...
- Task: Communicate sources or make decision based on them

Network information flow questions



• What is the limit on communication?

Network information flow questions



- What is the limit on communication?
- What are the coding schemes/protocols/architectures that achieve this limit?

• Multiple sources and destinations

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- Function computation or collaborative decision making

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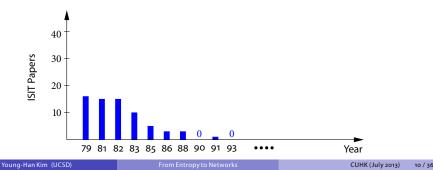
Network information flow questions

- Network capacity
- Optimal coding schemes/protocols/architectures

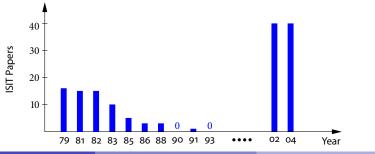
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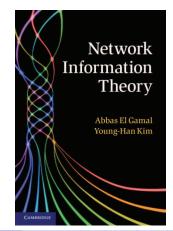
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 - Many basic problems open
 - S Little interest from practice



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- Significant research activities in 70s and early 80s
- Wireless communications and the Internet revived interest in mid 90s
 - Some progress on old open problems and many new problems
 - Very large number of papers in ISIT, T-IT, T-COM, T-WC, …
 - Results starting to have an impact on real-world networks



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- State of the theory: El Gamal-K (2011)



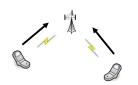
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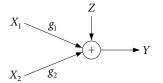
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 - Interference management
 - Multiple unicast
 - Wireless relaying

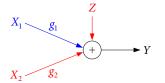
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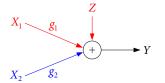
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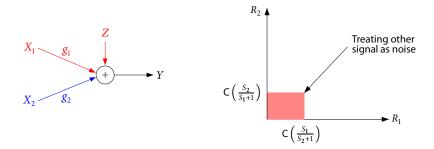
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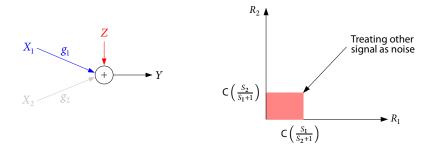


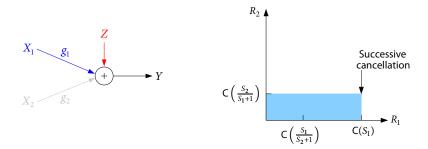


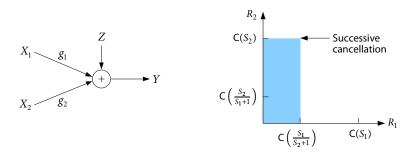


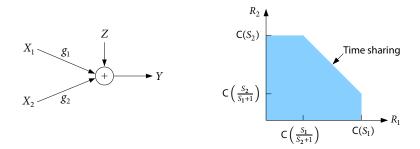


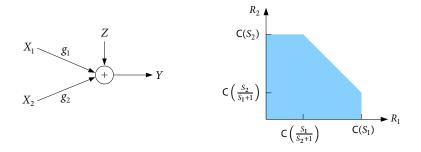




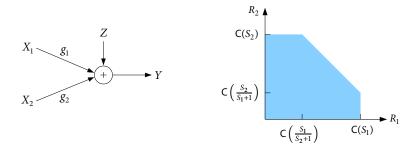




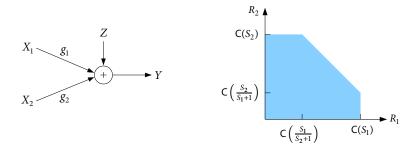




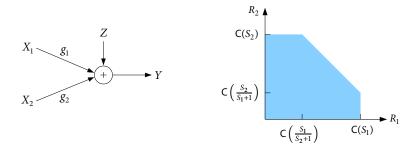
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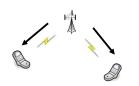
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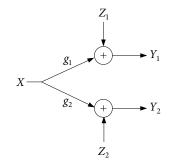


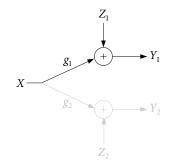
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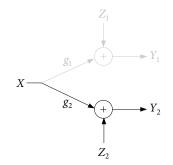


