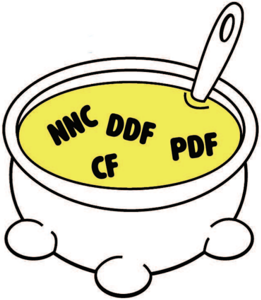
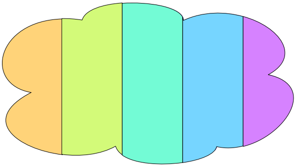
Coding for Wireless Relay Networks
Alphabet Soups and the Network Cutlet Bound

Young-Han Kim

University of California, San Diego



European School of Information Theory
Zandvoort, The Netherlands
April 20, 2015



Theorem 1

history of communication = history of wireless

Japanese scientists dug 50 meters underground and discovered small pieces of copper. After studying these pieces for a long time, Japan announced that the ancient Japanese 20,000 years ago had a nationwide telephone network.

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American scientists were outraged. They dug 50, 100 and 200 meters underground, but found nothing. They concluded that the ancient United States 40,000 years ago had cellular telephones.

Theorem 2

history of wireless = history of relaying

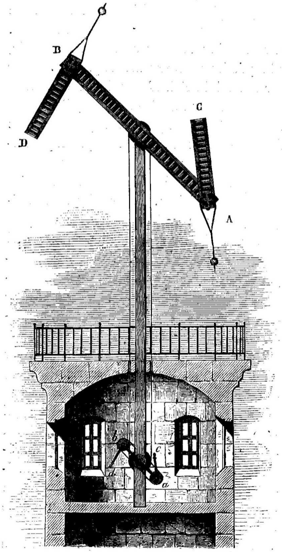
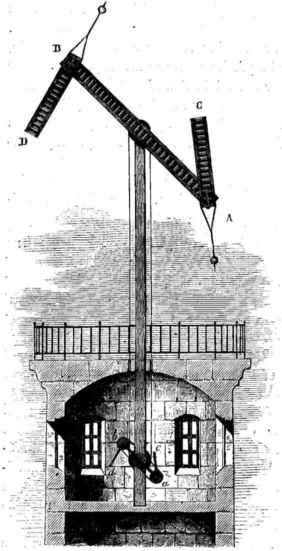


Fig. 19. — Télégraphe de Chappe.



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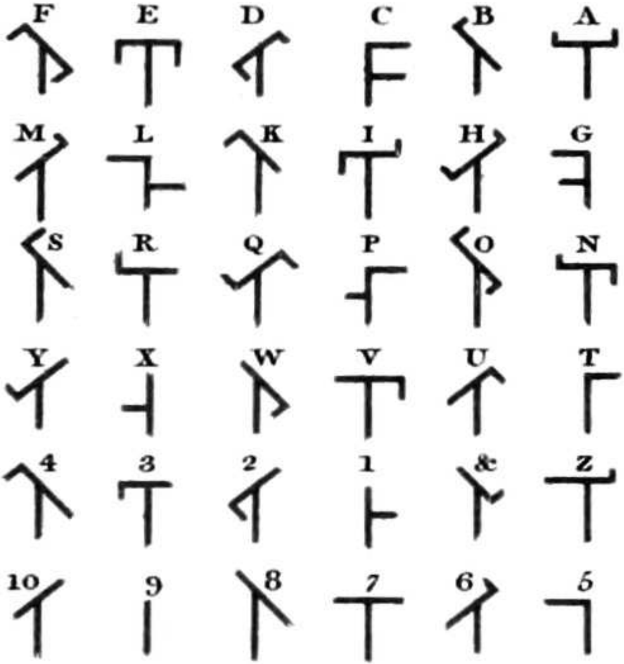

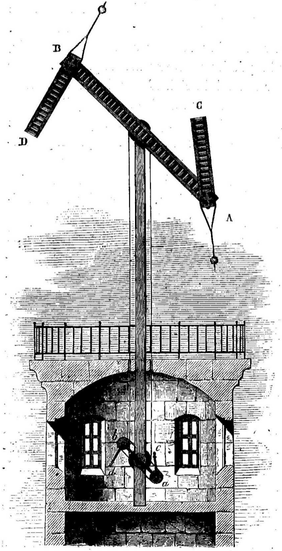

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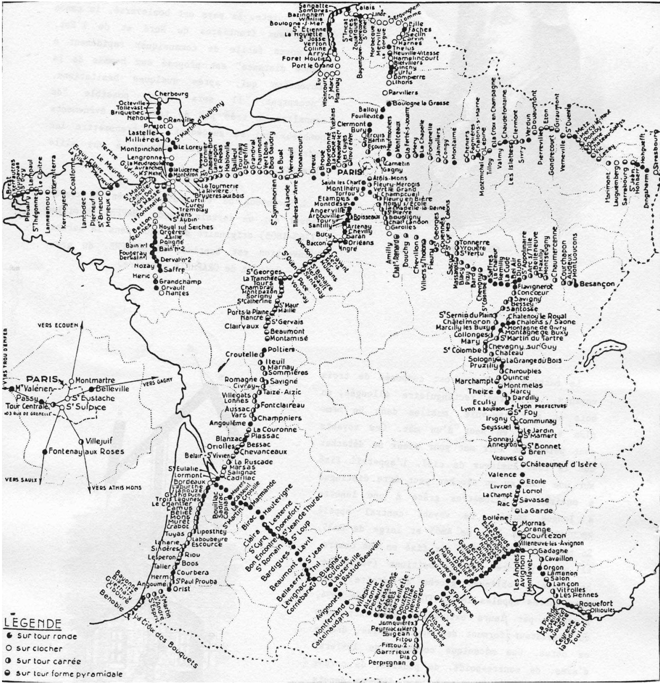

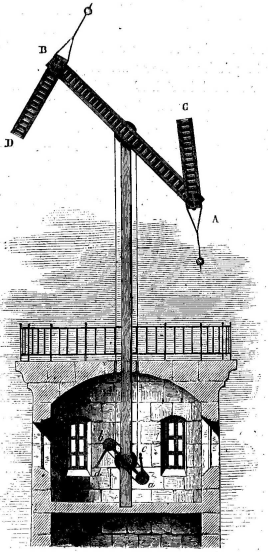

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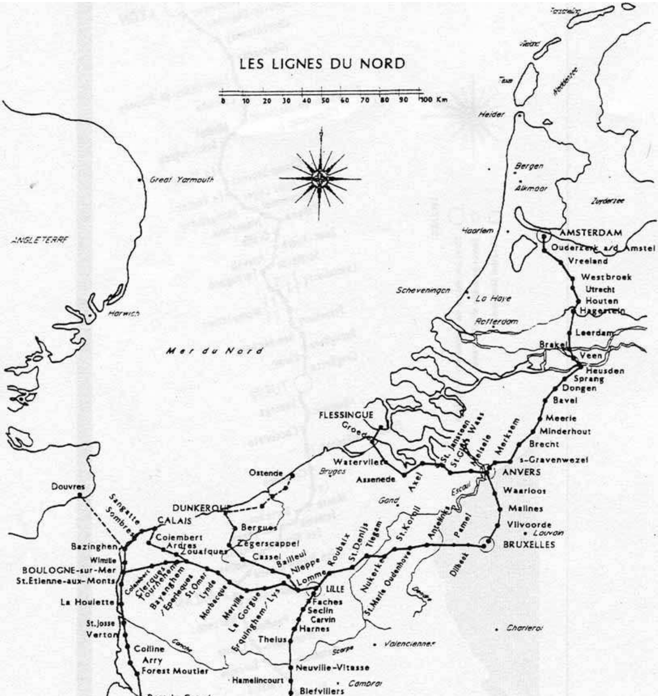

New skyway spans nation with words and pictures

BELL SYSTEM *Radio-Relay* BUILT FOR LONG DISTANCE CALLS AND TELEVISION

There's something new on the national horizon! Bell Telephone construction crews have completed the last link in a coast-to-coast *Radio-Relay* system that is unique in all the world. Today, communications ride on radio microwaves, flashed through the air from tower to tower.

It was an historic event in 1915, when wires first carried the human voice across three thousand miles of mountains and prairie. By 1942, telephone messages were carried across the United States by another means – cable, both underground and overhead. And now comes *Radio-Relay* to supplement wire and cable!

The new system is already in use for Long Distance telephone service and coast-to-coast television. This new skyway helps make America's vast communications network even stronger and more flexible. And it could hardly happen at a better time. The demands of defense are heavy and urgent.

BELL TELEPHONE SYSTEM

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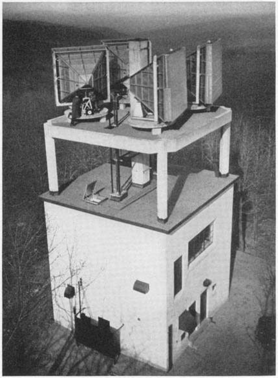
BELL TELEPHONE SYSTEM

Seven towers on seven hilltops



One of seven relay stations — to test use of radio "microwaves" for Long Distance calls.

Built by the Bell System, they will provide a new kind of Long Distance communication.

Each hilltop tower is a relay station between New York and Boston* for very short radio waves. These "microwaves" are free from static and most man-made interference. But they shoot off into space instead of following the earth's curve. So they have to be gathered into a beam and aimed at the next tower, about 30 miles away. That's the job of the four big, square, metal lenses on each tower. They focus microwaves very much as a magnifying glass focuses the sun's rays.

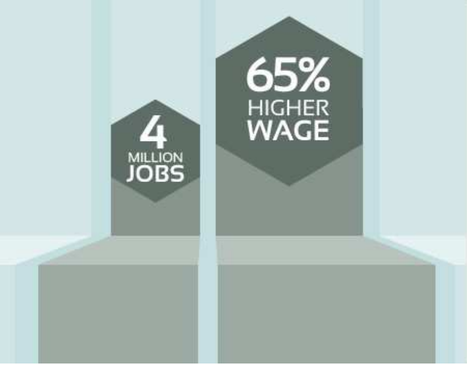
These radio relay systems may be used for Long Distance telephone calls and to transmit pictures, radio broadcasts and television programs.

This is another example of the Bell System's effort to provide more and better Long Distance service.

BELL TELEPHONE SYSTEM

**We have applied to the Federal Communications Commission for authority to start a similar link later between New York and Chicago*

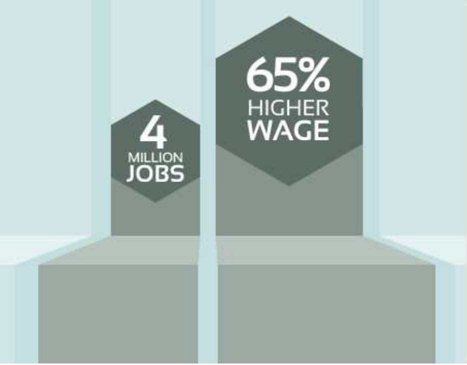

VICTIM
OF
WIRELESS
TECHNOLOGY



Wireless Jobs, Good Pay

Nearly four million U.S. jobs are directly or indirectly associated with wireless, and they pay 65% higher wages than the national average.

Source: Larry Summers, "Technological Opportunities, Job Creation, and Economic Growth," Remarks at the New America Foundation, June 28, 2010, at http://www.whitehouse.gov/administration/eop/nec/speeches/technological-opportunities-job-creation-economic-growth.



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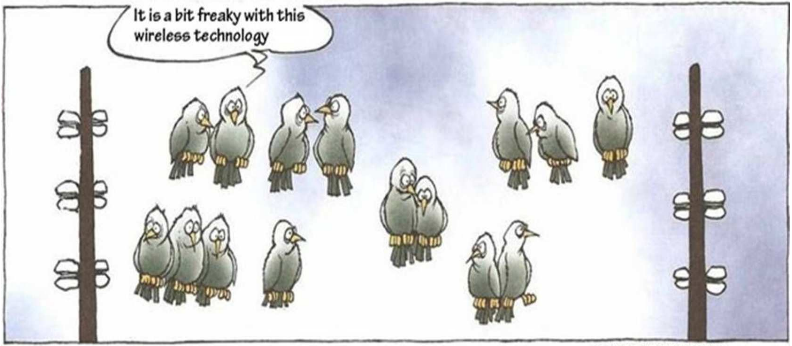
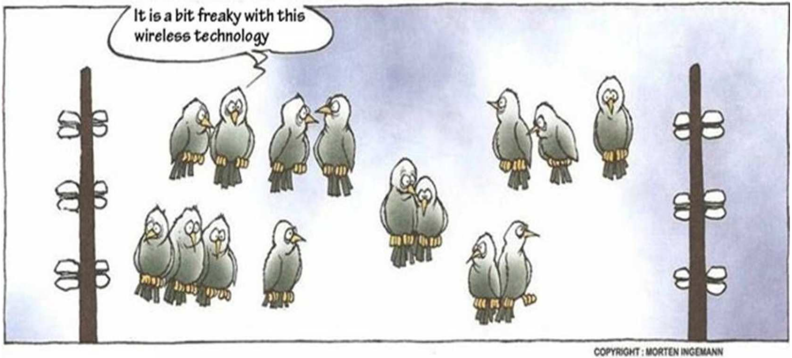
U.S. Wireless Major Industry

The U.S. wireless industry is valued at nearly $196 billion, which is larger than publishing, agriculture, hotels and lodging, air transportation, and motor vehicle manufacturing segments.

Source: Roger Entner, *The Wireless Industry: The Essential Engine of US Economic Growth*, Recon Analytics, April 2012, *available at* http://reconanalytics.com/wp-content/uploads/2012/04/Wireless-The-Ubiquitous-Engine-by-Recon-Analytics-1.pdf. [at p.1]


It is a bit freaky with this wireless technology


It is a bit freaky with this wireless technology
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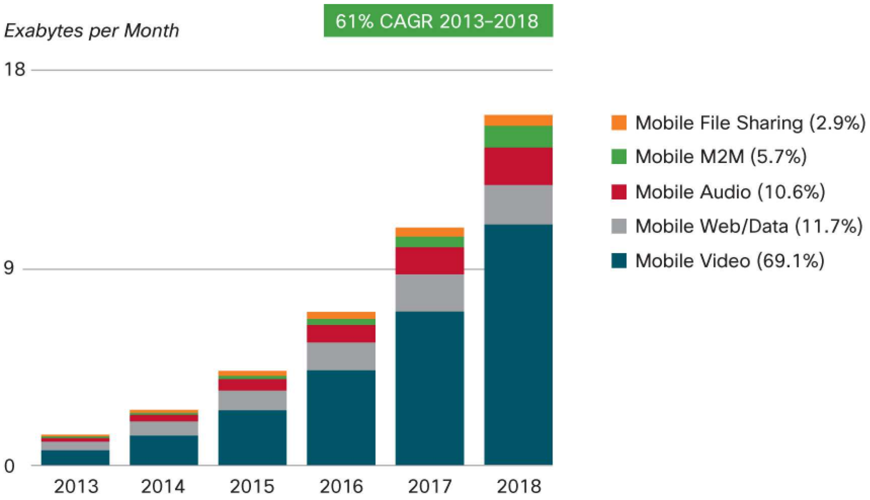
What the hell, Samuel! You're so old fashioned Nowadays everything is wireless!

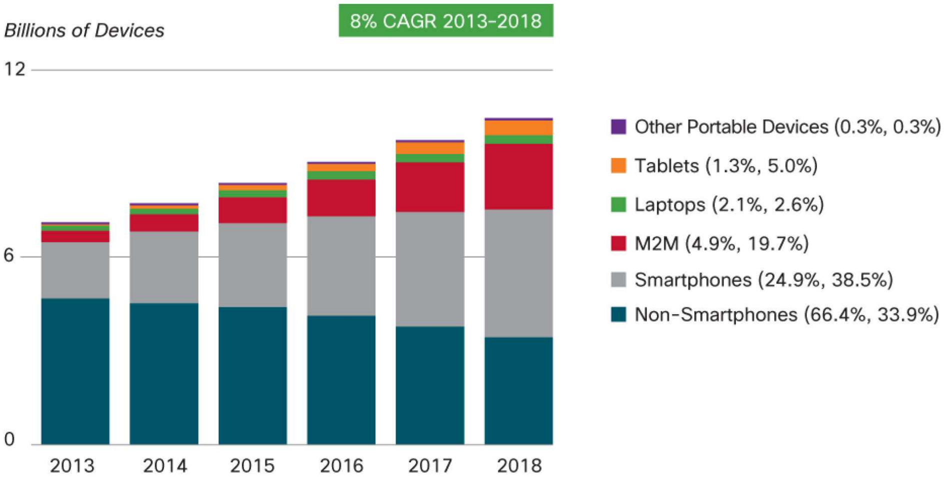
Where is wireless going?



Figures in parentheses refer to traffic share in 2018.

Source: Cisco VNI Mobile, 2014

Where is wireless going?

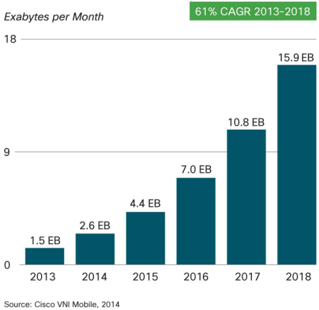


Figures in parentheses refer to device or connections share in 2013, 2018.

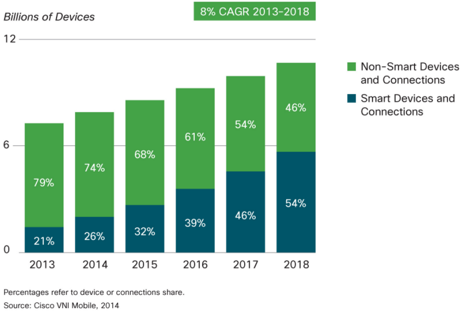
Source: Cisco VNI Mobile, 2014

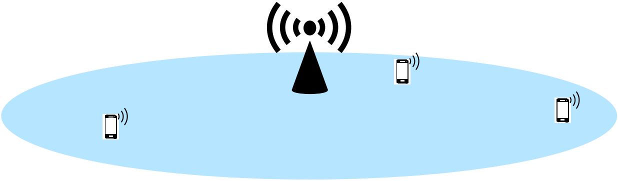
Where is wireless going?

