

Multiple Description Coding Over Many Channels

Raman Venkataramani
Harvard University, Cambridge, MA
raman@deas.harvard.edu

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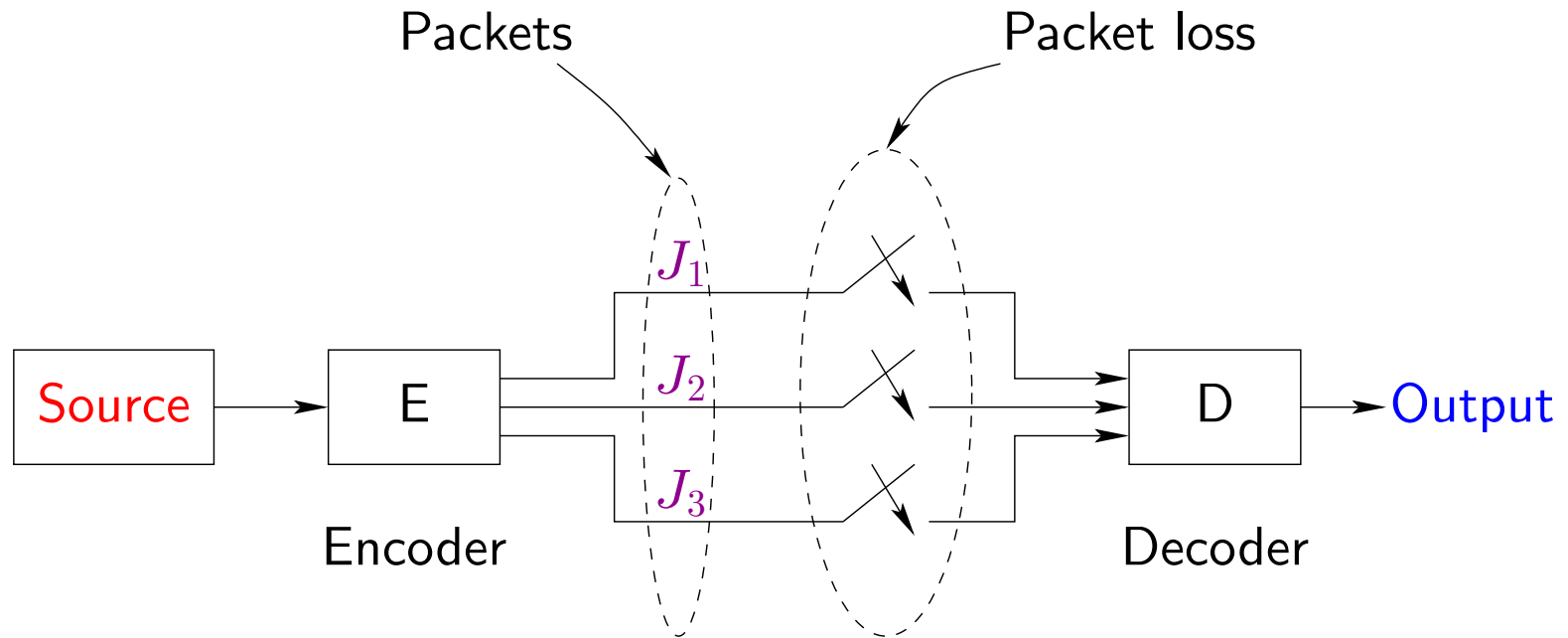
Outline

- Motivation for *Multiple Description Coding*
- Problem statement
- Review of known results (*Two-Description Coding*)
- New results for L -description coding ($L > 2$)
- Special cases and examples
- Summary

Underwater Wireless Channels

- Limited bandwidth and severe multipath propagation.
- Long delays (speed of sound in water is 1500m/s)
- Need for transmission of **real-time** data:
 - e.g., images or video data to/from divers.
- Poor propagation occasionally cause weak SNRs.

Erasure Channels



- Packets are either **lost completely** (erased) or **received error-free**

How do we deal with erasures?