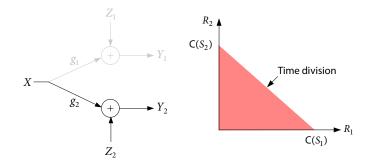
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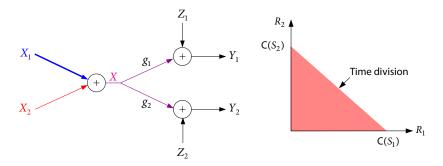


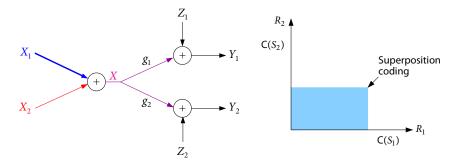


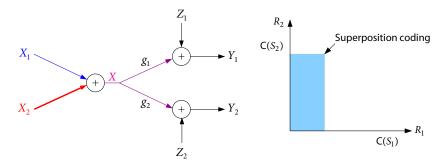


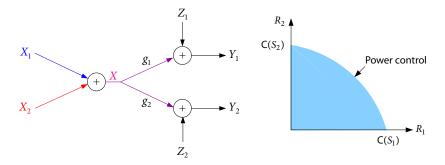


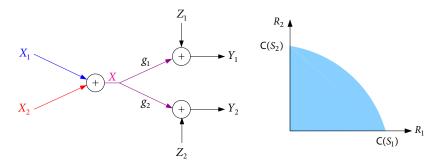




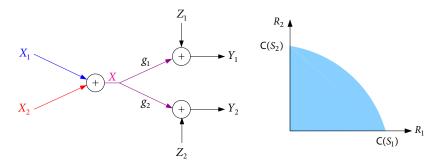






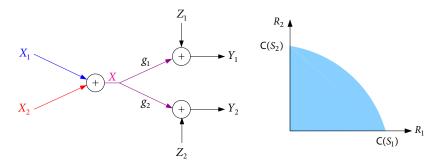


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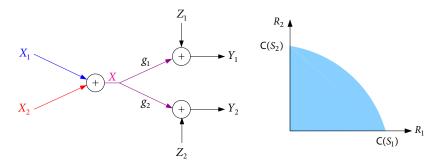
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Broadcast (downlink) communication

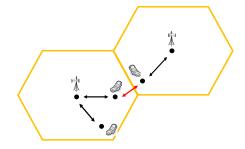


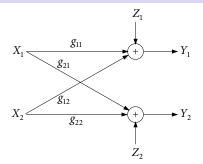
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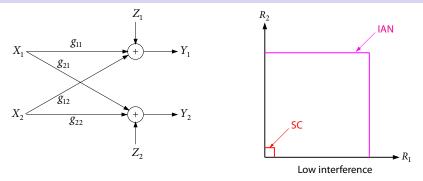
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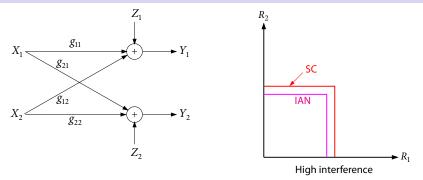
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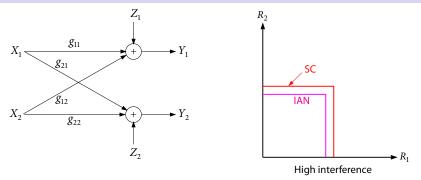




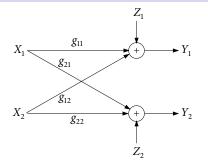
- Low interference: treating interference as noise (IAN) performs well
- High interference: successive cancellation (SC) performs well

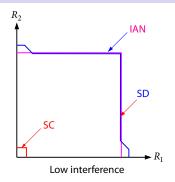


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- Neither is optimal in general; nor is time division

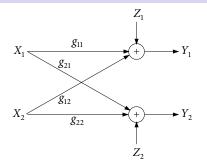


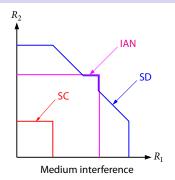
- Low interference: treating interference as noise (IAN) performs well
- High interference: successive cancellation (SC) performs well
- Neither is optimal in general; nor is time division
- In general, we need more than good point-to-point codes + signal processing