Mobile data per month



Number of devices





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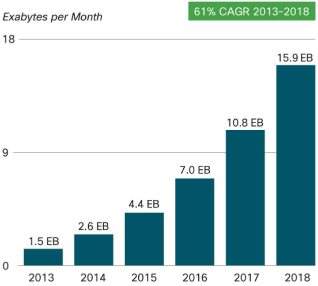

Source: Cisco VNI Mobile, 2014

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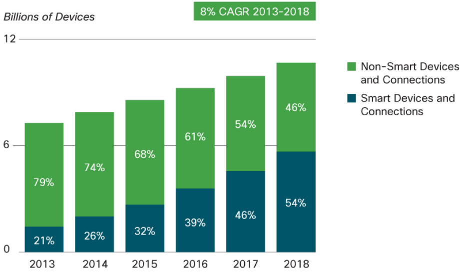

Percentages refer to device or connections share.
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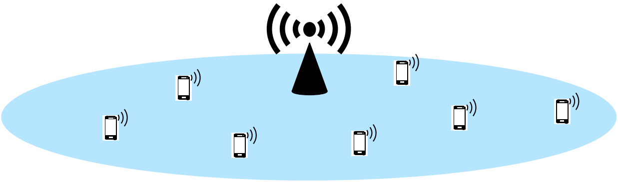

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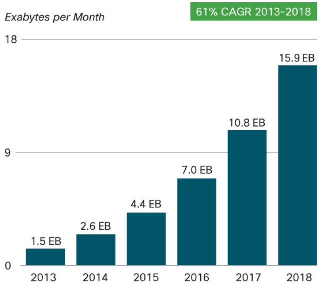

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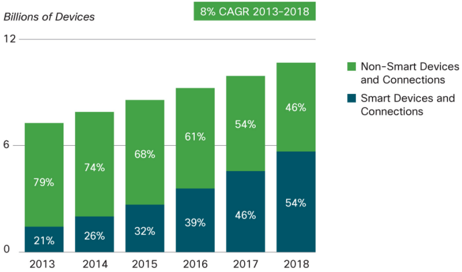

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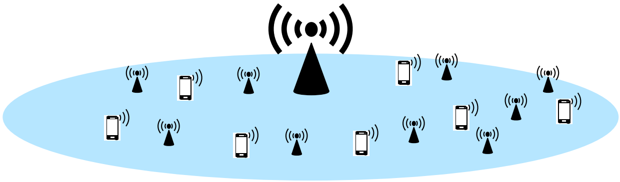
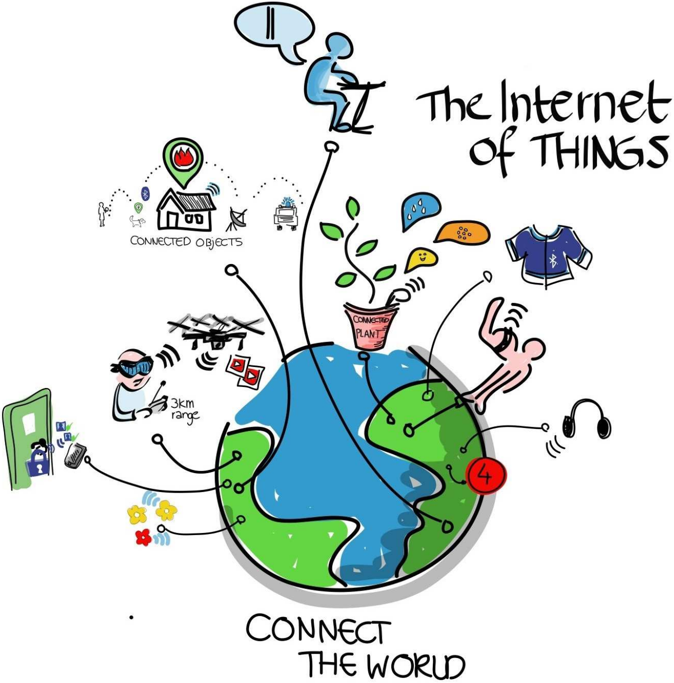
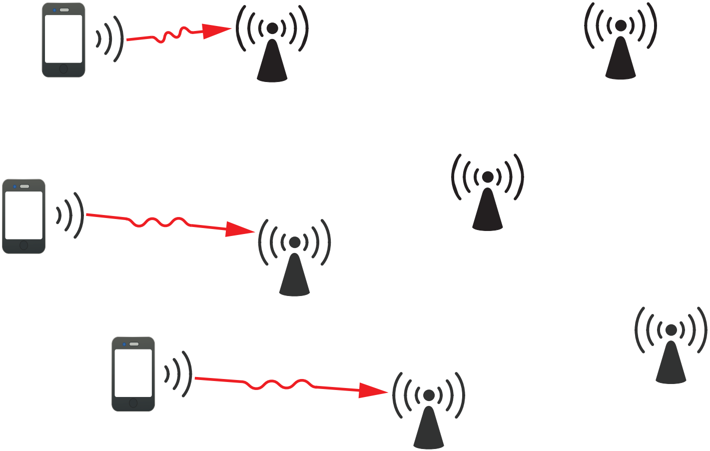




The Internet of THINGS
CONNECTED OBJECTS
CONNECTED PLANT
3km range
4
CONNECT THE WORLD

Cloud radio access network (C-RAN)



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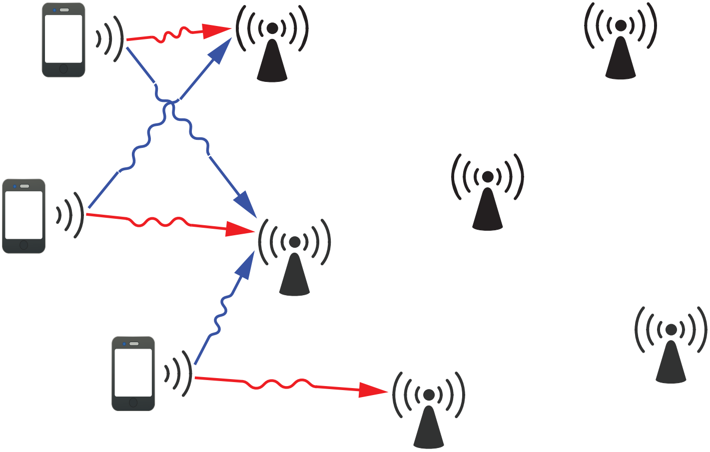

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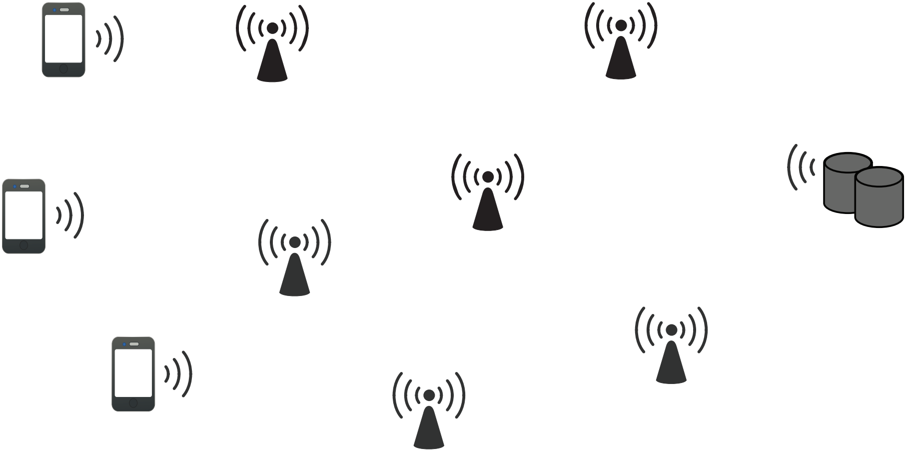

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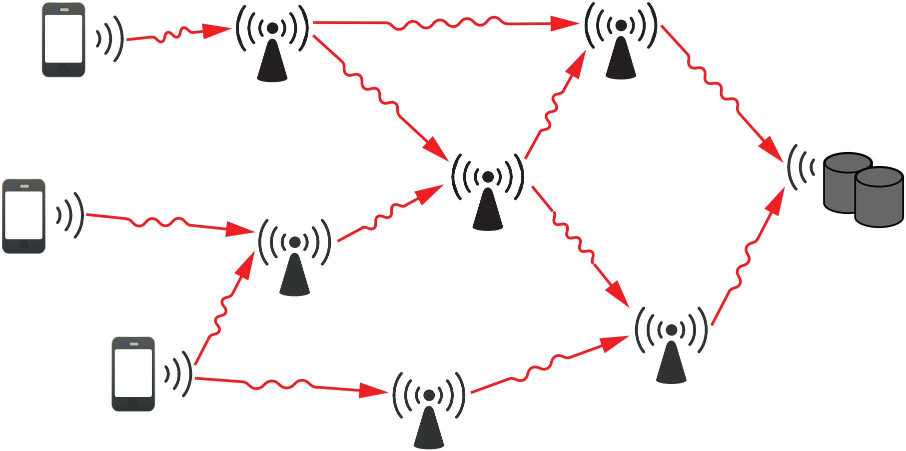

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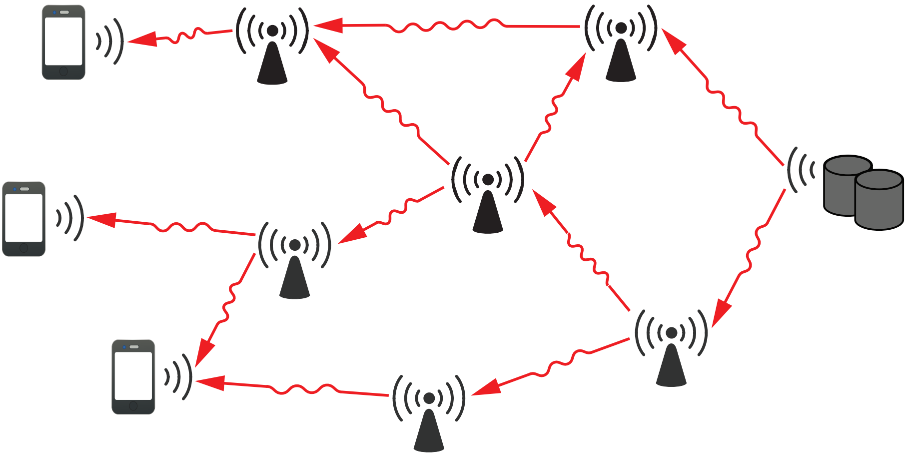
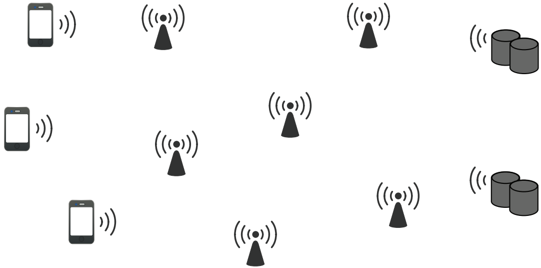
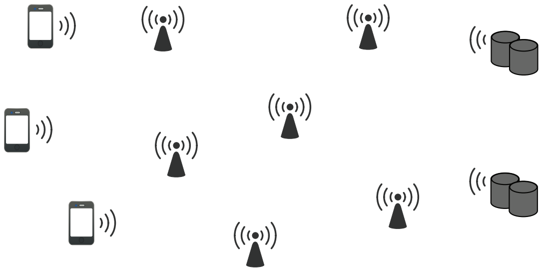


About this tutorial



- Information theory for relay networks
  - What is the limit on the amount of reliable communication?
  - What are the coding schemes that achieve this limit?

About this tutorial



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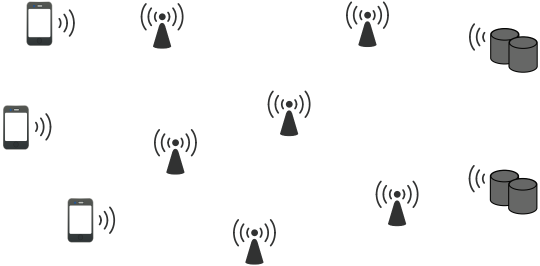
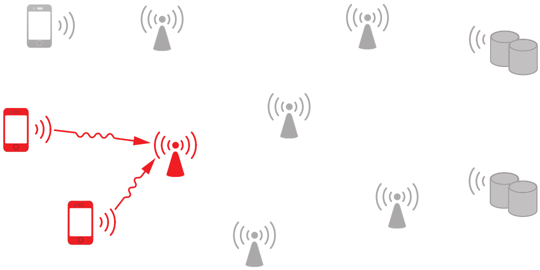
  - What is the limit on the amount of reliable communication?
  - What are the coding schemes that achieve this limit?
- Personal angle on the topic (not a comprehensive survey of the literature)

Organization



Organization

- Single-hop communication
  - Multiple access (many-to-one)



Organization

- Single-hop communication
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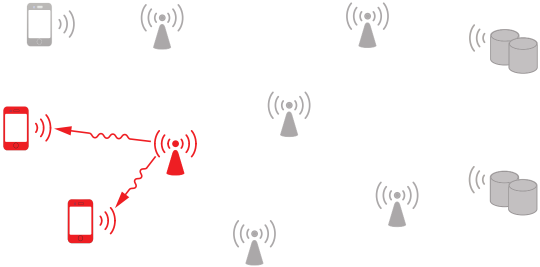
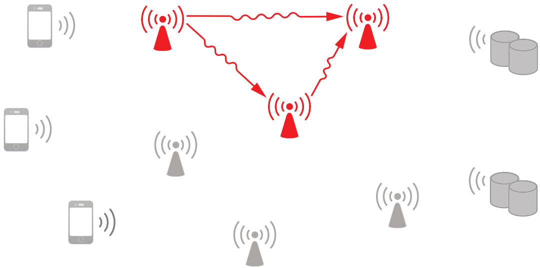
  - Broadcast (one-to-many)



Organization

- Single-hop communication
  - Multiple access (many-to-one)
  - Broadcast (one-to-many)
- Basic relaying techniques
  - Partial decode–forward (PDF)
  - Compress–forward (CF)



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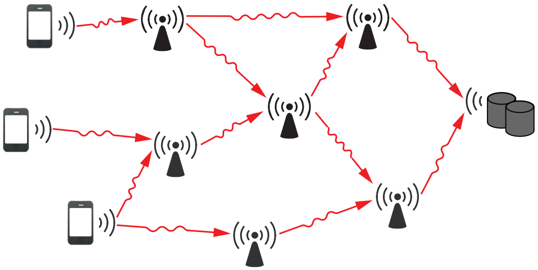

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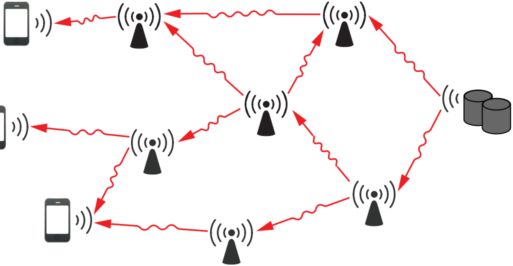
  - Distributed decode–forward (DDF)



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- Basic relaying techniques
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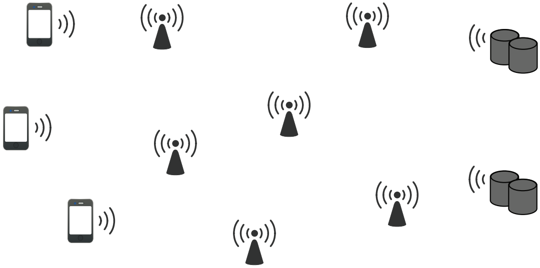
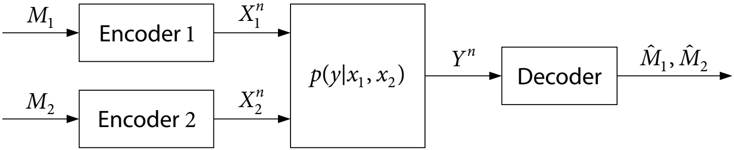
- Multihop communication
  - Noisy network coding (NNC)
  - Distributed decode–forward (DDF)
- Limit on communication
  - Cutset bound
  - Directed cutset (cutlet) bound



$$C \leq \ldots I(X;Y) \ldots$$

Multiple access channel



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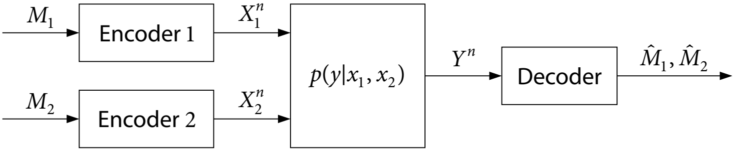

- A $(2^{nR_1}, 2^{nR_2}, n)$ code:
  - Message sets: $[1:2^{nR_1}]$ and $[1:2^{nR_2}]$
  - Encoder $j = 1, 2$: $x_j^n(m_j)$
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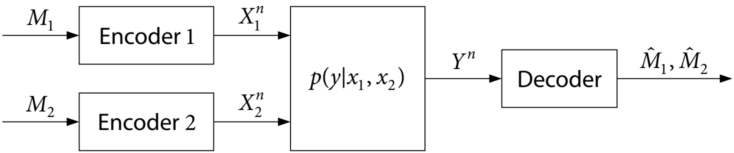
Multiple access channel



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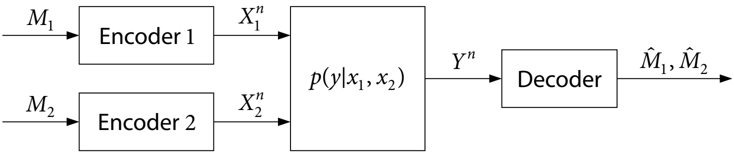

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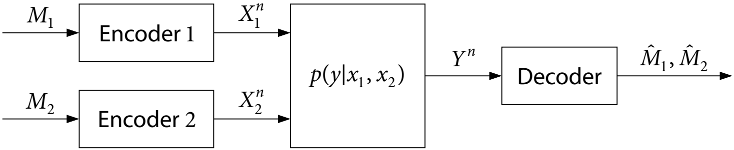

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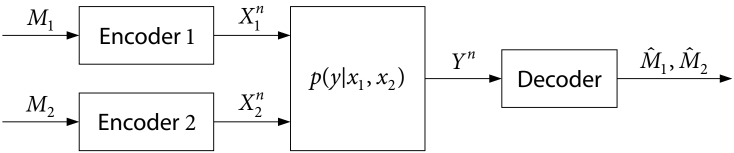
Multiple access channel



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- (R_1, R_2) achievable if $\exists$ $(2^{nR_1}, 2^{nR_2}, n)$ codes such that $\lim_{n\to\infty} P_e^{(n)} = 0$
- Capacity region $\mathscr{C}$: Closure of the set of achievable rate pairs (R_1, R_2)

Random coding and simultaneous decoding

- Codebook generation:
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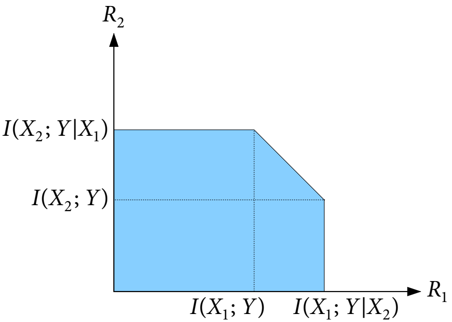

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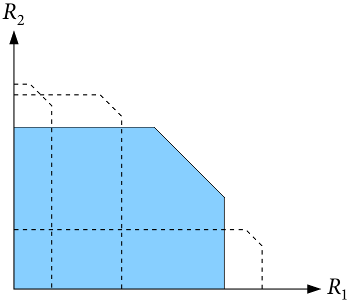

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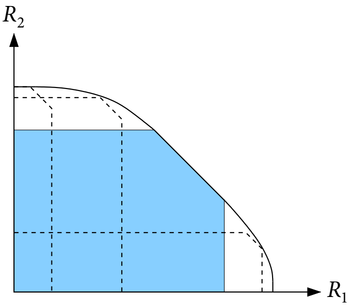

