

The Infinite Message Limit of Interactive Source Coding

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Joint work with Nan Ma

Electrical and Computer Engineering

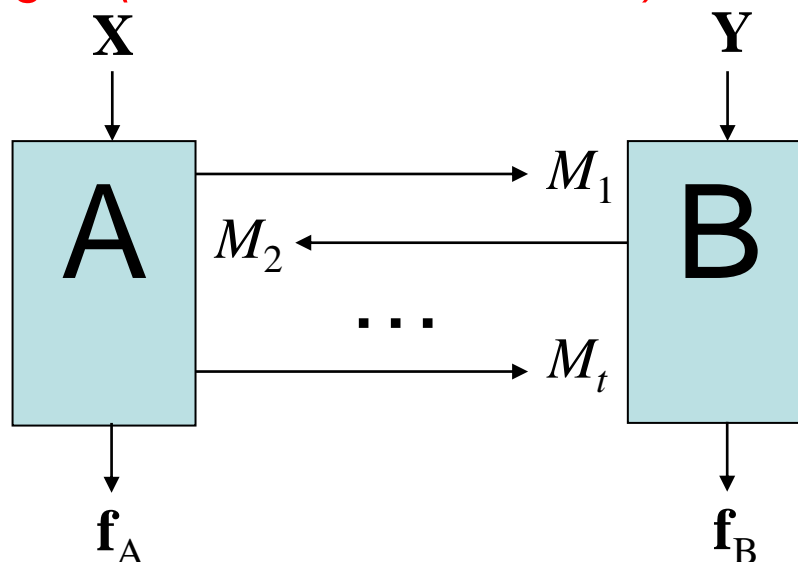


Research supported by



Motivation

- Degrees of freedom “resources” in distributed source coding and computation:
 - Blocklength (delay)
 - Rate (speed)
 - Quantizer step-size (quality)
 - Network size
 - Alphabet size ...
 - *Number of messages (Multiround interaction)*

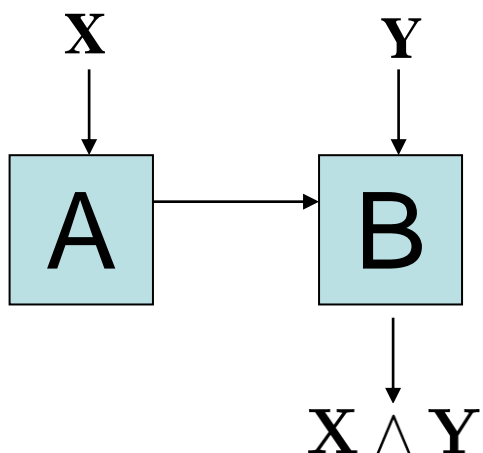


- Main goal: Characterize the *ultimate* benefit of interaction

Example: compute AND at terminal B

- Sources: $X \perp\!\!\!\perp Y, X \sim \text{Ber}(p), Y \sim \text{Ber}(q)$, Compute: $X \wedge Y$ at B

- One message: min rate = ?



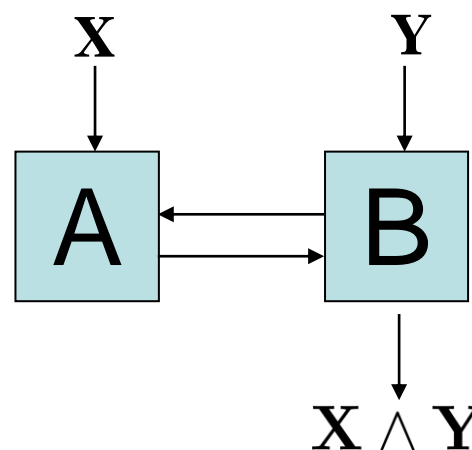
- 1-msg rate = $h(p)$

[Yamamoto, IT'82]

- 3-msgs.? 4-msgs.? ...

- What about ∞ -msgs.?

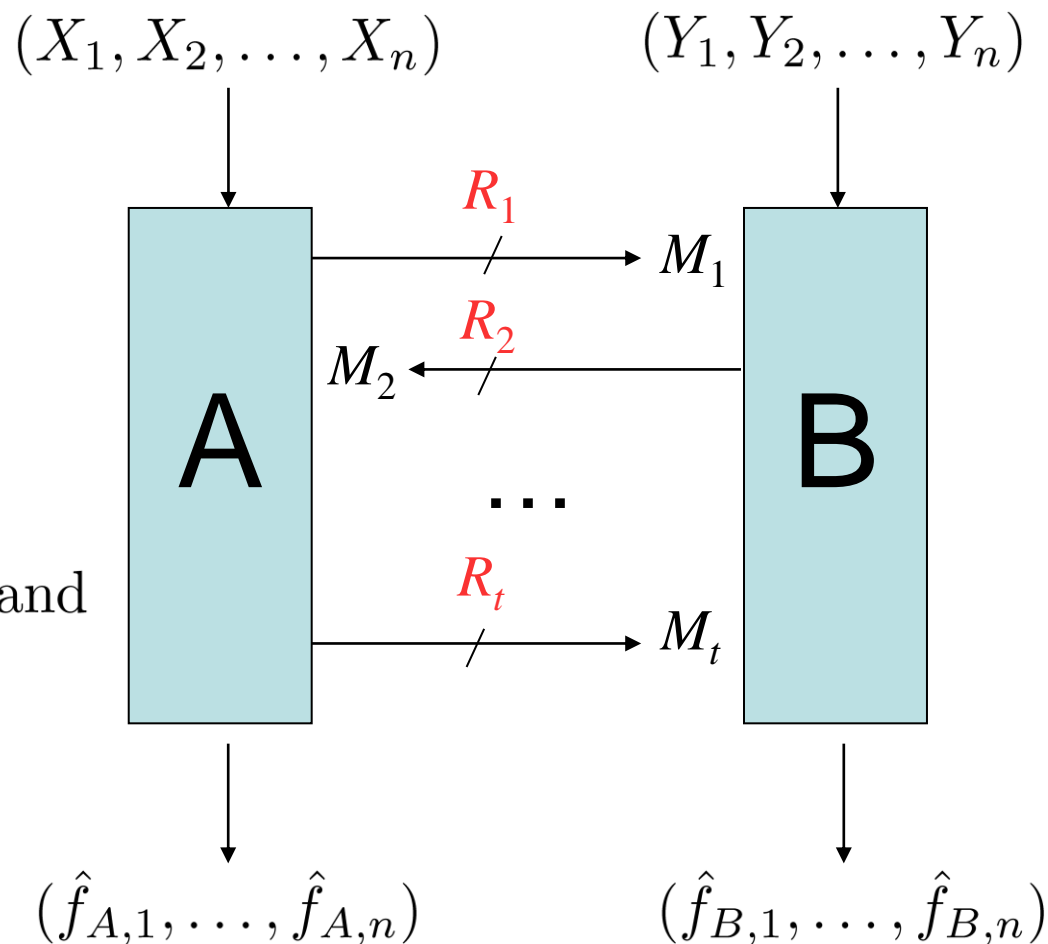
- Two messages min rate = ?



- 2-msg sum-rate $\leq h(q) + q$

General 2-terminal interactive function computation

- n samples $(X_i, Y_i) \sim \text{iid } p_{XY}$
- Samplewise function computation at A and B
- t alternating messages
- (R_1, \dots, R_t) is admissible if there exists a sequence of codes: as $n \rightarrow \infty$, $(\# \text{ bits msg } j)/n \rightarrow R_j$ and $\Pr[\text{comp. error}] \rightarrow 0$ (lossless, can extend to lossy)



- Minimum sum-rate:

$$R_{sum,t}^A = \min \sum_{i=1}^t R_t$$

need $(f_A(X_1, Y_1), \dots, f_A(X_n, Y_n)), (f_B(X_1, Y_1), \dots, f_B(X_n, Y_n))$

Characterize and compute

$$R_{sum,\infty} := \lim_{t \rightarrow \infty} R_{sum,t}^A$$

- Understand ultimate benefit of cooperative interaction
- “Unexplored” dimension for asymptotic analysis:
 - (possibly) **infinite messages with infinitesimal rate**
- “Untapped” resource: Multi-round interaction

Related work

- Communication complexity (Yao, Ahlswede, Cai, Orlitsky, Kushilevitz, Nisan, ..., Braverman, A. Rao, ...)
 - Usual focus on worst-case and comp. error = 0
 - Usually about **bits** not rate
- Two-way source coding [Kaspi IT'86]
 - Source **reproduction**
- Coding for computing [Orlitsky & Roche IT'00]
 - **Two messages**

$R_{sum,t}$ for finite t : Solved

Single-letter characterization [Ma, Ishwar: ISIT'08, IT'11]

$$R_{sum,t}^A = \min_{U^t} [I(X; U^t | Y) + I(Y; U^t | X)]$$

constraints

Markov chains

$$U_i - (X, U^{i-1}) - Y, \quad i \text{ odd}$$

$$U_i - (Y, U^{i-1}) - X, \quad i \text{ even}$$

Entropy constraints

$$H(f_A(X, Y) | X, U^t) = 0$$

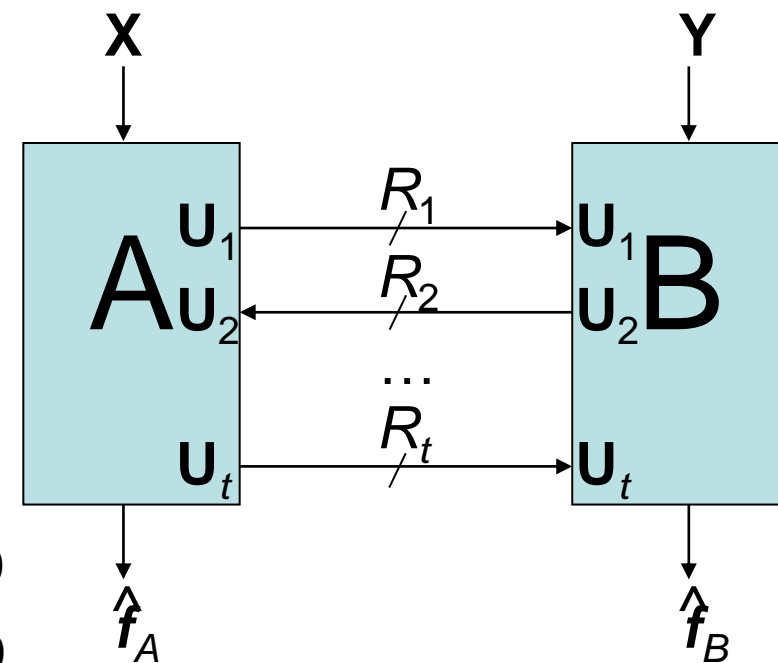
$$H(f_B(X, Y) | Y, U^t) = 0$$

Cardinality bounds

$$|U_j| \leq \begin{cases} |\mathcal{X}| \left(\prod_{i=1}^{j-1} |U_i| \right) + t - j + 3, & j \text{ odd,} \\ |\mathcal{Y}| \left(\prod_{i=1}^{j-1} |U_i| \right) + t - j + 3, & j \text{ even.} \end{cases}$$

Achievability: (sequence of Wyner-Ziv codes)

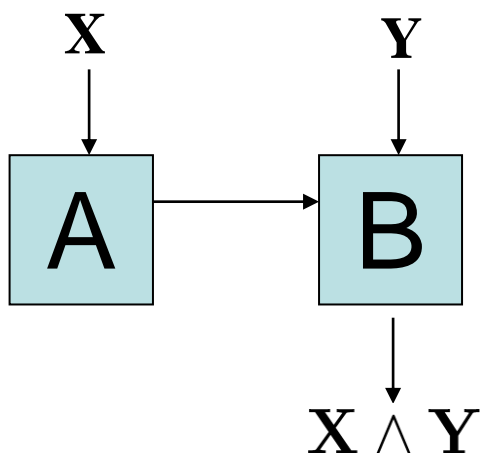
- 1st msg: Quantizes \mathbf{X} to \mathbf{U}_1 with side info \mathbf{Y}
 $R_1 = I(X; U_1 | Y), \quad U_1 - X - Y$
- 2nd msg: Quantizes $(\mathbf{Y}, \mathbf{U}_1)$ to \mathbf{U}_2 with side info $(\mathbf{X}, \mathbf{U}_1)$
 $R_2 = I(Y; U_2 | X, U_1), \quad U_2 - (Y, U_1) - X$
-
- Recover f_A based on $(\mathbf{X}, \mathbf{U}_1, \dots, \mathbf{U}_t)$: $H(f_A | X, U_1, \dots, U_t) = 0$
- Recover f_B based on $(\mathbf{Y}, \mathbf{U}_1, \dots, \mathbf{U}_t)$: $H(f_B | Y, U_1, \dots, U_t) = 0$



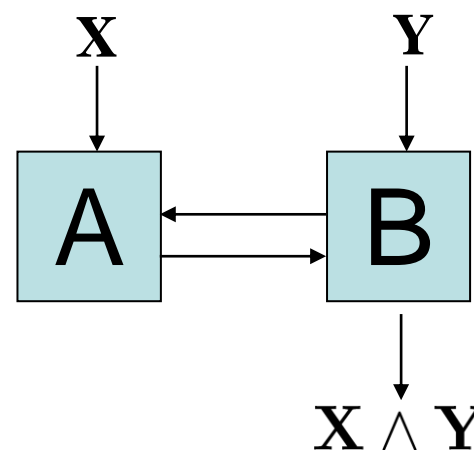
Example: compute AND at terminal B

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- One message: min rate = ?



- Two messages: min rate = ?

