

# Coding for Computing

Jie Ren

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ASPITRG, Drexel University

# Outline

- Background and definitions
- Main Results
- Examples and Analysis
- Proofs

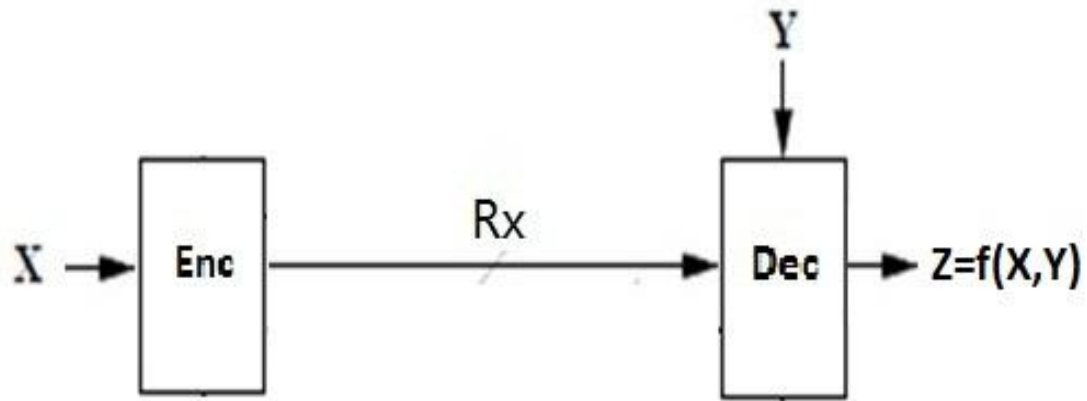


# Background and definitions

- Problem Setup
- Graph Entropy
- Conditional Graph Entropy
- Characteristic Graph
- Wyner-Ziv



# Problem Setup



# Problem Setup

- Encode by block

Transmits after observing the whole block of  $X$

- Decode by block

$f(X,Y)$  must be determined for a block of  $(X,Y)$

- Vanishing block-error probability

$$p(Z^N \neq f(X^N, Y^N)) \leq \epsilon$$

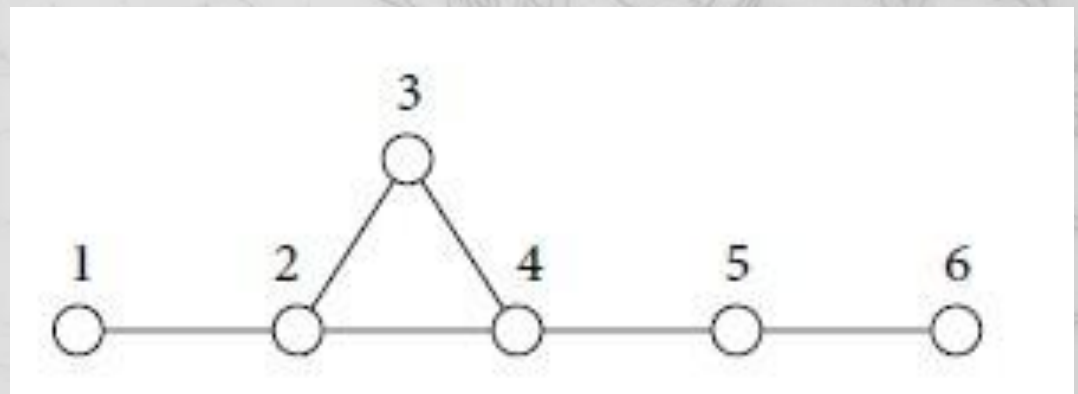
- Problem's rate

$$L_f(X|Y) = R/N$$

# Graph Entropy

Maximum independent sets of  $G(V,E)$

$$\Gamma(G) = \{\{1,3,5\}, \{1,3,6\}, \{1,4,6\}, \{2,5\}, \{2,6\}\}$$

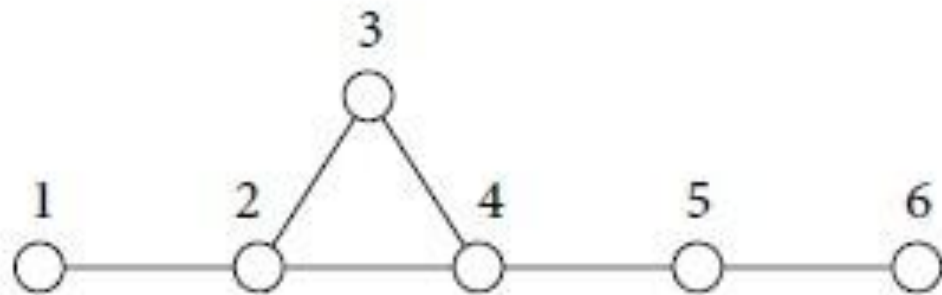


# Graph Entropy

Define random viable  $W$ ,  $W \in \Gamma(G)$

$p(w|x) = 0$  if  $x$  is not in  $w$

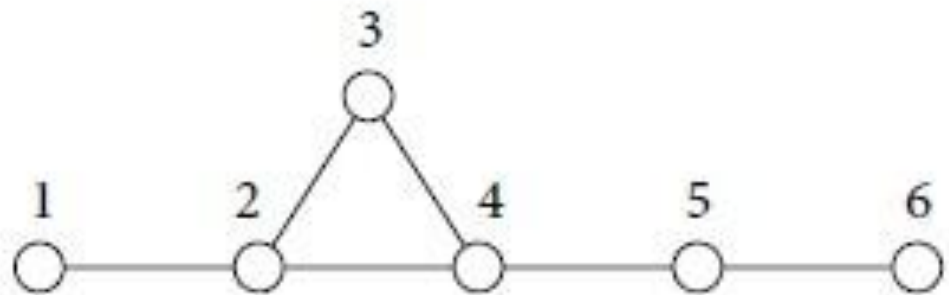
$$\sum_{w:x \in w} p(w|x) = 1$$



# Graph Entropy

Graph Entropy:

$$H_G(X) = \min_{p(w|x)} I(X; W)$$





# Conditional Graph Entropy

Extend the definition of graph entropy:

Let  $(X,Y)$  be a random pair and let a graph  $G$  be defined over the support set of  $X$

$$H_G(X|Y) = \min_{\substack{W-X-Y \\ X \in W \in \Gamma(G)}} I(W; X|Y)$$



# Characteristic Graph

- $G(V,E)$  and  $f$
- The vertex set is the support set of  $X$
- The edge  $(x,x')$  exists if there is a  $y$  such that

$$p(x,y) > 0 \quad p(x',y) > 0$$

$$f(x,y) \neq f(x',y)$$

# Outline

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- Main Results
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- Proofs



# Main Results

- Naive bound
- Theorem 1
- Theorem 2
- Theorem 3



# Main Results(Naive bounds)

Inner bound

$$L_f(X|Y) \geq H(f(X, Y)|Y)$$

Outer bound

$$L_f(X|Y) \leq \min \{H(g(X)|Y): g(X) \text{ and } Y \text{ determine } f\}$$

$$H(f(X, Y)|g(X), Y) = 0$$

# Main Results(Naive bounds)

- lower bound: number of bits required when  $P_x$  knows  $Y$  in advance
- upper bound : using slepian-wolf theorem
- Both bounds are tight in some special cases but not in general



# Main Results(Theorem1)

Theorem 1 for every  $X, Y$  and  $f$

$$L_f(X|Y) = H_G(X|Y)$$

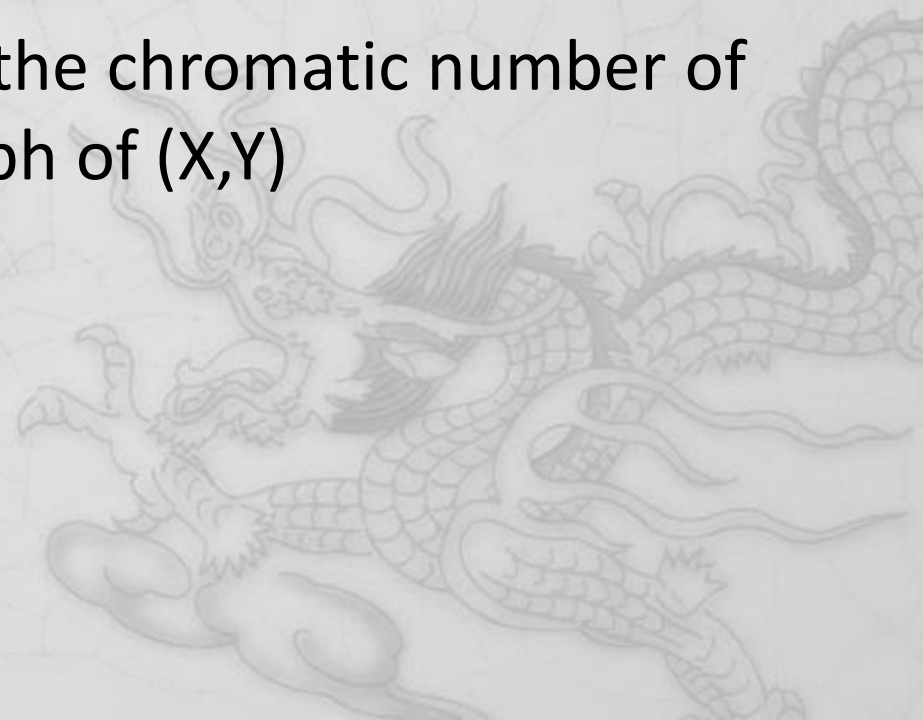


# Zero Error Worst Length Case

One-Way Complexity

$$\hat{C}_1(X|Y) = \lceil \log \omega(G(X|Y)) \rceil$$

The one-way complexity is the chromatic number of the characteristic hypergraph of  $(X,Y)$