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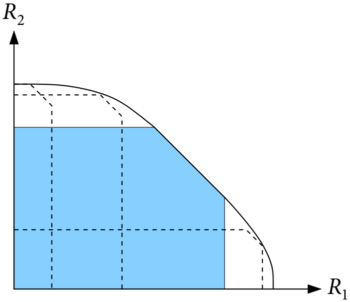
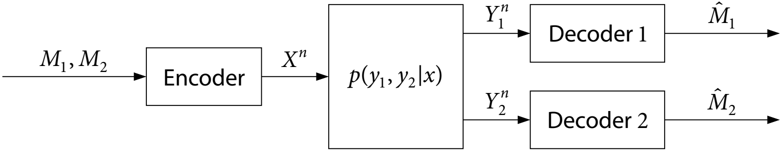


- Capacity region (Ahlswede 1971, Liao 1972)

Broadcast channel



- $(2^{nR_1}, 2^{nR_2}, n)$ code, $P_e^{(n)}$, achievability, $\mathscr{C}$: Same as before

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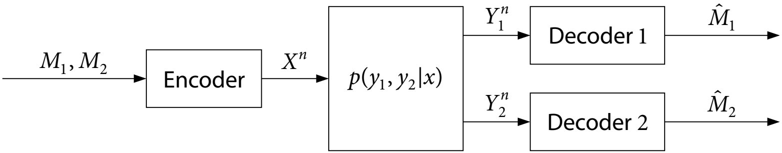
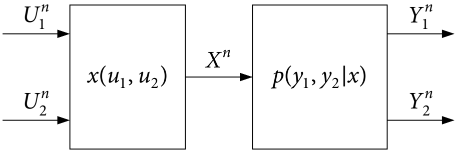


- $(2^{nR_1}, 2^{nR_2}, n)$ code, $P_e^{(n)}$, achievability, $\mathscr{C}$: Same as before
- Capacity region is not known in general

Superposition coding



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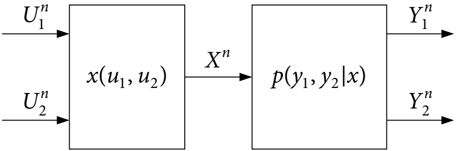
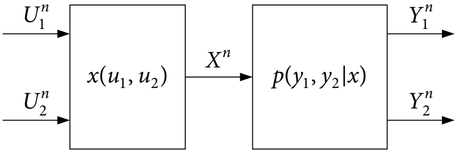


- Independent $U_1^n(m_1)$ and $U_2^n(m_2)$ (as in MAC)

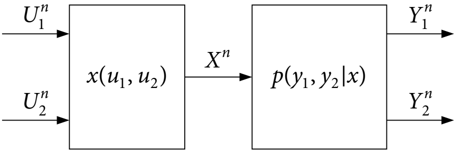


- Independent $U_1^n(m_1)$ and $U_2^n(m_2)$ (as in MAC)
- A simple inner bound: (R_1, R_2) is achievable if

$$R_1 < I(U_1; Y_1),$$
$$R_2 < I(U_2; Y_2)$$

for some $p(u_1)p(u_2)$ and function $x(u_1, u_2)$

Marton coding



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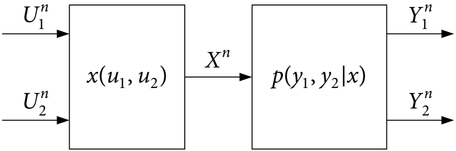

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Marton (1979)

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$$
\begin{aligned}
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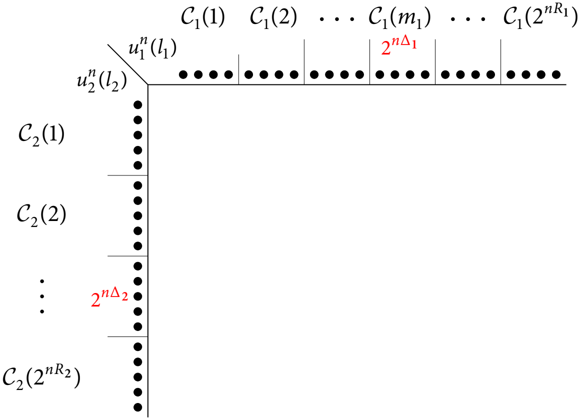
Multicoding and joint typicality encoding

- For each m_j, generate a subcodebook $\mathcal{C}_j(m_j)$ consisting of $u_j^n(l_j) \sim \prod_{i=1}^n p_{U_j}(u_{ji})$
- For each (m_1, m_2), find jointly typical $(u_1^n(l_1), u_2^n(l_2)) \in \mathcal{C}_1(m_1) \times \mathcal{C}_2(m_2)$



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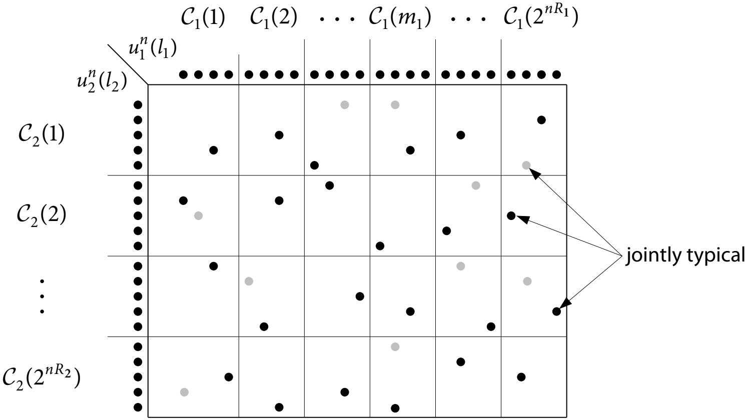

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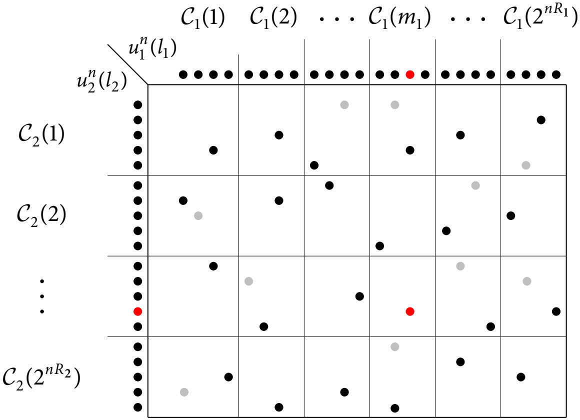

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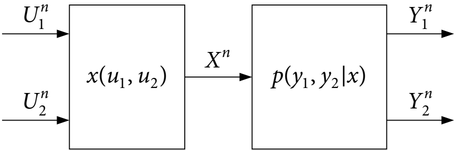

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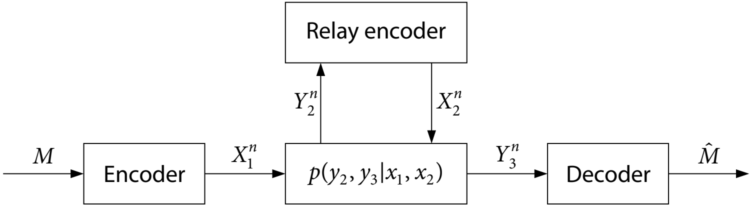
- Essentially the best known scheme for broadcast (dirty paper coding)

Relay channel



- A $(2^{nR}, n)$ code:
  - Message set: $[1 : 2^{nR}]$
  - Encoder: $x_1^n(m)$, $m \in [1 : 2^{nR}]$
  - Decoder: $\hat{m}(y_3^n)$

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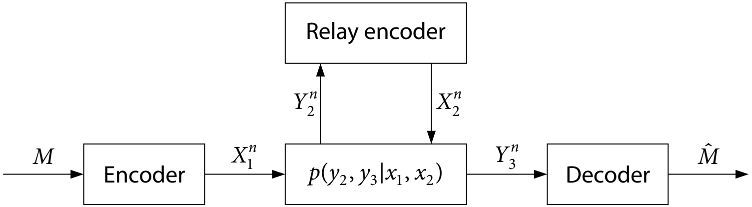

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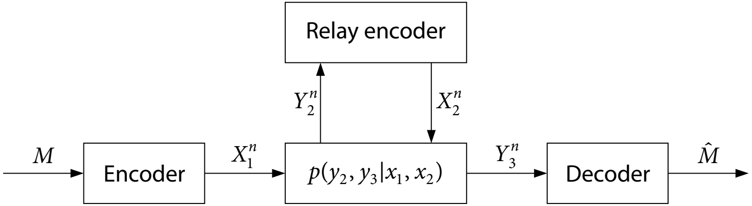

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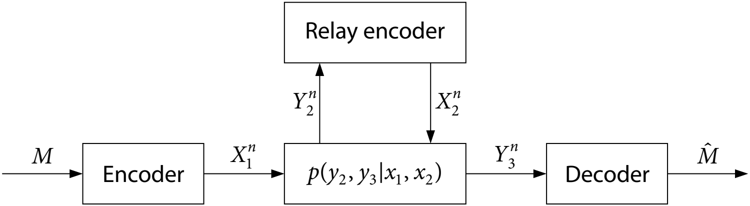

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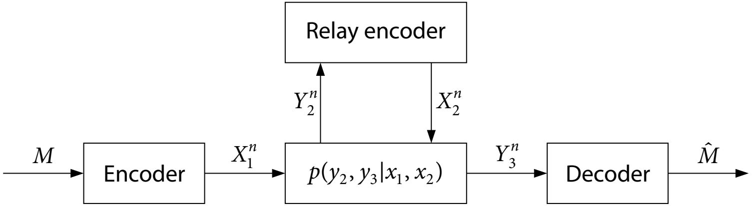

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- Capacity C: the supremum of all achievable rates (not known in general)

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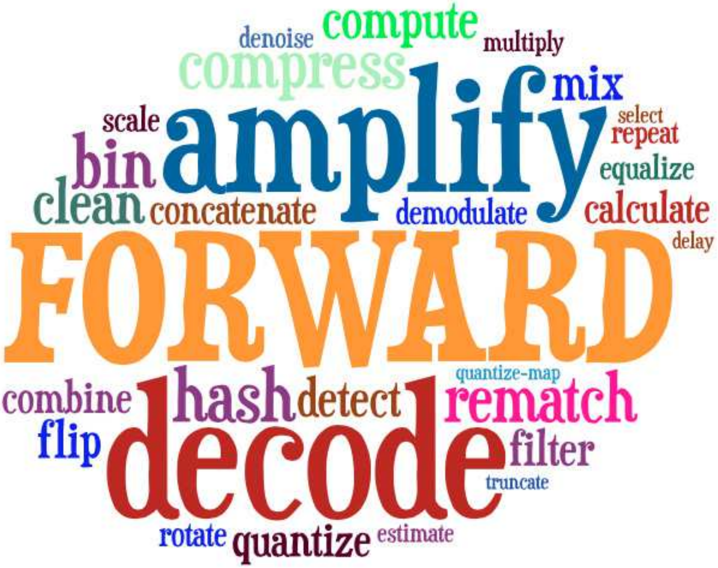

http://circuit.ucsd.edu/~yhk/relaying.html

Cutset upper bound

Cover–El Gamal (1979)

$$C \leq R_{\text{CS}} = \max_{p(x_1,x_2)} \min\left\{I(X_1, X_2; Y_3),\ I(X_1; Y_2, Y_3 | X_2)\right\}$$

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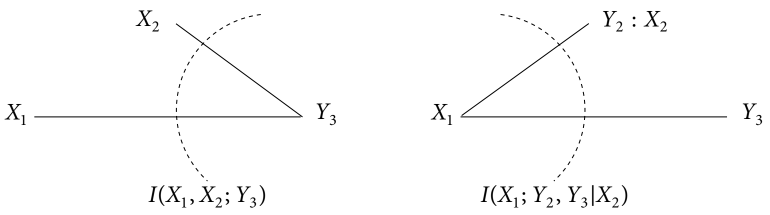
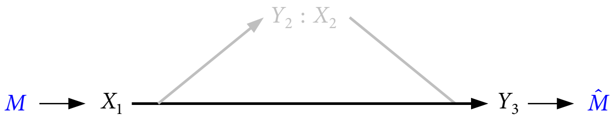


Cooperative MAC bound

Cooperative BC bound

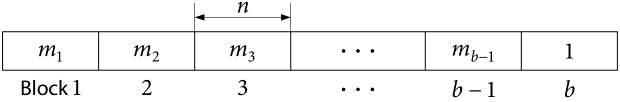
Direct transmission



$$C \geq \max_{p(x_1),\, x_2} I(X_1; Y_3 | X_2 = x_2)$$

Multihop

- Block Markov coding: Send $b-1$ messages over b n-transmission blocks



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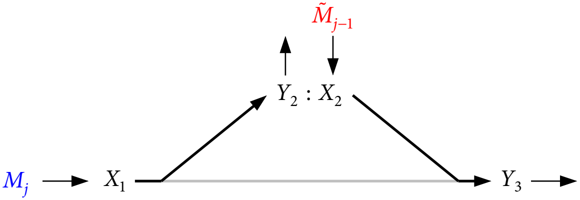
n

m_1	m_2	m_3	$\cdots$	m_{b-1}	1
Block 1	2	3	$\cdots$	$b-1$	b



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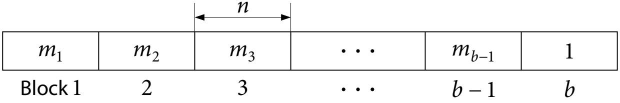

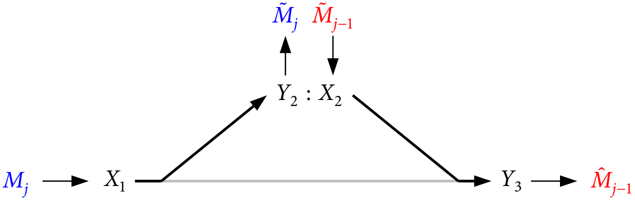

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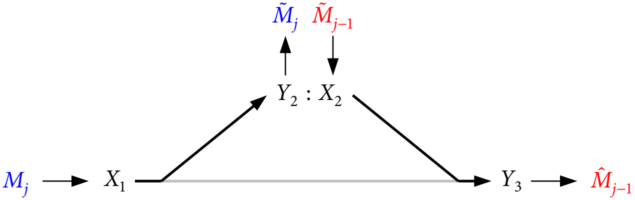
(block length: n)



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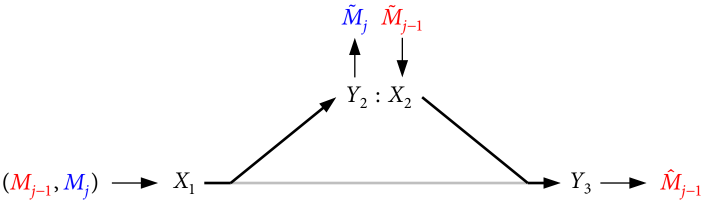

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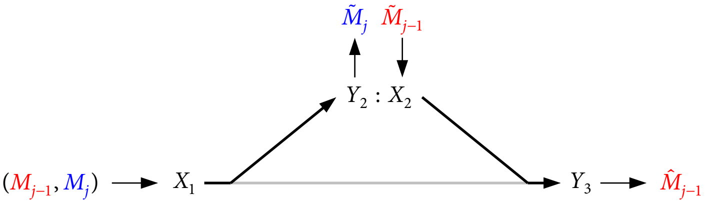

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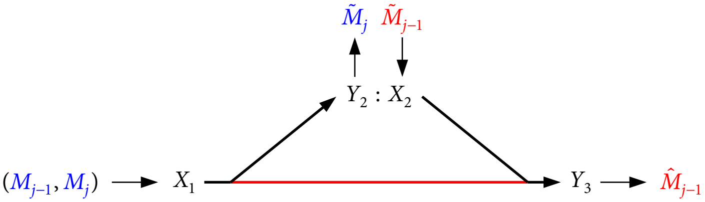
Decode–forward

- Also utilize the direct path via more sophisticated decoding



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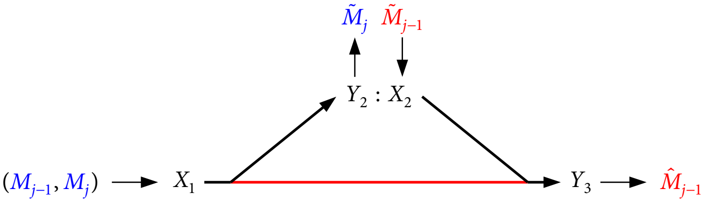

Block	1	2	3	...	$b-1$	b
X_1	$x_1^n(m_1\|1)$	$x_1^n(m_2\|m_1)$	$x_1^n(m_3\|m_2)$	...	$x_1^n(m_{b-1}\|m_{b-2})$	$x_1^n(1\|m_{b-1})$
Y_2						
X_2						
Y_3						

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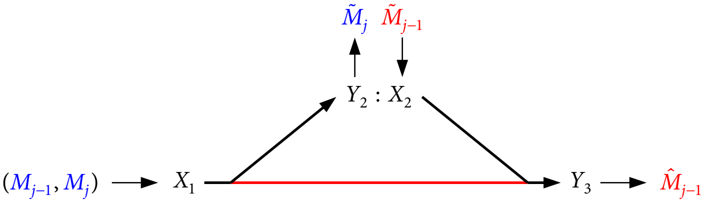

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X_1	$x_1^n(m_1 \vert 1)$	$x_1^n(m_2 \vert m_1)$	$x_1^n(m_3 \vert m_2)$	…	$x_1^n(m_{b-1} \vert m_{b-2})$	$x_1^n(1 \vert m_{b-1})$
Y_2	$\tilde{m}_1 \to$	$\tilde{m}_2 \to$	$\tilde{m}_3 \to$	…	$\tilde{m}_{b-1}$	$\emptyset$
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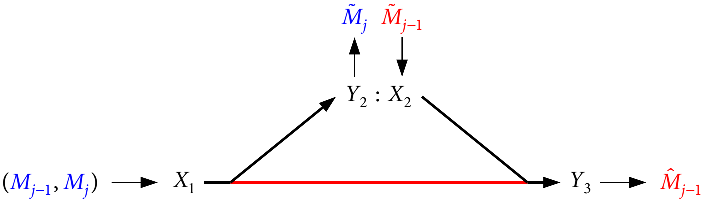
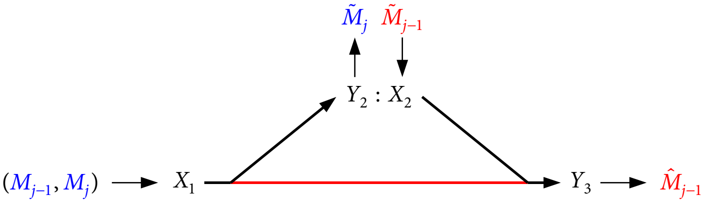


Block	1	2	3	…	$b-1$	b
X_1	$x_1^n(m_1\|1)$	$x_1^n(m_2\|m_1)$	$x_1^n(m_3\|m_2)$	…	$x_1^n(m_{b-1}\|m_{b-2})$	$x_1^n(1\|m_{b-1})$
Y_2	$\tilde{m}_1 \rightarrow$	$\tilde{m}_2 \rightarrow$	$\tilde{m}_3 \rightarrow$	…	$\tilde{m}_{b-1}$	$\emptyset$
X_2	$x_2^n(1)$	$x_2^n(\tilde{m}_1)$	$x_2^n(\tilde{m}_2)$	…	$x_2^n(\tilde{m}_{b-2})$	$x_2^n(\tilde{m}_{b-1})$
Y_3						