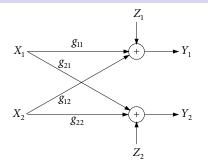


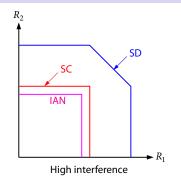
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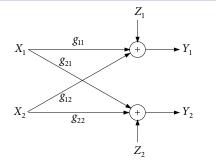


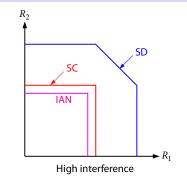
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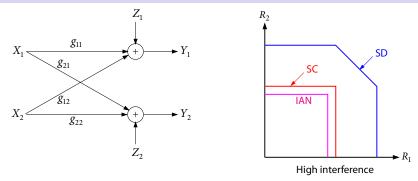


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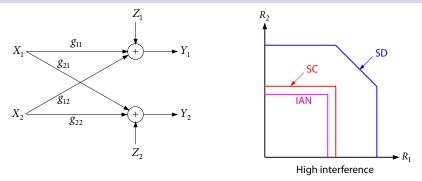




- Simultaneous decoding
 - Always better than treating interference as noise and successive cancellation

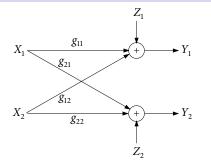


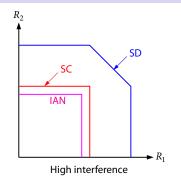
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Simultaneous decoding

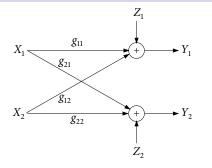
- Always better than treating interference as noise and successive cancellation
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- + superposition coding: optimal within 1/2 bit (Etkin–Tse–Wang 2008)

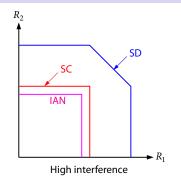




Conclusion 1

Interference-aware coding schemes can boost performance tremendously





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Challenge 1

Low-complexity implementation of simultaneous decoding

Young-Han Kim (UCSD)

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More on simultaneous decoding

• Maximum likelihood decoding (MLD): At receiver 1, find

$$\arg \max_{\hat{m}_{1}} f(y_{1}^{n} | x_{1}^{n}(\hat{m}_{1}))$$

= $\arg \max_{\hat{m}_{1}} \sum_{m_{2}} f(y_{1}^{n} | x_{1}^{n}(\hat{m}_{1}), x_{2}^{n}(m_{2}))$
= $\arg \max_{\hat{m}_{1}} \sum_{m_{2}} \exp(-\|y^{n} - g_{11}x_{1}^{n}(\hat{m}_{1}) - g_{12}x_{2}^{n}(m_{2})\|^{2}/2)$

• Simultaneous decoding (Bandemer–El Gamal–K 2012):

$$\approx \arg \max_{\hat{m}_{1}} \max_{m_{2}} \exp(-\|y_{1}^{n} - g_{11}x_{1}^{n}(\hat{m}_{1}) - g_{12}x_{2}^{n}(m_{2})\|^{2}/2)$$

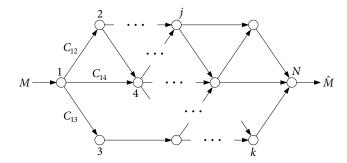
=
$$\arg \max_{\hat{m}_{1}} \max_{m_{2}} -\|y_{1}^{n} - g_{11}x_{1}^{n}(\hat{m}_{1}) - g_{12}x_{2}^{n}(m_{2})\|$$

=
$$\arg \min_{\hat{m}_{1}} \min_{m_{2}} \|y_{1}^{n} - g_{11}x_{1}^{n}(\hat{m}_{1}) - g_{12}x_{2}^{n}(m_{2})\|$$

Outline of the talk

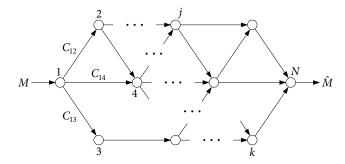
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 - Wireless relaying
- Solutions from network information theory
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Graphical unicast networks



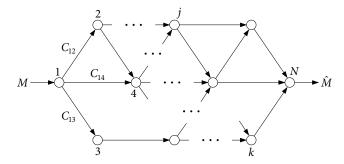
Model for wired networks (Internet, distributed storage, ...)