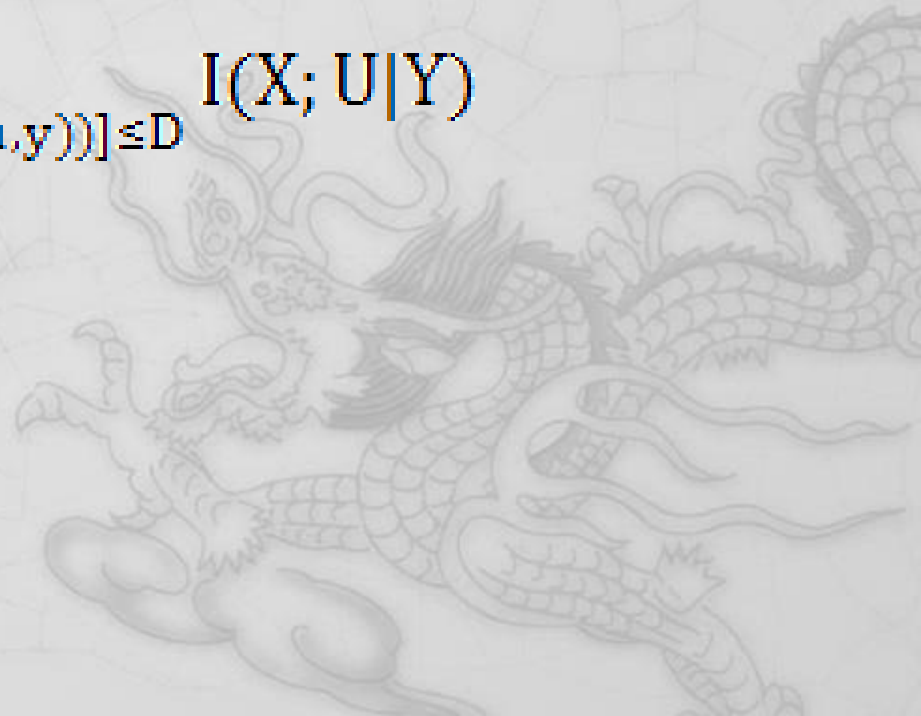




# Main Results(Theorem2)

- An extension of Wyner-Ziv
- Allowing the distortion to depend on  $Y$  as well as on  $X$  and  $Z$

$$R_D = \min_{\substack{p(u|x), g | E[d(x,y,g(u,y))] \leq D \\ U \leftrightarrow X \leftrightarrow Y}} I(X; U|Y)$$



# Main Results(Theorem2)

- Applicant in our problem

$$d(x, y, z) = \begin{cases} 0, & \text{if } z = f(x, y) \\ 1, & \text{otherwise} \end{cases}$$

$$d(X, Y, Z) = p(Z \neq f(X, Y))$$

$$d(X^N, Y^N, Z^N) = \frac{1}{N} \sum_{i=1}^N p(Z_i \neq f(X_i, Y_i))$$

# Main Results(Theorem 2)

Lemma: For every  $X, Y$  and  $f$

$$L_f(X|Y) \geq R(0)$$

$$p(Z^N \neq f(X^N, Y^N)) \geq \frac{1}{N} \sum_{i=1}^N p(Z_i \neq f(X_i, Y_i))$$

hence every rate achievable with vanishing block error is also achievable with zero distortion