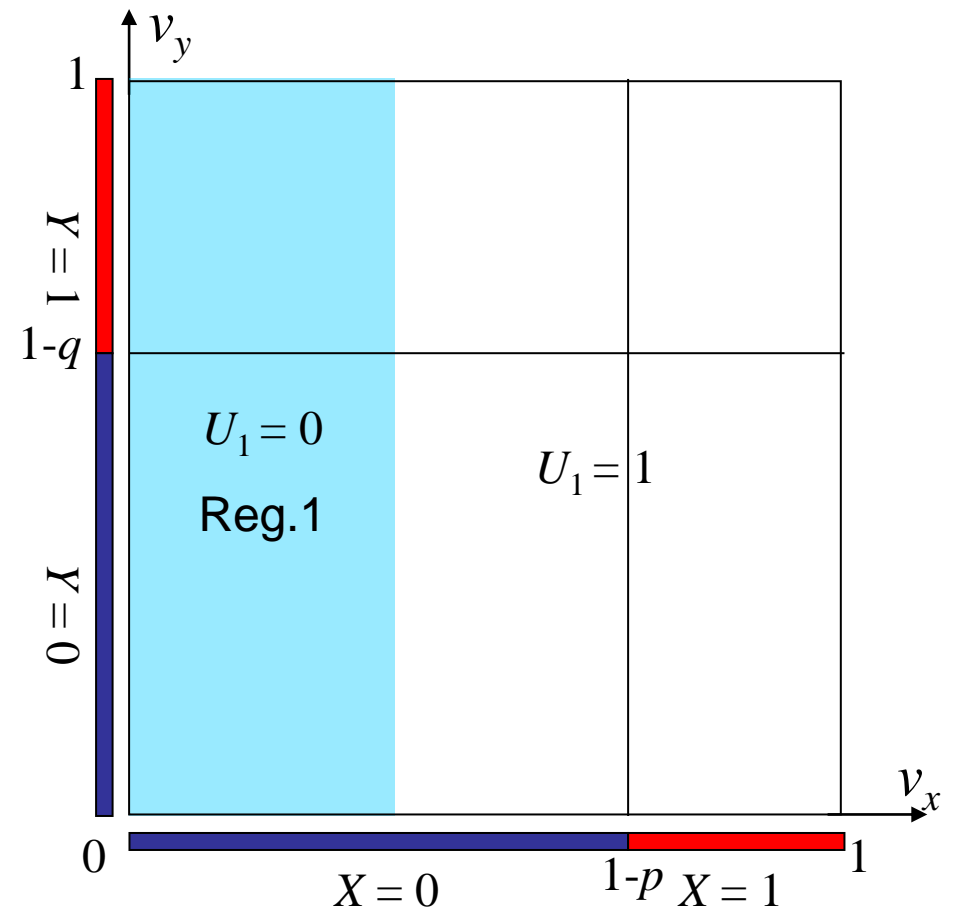


- 1-msg rate = $h(p)$
- 3-msgs.? 4-msgs.? ...
- What about ∞ -msgs.?

$$2\text{-msg sum-rate} \leq h(q) + q$$

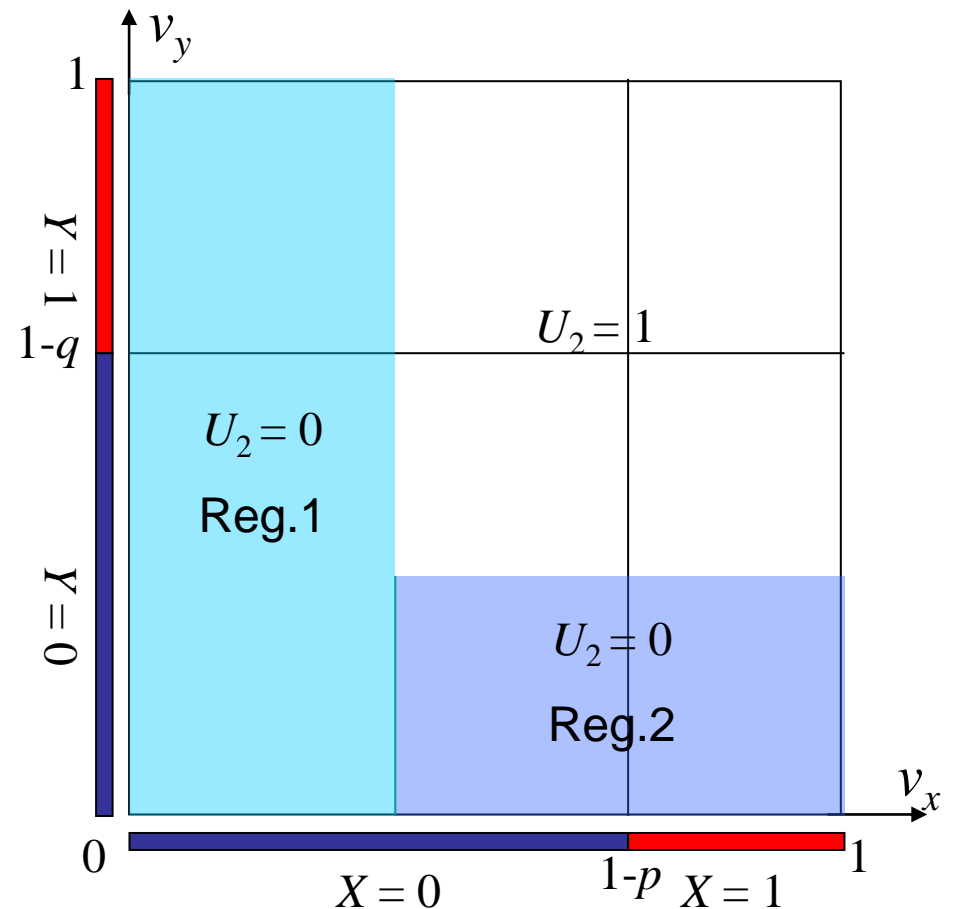
∞ -msg interaction

- Auxiliary random variables
 - Hidden variables: $(V_x, V_y) \sim \text{Uniform}(\text{unit-sq.})$
 - X, Y as functions of (V_x, V_y) .
 - $U_1 = 0$ in Reg.1; = 1 otherwise



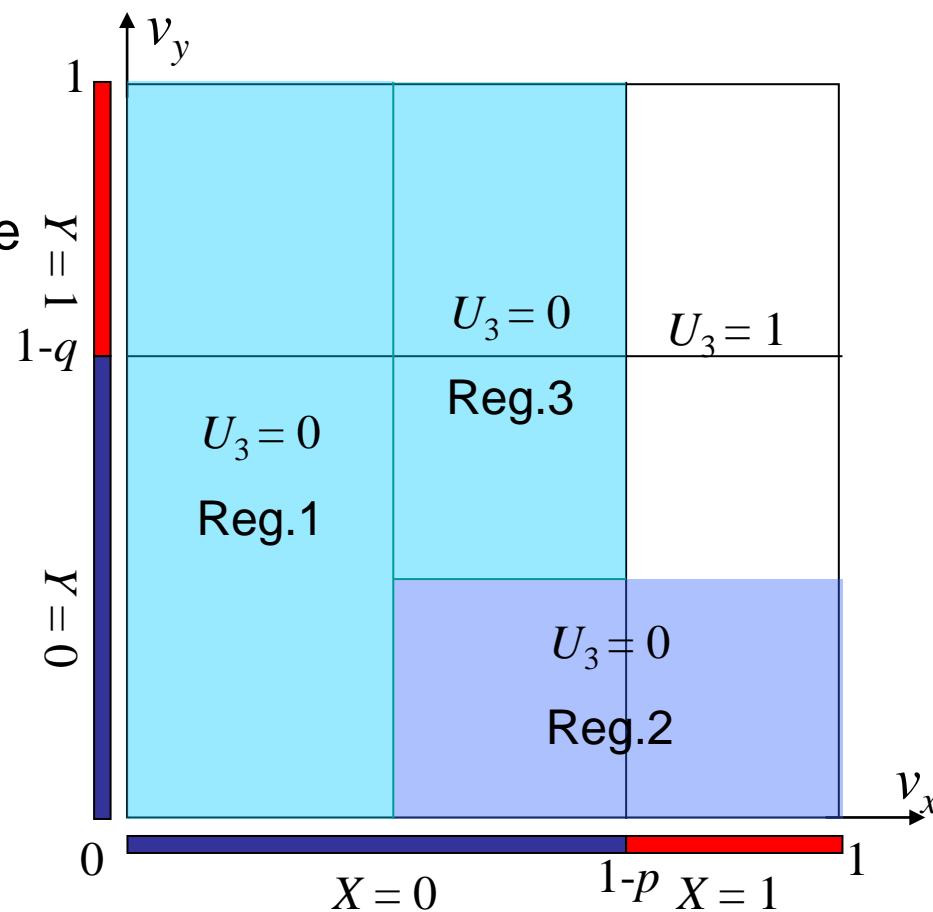
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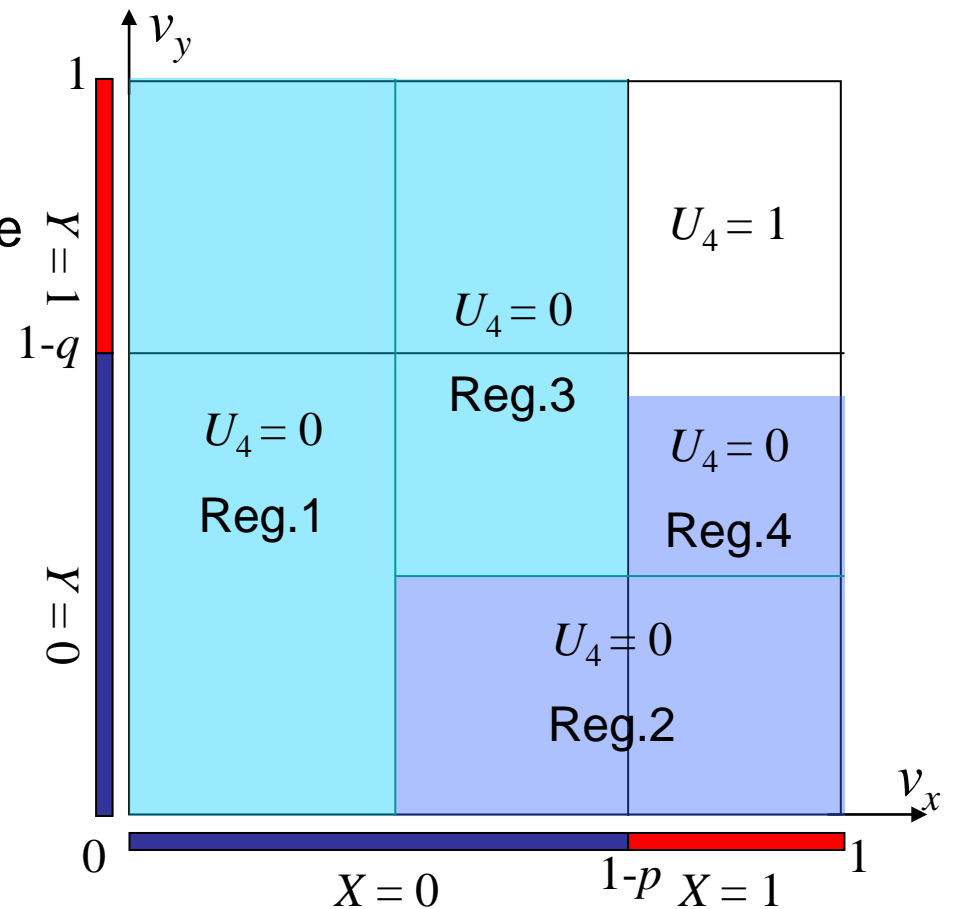
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∞ -msg interaction

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 - Hidden variables: $(V_x, V_y) \sim \text{Uniform}(\text{unit-sq.})$
 - X, Y as functions of (V_x, V_y) .
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 - $U_2 = 0$ in Reg.1 U 2; = 1 otherwise
 - $U_3 = 0$ in Reg.1 U 2 U 3; = 1 otherwise
 - $U_4 = 0$ in Reg.1 U 2 U 3 U 4;
= 1 otherwise ($X \wedge Y = U_4 \wedge Y$)