Decode–forward

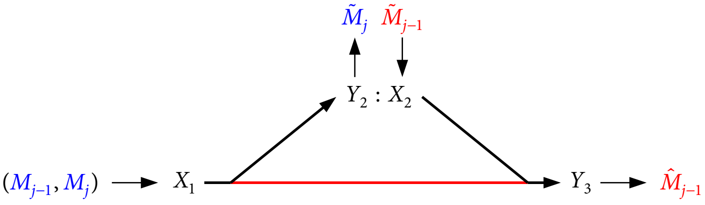
- Also utilize the direct path via more sophisticated decoding



Block	1	2	3	…	$b-1$	b
X_1	$x_1^n(m_1\|1)$	$x_1^n(m_2\|m_1)$	$x_1^n(m_3\|m_2)$	…	$x_1^n(m_{b-1}\|m_{b-2})$	$x_1^n(1\|m_{b-1})$
Y_2	$\tilde{m}_1 \rightarrow$	$\tilde{m}_2 \rightarrow$	$\tilde{m}_3 \rightarrow$	…	$\tilde{m}_{b-1}$	$\emptyset$
X_2	$x_2^n(1)$	$x_2^n(\tilde{m}_1)$	$x_2^n(\tilde{m}_2)$	…	$x_2^n(\tilde{m}_{b-2})$	$x_2^n(\tilde{m}_{b-1})$
Y_3	$\emptyset$	$\hat{m}_1$	$\leftarrow \hat{m}_2$	…	$\leftarrow \hat{m}_{b-2}$	$\leftarrow \hat{m}_{b-1}$

Decode–forward

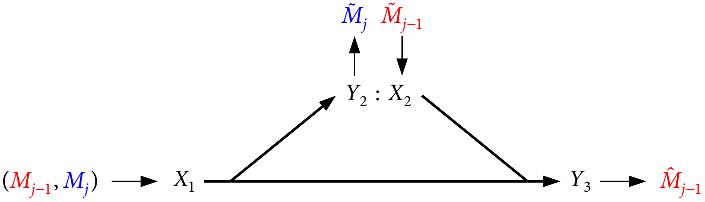
- Also utilize the direct path via more sophisticated decoding



Cover–El-Gamal (1979)

$$C \geq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2 | X_2)\}$$

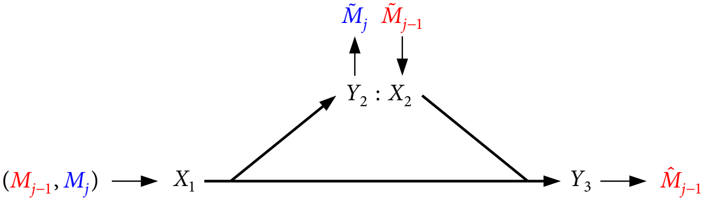
Decode–forward



- Decode–forward: $C \geq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2|X_2)\}$

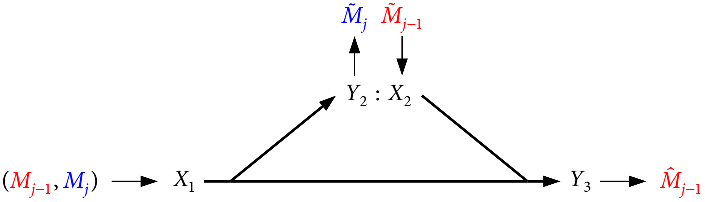
  Cutset bound: $C \leq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2, Y_3|X_2)\}$

Decode–forward



- Decode–forward: $C \geq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2|X_2)\}$
  Cutset bound: $C \leq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2, Y_3|X_2)\}$
- Performs well when the relay is stronger than the receiver

Decode–forward



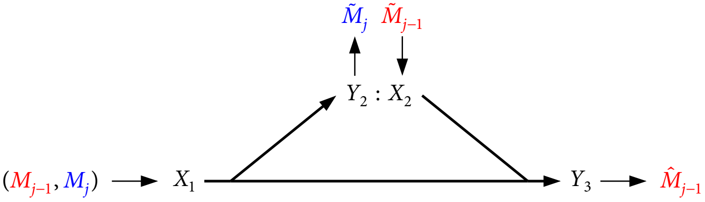
- Decode–forward: $C \geq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2|X_2)\}$

  Cutset bound: $C \leq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2, Y_3|X_2)\}$

- Performs well when the relay is stronger than the receiver

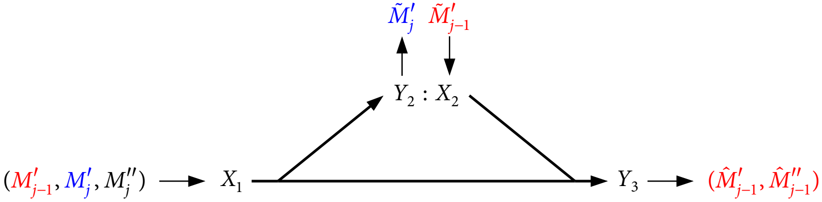
- But worse than even direct transmission if the relay is weaker

Decode–forward



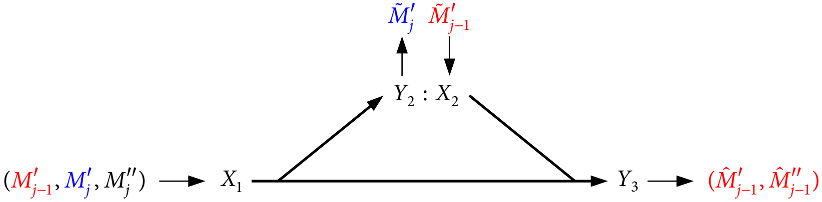
- Decode–forward: $C \geq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2|X_2)\}$
  Cutset bound: $C \leq \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2, Y_3|X_2)\}$
- Performs well when the relay is stronger than the receiver
- But worse than even direct transmission if the relay is weaker
- Solutions:
  - Partial decode–forward: Recover only part of the message
  - Compress–forward: Do not recover the message at all

Partial decode–forward



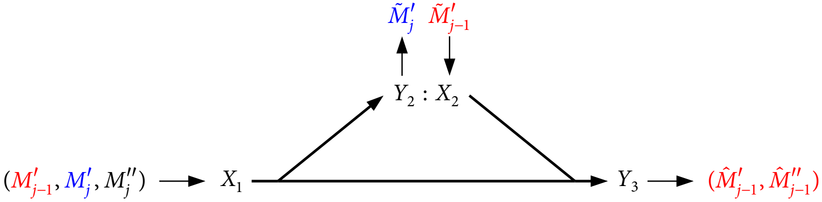
- Codebook structure: $(u^n(m'_j|m'_{j-1}), x_1^n(m'_j, m''_j|m'_{j-1}), x_2^n(m'_{j-1}))$

Partial decode–forward



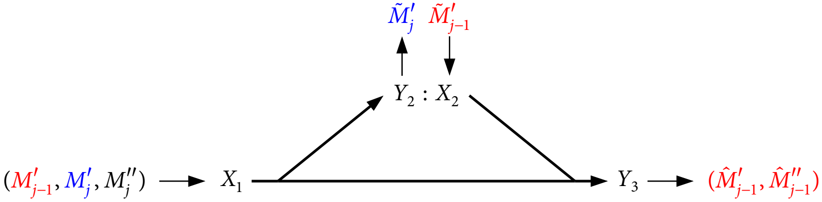
- Codebook structure: $(u^n(m'_j|m'_{j-1}), x_1^n(m'_j, m''_j|m'_{j-1}), x_2^n(m'_{j-1}))$
- Decode–forward of M'_j over $p(y_2, y_3|u, x_2)$

Partial decode–forward



- Codebook structure: $(u^n(m'_j|m'_{j-1}), x_1^n(m'_j, m''_j|m'_{j-1}), x_2^n(m'_{j-1}))$
- Decode–forward of M'_j over $p(y_2, y_3|u, x_2)$
- Direct transmission of M''_j over $p(y_3|x_1, u, x_2)$

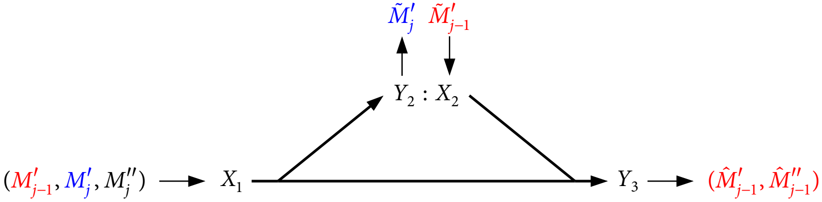
Partial decode–forward



Cover–El Gamal (1979)

$$C \geq R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(U; Y_2|X_2) + I(X_1; Y_3|X_2, U)\}$$

Partial decode–forward



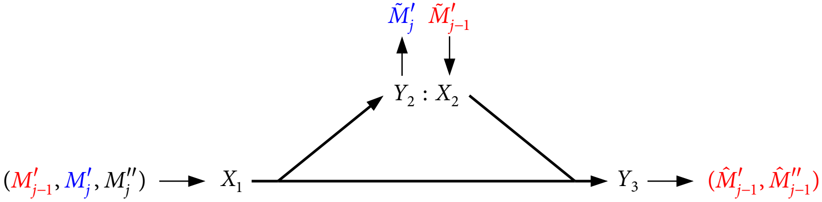
Cover–El Gamal (1979)

$$C \geq R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(U; Y_2|X_2) + I(X_1; Y_3|X_2, U)\}$$

- Direct transmission ($U = \emptyset, M' = \emptyset$):

$$R_{\text{DT}} = \max_{p(x_1)} I(X_1; Y_3)$$

Partial decode–forward



Cover–El Gamal (1979)

$$C \geq R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(U; Y_2|X_2) + I(X_1; Y_3|X_2, U)\}$$

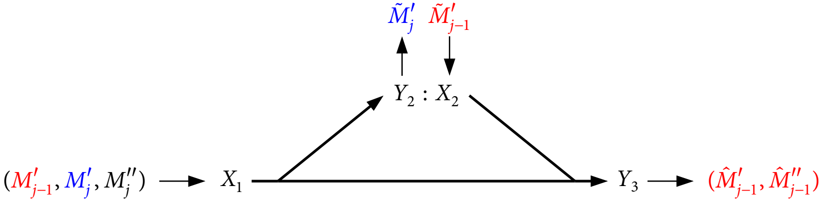
- Direct transmission ($U = \emptyset$, $M' = \emptyset$):

$$R_{\text{DT}} = \max_{p(x_1)} I(X_1; Y_3)$$

- Decode–forward ($U = X_1$, $M'' = \emptyset$):

$$R_{\text{DF}} = \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2|X_2)\}$$

Partial decode–forward



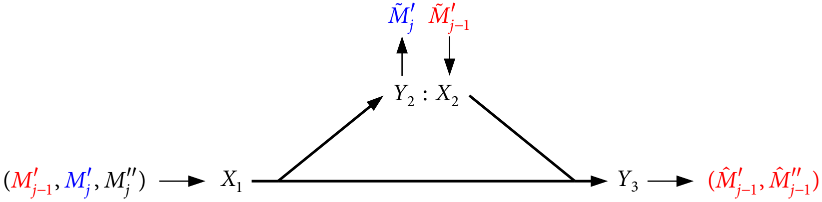
Cover–El Gamal (1979)

$$C \geq R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(U; Y_2 | X_2) + I(X_1; Y_3 | X_2, U)\}$$

- Alternative representation (El Gamal–Aref 1982):

$$R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; U, Y_3 | X_2) - I(U; X_1 | X_2, Y_2)\}$$

Partial decode–forward



Cover–El Gamal (1979)

$$C \geq R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(U; Y_2|X_2) + I(X_1; Y_3|X_2, U)\}$$

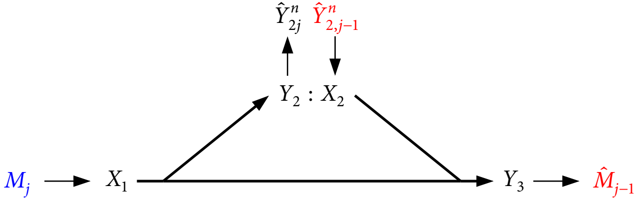
- Alternative representation (El Gamal–Aref 1982):

$$R_{\text{PDF}} = \max_{p(u,x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; U, Y_3|X_2) - I(U; X_1|X_2, Y_2)\}$$

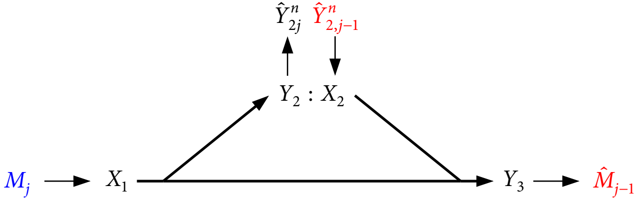
- Comparison to the cutset bound:

$$R_{\text{CS}} = \max_{p(x_1,x_2)} \min\{I(X_1, X_2; Y_3),\ I(X_1; Y_2, Y_3|X_2)\}$$

Compress–forward

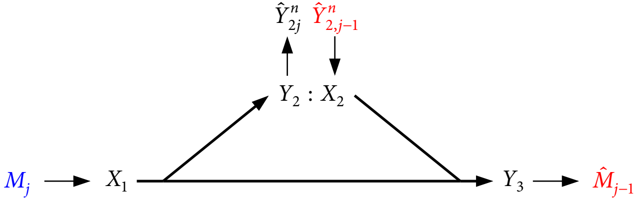


Compress–forward



- Codebook structure: $(x_1^n(m_j), x_2^n(l_{j-1}), \hat{y}_2^n(k_j | l_{j-1}))$

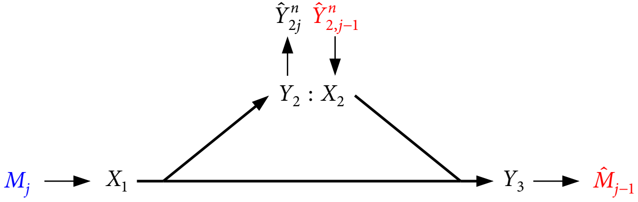
Compress–forward



- Codebook structure: $(x_1^n(m_j), x_2^n(l_{j-1}), \hat{y}_2^n(k_j|l_{j-1}))$

Block	1	2	3	...	$b-1$	b
X_1	$x_1^n(m_1)$	$x_1^n(m_2)$	$x_1^n(m_3)$	...	$x_1^n(m_{b-1})$	$x_1^n(1)$
Y_2						
X_2						
Y_3						

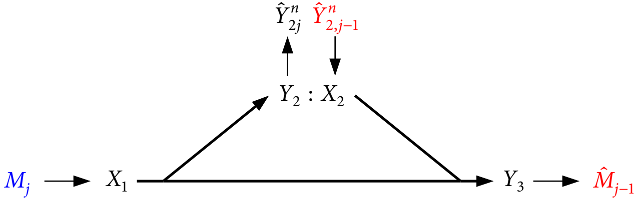
Compress–forward



- Codebook structure: $(x_1^n(m_j), x_2^n(l_{j-1}), \hat{y}_2^n(k_j|l_{j-1}))$

Block	1	2	3	...	$b-1$	b
X_1	$x_1^n(m_1)$	$x_1^n(m_2)$	$x_1^n(m_3)$	...	$x_1^n(m_{b-1})$	$x_1^n(1)$
Y_2	$\hat{y}_2^n(k_1\|1), l_1$	$\hat{y}_2^n(k_2\|l_1), l_2$	$\hat{y}_2^n(k_3\|l_2), l_3$	...	$\hat{y}_2^n(k_{b-1}\|l_{b-2}), l_{b-1}$	$\emptyset$
X_2						
Y_3						

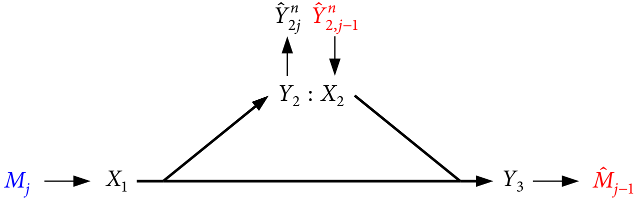
Compress–forward



- Codebook structure: $(x_1^n(m_j), x_2^n(l_{j-1}), \hat{y}_2^n(k_j|l_{j-1}))$

Block	1	2	3	...	$b-1$	b
X_1	$x_1^n(m_1)$	$x_1^n(m_2)$	$x_1^n(m_3)$	...	$x_1^n(m_{b-1})$	$x_1^n(1)$
Y_2	$\hat{y}_2^n(k_1|1), l_1$	$\hat{y}_2^n(k_2|l_1), l_2$	$\hat{y}_2^n(k_3|l_2), l_3$	...	$\hat{y}_2^n(k_{b-1}|l_{b-2}), l_{b-1}$	$\emptyset$
X_2	$x_2^n(1)$	$x_2^n(l_1)$	$x_2^n(l_2)$	...	$x_2^n(l_{b-2})$	$x_2^n(l_{b-1})$
Y_3						

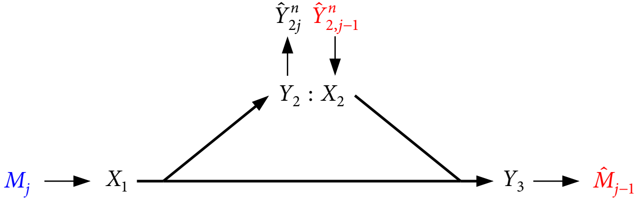
Compress–forward



- Codebook structure: $(x_1^n(m_j), x_2^n(l_{j-1}), \hat{y}_2^n(k_j|l_{j-1}))$

Block	1	2	3	…	$b-1$	b
X_1	$x_1^n(m_1)$	$x_1^n(m_2)$	$x_1^n(m_3)$	…	$x_1^n(m_{b-1})$	$x_1^n(1)$
Y_2	$\hat{y}_2^n(k_1\|1), l_1$	$\hat{y}_2^n(k_2\|l_1), l_2$	$\hat{y}_2^n(k_3\|l_2), l_3$	…	$\hat{y}_2^n(k_{b-1}\|l_{b-2}), l_{b-1}$	$\emptyset$
X_2	$x_2^n(1)$	$x_2^n(l_1)$	$x_2^n(l_2)$	…	$x_2^n(l_{b-2})$	$x_2^n(l_{b-1})$
Y_3	$\emptyset$	$\hat{l}_1, \hat{k}_1, \hat{m}_1$	$\hat{l}_2, \hat{k}_2, \hat{m}_2$	…	$\hat{l}_{b-2}, \hat{k}_{b-2}, \hat{m}_{b-2}$	$\hat{l}_{b-1}, \hat{k}_{b-1}, \hat{m}_{b-1}$

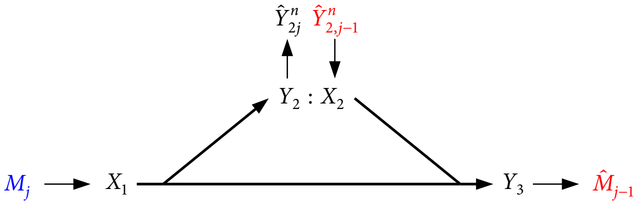
Compress–forward



Cover–El Gamal (1979)

$$C \geq R_{\mathrm{CF}} = \max_{p(x_1)p(x_2)p(\hat{y}_2|y_2,x_2):I(X_2;Y_3)\geq I(Y_2;\hat{Y}_2|X_2,Y_3)} I(X_1;\hat{Y}_2,Y_3|X_2)$$