Graphical unicast networks



- Model for wired networks (Internet, distributed storage, ...)
- Directed weighted graph $(\mathcal{N}, \mathcal{E})$ with link capacities C_{ik}

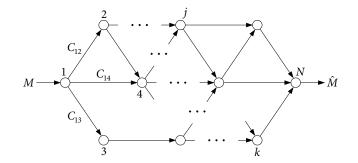
Graphical unicast networks

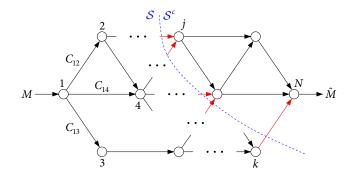


- Model for wired networks (Internet, distributed storage, ...)
- Directed weighted graph $(\mathcal{N}, \mathcal{E})$ with link capacities C_{jk}

Network information flow questions

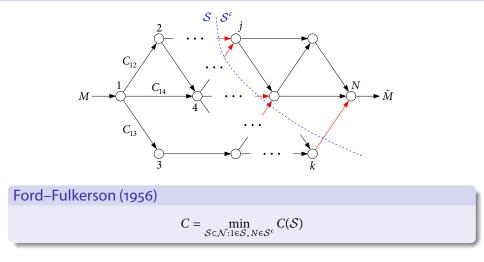
- Network capacity
- Optimal coding schemes/protocols/architectures



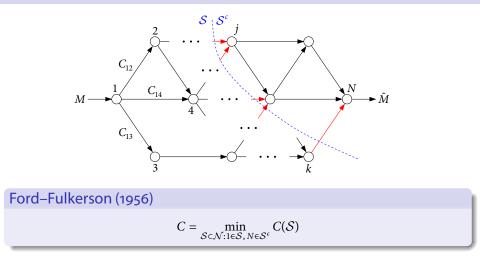


Ford–Fulkerson (1956)

$$C = \min_{\mathcal{S} \subset \mathcal{N}: 1 \in \mathcal{S}, N \in \mathcal{S}^{c}} C(\mathcal{S})$$



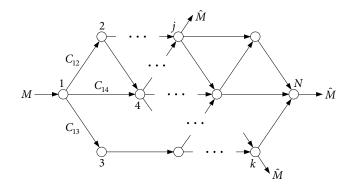
Achieved error-free using simple routing (Ford–Fulkerson algorithm)



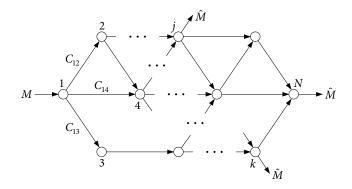
- Achieved error-free using simple routing (Ford–Fulkerson algorithm)
- Information treated as commodity flow

Young-Han Kim (UCSD)

Graphical multicast network

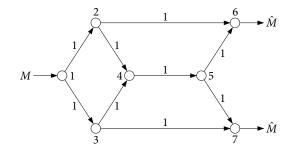


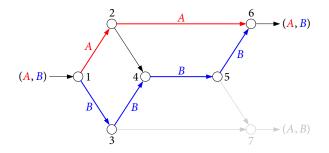
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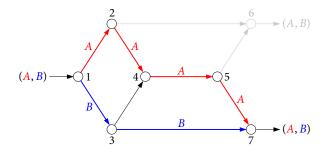


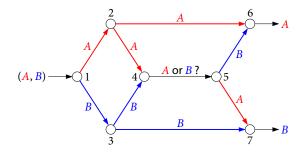
Network information flow questions

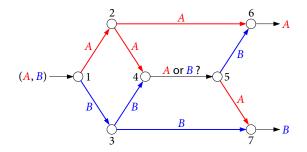
- Network capacity?
- Is routing sufficient?