∞ -msg interaction

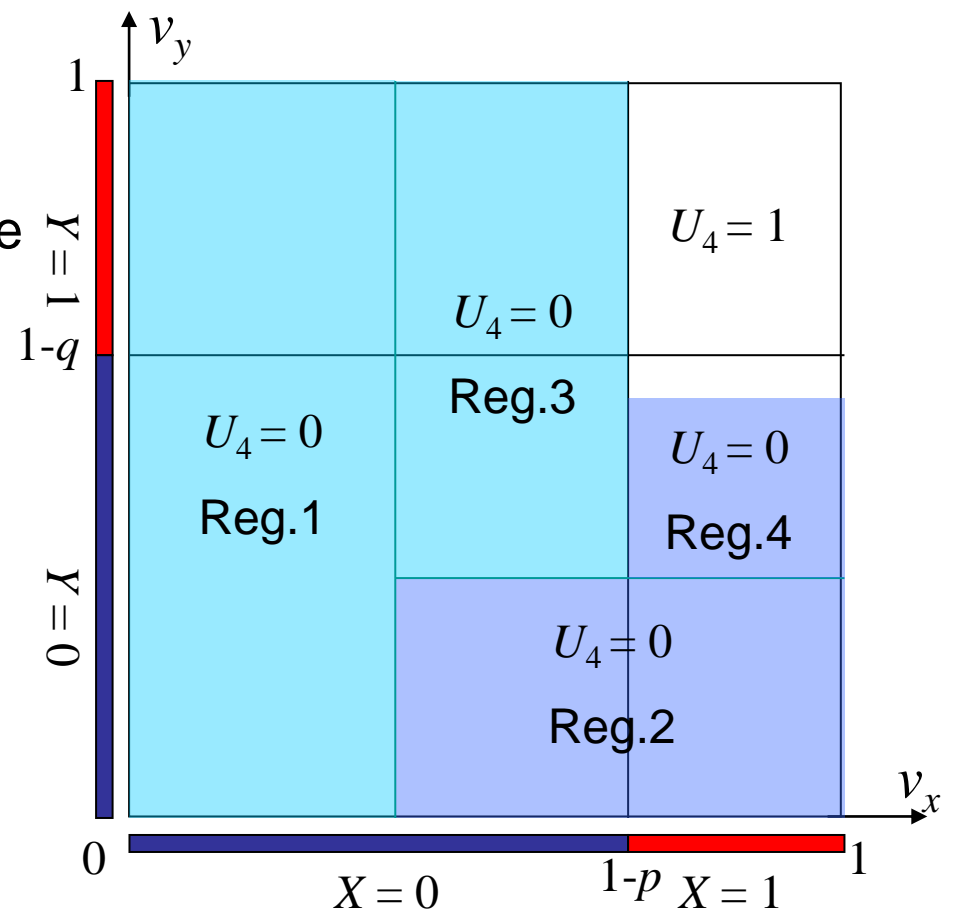
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 - X, Y as functions of (V_x, V_y) .
 - $U_1 = 0$ in Reg.1; = 1 otherwise
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 - $U_3 = 0$ in Reg.1 U 2 U 3; = 1 otherwise
 - $U_4 = 0$ in Reg.1 U 2 U 3 U 4; = 1 otherwise ($X \wedge Y = U_4 \wedge Y$)
 - Satisfy constraints

$$U_i - (X, U^{i-1}) - Y, \quad i \text{ odd}$$

$$U_i - (Y, U^{i-1}) - X, \quad i \text{ even}$$

$$H(f_A(X, Y) | X, U^t) = 0$$

$$H(f_B(X, Y) | Y, U^t) = 0$$



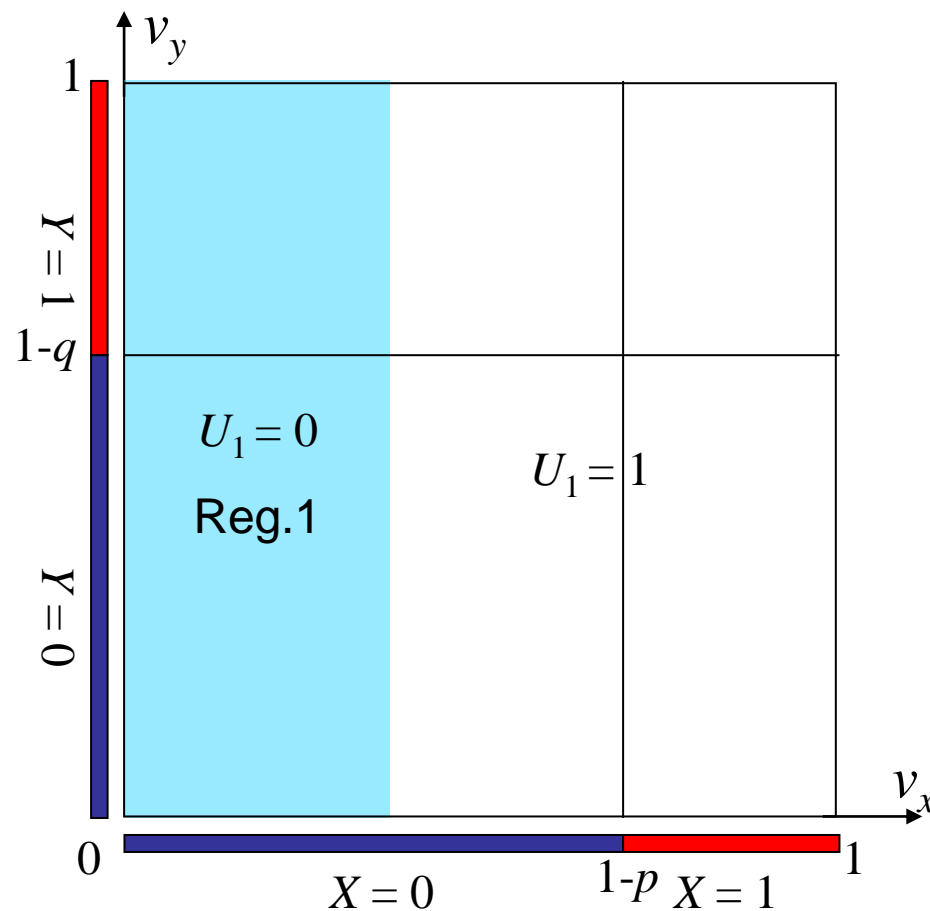
∞ -msg interaction

- Admissible rates

- 1st msg:

$$I(X; U_1 | Y) = \int_{\text{Reg.1}} w_x(v_x, p) dv_x dv_y$$

$$w_x(v_x, p) = \log \frac{1-x}{1-p-x}$$



∞ -msg interaction

- Admissible rates

- 1st msg:

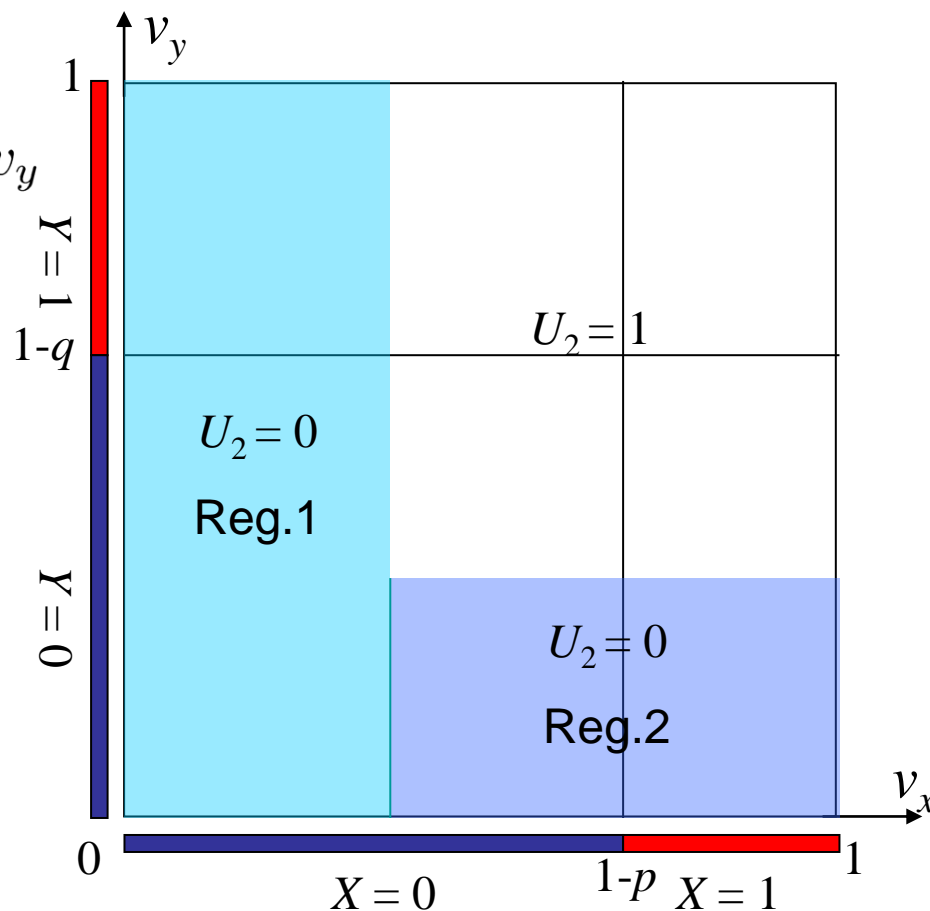
$$I(X; U_1 | Y) = \int_{\text{Reg.1}} w_x(v_x, p) dv_x dv_y$$

- 2nd msg:

$$I(Y; U_2 | XU_1) = \int_{\text{Reg.2}} w_y(v_y, q) dv_x dv_y$$

$$w_x(v_x, p) = \log \frac{1-x}{1-p-x}$$

$$w_y(v_y, q) = \log \frac{1-y}{1-q-y}$$



∞ -msg interaction

- Admissible rates

- 1st msg:

$$I(X; U_1 | Y) = \int_{\text{Reg.1}} w_x(v_x, p) dv_x dv_y$$

- 2nd msg:

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.....

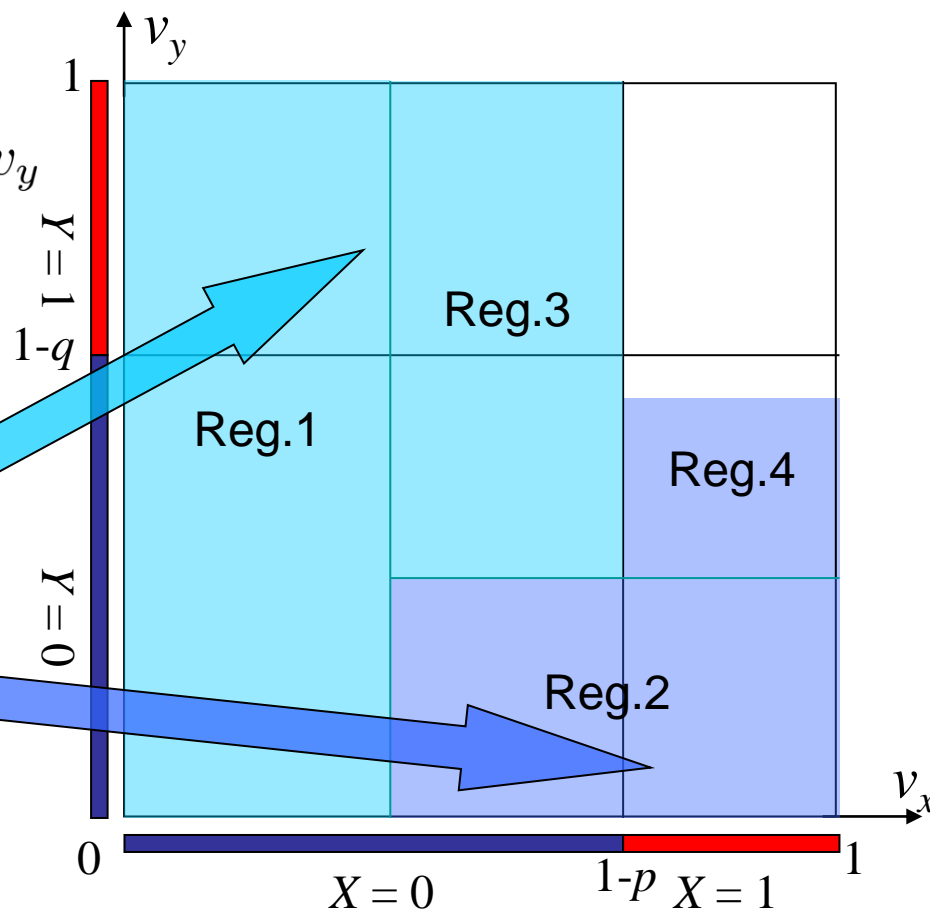
- Sum-rate:

$$\int_{\text{vertical bars}} w_x(v_x, p) dv_x dv_y$$

$$+ \int_{\text{horizontal bars}} w_y(v_y, q) dv_x dv_y$$

$$w_x(v_x, p) = \log \frac{1-x}{1-p-x}$$

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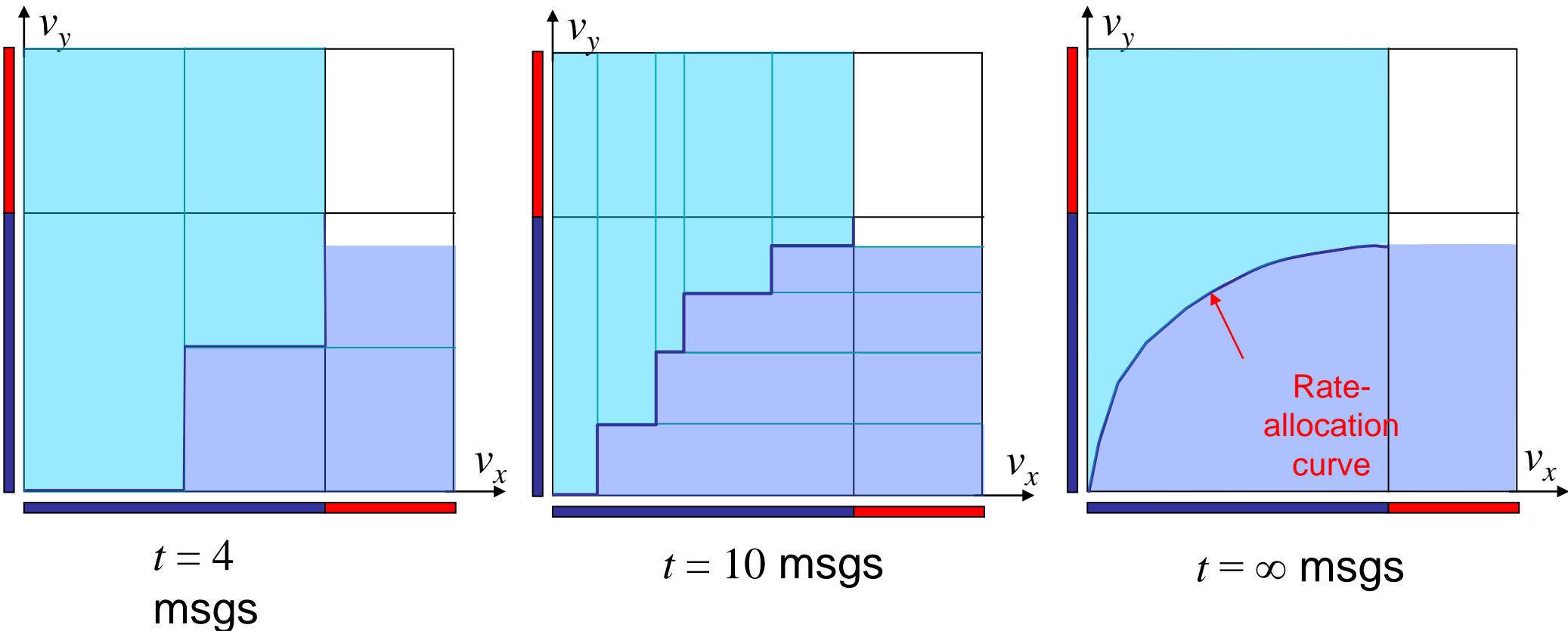
∞ -msg interaction