# Main Results(Theorem 2)

Theorem 2 For every  $X, Y$  and  $f$

$$R(0) = H_G(X|Y)$$

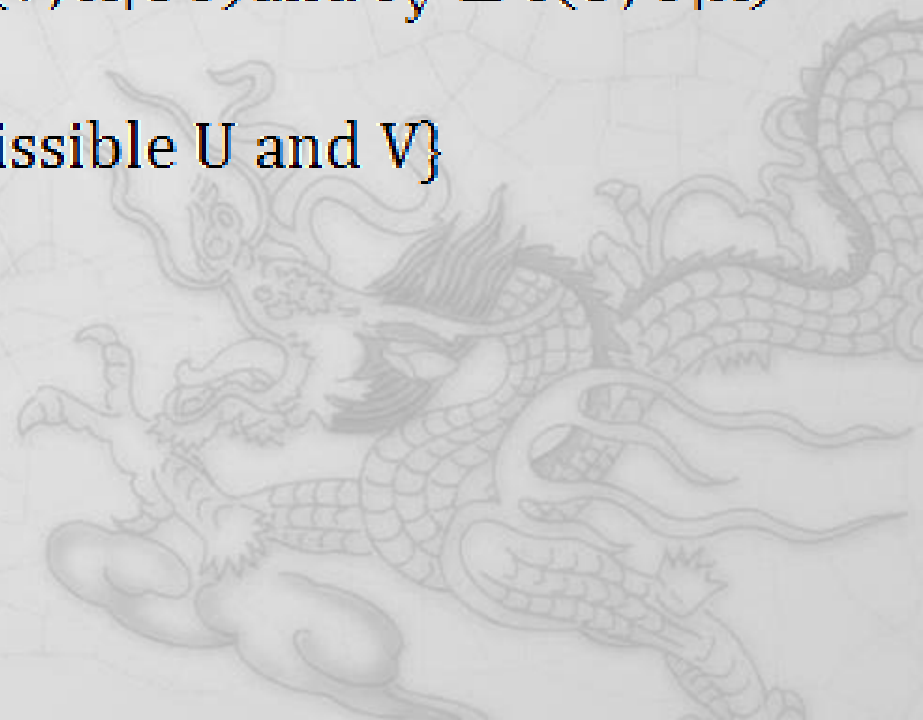


# Main Results

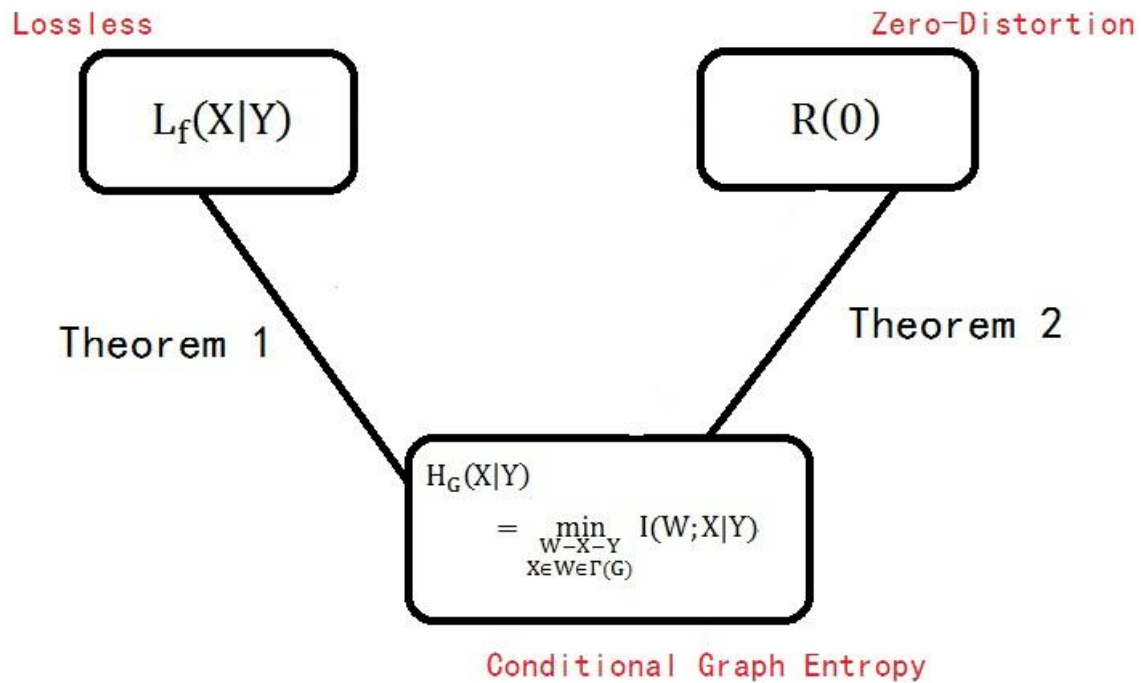
Theorem 3 For every  $X, Y$  and  $f$

$$R_f^2(X|Y) = \{(r_x, r_y) : r_x \geq I(V; X|UY) \text{ and } r_y \geq I(U; Y|X)\}$$