Compress–forward



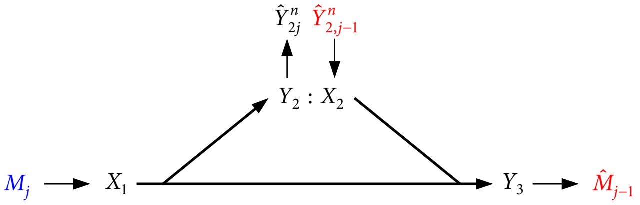
Cover–El Gamal (1979)

$$C \geq R_{\text{CF}} = \max_{p(x_1)p(x_2)p(\hat{y}_2|y_2,x_2):I(X_2;Y_3)\geq I(Y_2;\hat{Y}_2|X_2,Y_3)} I(X_1;\hat{Y}_2,Y_3|X_2)$$

- Alternative representation (El-Gamal–Mohseni–Zahedi 2006):

$$R_{\text{CF}} = \max_{p(x_1)p(x_2)p(\hat{y}_2|y_2,x_2)} \min\{I(X_1,X_2;Y_3) - I(Y_2;\hat{Y}_2|X_1,X_2,Y_3),\ I(X_1;\hat{Y}_2,Y_3|X_2)\}$$

Compress–forward



Cover–El Gamal (1979)

$$C \geq R_{\text{CF}} = \max_{p(x_1)p(x_2)p(\hat{y}_2|y_2,x_2):I(X_2;Y_3)\geq I(Y_2;\hat{Y}_2|X_2,Y_3)} I(X_1;\hat{Y}_2,Y_3|X_2)$$

- Alternative representation (El-Gamal–Mohseni–Zahedi 2006):

$$R_{\text{CF}} = \max_{p(x_1)p(x_2)p(\hat{y}_2|y_2,x_2)} \min\{I(X_1,X_2;Y_3) - I(Y_2;\hat{Y}_2|X_1,X_2,Y_3),\ I(X_1;\hat{Y}_2,Y_3|X_2)\}$$

- Comparison to the cutset bound:

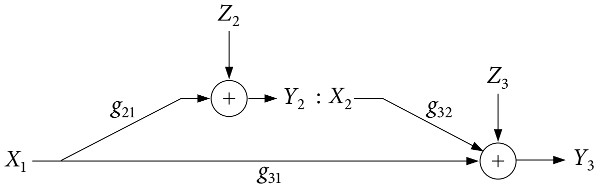
$$R_{\text{CS}} = \max_{p(x_1,x_2)} \min\{I(X_1,X_2;Y_3),\ I(X_1;Y_2,Y_3|X_2)\}$$

Partial decode–forward vs. compress–forward

$$
\begin{aligned}
R_{\text{PDF}} &= \max\min\{I(X_1, X_2; Y_3), \quad -I(U; X_1 | X_2, Y_2) + I(X_1; U,\ Y_3 | X_2)\},\\
R_{\text{CF}} &= \max\min\{I(X_1, X_2; Y_3) - I(Y_2; \hat{Y}_2 | X_1, X_2, Y_3), \quad I(X_1; \hat{Y}_2, Y_3 | X_2)\},\\
R_{\text{CS}} &= \max\min\{I(X_1, X_2; Y_3), \quad I(X_1; Y_2, Y_3 | X_2)\}
\end{aligned}
$$

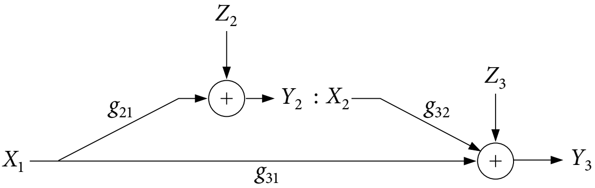
Partial decode–forward vs. compress–forward

$$
\begin{aligned}
R_{\mathrm{PDF}} &= \max\min\{I(X_1, X_2; Y_3), \quad -I(U; X_1 | X_2, Y_2) + I(X_1; U, Y_3 | X_2)\}, \\
R_{\mathrm{CF}} &= \max\min\{I(X_1, X_2; Y_3) - I(Y_2; \hat{Y}_2 | X_1, X_2, Y_3), \quad I(X_1; \hat{Y}_2, Y_3 | X_2)\}, \\
R_{\mathrm{CS}} &= \max\min\{I(X_1, X_2; Y_3), \quad I(X_1; Y_2, Y_3 | X_2)\}
\end{aligned}
$$



Partial decode–forward vs. compress–forward

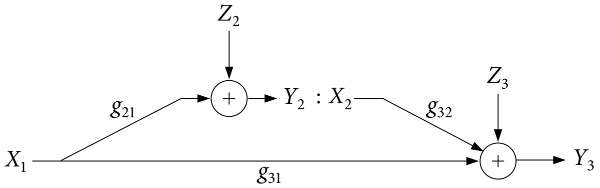
$$\begin{aligned}
R_{\text{PDF}} &= \max\min\{I(X_1, X_2; Y_3), \quad -I(U; X_1 | X_2, Y_2) + I(X_1; U, Y_3 | X_2)\},\\
R_{\text{CF}} &= \max\min\{I(X_1, X_2; Y_3) - I(Y_2; \hat{Y}_2 | X_1, X_2, Y_3), \quad I(X_1; \hat{Y}_2, Y_3 | X_2)\},\\
R_{\text{CS}} &= \max\min\{I(X_1, X_2; Y_3), \quad I(X_1; Y_2, Y_3 | X_2)\}
\end{aligned}$$



- $\Delta_{\text{PDF}} = R_{\text{CS}} - R_{\text{PDF}} \leq 1/2$ and $\Delta_{\text{CF}} = R_{\text{CS}} - R_{\text{CF}} \leq 1/2$ (cf. Jin–Kim 2014)

Partial decode–forward vs. compress–forward

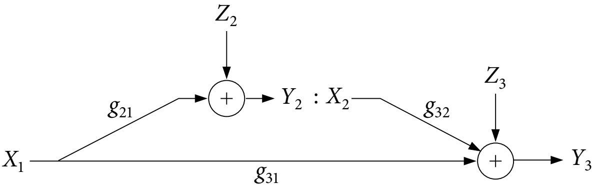
$$R_{\text{PDF}} = \max\min\{I(X_1, X_2; Y_3), \quad -I(U; X_1 | X_2, Y_2) + I(X_1; U, Y_3 | X_2)\},$$
$$R_{\text{CF}} = \max\min\{I(X_1, X_2; Y_3) - I(Y_2; \hat{Y}_2 | X_1, X_2, Y_3), \quad I(X_1; \hat{Y}_2, Y_3 | X_2)\},$$
$$R_{\text{CS}} = \max\min\{I(X_1, X_2; Y_3), \quad I(X_1; Y_2, Y_3 | X_2)\}$$



- $\Delta_{\text{PDF}} = R_{\text{CS}} - R_{\text{PDF}} \leq 1/2$ and $\Delta_{\text{CF}} = R_{\text{CS}} - R_{\text{CF}} \leq 1/2$ (cf. Jin–Kim 2014)
- Choice of U

Partial decode–forward vs. compress–forward

$$R_{\text{PDF}} = \max\min\{I(X_1, X_2; Y_3), \quad -I(U; X_1|X_2, Y_2) + I(X_1; U, Y_3|X_2)\},$$
$$R_{\text{CF}} = \max\min\{I(X_1, X_2; Y_3) - I(Y_2; \hat{Y}_2|X_1, X_2, Y_3), \quad I(X_1; \hat{Y}_2, Y_3|X_2)\},$$
$$R_{\text{CS}} = \max\min\{I(X_1, X_2; Y_3), \quad I(X_1; Y_2, Y_3|X_2)\}$$



- $\Delta_{\text{PDF}} = R_{\text{CS}} - R_{\text{PDF}} \le 1/2$ and $\Delta_{\text{CF}} = R_{\text{CS}} - R_{\text{CF}} \le 1/2$ (cf. Jin–Kim 2014)
- Choice of U: $U = g_{21}X_1 + \mathrm{N}(0,1) \sim Y_2$ (cf. Lim–Kim–Kim 2014a)

Partial decode–forward vs. compress–forward

$$\begin{aligned}
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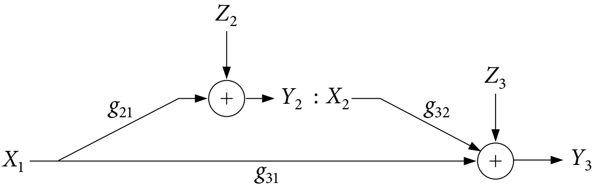
\end{aligned}$$



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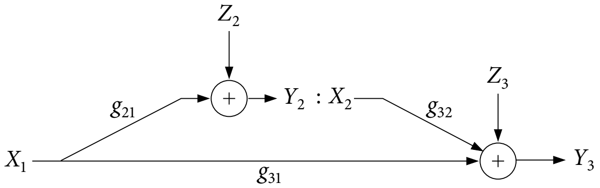

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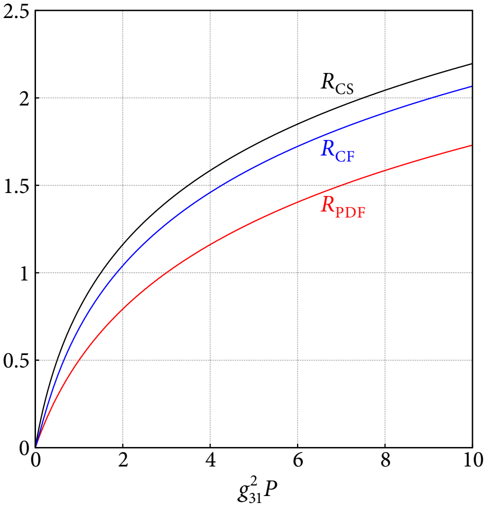
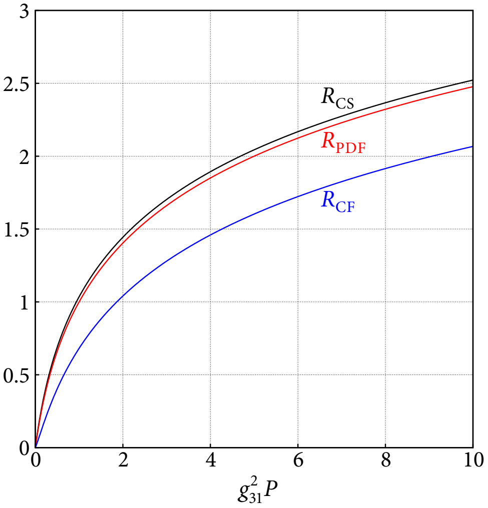
- Choice of $\hat{Y}_2$: $\hat{Y}_2 = Y_2 + \text{N}(0,1)$ (cf. Avestimehr–Diggavi–Tse 2011)

Comparison of coding schemes



$$g_{21} = g_{31}, g_{32} = 2g_{31}$$

Comparison of coding schemes



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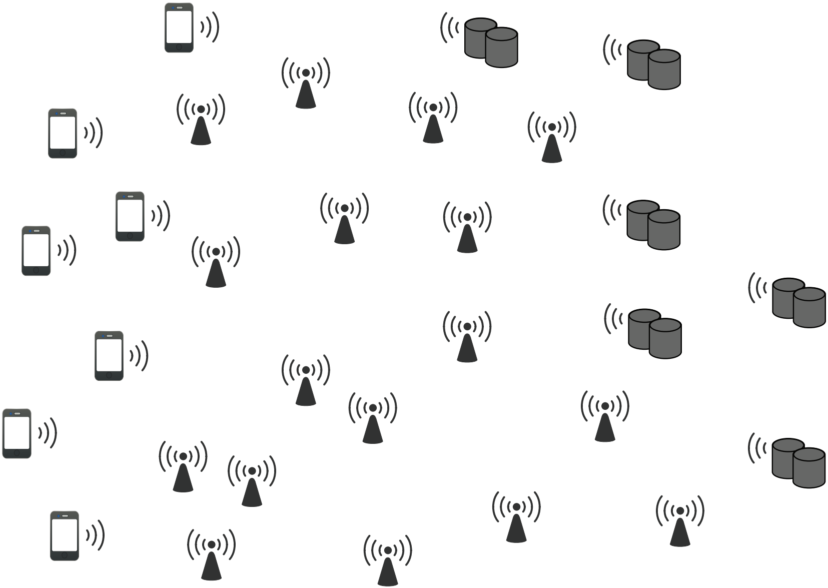
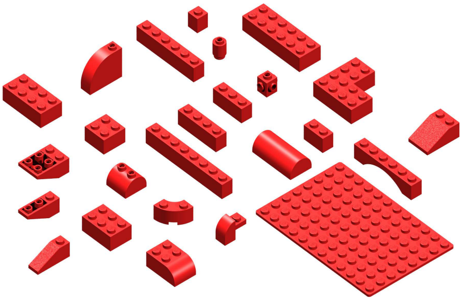
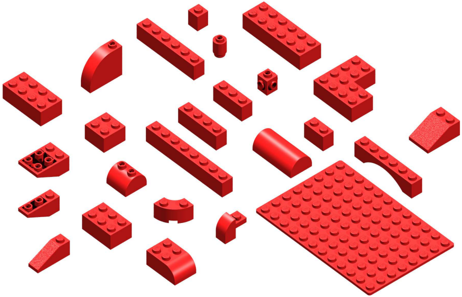


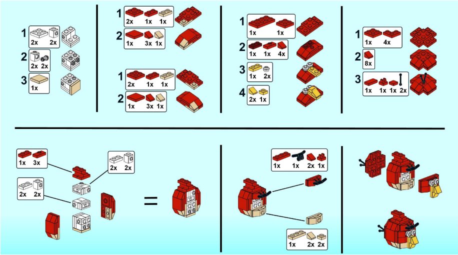
Relaying for networks



- Simultaneous decoding
- Superposition coding
- Multicoding
- Block Markov coding
- Decode–forward
- Compress–forward

Relaying for networks





- Simultaneous decoding
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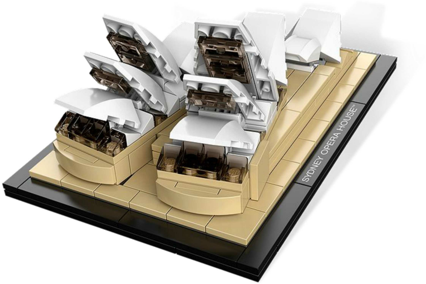
- Noisy network coding
- Distributed decode–forward

Putting things together

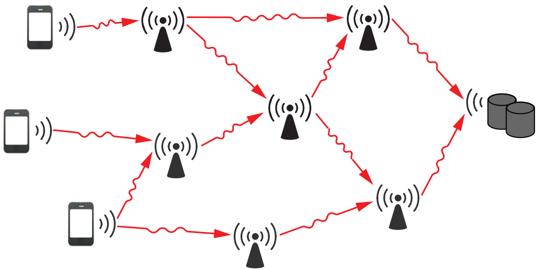
- Often vanilla extensions do not work

Putting things together

- Often vanilla extensions do not work
- Building blocks should be refined



Multiple access relay network



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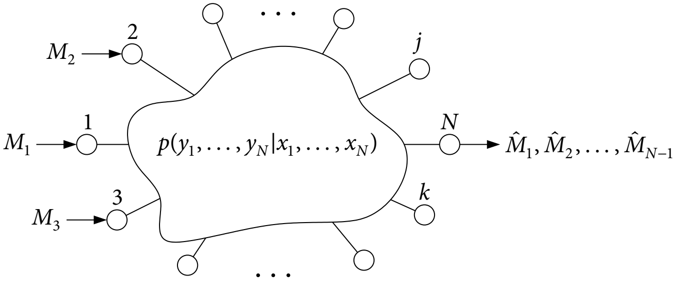

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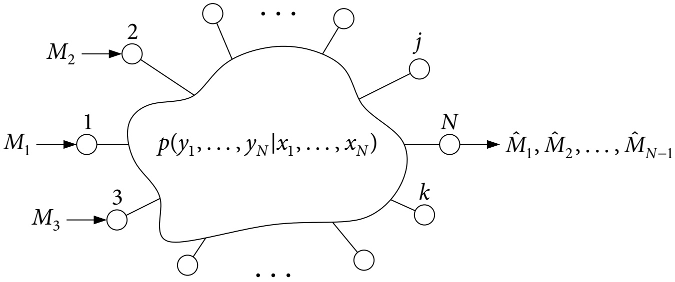
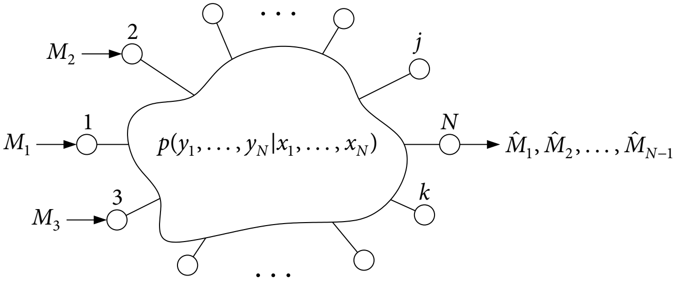


- Multihop uplink communication

Multiple access relay network



- Multihop uplink communication
- A $(2^{nR_1}, \ldots, 2^{nR_{N-1}}, n)$ code:
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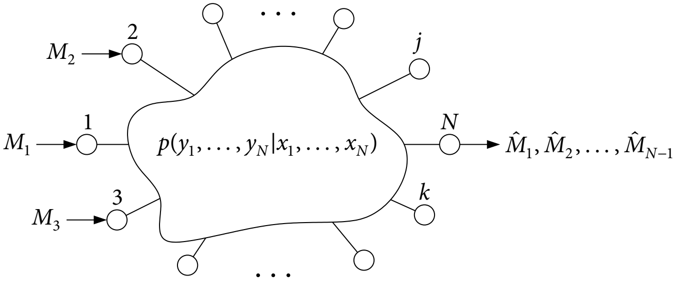
Multiple access relay network



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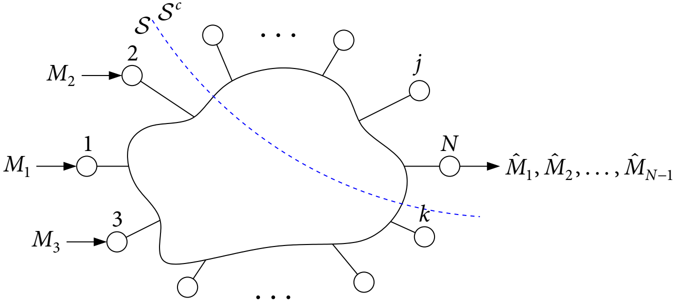
Cutset outer bound