- Admissible sum-rate:

$$\int_{\text{vertical bars}} w_x(v_x, p) dv_x dv_y + \int_{\text{horizontal bars}} w_y(v_y, q) dv_x dv_y$$

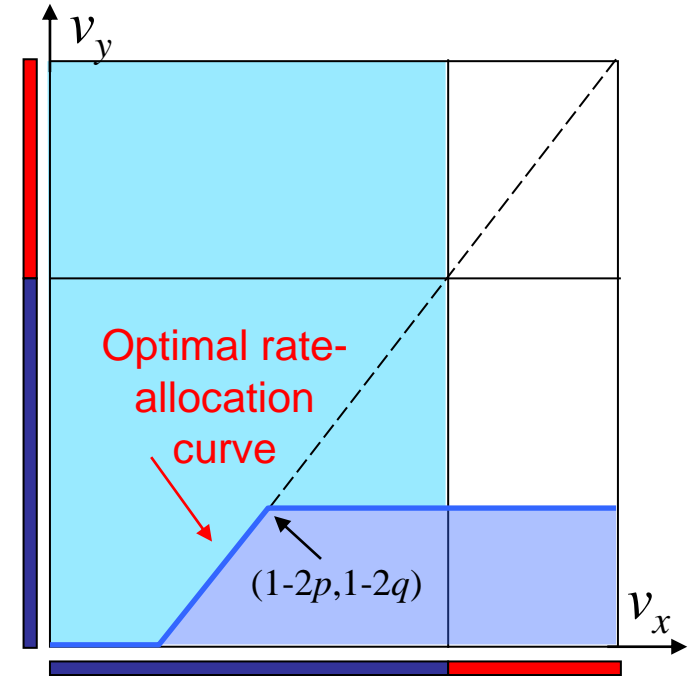
- $t \rightarrow \infty$: $\int_{\text{upper region}} w_x(v_x, p) dv_x dv_y + \int_{\text{lower region}} w_y(v_y, q) dv_x dv_y$

- Rate-allocation curve: design parameter



∞ -msg interaction

- Optimize the rate-allocation curve:
- Cannot be achieved by finite msgs using this family of aux.r.v.'s (for general p, q)



- Admissible sum-rate in closed form:

$$R^*(p, q) = \begin{cases} h_2(p) + ph_2(q) + p \log_2 q + p(1 - 2q) \log_2 e, & \text{if } 0 \leq p \leq q \leq 1/2, \\ R^*(q, p), & \text{if } 0 \leq q \leq p \leq 1/2, \\ R^*(1 - p, q), & \text{if } 0 \leq q \leq 1/2 \leq p \leq 1, \\ h_2(p), & \text{if } 1/2 \leq q \leq 1. \end{cases}$$

$R_{sum,t}$ for finite t : Solved

Single-letter characterization [Ma, Ishwar: ISIT'08, IT'11]

$$R_{sum,t}^A = \min_{U^t} [I(X; U^t | Y) + I(Y; U^t | X)]$$

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Cardinality bounds

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- Achievability: sequence of Wyner-Ziv-like codes
- Finite dimensional optimization problem

How to compute $R_{sum,\infty}$?

- Idea 1:
 - Pick a large t , compute $R_{sum,t}^A$, pray this is a good approximation
 - ☺ Finite-dimensional optimization problem
 - ☹ **How large t ?**
 - ☹ **Dimension explodes** exponentially with t
- Idea 2:
 - Compute $R_{sum,t}^A$ for $t = 1, 2, \dots$ till change is “negligible”
 - ☺ Finite-dimensional optimization problem (for each t)
 - ☹ **Multiple problems, solve from scratch**
 - ☹ **Dimension explodes** exponentially with t
- All this effort **only for one p_{XY}**
- Any hope?

A new approach

- View $R_{sum,\infty}(p_{XY})$ as a **functional** of p_{XY}
- New convex-geometric “limit-free” characterization of $R_{sum,\infty}$
 - For entire **functional** $R_{sum,\infty}(\cdot)$ (not for only one fixed p_{XY})
 - Provides **optimality test** for admissible sum-rate functionals
 - Family of **lower bounds** for $R_{sum,\infty}(\cdot)$
- Alternating “concavification” algorithm for $R_{sum,\infty}$
 - Each iteration uses “same amount” of computation
 - **Reuses** results from previous steps
 - Works with the entire $R_{sum,\infty}(p_{XY})$ “surface”

A new approach

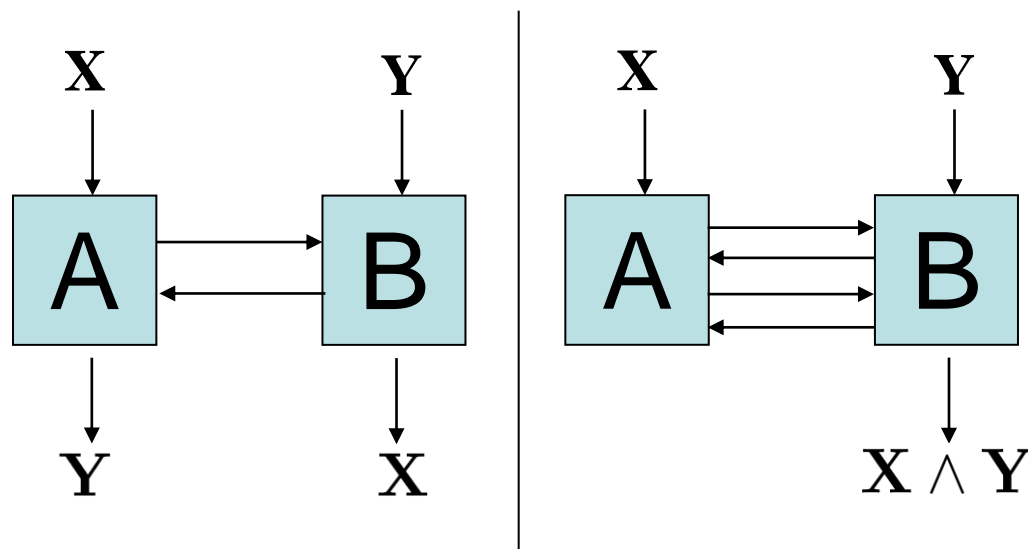
Rest of this talk:

- Illustrate new approach through **one simple example** for **2-terminal lossless** function computation
- Extension to general **lossy** two-terminal problem (brief)
- Extension to multi-terminal problems (very brief)

Example: compute AND at terminal B

- Sources: $X \perp\!\!\!\perp Y, X \sim \text{Ber}(p), Y \sim \text{Ber}(q)$
- Only B computes: $f_A(X, Y) = 0, f_B(X, Y) = X \wedge Y$ (AND)
- Goal: Characterize $R_{sum, \infty}(p, q)$ as a function of (p, q)
- Rate reduction functional:

$$\rho_t^A := H(X|Y) + H(Y|X) - R_{sum,t}^A = h(p) + h(q) - R_{sum,t}^A$$

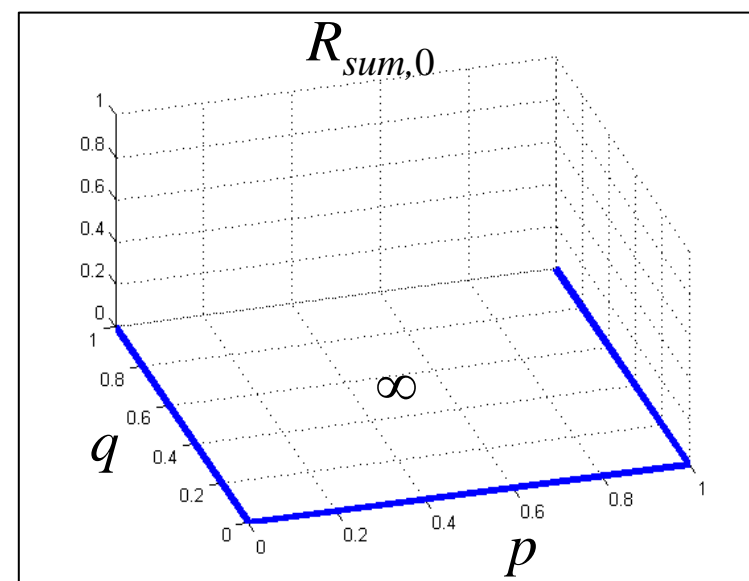
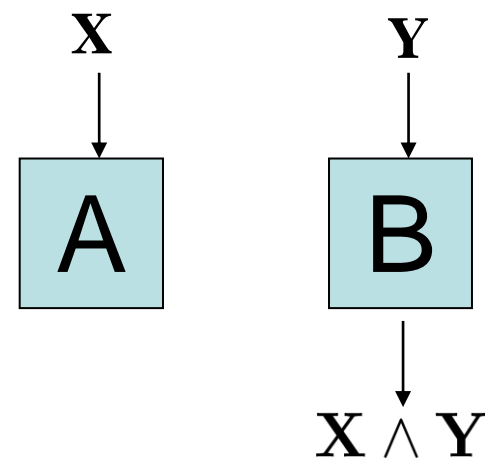


- Find $R_{sum, \infty} \iff$ find $\rho_{\infty} := H(X|Y) + H(Y|X) - R_{sum, \infty}$

Example: compute AND at terminal B

- Consider $t = 0$:
 - Feasible for special **boundary** distributions
 - If infeasible, define rate $:= \infty$

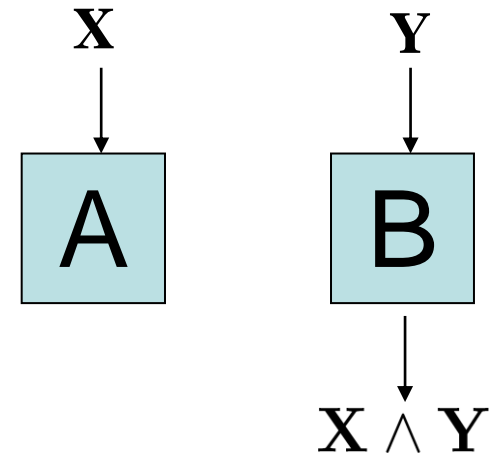
$$R_{sum,0}(p, q) = \begin{cases} 0, & \text{if } p = 0 \text{ or } q = 0 \quad (X \wedge Y = 0) \\ & \text{or } p = 1 \quad (X \wedge Y = Y) \\ \infty, & \text{otherwise } (X \wedge Y \text{ not determined}) \end{cases}$$



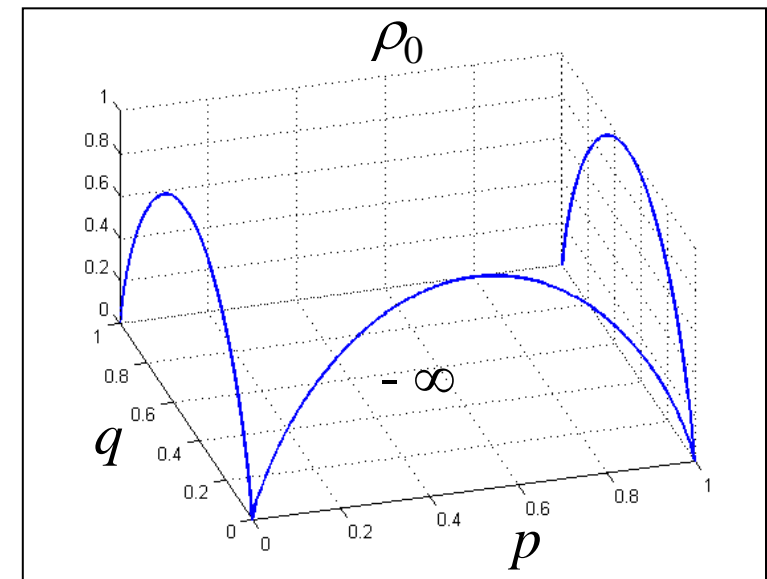
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$$\rho_0(p, q) = \begin{cases} h(p) + h(q), & \text{if } p = 0 \text{ or } q = 0 \\ & \text{or } p = 1 \\ -\infty, & \text{otherwise} \end{cases}$$