for some admissible  $U$  and  $V$



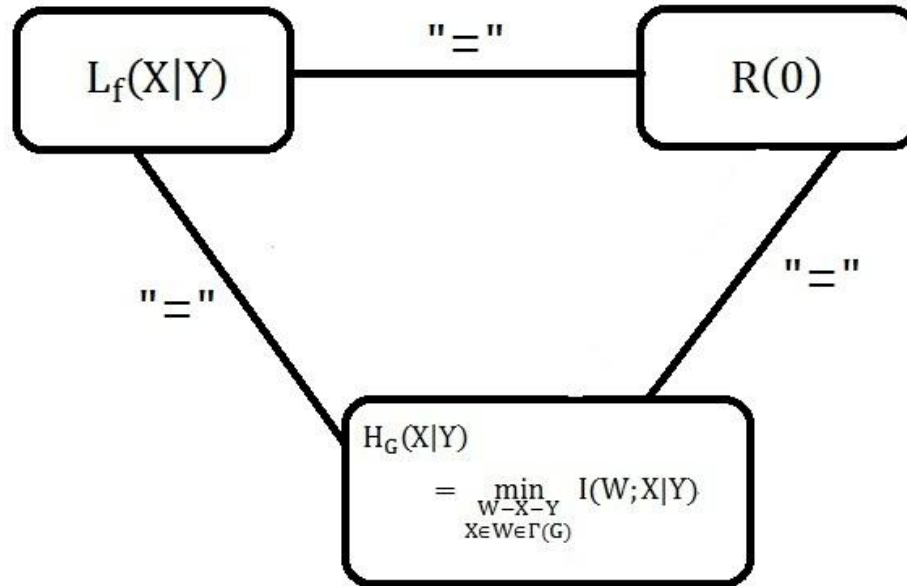
# Proof of Theorem 1&2



# Proof of Theorem 1&2

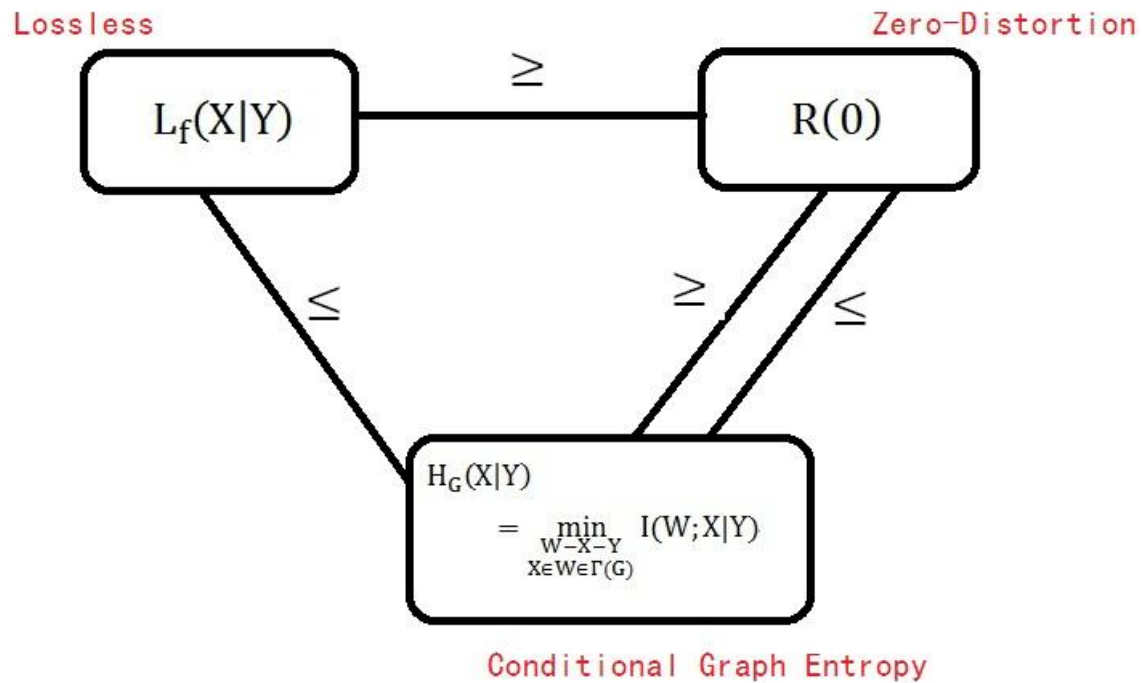
Lossless

Zero-Distortion



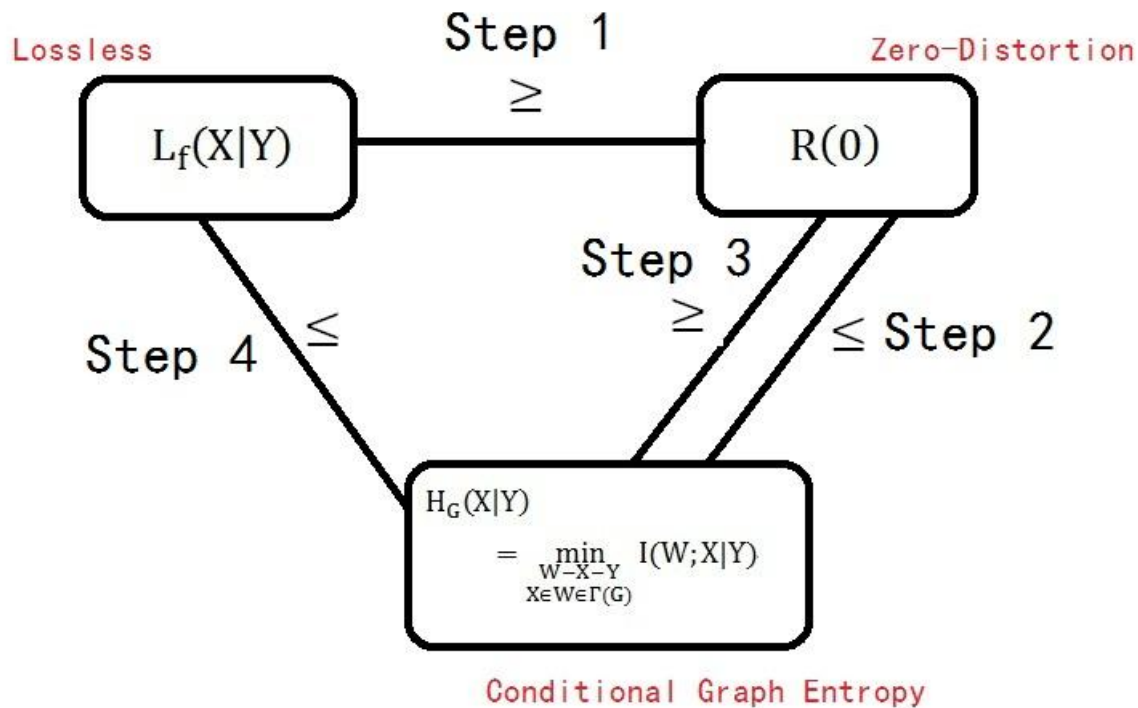
Conditional Graph Entropy

# Proof of Theorem 1&2

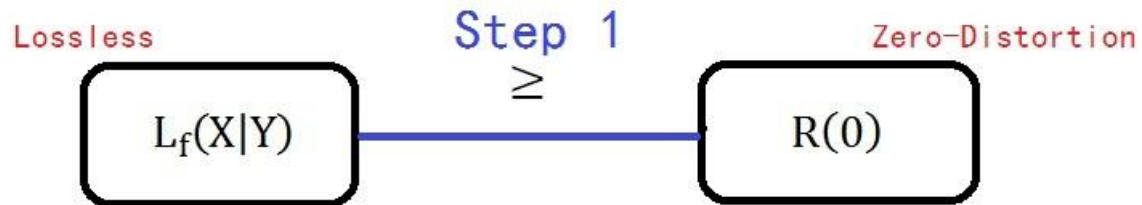




# Proof of Theorem 1&2



# Proof of Theorem 1&2



$H_G(X|Y)$

$$= \min_{\substack{W-X-Y \\ X \in W \in \Gamma(G)}} I(W; X|Y)$$

Conditional Graph Entropy

# Step 1

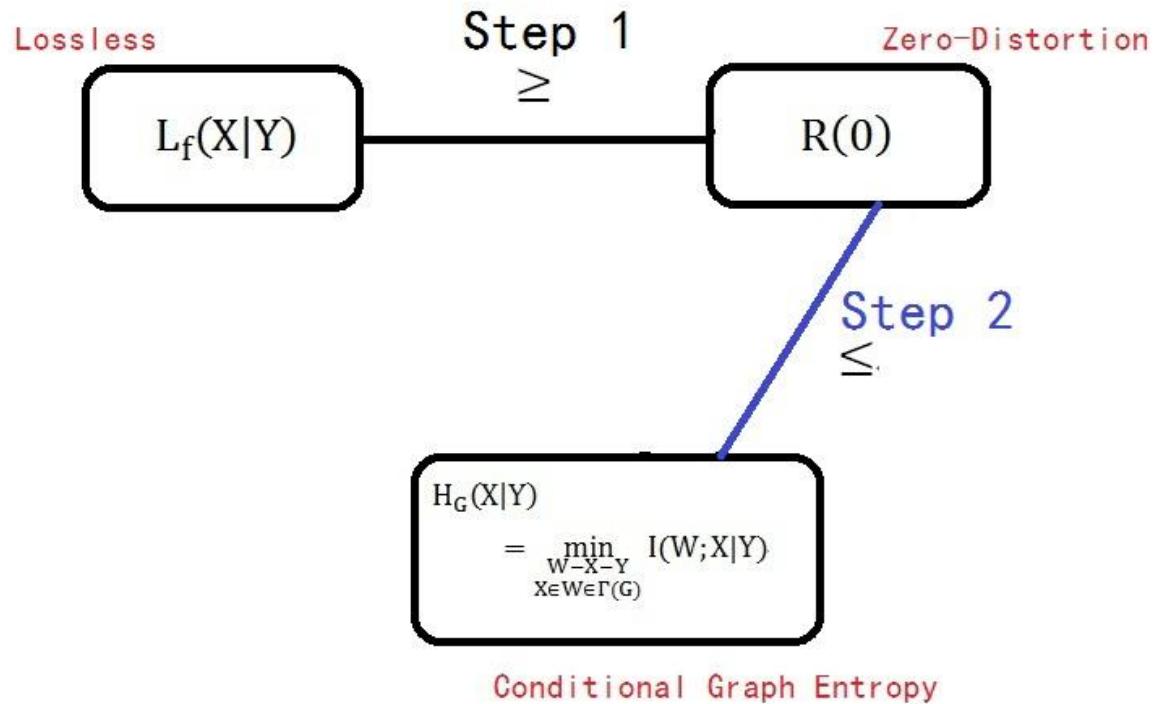