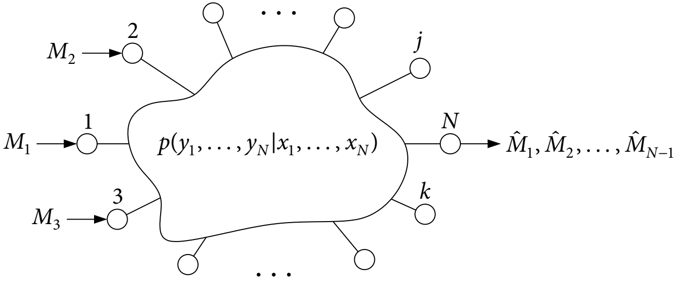


El Gamal (1981)

$$R(\mathcal{S}) := \sum_{j \in \mathcal{S}} R_j \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $p(x^N)$

Noisy network coding = CF for networks



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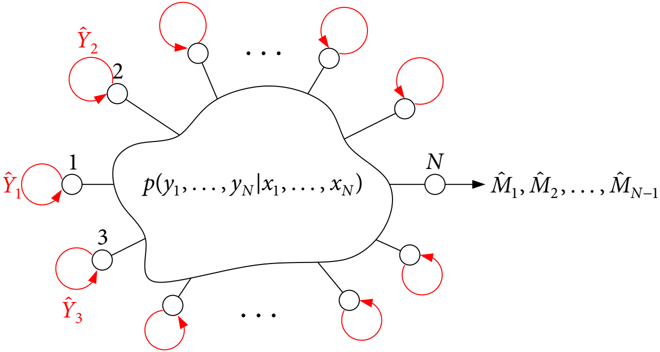

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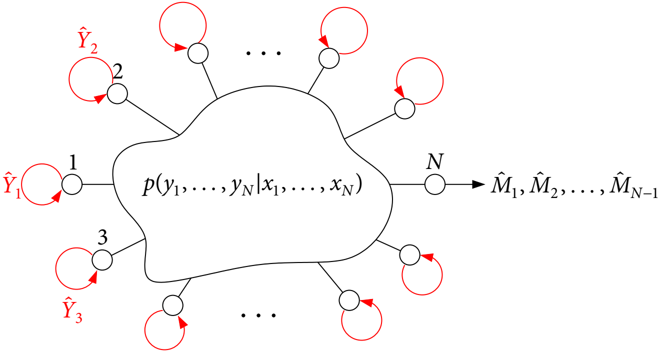

- Block Markov coding:
- Relay compression:
- Auxiliary index:
- Simultaneous decoding:

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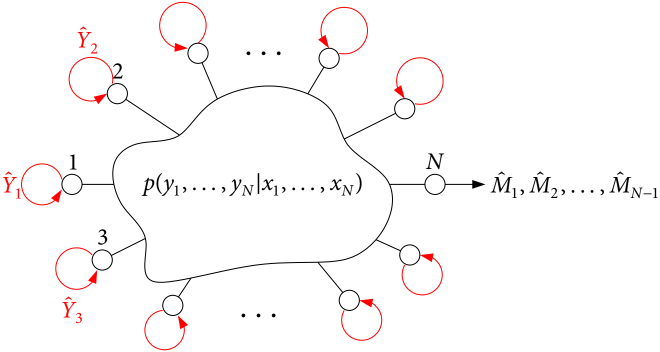

- Block Markov coding: Splitting of M_* over b blocks of n transmissions
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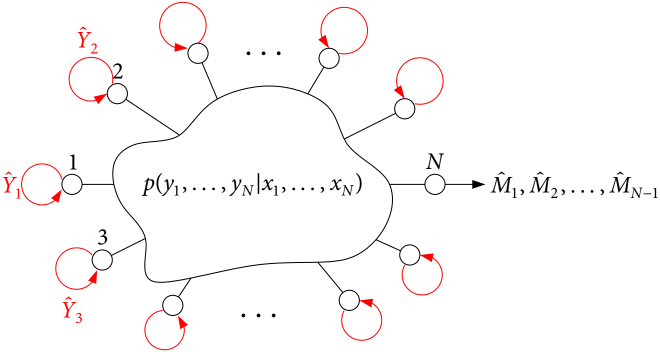

- Block Markov coding: Splitting of M_* over b blocks of n transmissions
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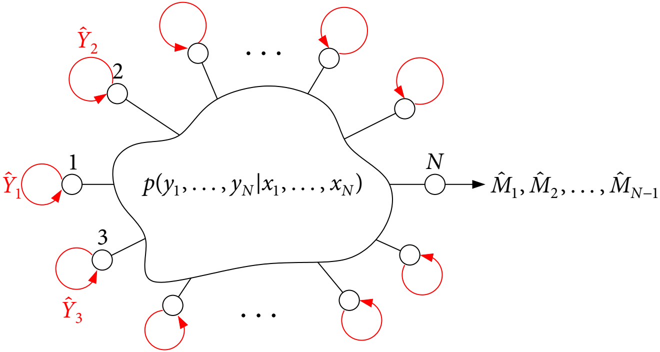

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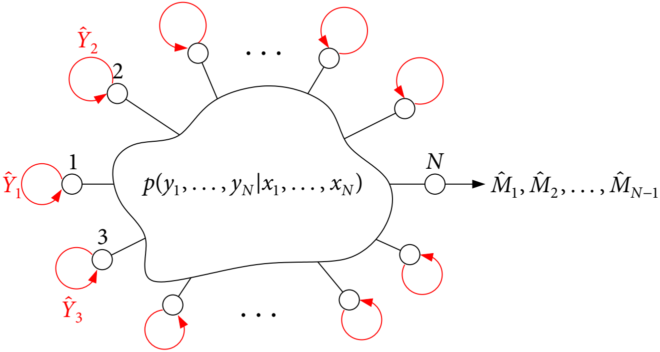

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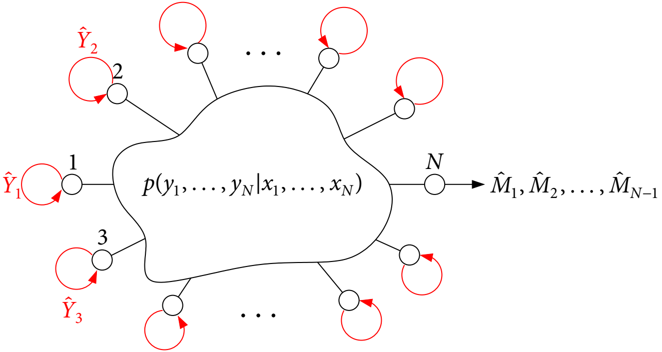
- Simultaneous decoding: Y_N recovers (M_*, L_*)

Noisy network coding = CF for networks

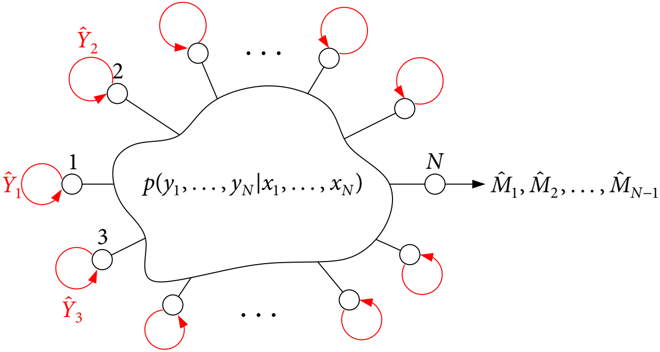


Yassaee–Aref, Lim–Kim–El Gamal–Chung (2011), Hou–Kramer (2012)

$$R(\mathcal{S}) \le I(X(\mathcal{S}); \hat{Y}(\mathcal{S}^c) | X(\mathcal{S}^c)) - I(Y(\mathcal{S}); \hat{Y}(\mathcal{S}) | X^N, \hat{Y}(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $\prod_{k=1}^N p(x_k) p(\hat{y}_k | y_k, x_k)$

Noisy network coding = CF for networks



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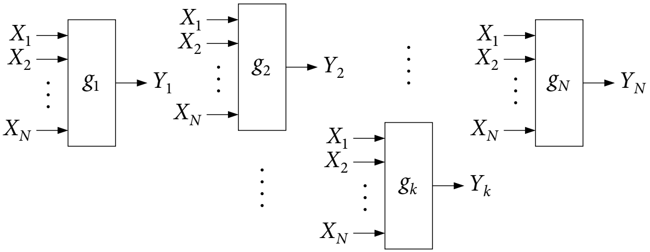
Deterministic network (Avestimehr–Diggavi–Tse 2011)



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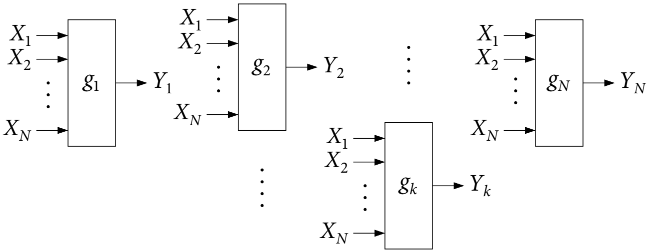

Cutset

$$R(\mathcal{S}) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c))$$

Noisy network coding

$$R(\mathcal{S}) \le I(X(\mathcal{S}); \hat{Y}(\mathcal{S}^c) | X(\mathcal{S}^c)) - I(Y(\mathcal{S}); \hat{Y}(\mathcal{S}) | X^N, \hat{Y}(\mathcal{S}^c))$$

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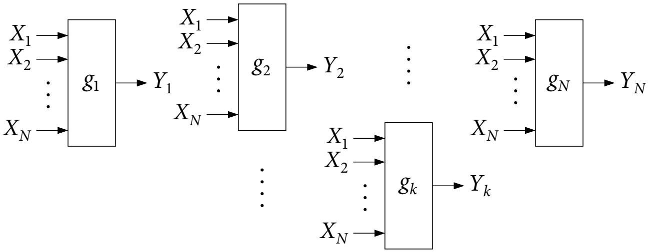

Cutset

$$R(\mathcal{S}) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c))$$
$$= H(Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $p(x^N)$

Noisy network coding

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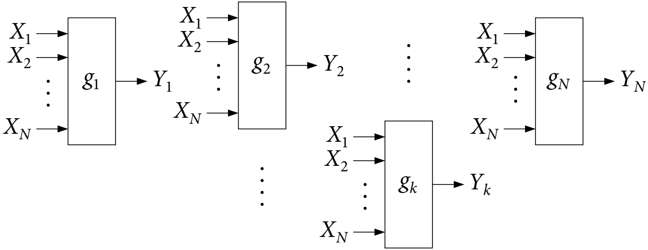

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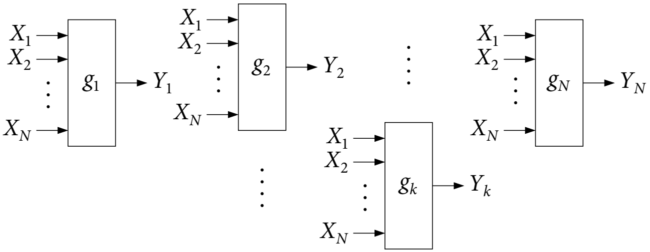

Cutset

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Noisy network coding

$$\begin{aligned} R(\mathcal{S}) &\le I(X(\mathcal{S}); \hat{Y}(\mathcal{S}^c) | X(\mathcal{S}^c)) \\ &\quad - I(Y(\mathcal{S}); \hat{Y}(\mathcal{S}) | X^N, \hat{Y}(\mathcal{S}^c)) \\ &= H(Y(\mathcal{S}^c)|X(\mathcal{S}^c)), \quad \forall \mathcal{S} \end{aligned}$$

for some $\prod_{k=1}^N p(x_k)$

- Tight for graphical networks and deterministic networks with no interference

Gaussian network

- Channel model:

$$Y_k = \sum_j g_{kj} X_j + Z_k, \quad k \in [1:N]$$

Gaussian network

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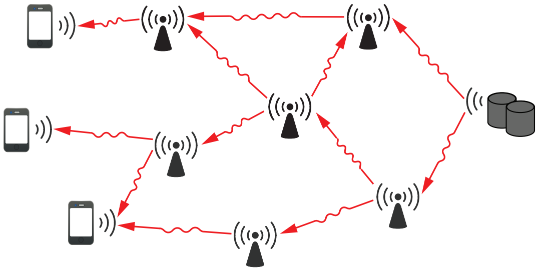
Avestimehr–Diggavi–Tse (2011), Lim–Kim–El Gamal–Chung (2011)

If $(R_1, \ldots, R_N) \in \mathscr{R}_{\mathrm{CS}}$, then $(R_1 - 0.63N, \ldots, R_N - 0.63N) \in \mathscr{R}_{\mathrm{NNC}}$

Broadcast relay network



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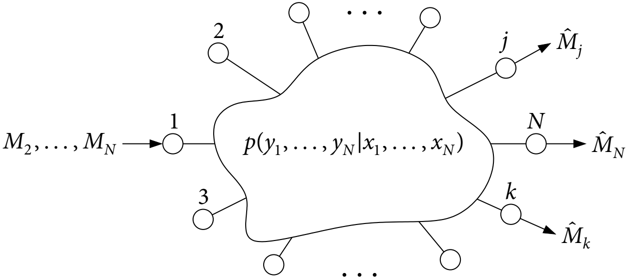

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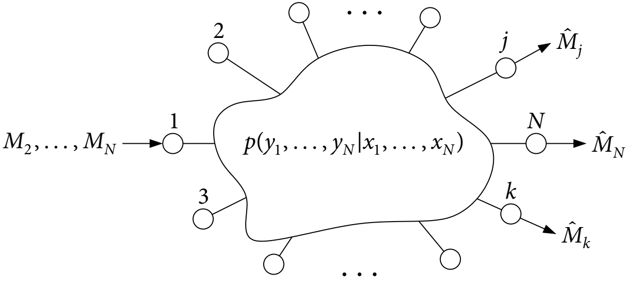
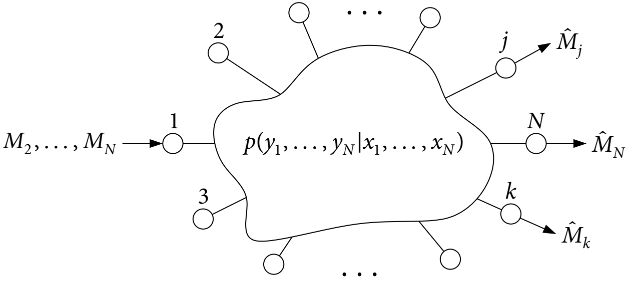


- Multihop downlink communication

Broadcast relay network



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  - Message set: $[1:2^{nR_2}] \times \cdots \times [1:2^{nR_N}]$
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  - Relay encoder: $x_{ji}(y_j^{i-1})$, $j \in [2:N]$, $i \in [1:n]$
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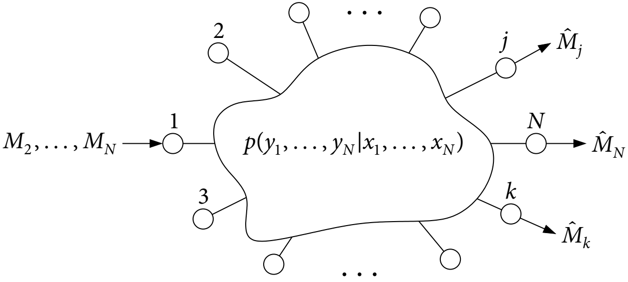
Broadcast relay network



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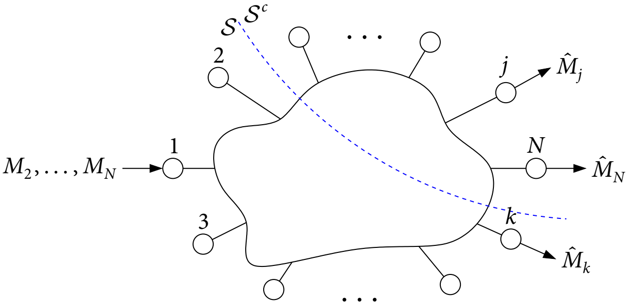
Cutset outer bound

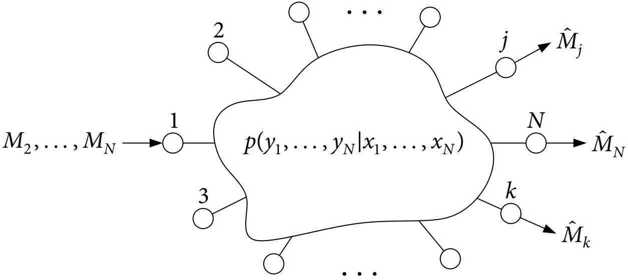


El Gamal (1981)

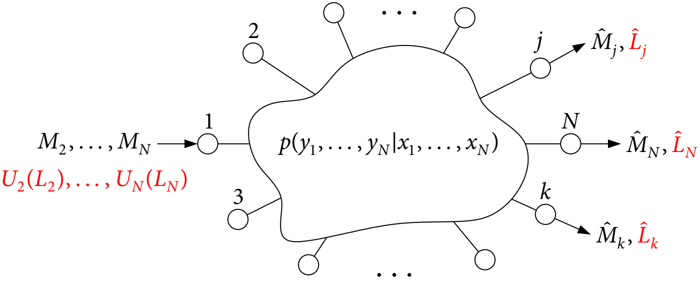
$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $p(x^N)$

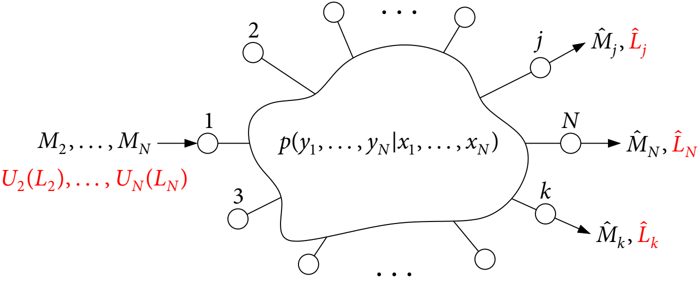
Distributed decode–forward = PDF for networks



Distributed decode–forward = PDF for networks

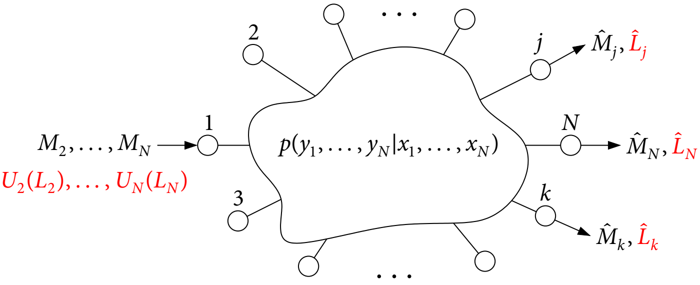


Distributed decode–forward = PDF for networks



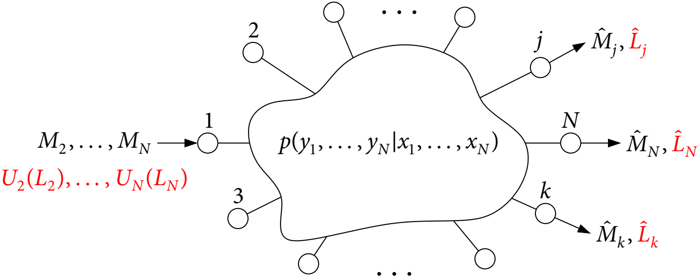
- Block Markov coding:
- Backward encoding:
- Partial decoding:
- Auxiliary index:

Distributed decode–forward = PDF for networks



- Block Markov coding: Splitting of M_* over b blocks of n transmissions
- Backward encoding:
- Partial decoding:
- Auxiliary index:

Distributed decode–forward = PDF for networks



- Block Markov coding: Splitting of M_* over b blocks of n transmissions
- Backward encoding: Multicoding of $X_*(M_*)$ and $U_2(L_2), \dots, U_N(L_N)$
- Partial decoding:
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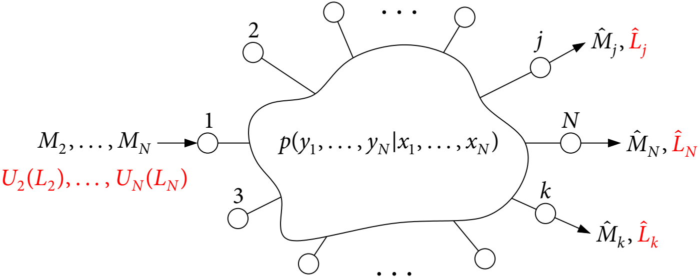
Distributed decode–forward = PDF for networks



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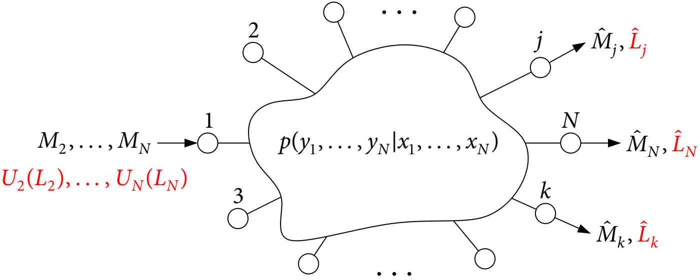
Distributed decode–forward = PDF for networks



- Block Markov coding: Splitting of M_* over b blocks of n transmissions
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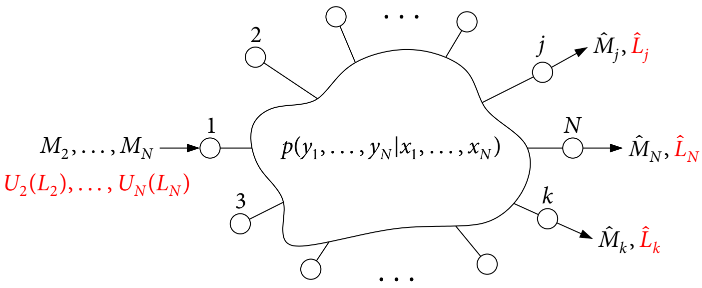
- Partial decoding: Relay k decodes Y_k to recover $U_k(L_k)$
- Auxiliary index: $(U_2(L_2), \ldots, U_N(L_N))$ conveys the messages implicitly

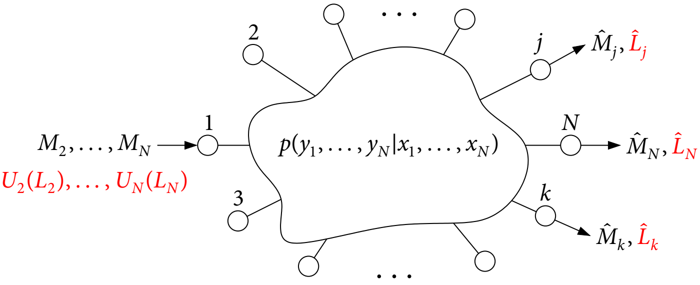
Distributed decode–forward = PDF for networks



- Codebook structure: $x_1^n(\mathbf{m}_j | \mathbf{l}_j, \mathbf{l}_{j-1}), x_k^n(l_{k,j-1}), u_k^n(m_{kj}, l_{kj} | l_{k,j-1})$

Block	1	2	3	$\cdots$	b
X_1					
X_k					
Y_k					

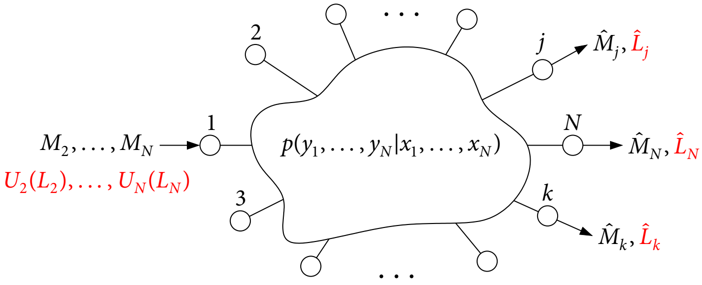
Distributed decode–forward = PDF for networks



- Codebook structure: $x_1^n(\mathbf{m}_j | \mathbf{l}_j, \mathbf{l}_{j-1}), x_k^n(l_{k,j-1}), u_k^n(m_{kj}, l_{kj} | l_{k,j-1})$

Block	1	2	3	$\cdots$	b
X_1	$\mathbf{l}_0$	$\leftarrow \mathbf{l}_1$	$\leftarrow \mathbf{l}_2$	$\cdots$	$\leftarrow \mathbf{l}_{b-1}$
X_k					
Y_k					

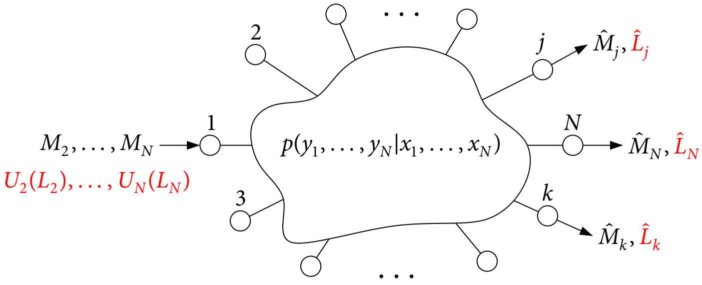
Distributed decode–forward = PDF for networks



- Codebook structure: $x_1^n(\mathbf{m}_j|\mathbf{l}_j, \mathbf{l}_{j-1}), x_k^n(l_{k,j-1}), u_k^n(m_{kj}, l_{kj}|l_{k,j-1})$

Block	1	2	3	$\cdots$	b
X_1	$\mathbf{l}_0$	$\leftarrow \mathbf{l}_1$	$\leftarrow \mathbf{l}_2$	$\cdots$	$\leftarrow \mathbf{l}_{b-1}$
	$x_1^n(\mathbf{m}_1|\mathbf{l}_1, \mathbf{l}_0)$	$x_1^n(\mathbf{m}_2|\mathbf{l}_2, \mathbf{l}_1)$	$x_1^n(\mathbf{m}_3|\mathbf{l}_3, \mathbf{l}_2)$	$\cdots$	$x_1^n(\mathbf{m}_b|\mathbf{l}_b, \mathbf{l}_{b-1})$
X_k					
Y_k					

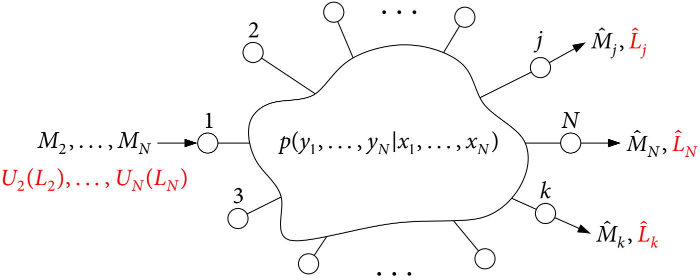
Distributed decode–forward = PDF for networks



- Codebook structure: $x_1^n(\mathbf{m}_j | \mathbf{l}_j, \mathbf{l}_{j-1}), x_k^n(l_{k,j-1}), u_k^n(m_{kj}, l_{kj} | l_{k,j-1})$

Block	1	2	3	$\cdots$	b
X_1	$\mathbf{l}_0$	$\leftarrow \mathbf{l}_1$	$\leftarrow \mathbf{l}_2$	$\cdots$	$\leftarrow \mathbf{l}_{b-1}$
	$x_1^n(\mathbf{m}_1 \vert \mathbf{l}_1, \mathbf{l}_0)$	$x_1^n(\mathbf{m}_2 \vert \mathbf{l}_2, \mathbf{l}_1)$	$x_1^n(\mathbf{m}_3 \vert \mathbf{l}_3, \mathbf{l}_2)$	$\cdots$	$x_1^n(\mathbf{m}_b \vert \mathbf{l}_b, \mathbf{l}_{b-1})$
X_k	$x_k^n(\hat{l}_{k0})$				
Y_k					

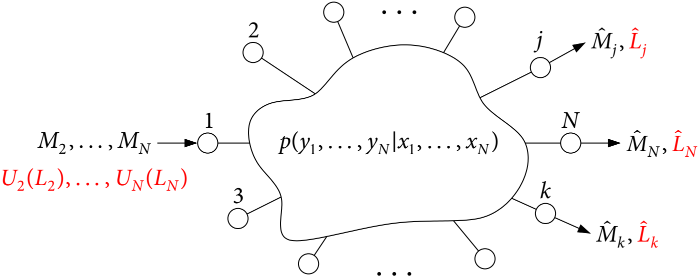
Distributed decode–forward = PDF for networks



- Codebook structure: $x_1^n(\mathbf{m}_j | \mathbf{l}_j, \mathbf{l}_{j-1}), x_k^n(l_{k,j-1}), u_k^n(m_{kj}, l_{kj} | l_{k,j-1})$

Block	1	2	3	$\cdots$	b
X_1	$\mathbf{l}_0$	$\leftarrow \mathbf{l}_1$	$\leftarrow \mathbf{l}_2$	$\cdots$	$\leftarrow \mathbf{l}_{b-1}$
	$x_1^n(\mathbf{m}_1 \vert \mathbf{l}_1, \mathbf{l}_0)$	$x_1^n(\mathbf{m}_2 \vert \mathbf{l}_2, \mathbf{l}_1)$	$x_1^n(\mathbf{m}_3 \vert \mathbf{l}_3, \mathbf{l}_2)$	$\cdots$	$x_1^n(\mathbf{m}_b \vert \mathbf{l}_b, \mathbf{l}_{b-1})$
X_k	$x_k^n(\hat{l}_{k0})$	$x_k^n(\hat{l}_{k1})$			
Y_k	$\hat{m}_{k1}, \hat{l}_{k1}$				

Distributed decode–forward = PDF for networks



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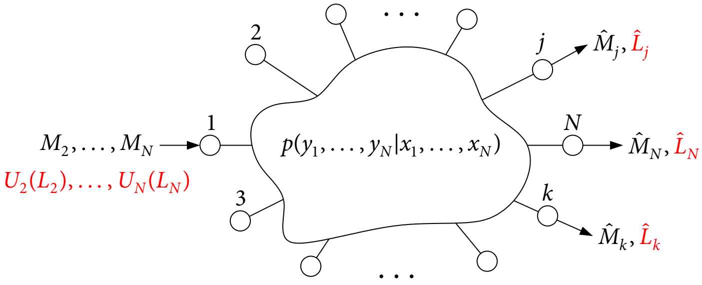
Block	1	2	3	$\cdots$	b
X_1	$\mathbf{l}_0$	$\leftarrow \mathbf{l}_1$	$\leftarrow \mathbf{l}_2$	$\cdots$	$\leftarrow \mathbf{l}_{b-1}$
	$x_1^n(\mathbf{m}_1|\mathbf{l}_1, \mathbf{l}_0)$	$x_1^n(\mathbf{m}_2|\mathbf{l}_2, \mathbf{l}_1)$	$x_1^n(\mathbf{m}_3|\mathbf{l}_3, \mathbf{l}_2)$	$\cdots$	$x_1^n(\mathbf{m}_b|\mathbf{l}_b, \mathbf{l}_{b-1})$
X_k	$x_k^n(\hat{l}_{k0})$	$x_k^n(\hat{l}_{k1})$	$x_k^n(\hat{l}_{k2})$	$\cdots$	$x_k^n(\hat{l}_{k,b-1})$
Y_k	$\hat{m}_{k1}, \hat{l}_{k1}$	$\hat{m}_{k2}, \hat{l}_{k2}$	$\hat{m}_{k3}, \hat{l}_{k3}$	$\cdots$	$\hat{m}_{kb}, \hat{l}_{kb}$

Distributed decode–forward = PDF for networks

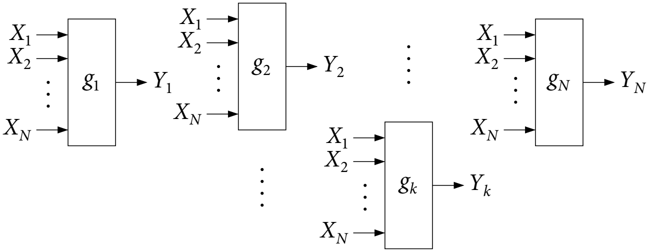


Lim–Kim–Kim (2014)

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); U(\mathcal{S}^c) | X(\mathcal{S}^c)) - \sum_{k \in \mathcal{S}^c} I(U_k; U(\mathcal{S}_k^c), X^N | X_k, Y_k), \quad \forall \mathcal{S}$$

for some $(\prod_{k=2}^N p(x_k)) p(x_1, u_2^N | x_2^N)$

Deterministic network (Kannan–Raja–Viswanath 2012)



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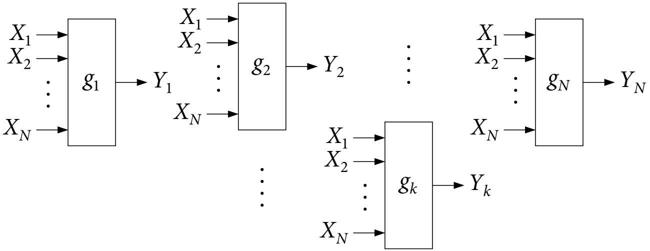

Cutset

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c))$$

Distributed decode–forward

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); U(\mathcal{S}^c) | X(\mathcal{S}^c)) - \sum_{k \in \mathcal{S}^c} I(U_k; U(\mathcal{S}_k^c), X^N | X_k, Y_k)$$

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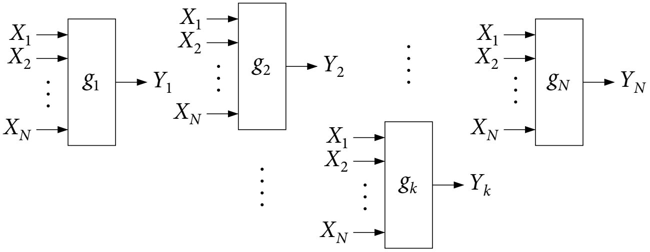

Cutset

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c))$$
$$= H(Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $p(x^N)$

Distributed decode–forward

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); U(\mathcal{S}^c) | X(\mathcal{S}^c)) - \sum_{k \in \mathcal{S}^c} I(U_k; U(\mathcal{S}_k^c), X^N | X_k, Y_k)$$

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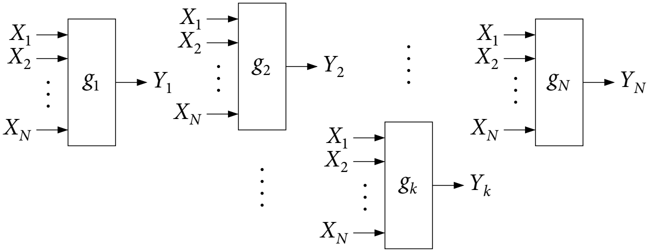

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$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c))$$
$$= H(Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

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Distributed decode–forward

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); U(\mathcal{S}^c) | X(\mathcal{S}^c))$$
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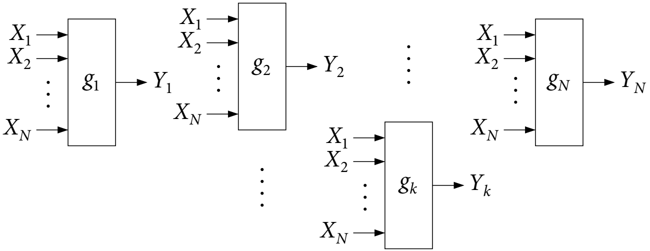
$$= H(Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $(\prod_{k=2}^N p(x_k)) p(x_1 | x_2^N)$

Deterministic network (Kannan–Raja–Viswanath 2012)



Cutset

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); Y(\mathcal{S}^c) | X(\mathcal{S}^c)) = H(Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $p(x^N)$

Distributed decode–forward

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); U(\mathcal{S}^c) | X(\mathcal{S}^c)) - \sum_{k \in \mathcal{S}^c} I(U_k; U(\mathcal{S}_k^c), X^N | X_k, Y_k) = H(Y(\mathcal{S}^c) | X(\mathcal{S}^c)), \quad \forall \mathcal{S}$$

for some $(\prod_{k=2}^N p(x_k)) p(x_1 | x_2^N)$

- Tight for graphical networks and deterministic networks with no interference

Gaussian network

- Channel model:

$$Y_k = \sum_j g_{kj} X_j + Z_k, \quad k \in [1:N]$$

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Gaussian network

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- Distributed decode–forward bound:

$$R(\mathcal{S}^c) \le I(X(\mathcal{S}); U(\mathcal{S}^c) | X(\mathcal{S}^c)) - \sum_{k \in \mathcal{S}^c} I(U_k; U(\mathcal{S}_k^c), X^N | X_k, Y_k)$$

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- Set $U_k = \sum_j g_{kj} X_j + \hat{Z}_k \sim Y_k$, where $\hat{Z}_k \sim \mathrm{N}(0,1)$

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Kannan–Raja–Viswanath (2012), Lim–Kim–Kim (2014)

If $(R_1, \ldots, R_N) \in \mathscr{R}_{\mathrm{CS}}$, then $(R_1 - 0.5N, \ldots, R_N - 0.5N) \in \mathscr{R}_{\mathrm{DDF}}$

Distributed decode–forward (DDF)

Broadcast

Noisy network coding (NNC)

Multiple access

Distributed decode–forward (DDF)

Broadcast

Simple decoder

Noisy network coding (NNC)

Multiple access

Simple encoder

Distributed decode–forward (DDF)	Noisy network coding (NNC)
Broadcast	Multiple access
Simple decoder	Simple encoder
Marton's inner bound	MAC capacity region