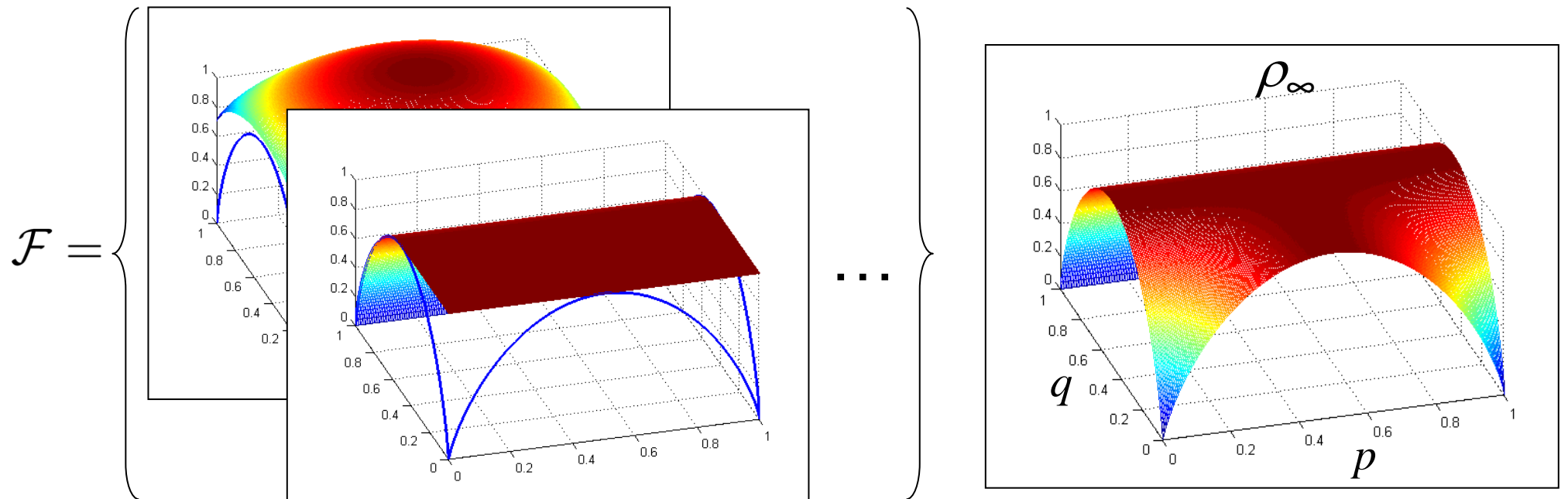


New characterization of ρ_∞

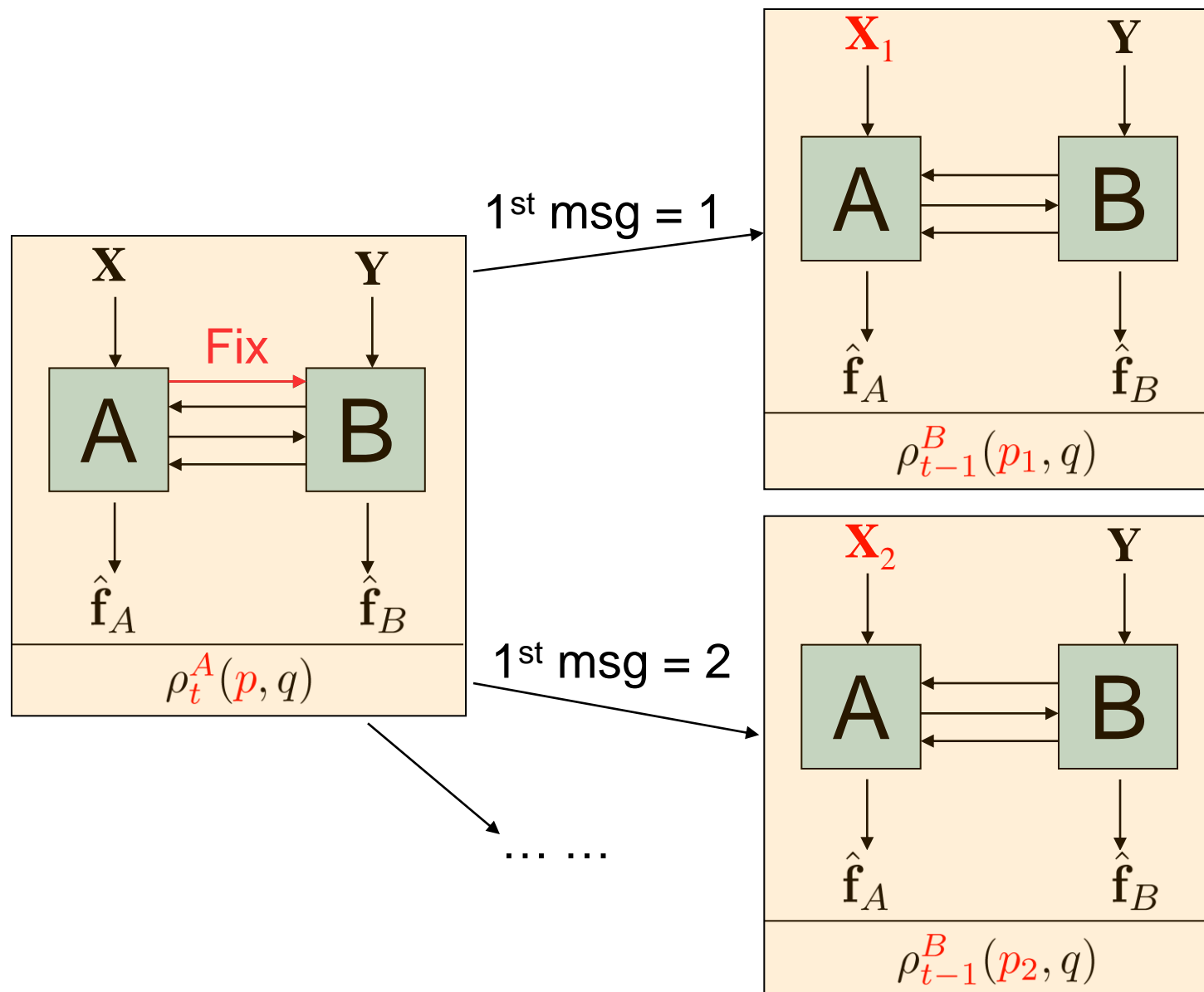
“Limit-free” characterization of ρ_∞ [Ma, Ishwar: Allerton 09]

ρ_∞ is the least element of \mathcal{F} , where

$$\mathcal{F} := \left\{ \rho(p, q) \left| \begin{array}{l} 1. \text{ For all } (p, q), \rho(p, q) \geq \rho_0(p, q) \\ 2. \text{ For all } q, \rho(p, q) \text{ is concave w.r.t. } p \\ 3. \text{ For all } p, \rho(p, q) \text{ is concave w.r.t. } q \end{array} \right. \right\}$$



Key insight: the subproblem viewpoint

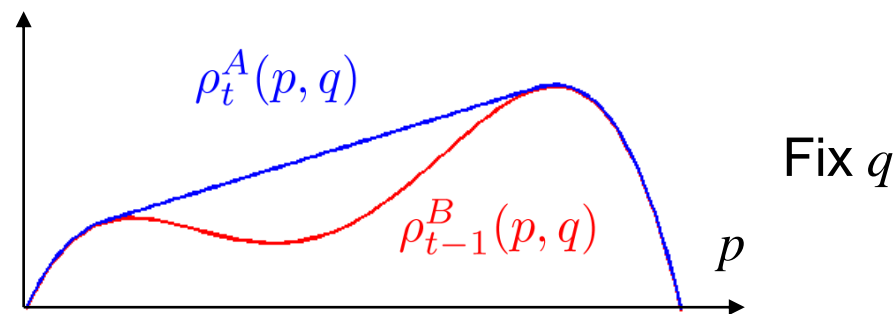


Connection between ρ_t^A and ρ_{t-1}^B ?

Subproblem viewpoint (continued)

$$\begin{aligned} R_{sum,t}^A &= \min_{U^t} [I(X; U^t | Y) + I(Y; U^t | X)] \\ \Rightarrow \rho_t^A &= \max_{U^t} [H(X|Y, U_2^t, U_1) + H(Y|X, U_2^t, U_1)] \\ &= \max_{U_1} \left[\sum_{u_1} p_{U_1}(u_1) \rho_{t-1}^B(p_{u_1}, q) \right] \end{aligned}$$

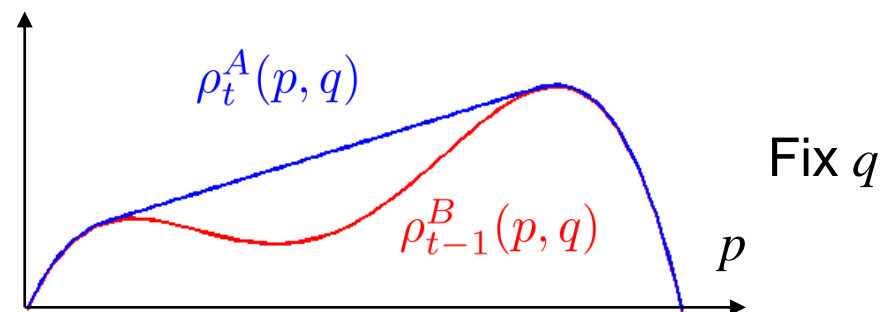
- $\rho_t^A(p, q) = \max[\text{convex combination of } \rho_{t-1}^B(p_1, q), \rho_{t-1}^B(p_2, q), \dots]$



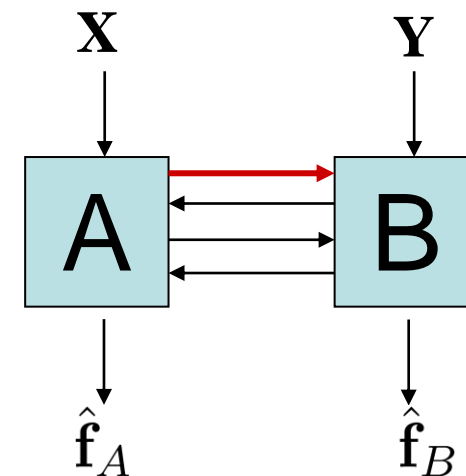
“Concavify”

- $\rho_t^A =$ the smallest concave function above ρ_{t-1}^B
- $\text{hypo}(\rho_t^A) =$ convex hull of $\text{hypo}(\rho_{t-1}^B)$

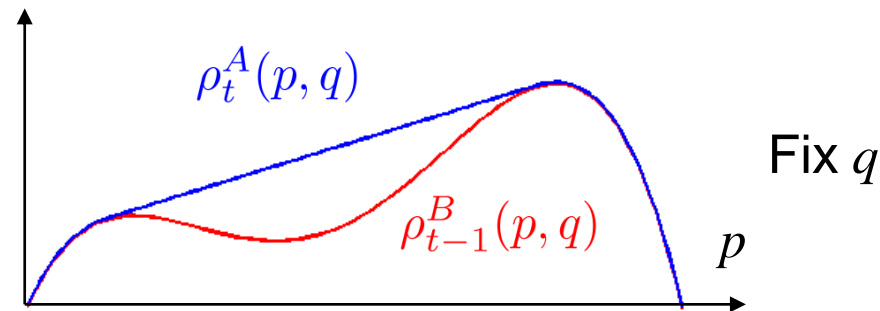
Concavity properties



- ρ_t^A concave in p , ρ_t^B concave in q
- ρ_{t-1}^B not concave in $p \iff \rho_t^A > \rho_{t-1}^B$
 \iff beneficial to add a message $A \rightarrow B$
- ρ_∞ : not beneficial to add any message
 \iff concave in p and q , respectively



Optimality Test



- $\rho_t^B = \rho_\infty \Leftrightarrow$
- ρ_t^B is concave in p (for all q) \Leftrightarrow
- $\rho_t^B = \rho_{t+1}^A$ (for all p and q)
- Similar results by interchanging (A, p) and (B, q)
- **Optimality Test:** If ρ is an achievable rate-reduction function and:
 1. For all (p, q) , $\rho(p, q) \geq \rho_0(p, q)$
 2. For all q , $\rho(p, q)$ is concave w.r.t. p
 3. For all p , $\rho(p, q)$ is concave w.r.t. qthen $\rho = \rho_\infty$

Closed form expression of ρ_∞

- **Optimality Test:** If ρ is an achievable rate-reduction function and:
 1. For all (p, q) , $\rho(p, q) \geq \rho_0(p, q)$
 2. For all q , $\rho(p, q)$ is concave w.r.t. p
 3. For all p , $\rho(p, q)$ is concave w.r.t. q

then $\rho = \rho_\infty$

- Admissible sum-rate R^* (from beginning of talk):

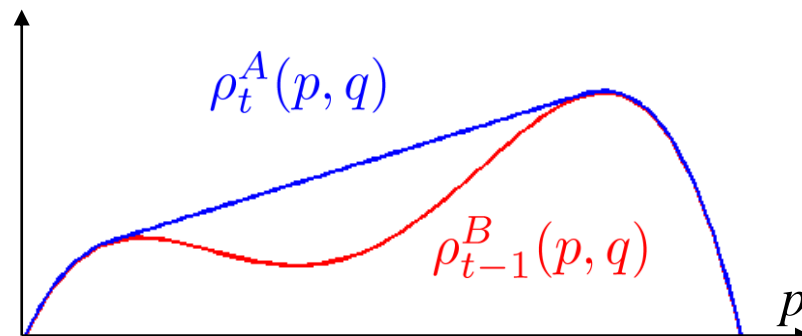
$$R^*(p, q) = \begin{cases} h_2(p) + ph_2(q) + p \log_2 q + p(1 - 2q) \log_2 e, & \text{if } 0 \leq p \leq q \leq 1/2, \\ R^*(q, p), & \text{if } 0 \leq q \leq p \leq 1/2, \\ R^*(1 - p, q), & \text{if } 0 \leq q \leq 1/2 \leq p \leq 1, \\ h_2(p), & \text{if } 1/2 \leq q \leq 1. \end{cases}$$