Lemma 1: For every  $X, Y$  and  $f$

$$L_f(X|Y) \geq R(0)$$

$$p(Z^N \neq f(X^N, Y^N)) \geq \frac{1}{N} \sum_{i=1}^N p(Z_i \neq f(X_i, Y_i))$$

hence every rate achievable with vanishing block error is also achievable with zero distortion

# Proof of Theorem 1&2



# Step 2

To show

$$\min_{\substack{V-X-Y \\ g | E[d(X, Y, g(V, Y)) \leq 0]}} I(V; X|Y) \leq \min_{\substack{W-X-Y \\ X \in W \in \Gamma(G)}} I(W; X|Y)$$

Need to show that if  $X \in W \in \Gamma(G)$

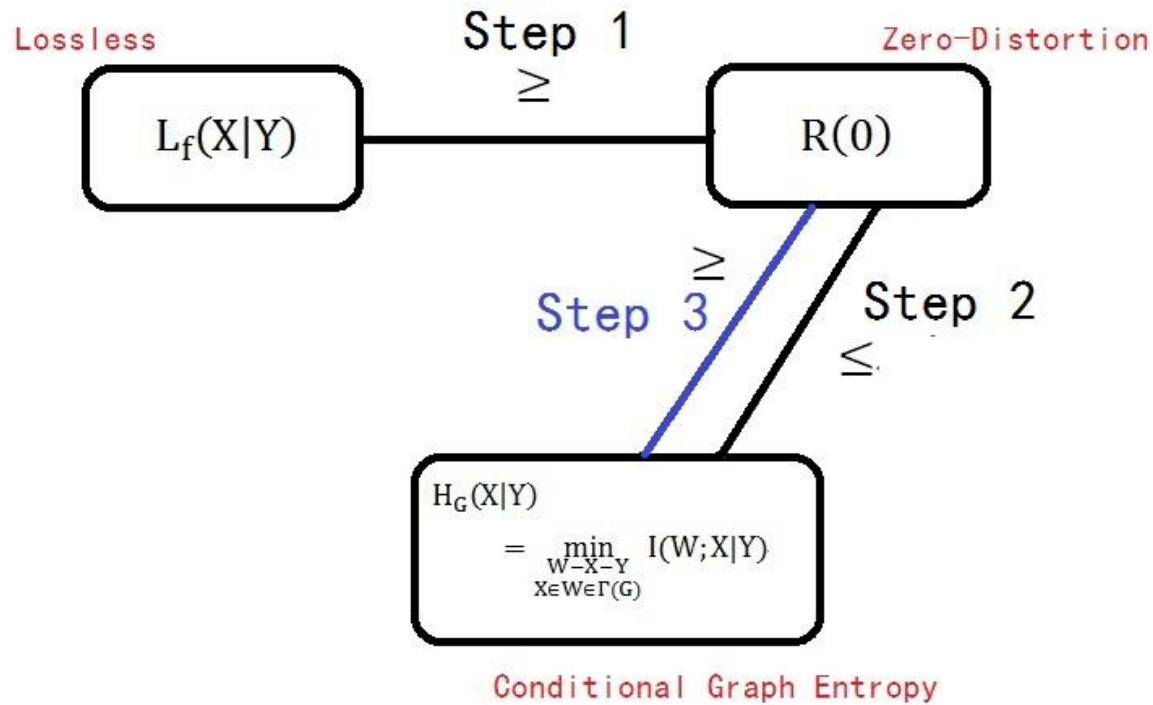
Then, there is a function  $g$  over  $\Gamma(G) \times Y$  s.t.