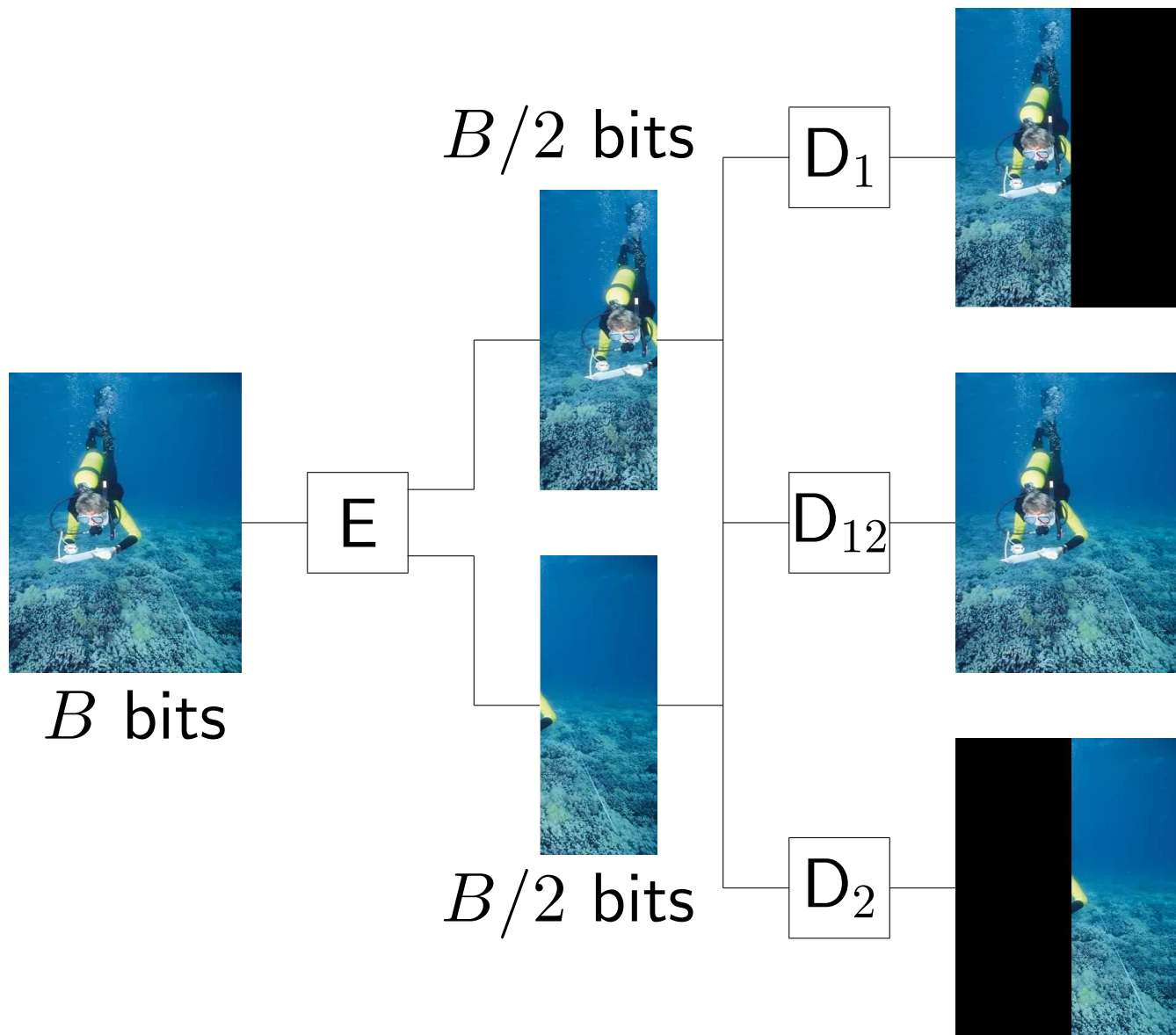
- Request a **retransmission**.
 - Ideal for loss-less transmission.
 - Not feasible for **real time** data such as **voice** and **video**.

Alternate Approach:

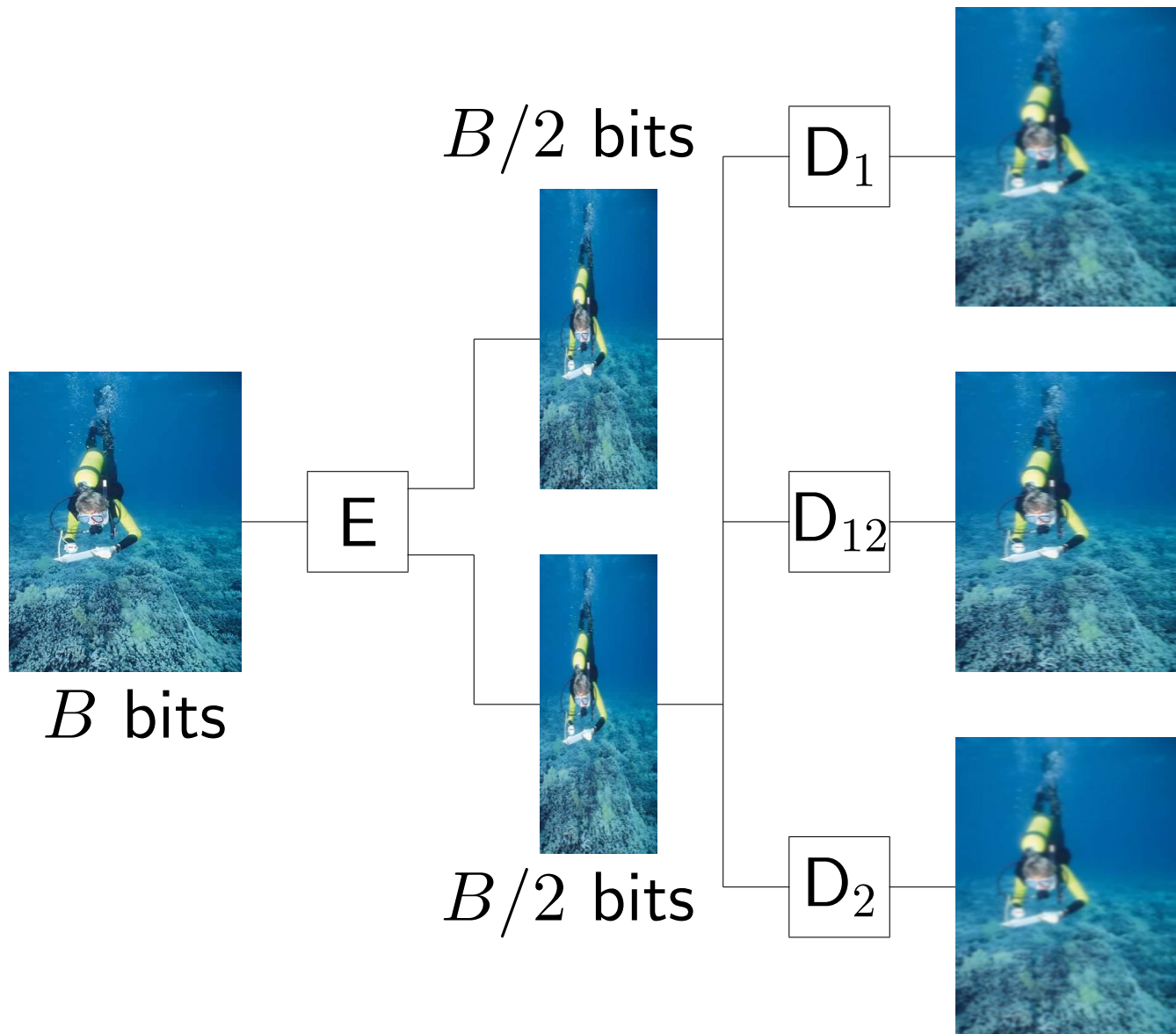
- Reconstruct using available packets
 - Requires adding **redundancy** to packets, i.e., **coding** across packets.

This approach is called **Multiple Description Coding**.