- Passes the optimality test!
- Therefore $R^* = R_{sum, \infty}$

Alternating concavification algorithm

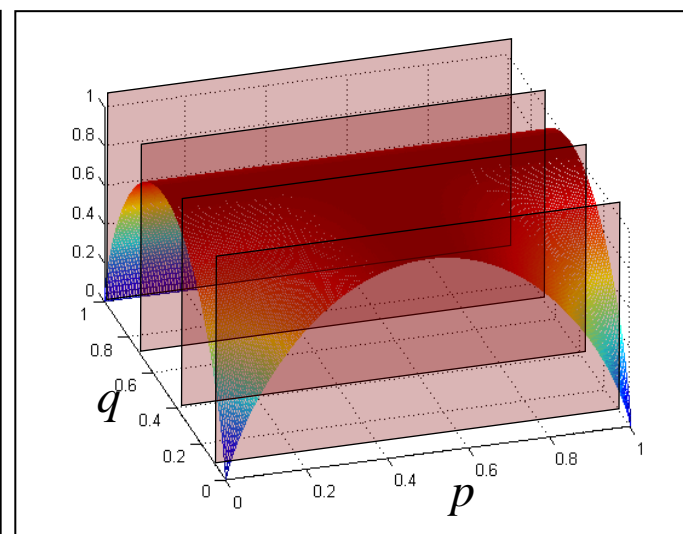
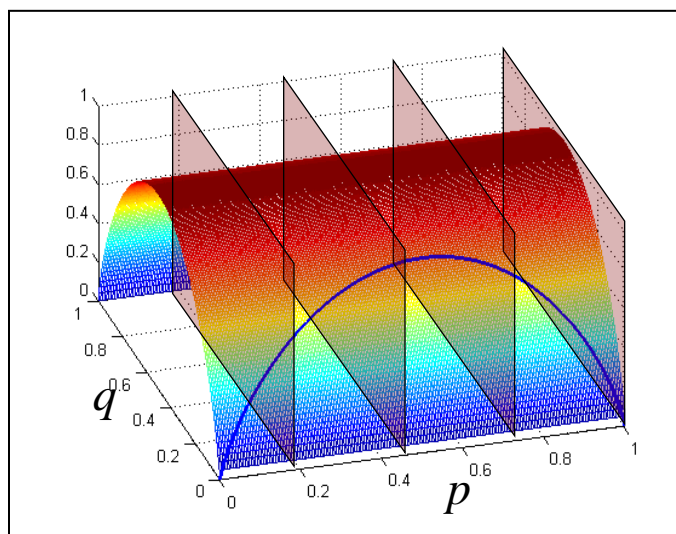
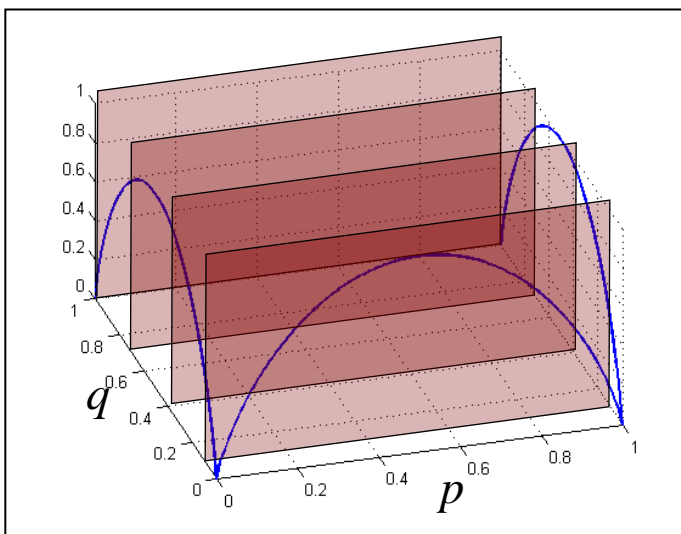
- Recall

$$\rho_{t-1}^B(p, q) \xrightarrow[\text{Fix } q]{\text{Concavify wrt } p} \rho_t^A(p, q)$$



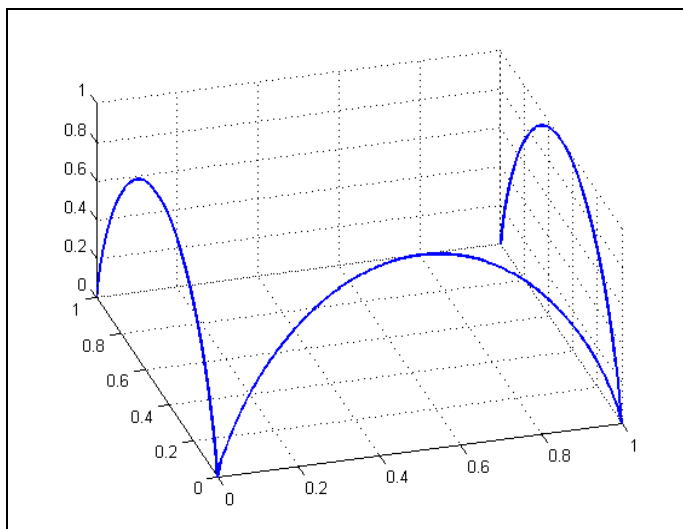
- Iterative algorithm

$$\rho_0(p, q) \xrightarrow[\text{Fix } q]{\text{Concavify wrt } p} \rho_1^A(p, q) \xrightarrow[\text{Fix } p]{\text{Concavify wrt } q} \rho_2^B(p, q) \longrightarrow \dots$$

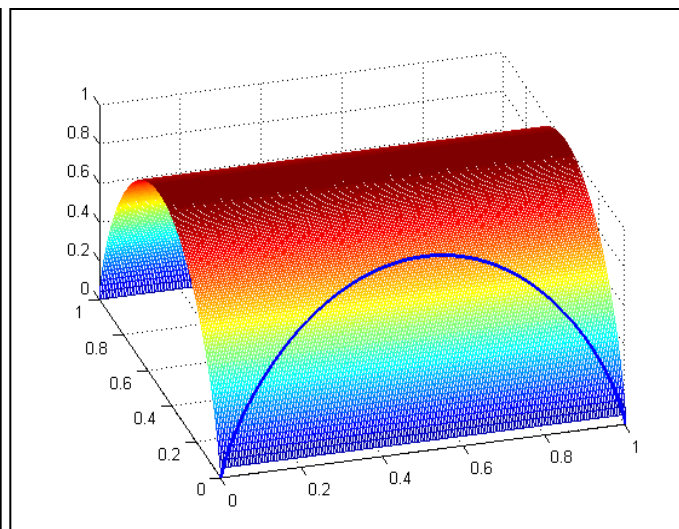


- Each iteration: “same amount” of computation
- Obtain the whole surface $\rho_t(p, q)$ for all (p, q)

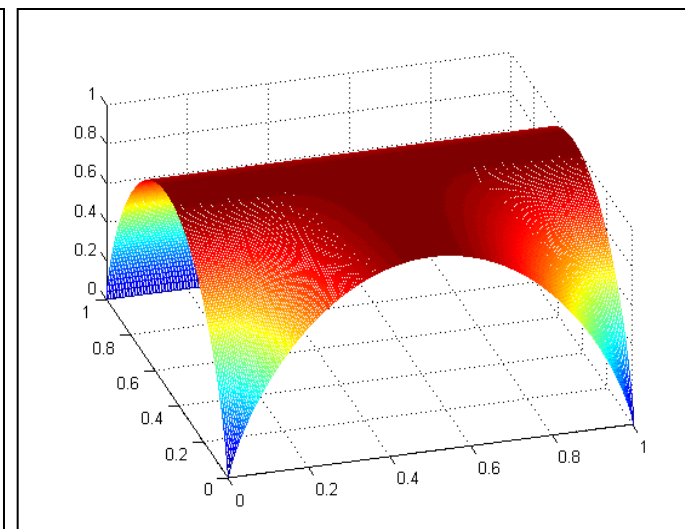
Alternating concavification algorithm



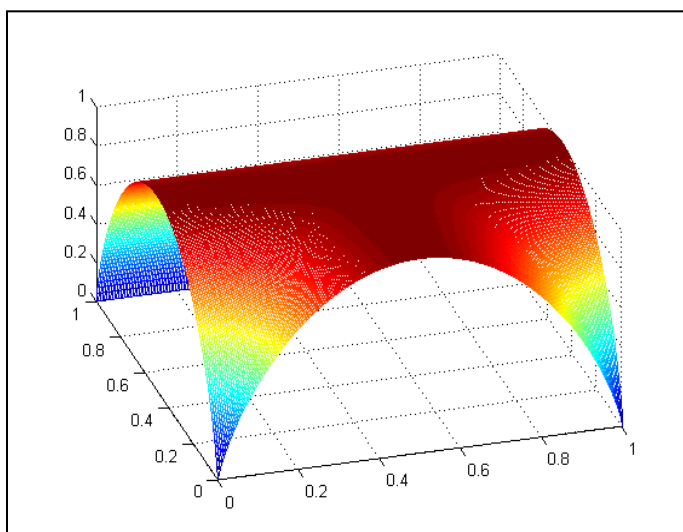
$t = 0$



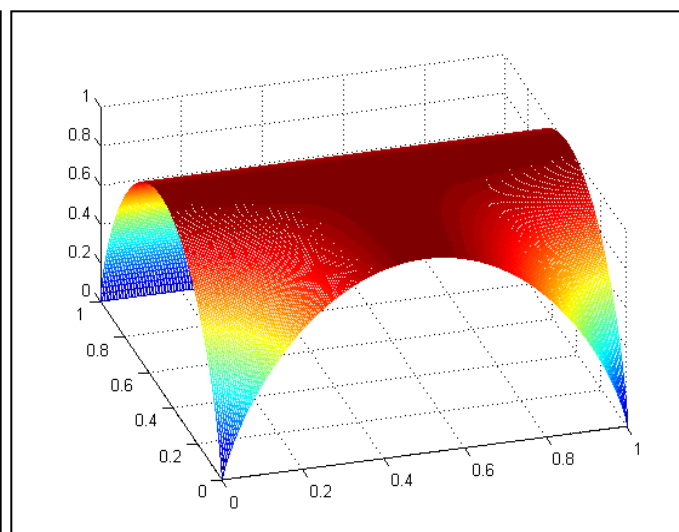
$t = 1$



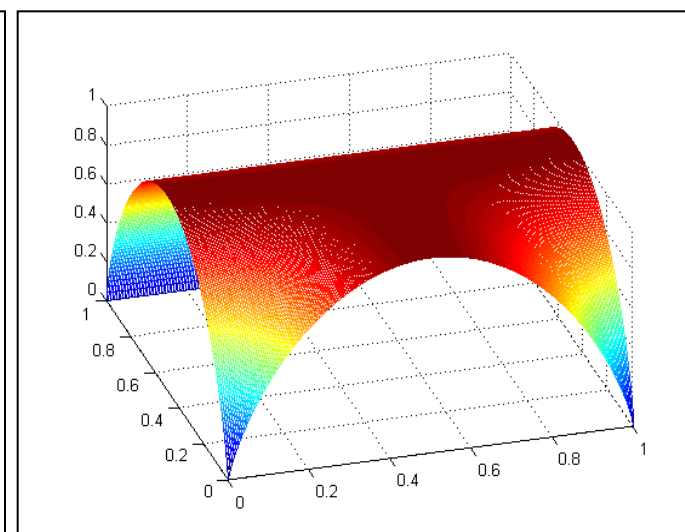
$t = 2$



$t = 3$



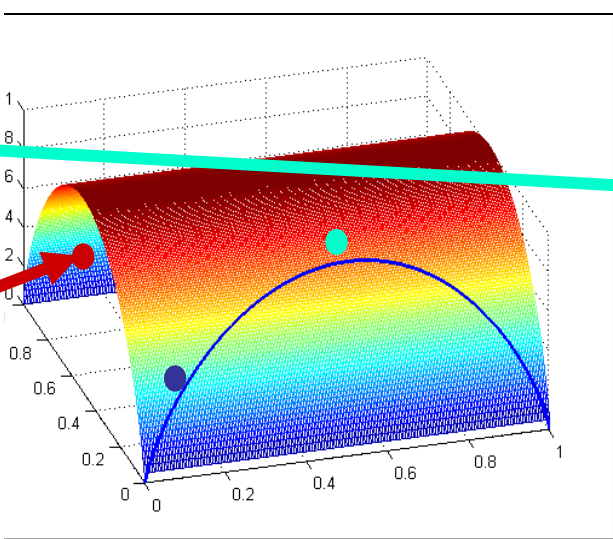
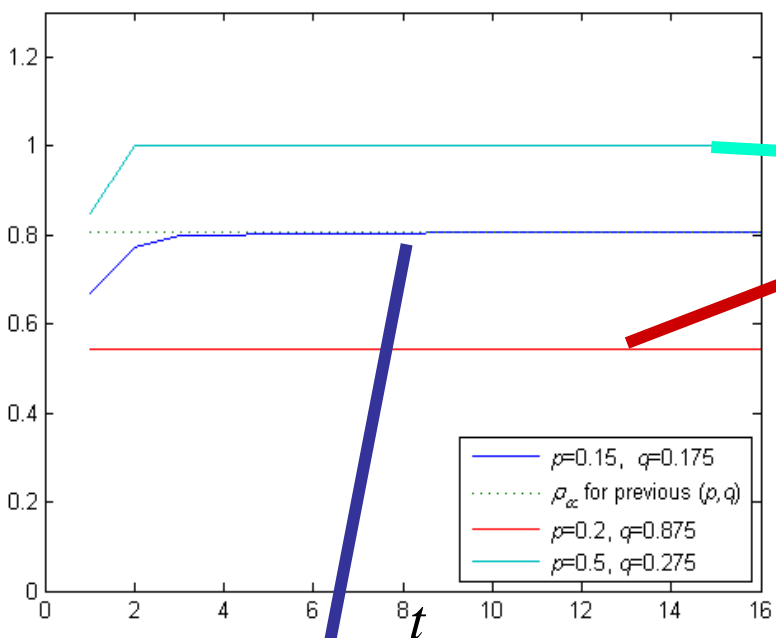
$t = 4$