$f(x, y) = g(w, y)$  whenever  $p(w, x, y) > 0$

Hence

$$E[d(X, Y, g(W, Y))] = 0$$

# Proof of Theorem 1&2



# Step 3

To show

$$\min_{\substack{V-X-Y \\ g | \text{Ed}(X, Y, g(V, Y)) \leq 0}} I(V; X|Y) \geq \min_{\substack{W-X-Y \\ X \in W \in \Gamma(G)}} I(W; X|Y)$$

Suppose that  $V-X-Y$  and that there exists  $g$  such that  $\text{Ed}(X, Y, g(V, Y)) \leq 0$ . We define  $W$  and prove:

1.  $X \in W \in \Gamma(G)$
2.  $W - X - Y$
3.  $I(W; X|Y) \leq I(V; X|Y)$

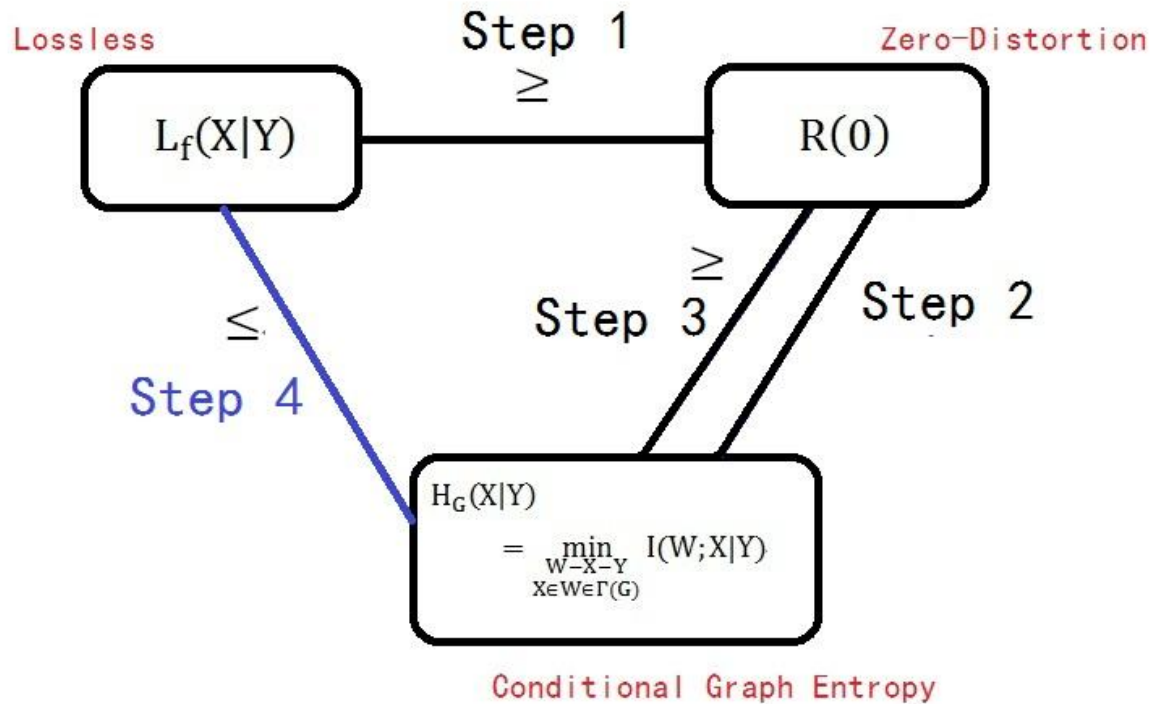
# Outline

- Finish the proof of theorem 1 and 2
- Analysis 2-Way Communication Model
  - Definitions
  - Results(Theorem 3)
  - Proof





# Proof of Theorem 1&2



# Step 4

Introduce Robust Typical Set

$$T_{\epsilon}^N(X) = \{\vec{x} \in X^N \mid \left| \frac{1}{N} N(x|\vec{x}) - p_X(x) \right| < \frac{\epsilon}{|X|} p_X(x)\}$$

Compare with Strong Typical Set

$$T_{\epsilon}^N(X) = \{\vec{x} \in X^N \mid \left| \frac{1}{N} N(x|\vec{x}) - p_X(x) \right| < \frac{\epsilon}{|X|}\}$$

# Step 4

High-level idea

Design a protocol based on conditional graph entropy

Define

$$J^* = \{j: (W^j, X) \in T_2\}$$

$$K^* = \{k: (W^k, Y) \in T_3 \text{ and } \Phi(k) = \Phi(j)\}$$

If  $|K^*| = 1$  then we can decode the message

# Step 4

Encoder:

If  $J^*$  is empty, transmits an error message. Otherwise, transmits  $\Phi(J)$

Decoder:

If  $|K^*| \neq 1$ , declares an error. Otherwise, proceeds to determine

$$g(W_i^K, Y_i) \text{ for all } i$$

# Step 4

Error Analysis:

$$E1: J^* = \emptyset$$

$$E2: J^* \neq \emptyset \text{ and } (W^J, Y) \text{ is not in } T_3$$

$$E3: J^* \neq \emptyset \text{ and } \exists k \neq J \text{ such that } (W^K, Y) \in T_3 \\ \text{and, } \Phi(k) = \Phi(J)$$

Need to show these errors are exponentially small

# Proof of Theorem 1&2

Lossless

$$L_f(X|Y)$$

Theorem 1

Zero-Distortion

$$R(0)$$

Theorem 2

$$H_G(X|Y) = \min_{\substack{W-X-Y \\ X \in W \in \Gamma(G)}} I(W; X|Y)$$

Conditional Graph Entropy

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- Finish the proof of theorem 1 and 2
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# Outline

- Finish the proof of theorem 1 and 2
- **Analysis 2-Way Communication Model**
  - Definitions
  - Results(Theorem 3)
  - Proof





# Definitions

Independent instances of a random pair  $(X,Y)$

$$\{(X_i, Y_i)\}_{i=1}^{\infty}$$

A two-message protocol consists of:

Y-encoding function  $\varphi : Y^n \rightarrow \{0,1\}^l$

X-encoding function  $\varepsilon : \{0,1\}^l \times X^n \rightarrow \{0,1\}^k$

Decoding function  $\Psi : \{0,1\}^l \times \{0,1\}^k \times X^n \rightarrow Z^n$

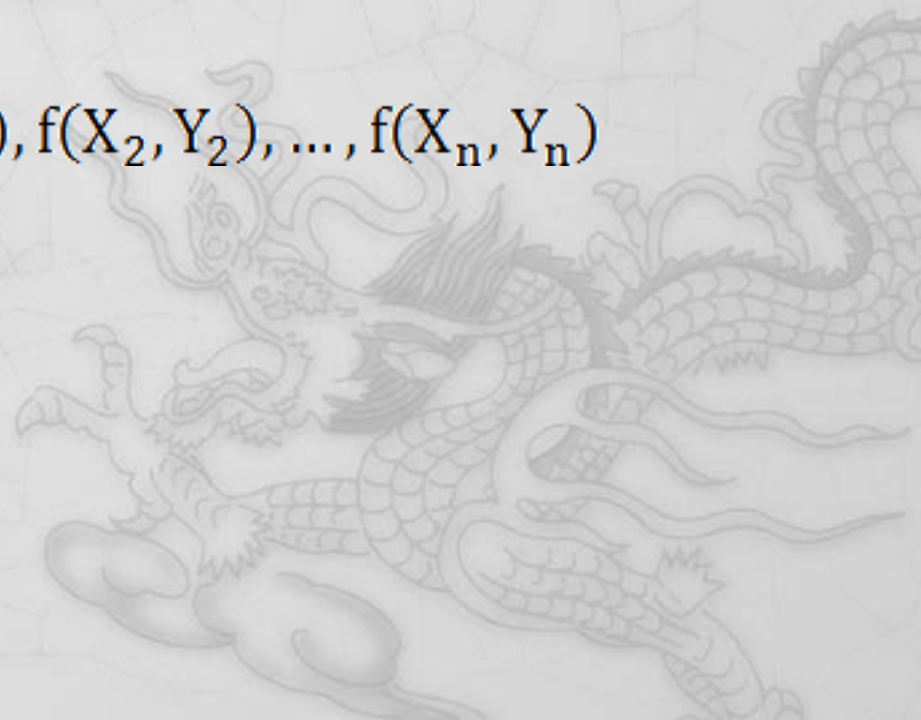
# Definitions

Block-error probability

$$p[\Psi(\varphi(Y^n), \varepsilon(\varphi(Y^n), X^n), Y^n) \neq f(X^n, Y^n)]$$

Where

$$f(X^n, Y^n) = f(X_1, Y_1), f(X_2, Y_2), \dots, f(X_n, Y_n)$$



# Definitions

Rate region

$$R_f^2(X|Y)$$

Two-message communication complexity

$$L_f^2(X|Y) = \min \{r_x + r_y : (r_x, r_y) \in R_f^2(X|Y)\}$$

# Outline

- Finish the proof of theorem 1 and 2
- **Analysis 2-Way Communication Model**
  - Definitions
  - Results(Theorem 3)**
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# Theorem 3

Two random variables  $U$  and  $V$  defined over finite alphabets are admissible if

1)  $U - Y - X$

2)  $V - UX - Y$

3)  $U, V,$  and  $Y$  determine  $f(X,Y)$




# Theorem 3

Theorem 3 :

For every  $(X,Y)$  and  $f$

$$R_f^2(X|Y) = \{r_x + r_y : r_x \geq I(V; X|UY)$$

and  $r_y \geq I(U; Y|X)$  for some admissible  $U$  and  $V\}$



# Outline

- Finish the proof of theorem 1 and 2
- **Analysis 2-Way Communication Model**
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  - Proof**



# Proof of Theorem 3

Converse:

1. Obtain an outer bound on  $S_3=(rx,ry,D)$ .

Show  $S_3 \subseteq S_3^*$

Where  $S_3^*$  defined in terms of  $X, Y, T, U, V$

2. Show  $S_3^*$  can be expressed in terms of only  $X, Y, U, V$

3. Show  $S_2$ , the 2-D slice with  $D=0$  is contained in the 2-D region  $S'_2$