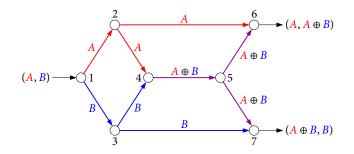




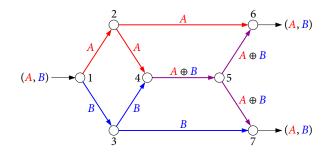




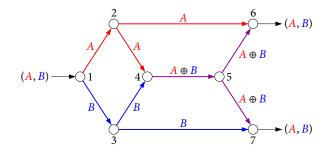
• **Routing:** *R* = 1



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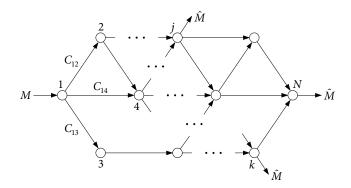


- Routing: R = 1
- Coding: *R* = 2 (capacity)



- Routing: R = 1
- Coding: R = 2 (capacity)
- Treating information as a commodity is not optimal in general

Network coding theorem

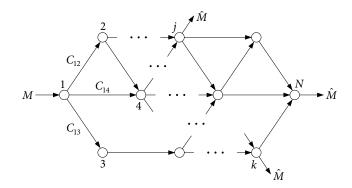


Ahlswede-Cai-Li-Yeung (2000)

$$C = \min_{j \in \mathcal{D}} \min_{\substack{\mathcal{S} \subset \mathcal{N} \\ 1 \in \mathcal{S}, j \in \mathcal{S}^{c}}} C(\mathcal{S})$$

Young-Han Kim	UCSD)
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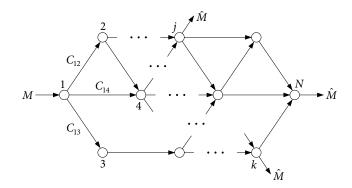
Network coding theorem





Achieved by linear network coding (Li–Yeung–Cai 2003, Koetter–Médard 2003)

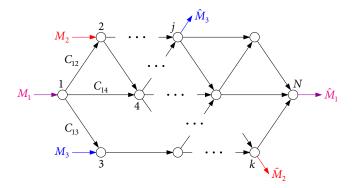
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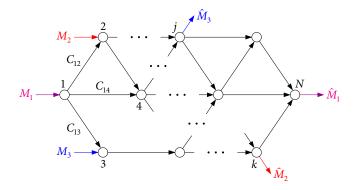




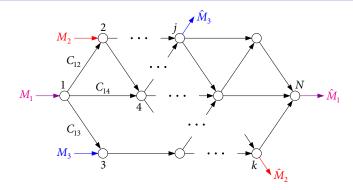
- Achieved by linear network coding (Li–Yeung–Cai 2003, Koetter–Médard 2003)
- Coded TCP: performance improvement for wi-fi and cellular networks

Young-Han Kim (UCSD)

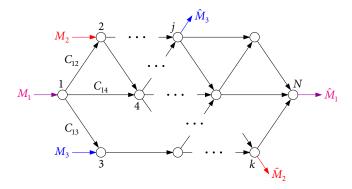




• Capacity not known in general even for two flows

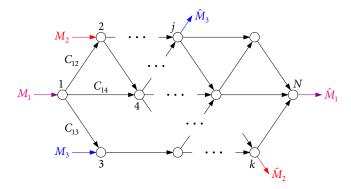


- Capacity not known in general even for two flows
- Nonlinear network coding > linear network coding > routing



Conclusion 2

Coding brings a new dimension to networking



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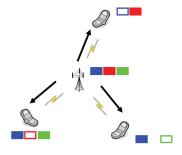
Challenge 2

Network capacity and optimal network coding

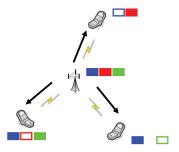
Young-Han Kim (UCSD)

From Entropy to Networks

Index coding

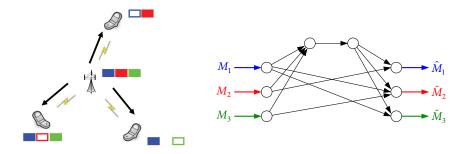


Index coding

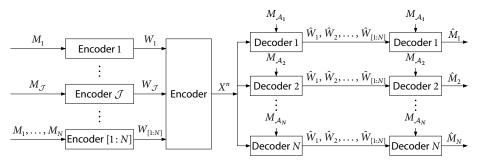


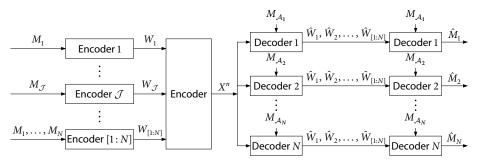
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- How many transmissions are needed?
- Which coding minimizes the number of transmissions?