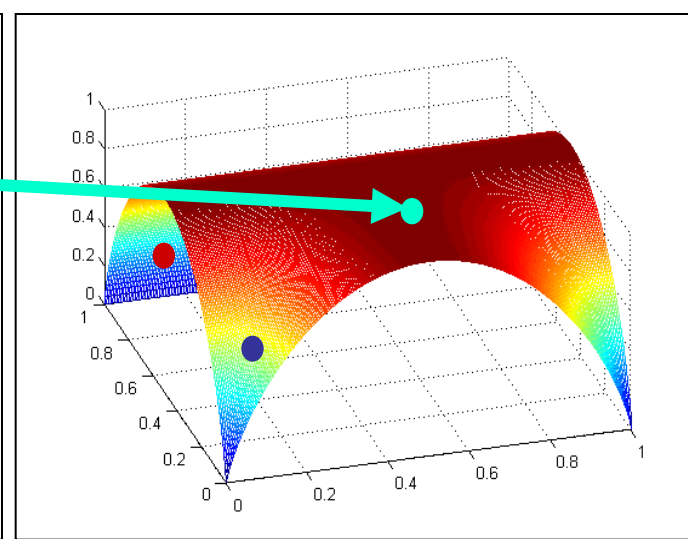
$t = \infty$ (from closed form)

How the surfaces evolve

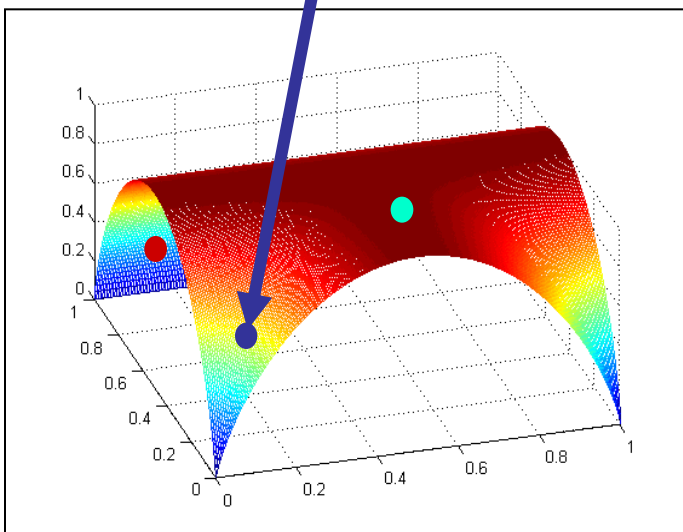
ρ_t



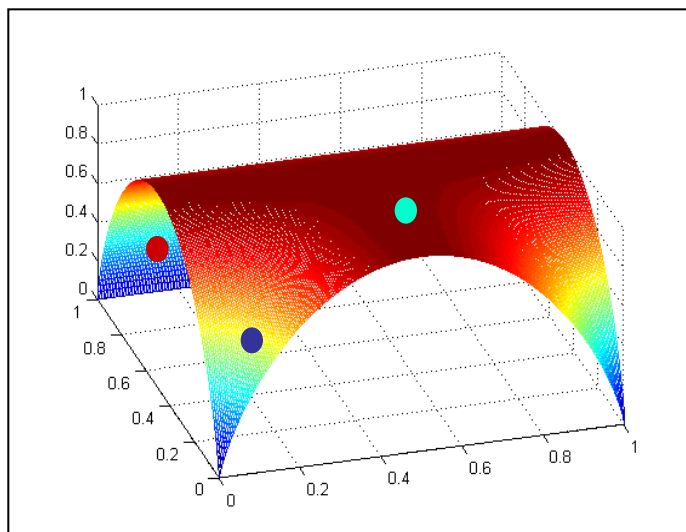
$t = 1$



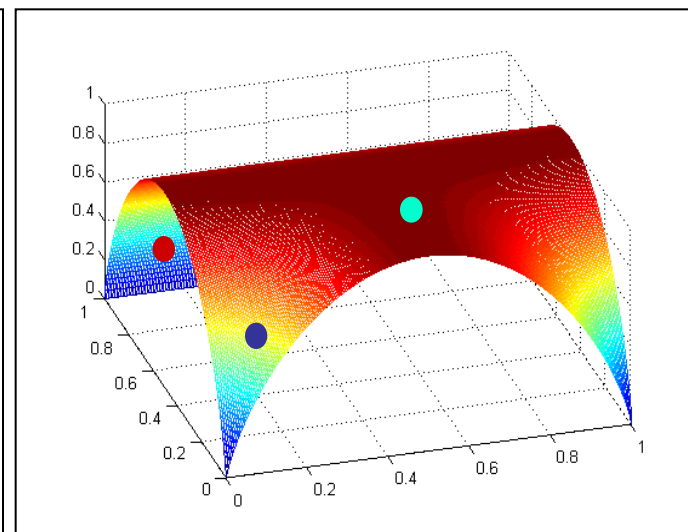
$t = 2$



$t = 3$



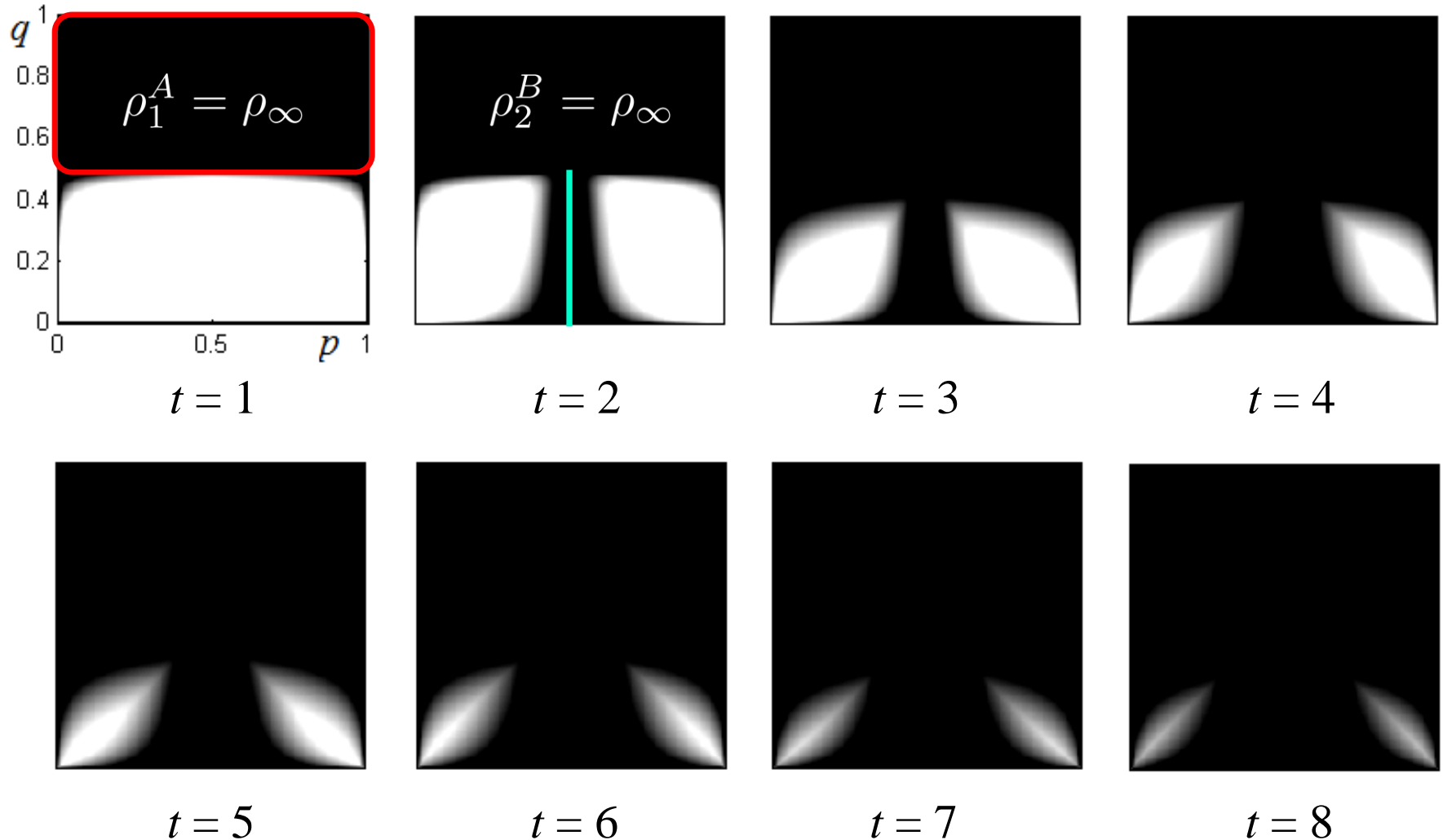
$t = 4$



$t = \infty$ (from closed form)