Distributed decode–forward (DDF)	Noisy network coding (NNC)
Broadcast	Multiple access
Simple decoder	Simple encoder
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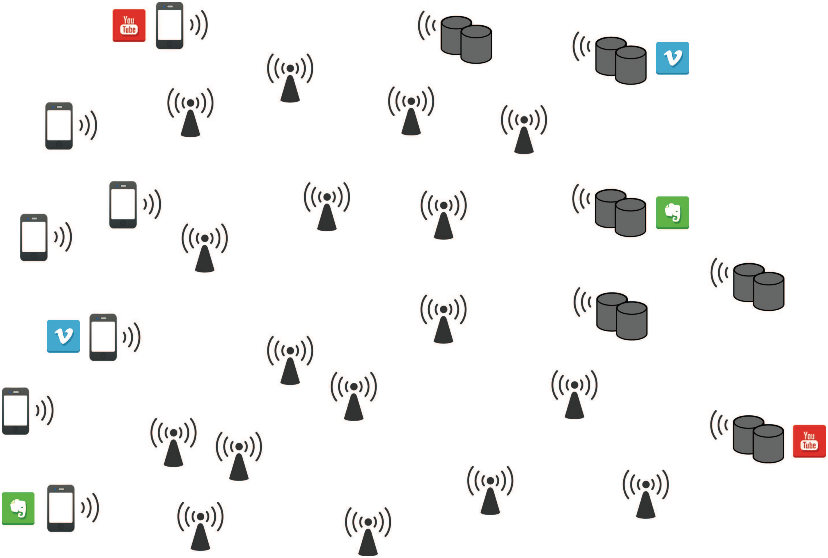
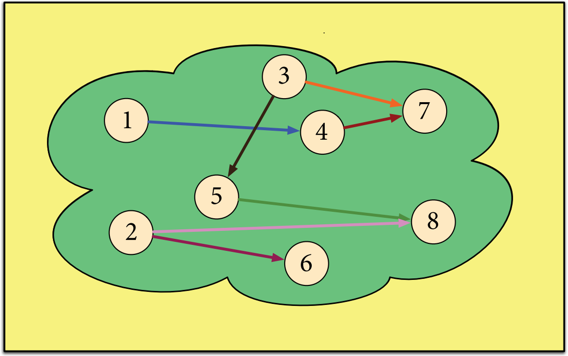
Partial decode–forward	Compress–forward

Distributed decode–forward (DDF)	Noisy network coding (NNC)
Broadcast	Multiple access
Simple decoder	Simple encoder
Marton's inner bound	MAC capacity region
Partial decode–forward	Compress–forward
$0.5N$	$0.63N$

Multiple unicast network



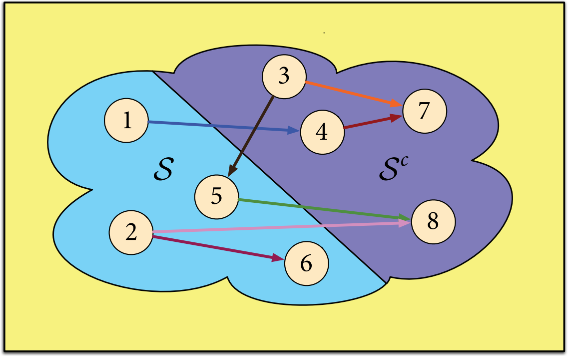
Multiple unicast network



- Capacity region $\mathscr{C}$: Closure of the set of achievable rate tuples

$$(R_{1\to4}, R_{2\to6}, R_{2\to8}, R_{3\to5}, R_{3\to7}, R_{4\to7}, R_{5\to8})$$

Cutset outer bound

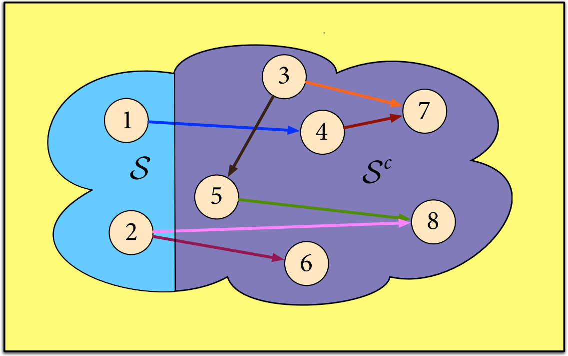


El Gamal (1981), Cover–Thomas (2006)

$$\sum_{j,k:j\in\mathcal{S},k\in\mathcal{S}^c} R_{j\to k} \le I(X(\mathcal{S});Y(\mathcal{S}^c)|X(\mathcal{S}^c)), \quad \forall \mathcal{S}\subseteq[1:8]$$

for some $p(x^N)$

Cutset outer bound



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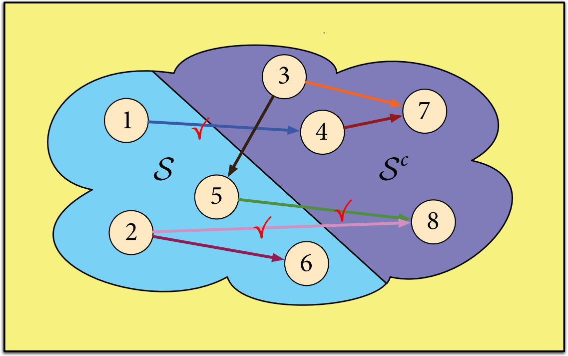
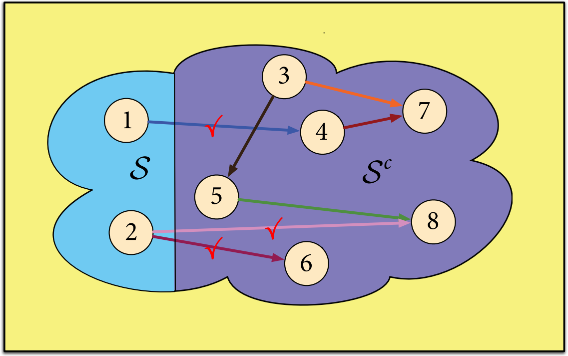


$$R_{1\to 4} + R_{2\to 8} + R_{5\to 8} \leq I(X_1, X_2, X_5, X_6; Y_3, Y_4, Y_7, Y_8 | X_3, X_4, X_7, X_8)$$

Cutset outer bound



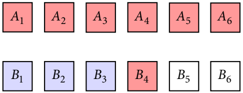
$$R_{1\to4} + R_{2\to8} + R_{5\to8} \le I(X_1, X_2, X_5, X_6; Y_3, Y_4, Y_7, Y_8 | X_3, X_4, X_7, X_8)$$

$$R_{1\to4} + R_{2\to6} + R_{2\to8} \le I(X_1, X_2; Y_3, Y_4, Y_5, Y_6, Y_7, Y_8 | X_3, X_4, X_5, X_6, X_7, X_8)$$

Directed information

- Mutual information

$$I(A_1, \ldots, A_N; B_1, \ldots, B_N) = \sum_{j=1}^{N} I(A^N; B_j | B^{j-1})$$



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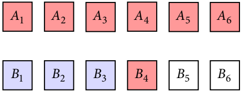
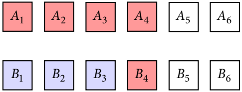


- Directed information
(Marko 1973, Massey 1990)

$$I(A_1, \ldots, A_N \to B_1, \ldots, B_N) = \sum_{j=1}^{N} I(A^j; B_j | B^{j-1})$$



Directed information

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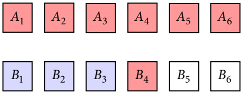

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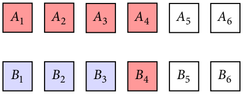
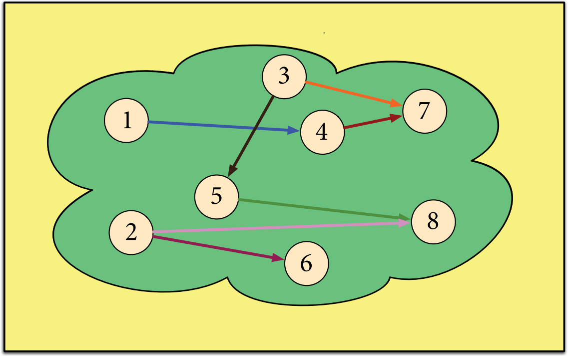
$$I(A_1,\ldots,A_N \to B_1,\ldots,B_N) = \sum_{j=1}^{N} I(A^j;B_j|B^{j-1})$$



- Amount of information A^N causally provides about B^N (Permuter et al. 2011)

Directed cutset outer bound

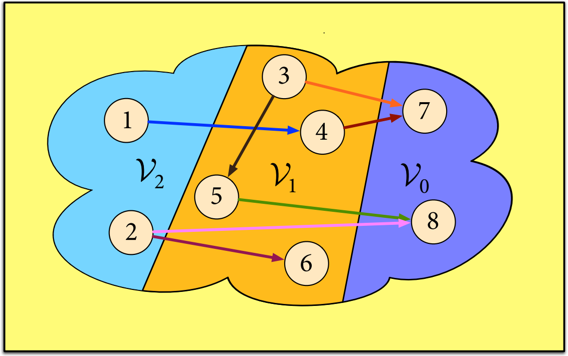


Kamath–Kim (2014)

$$\sum_{j,k: j \in \mathcal{V}_+, k \in \mathcal{V}_-} R_{j\to k} \le I(Y(\mathcal{V}_0), \ldots, Y(\mathcal{V}_{L-1}) \to X(\mathcal{V}_1), \ldots, X(\mathcal{V}_L) | X(\mathcal{V}_0)), \quad \forall \mathcal{V}_0, \ldots, \mathcal{V}_L$$

for some $p(x^N)$

Directed cutset outer bound



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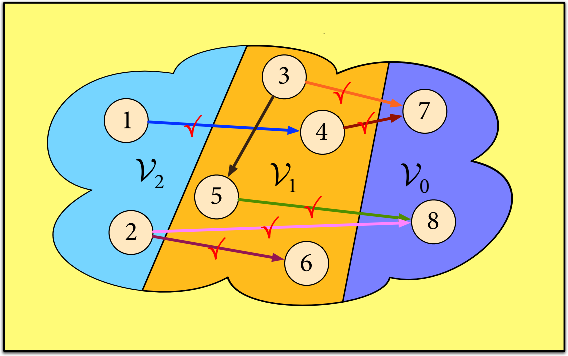
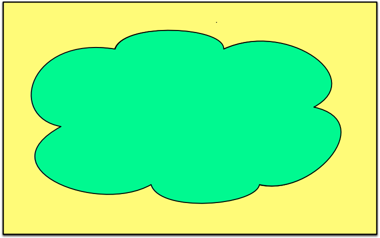
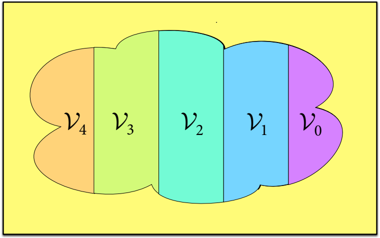


$$R_{1\to4} + R_{2\to6} + R_{2\to8} + R_{3\to7} + R_{4\to7} + R_{5\to8}$$
$$\le I\big((Y_7, Y_8), (Y_3, Y_4, Y_5, Y_6) \to (X_3, X_4, X_5, X_6), (X_1, X_2) \,|\, (X_7, X_8)\big)$$
$$= I(Y(7:8); X(3:6) \,|\, X(7:8)) + I(Y(3:6), Y(7:8); X(1:2) \,|\, X(3:6), X(7:8))$$

Finer partition (cutlet)

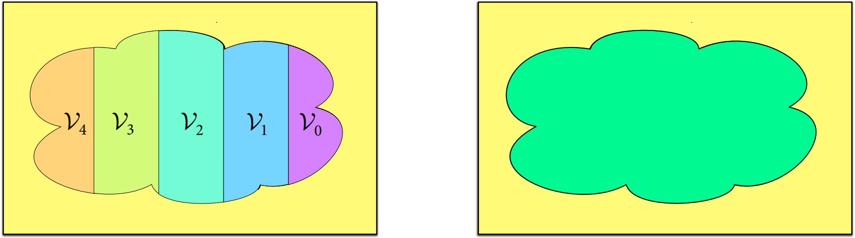


Finer partition (cutlet)



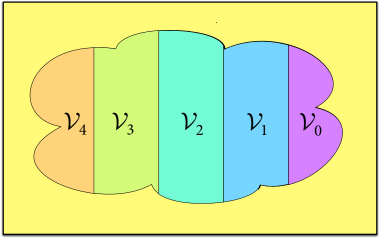
$$R_{\text{sum}} \le I\big(Y(\mathcal{V}_0), Y(\mathcal{V}_1), Y(\mathcal{V}_2), Y(\mathcal{V}_3) \\ \to X(\mathcal{V}_1), X(\mathcal{V}_2), X(\mathcal{V}_3), X(\mathcal{V}_4) \,|\, X(\mathcal{V}_0)\big)$$

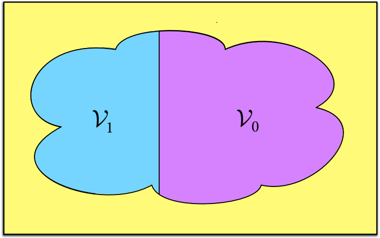
Finer partition (cutlet)



$$R_{\text{sum}} \le I\big(Y(\mathcal{V}_0), Y(\mathcal{V}_1), Y(\mathcal{V}_2), Y(\mathcal{V}_3) \\ \to X(\mathcal{V}_1), X(\mathcal{V}_2), X(\mathcal{V}_3), X(\mathcal{V}_4) \,|\, X(\mathcal{V}_0)\big)$$

Finer partition (cutlet)

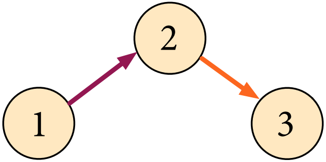




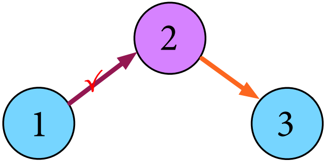
$$R_{\text{sum}} \le I\big(Y(\mathcal{V}_0), Y(\mathcal{V}_1), Y(\mathcal{V}_2), Y(\mathcal{V}_3) \to X(\mathcal{V}_1), X(\mathcal{V}_2), X(\mathcal{V}_3), X(\mathcal{V}_4) | X(\mathcal{V}_0)\big)$$

$$R_{\text{sum}} \le I(Y(\mathcal{V}_0) \to X(\mathcal{V}_1) | X(\mathcal{V}_0)) = I(X(\mathcal{V}_1); Y(\mathcal{V}_0) | X(\mathcal{V}_0))$$

Example: 3-node "not-relay" channel $p(y_2, y_3|x_1, x_2)$



Example: 3-node “not-relay” channel $p(y_2, y_3|x_1, x_2)$

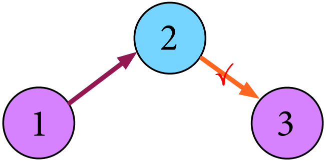


- Cutset bound

$$R_{1\to2} \le I(X_1; Y_2|X_2)$$

for some $p(x_1, x_2)$

Example: 3-node "not-relay" channel $p(y_2, y_3|x_1, x_2)$

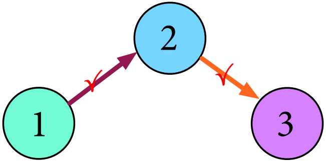


- Cutset bound

$$R_{1\to 2} \leq I(X_1; Y_2|X_2)$$
$$R_{2\to 3} \leq I(X_2; Y_3|X_1)$$

for some $p(x_1, x_2)$

Example: 3-node "not-relay" channel $p(y_2, y_3|x_1, x_2)$



- Cutset bound

$$R_{1\to 2} \le I(X_1; Y_2|X_2)$$
$$R_{2\to 3} \le I(X_2; Y_3|X_1)$$

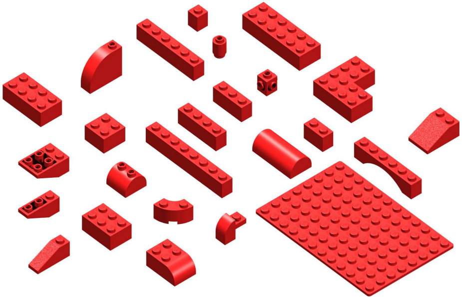
for some $p(x_1, x_2)$

- Cutlet bound

$$R_{1\to 2} \le I(X_1; Y_2|X_2)$$
$$R_{2\to 3} \le I(X_2; Y_3|X_1)$$
$$\begin{aligned} R_{1\to 2} + R_{2\to 3} &\le I(Y_3, Y_2 \to X_2, X_1) \\ &= I(X_2; Y_3) + I(X_1; Y_2, Y_3|X_2) \end{aligned}$$

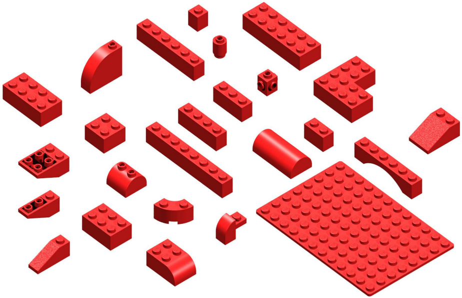
for some $p(x_1, x_2)$

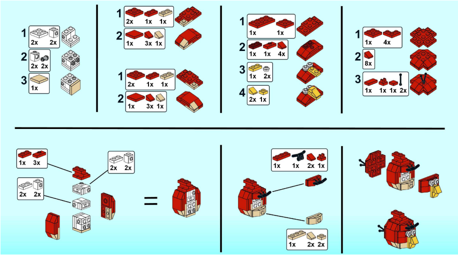
Coding for wireless relay networks



- Simultaneous decoding
- Superposition coding
- Multicoding
- Block Markov coding
- Decode–forward
- Compress–forward

Coding for wireless relay networks





- Simultaneous decoding
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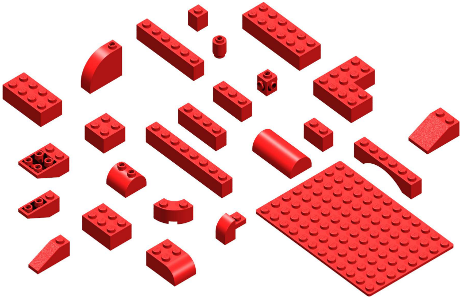
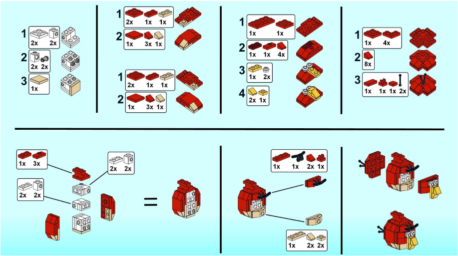
- Multiple access (NNC)
- Broadcast (DDF)

Coding for wireless relay networks





- Simultaneous decoding
- Superposition coding
- Multicoding
- Block Markov coding
- Decode–forward
- Compress–forward
- Practical codes??? (Thursday)

- Multiple access (NNC)
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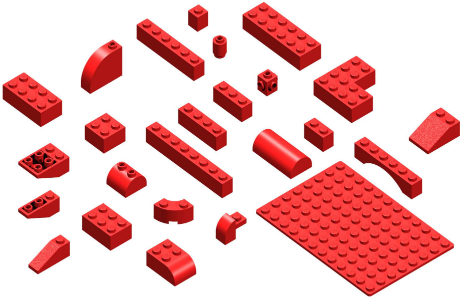
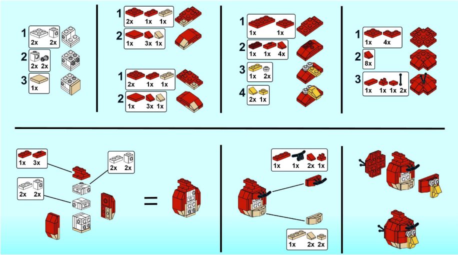
Coding for wireless relay networks





- Simultaneous decoding
- Superposition coding
- Multicoding
- Block Markov coding
- Decode–forward
- Compress–forward
- Practical codes??? (Thursday)

- Multiple access (NNC)
- Broadcast (DDF)
- Directed cutset bound
- Coding for multiple unicast??? (Tomorrow)

To learn more

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  In *Proc. IEEE Int. Symp. Inf. Theory*, Honolulu, HI

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