Index coding

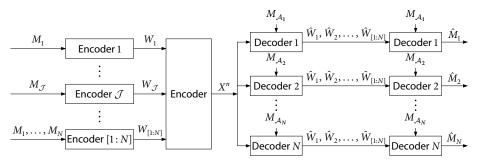


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- How many transmissions are needed?
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- Canonical network coding problem (broadcast and interference)



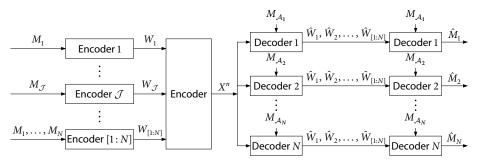


• All 218 index coding problems with N = 4



All 218 index coding problems with N = 4

 All 9608 index coding problems with N = 5 (Arbabjolfaei–Bandemer–K–Şaşoglu–Wang 2013)

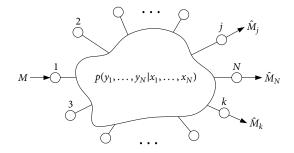


All 218 index coding problems with N = 4

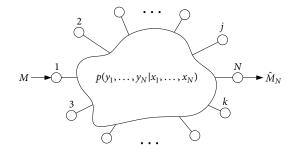
- All 9608 index coding problems with N = 5 (Arbabjolfaei–Bandemer–K–Şaşoglu–Wang 2013)
- Problems with circular symmetry (Maleki–Cadambe–Jafar 2012)

Outline of the talk

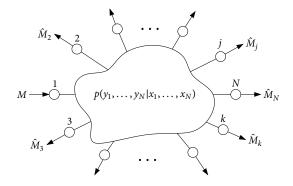
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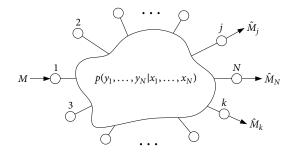
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- Network topology/noise: $p(y_1, \ldots, y_N | x_1, \ldots, x_N)$



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- Extended vocabulary: partial decode–forward, noncoherent decode–forward, coherent compress–forward, generalized amplify–forward

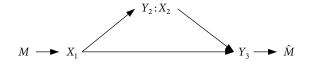
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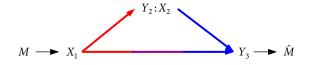
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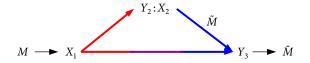
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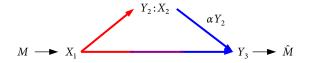




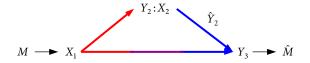




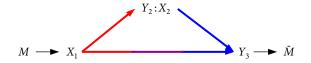
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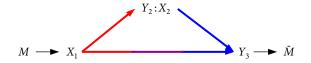
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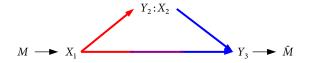
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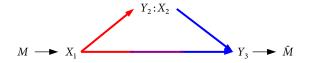
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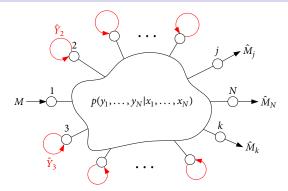


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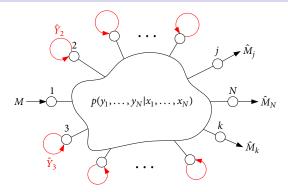
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Noisy network coding



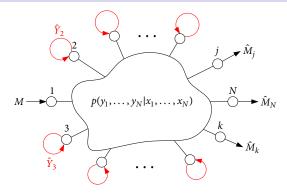
• Compress-forward + network coding

Noisy network coding



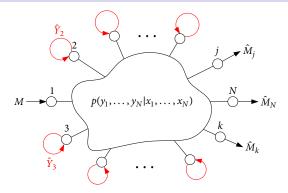
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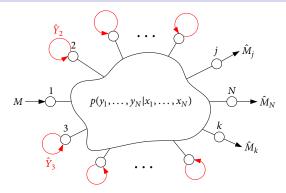
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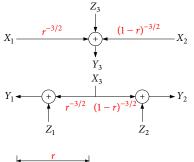