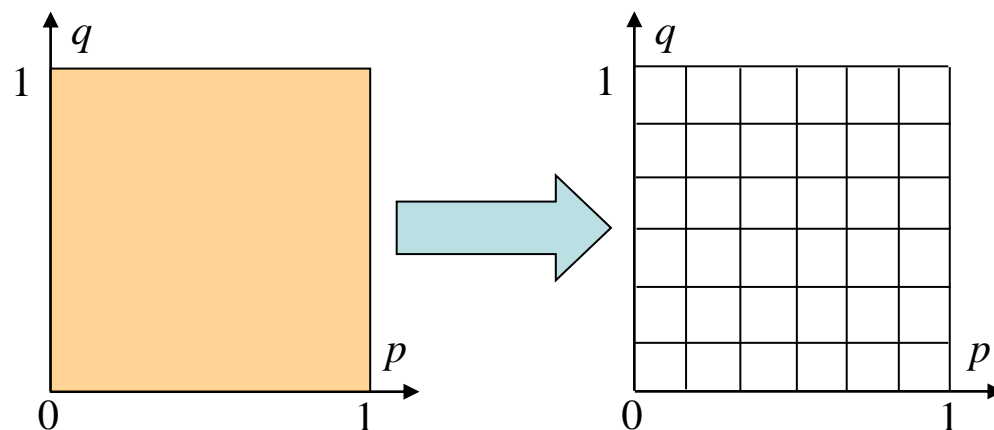
How the surfaces evolve

Brightness: $|\rho_t(p, q) - \rho_\infty(p, q)|$, black: $< 10^{-4}$

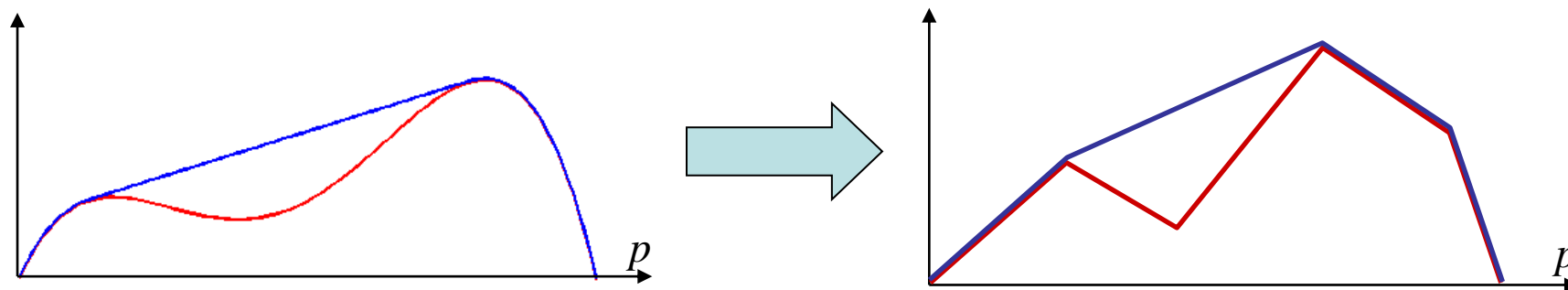


Discretization and convergence speed

- Numerical computation: discretize $[0,1]^2 \rightarrow N \times N$ grid

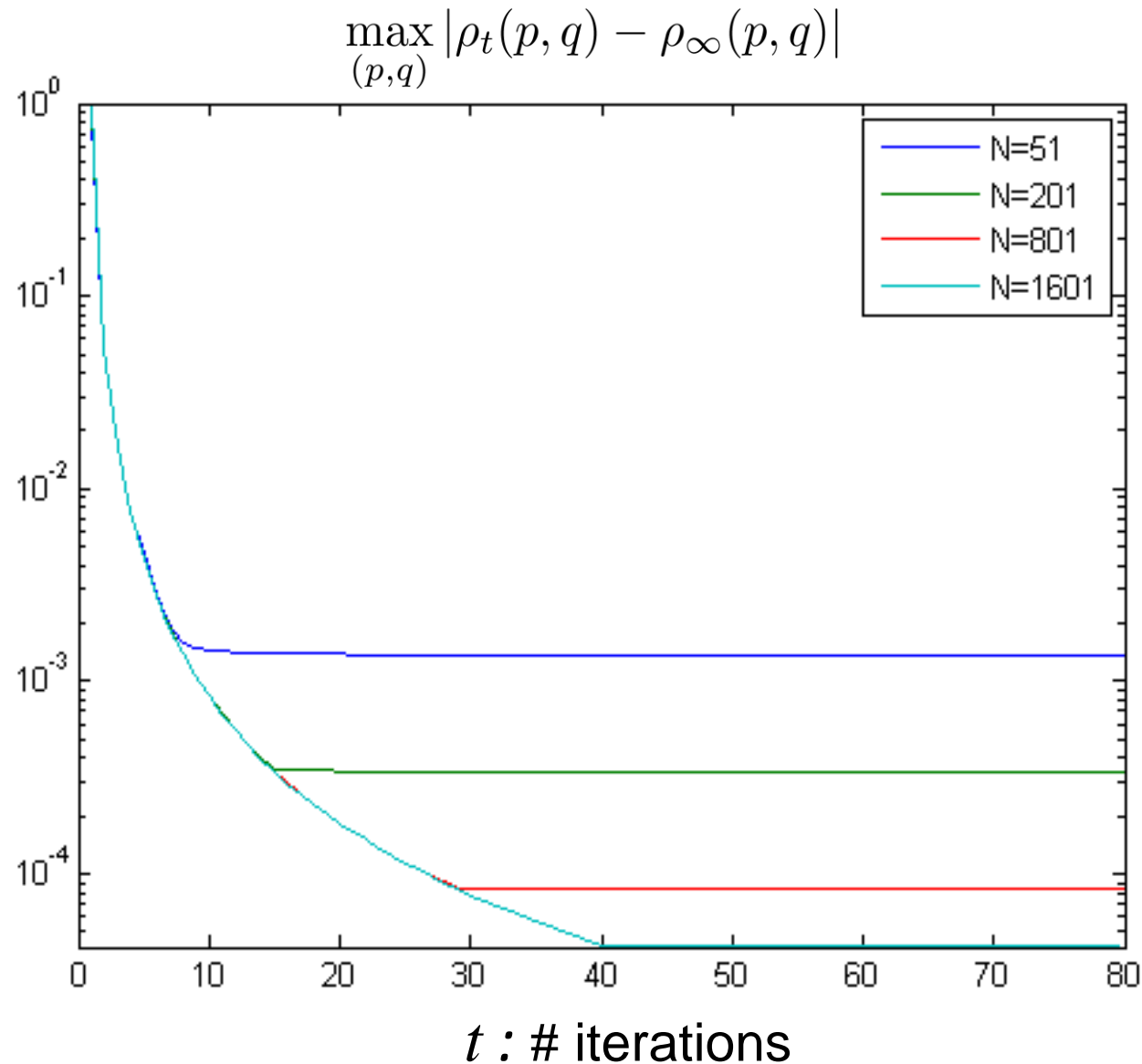


- Function “concavification” \rightarrow convex hull of N points



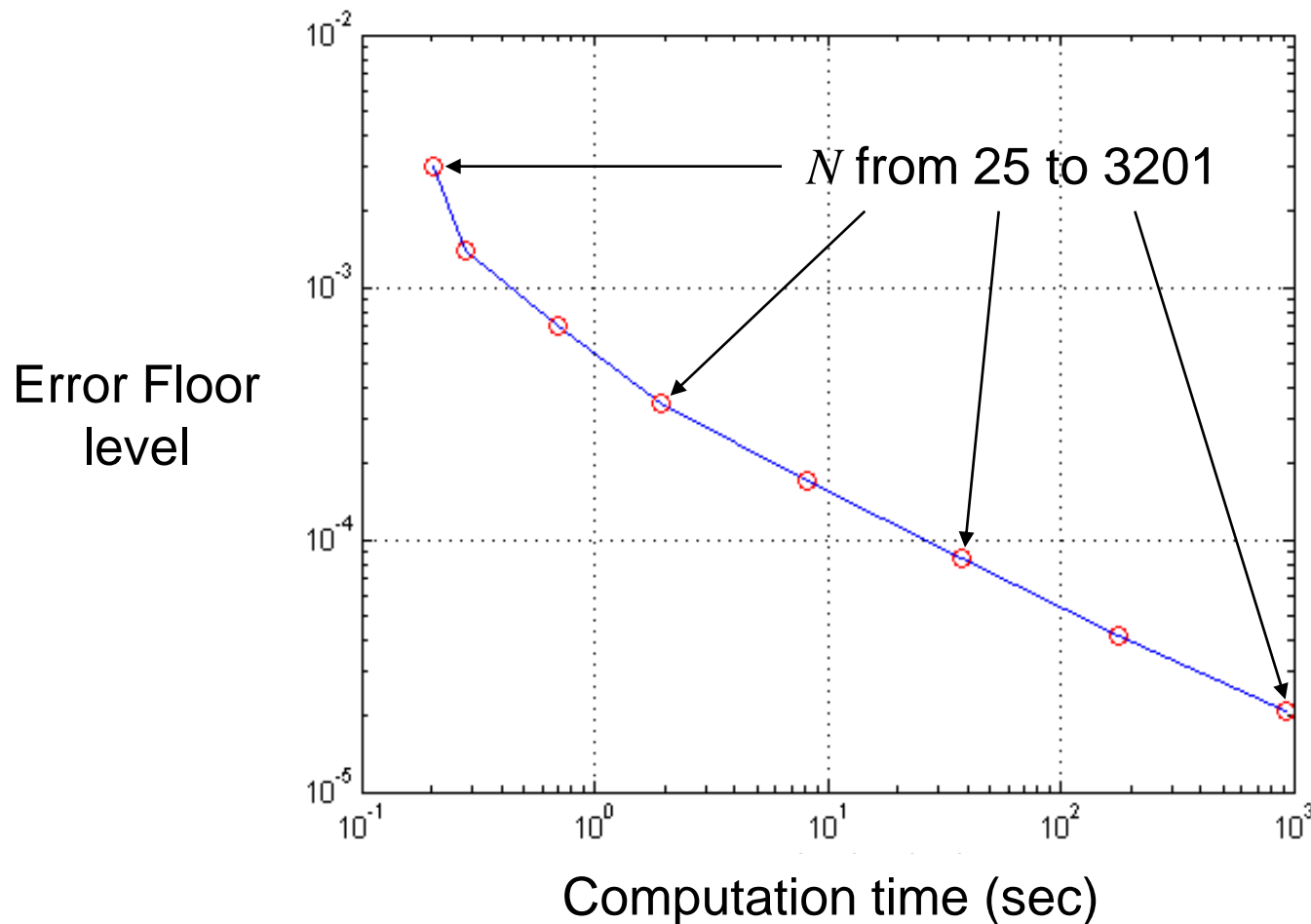
Discretization and convergence speed

- Maximum error v.s. t v.s. N



Discretization and convergence speed

- Error floor level v.s. computation time
 - Experimentally, as N doubled, error floor halved, comp. time x4



General p_{XY} and f_A, f_B with distortions D_A, D_B

“Limit-free” characterization of $\rho_\infty(p_{XY}, D_A, D_B)$ [arXiv Oct'09]

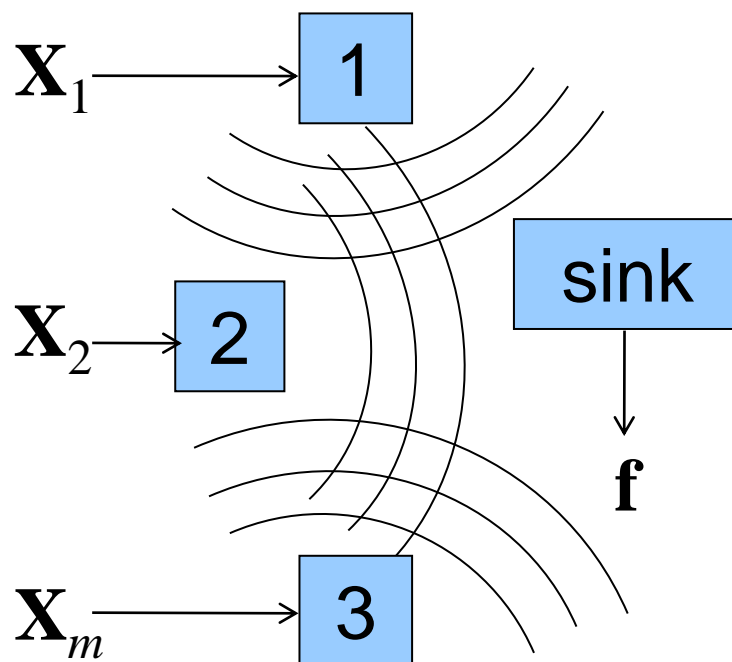
ρ_∞ is the least element of \mathcal{F} , where

$$\mathcal{F} := \left\{ \rho(p_{XY}, D_A, D_B) \right. \left. \begin{array}{l} 1. \quad \rho \geq \rho_0 \\ 2. \quad \text{For all } p_{Y|X}, \rho(p_X p_{Y|X}, D_A, D_B) \text{ is concave} \\ \text{w.r.t. } (p_X, D_A, D_B) \\ 3. \quad \text{For all } p_{X|Y}, \rho(p_Y p_{X|Y}, D_A, D_B) \text{ is concave} \\ \text{w.r.t. } (p_Y, D_A, D_B) \end{array} \right\}$$

- 2 messages can **strictly improve** the Wyner-Ziv R-D function [ISIT'10] + next talk by Nan Ma
 - Resolves a question in Kaspi's 1985 two-way source coding paper

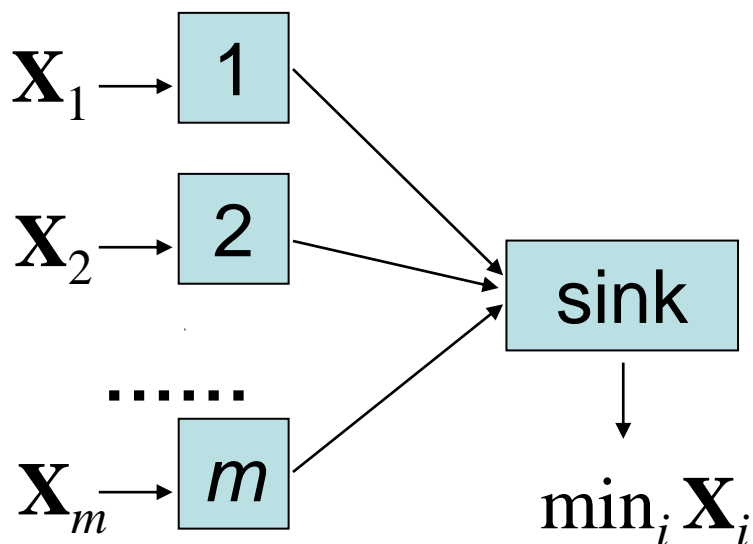
Collocated (broadcast) networks

- General independent sources, general function:
 - Single-letter characterization for $R_{sum,t}$ [ISIT'09]
 - “Limit-free” convex-geometric characterization for $R_{sum,\infty}$ [ISIT'10]



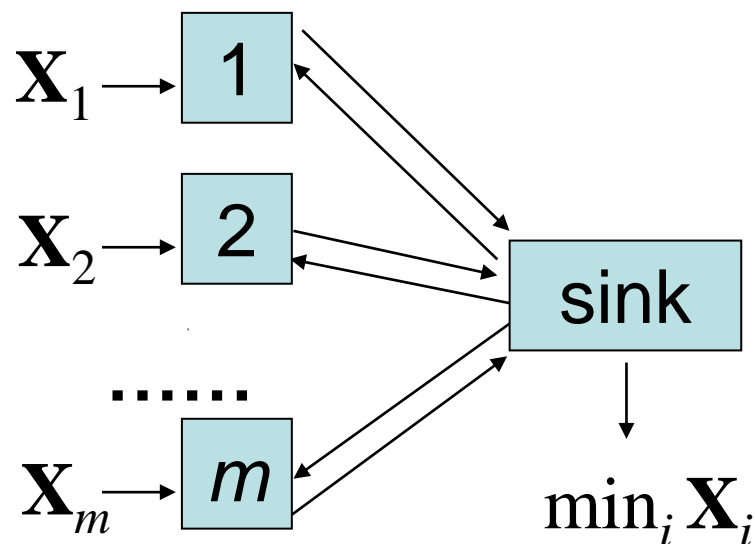
Interaction changes the scaling law in star networks

iid
Ber(1/2)



$$R_{sum}(m) = m$$

iid
Ber(1/2)



$$1 \leq R_{sum}(m) < 6$$

Concluding remarks

- Characterizing the ultimate limits of interaction:
 - New type of functional single-letter characterization of $R_{sum,\infty}$
 - Alternating “concavification” algorithm for computing $R_{sum,\infty}$
- A new way to construct an infinite sequence of auxiliary random variables that is also optimal
- Infinite messages with infinitesimal rate => “Calculus” for source coding?

Open Problems

- Determine the computational complexity as a function of grid (discretization) size
- Characterize the rate of convergence
- Provide a simple procedure to **directly determine** for each given joint pmf, the smallest number of messages needed to reach the infinite-message limit
- Develop the functional characterization from first principles without using the single-letter characterization (functional equations and inequalities may be useful here)
- Find channel coding counterparts