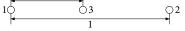
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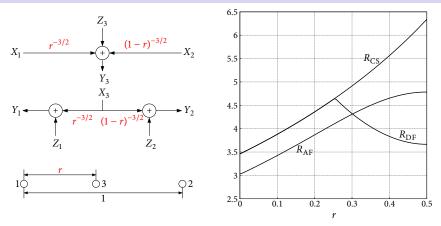
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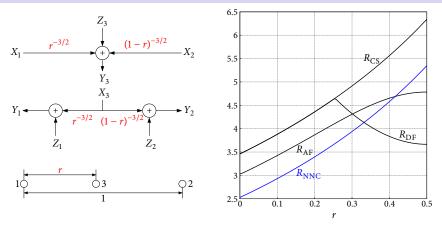
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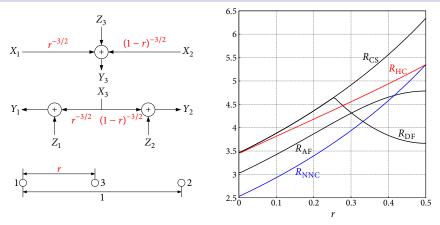




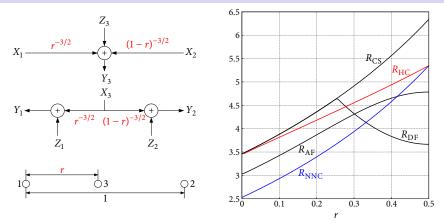
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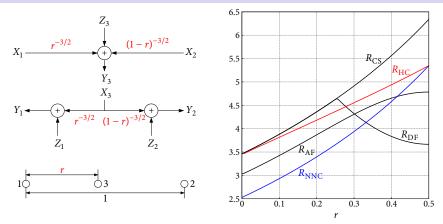


- Decode–forward and amplify–forward: unbounded gap from capacity in general
- Noisy network coding and hybrid coding: within 1 bit gap from capacity



Conclusion 3

Advanced relaying schemes can outperform traditional schemes



Conclusion 3

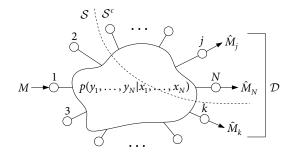
Advanced relaying schemes can outperform traditional schemes

Challenge 3

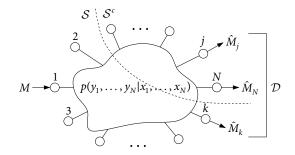
Practical coding, modulation, and signal processing techniques

Young-Han Kim (UCSD)

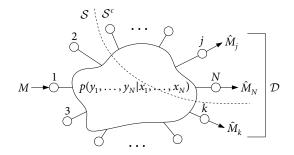
From Entropy to Networks



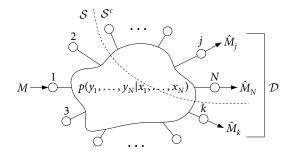




Cutset bound (El Gamal 1981) $C \leq \min_{k \in \mathcal{D}} \min_{\mathcal{S}: 1 \in \mathcal{S}, k \in \mathcal{S}^{c}} I(X(\mathcal{S}); Y(\mathcal{S}^{c}) | X(\mathcal{S}^{c}))$



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Noisy network coding bound (Lim-K-El Gamal-Chung 2011)

 $C \ge \max \min_{k \in \mathcal{D}} \min_{\mathcal{S}: 1 \in \mathcal{S}, k \in \mathcal{S}^{c}} (I(X(\mathcal{S}); \hat{Y}(\mathcal{S}^{c}), Y_{k} | X(\mathcal{S}^{c})) - I(Y(\mathcal{S}); \hat{Y}(\mathcal{S}) | X^{N}, \hat{Y}(\mathcal{S}^{c}), Y_{k})),$ where the maximum is over all $\prod_{k=1}^{N} p(x_{k}) p(\hat{y}_{k} | y_{k}, x_{k})$

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 - Emerging applications: small cells, device-to-device, ultra wideband, high mobility, ...

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— Claude E. Shannon

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