Transmitting Independent Packets



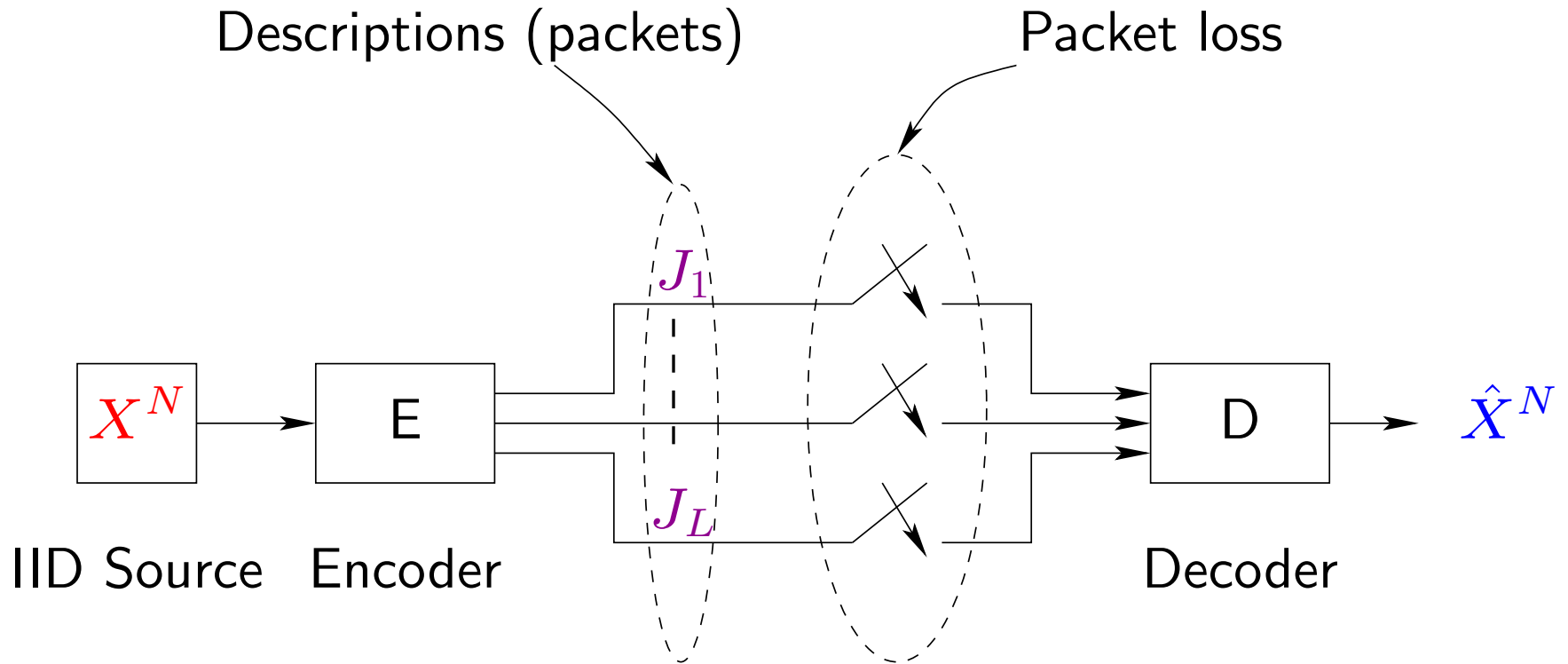
Transmitting Correlated Packets



Multiple Description (MD) Coding vs. Compression

- **Compression** (no coding across packets):
 - **Best reconstruction** when all packets are received.
 - Sharp degradation as packet losses increase.
- **Multiple Description Coding:**
 - **Graceful degradation** as packet losses increase.
 - Suboptimal when all packets are received.

Multiple Description Coding



Source: Length N vector X^N of i.i.d. random variables.

Encoder: $X^N \rightarrow \{J_1, \dots, J_L\}$ which are the L “descriptions” of X^N at rates R_1, \dots, R_L per source symbol.

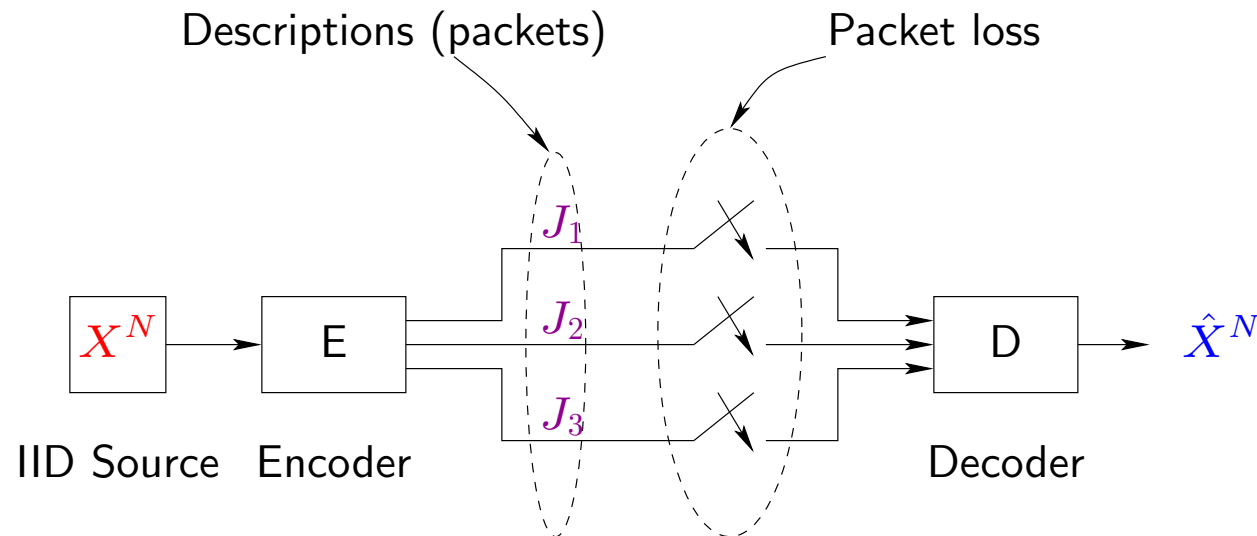
Descriptions:

$$J_l = f_l(X^N), \quad H(J_l) \leq NR_l, \quad l = 1, \dots, L.$$

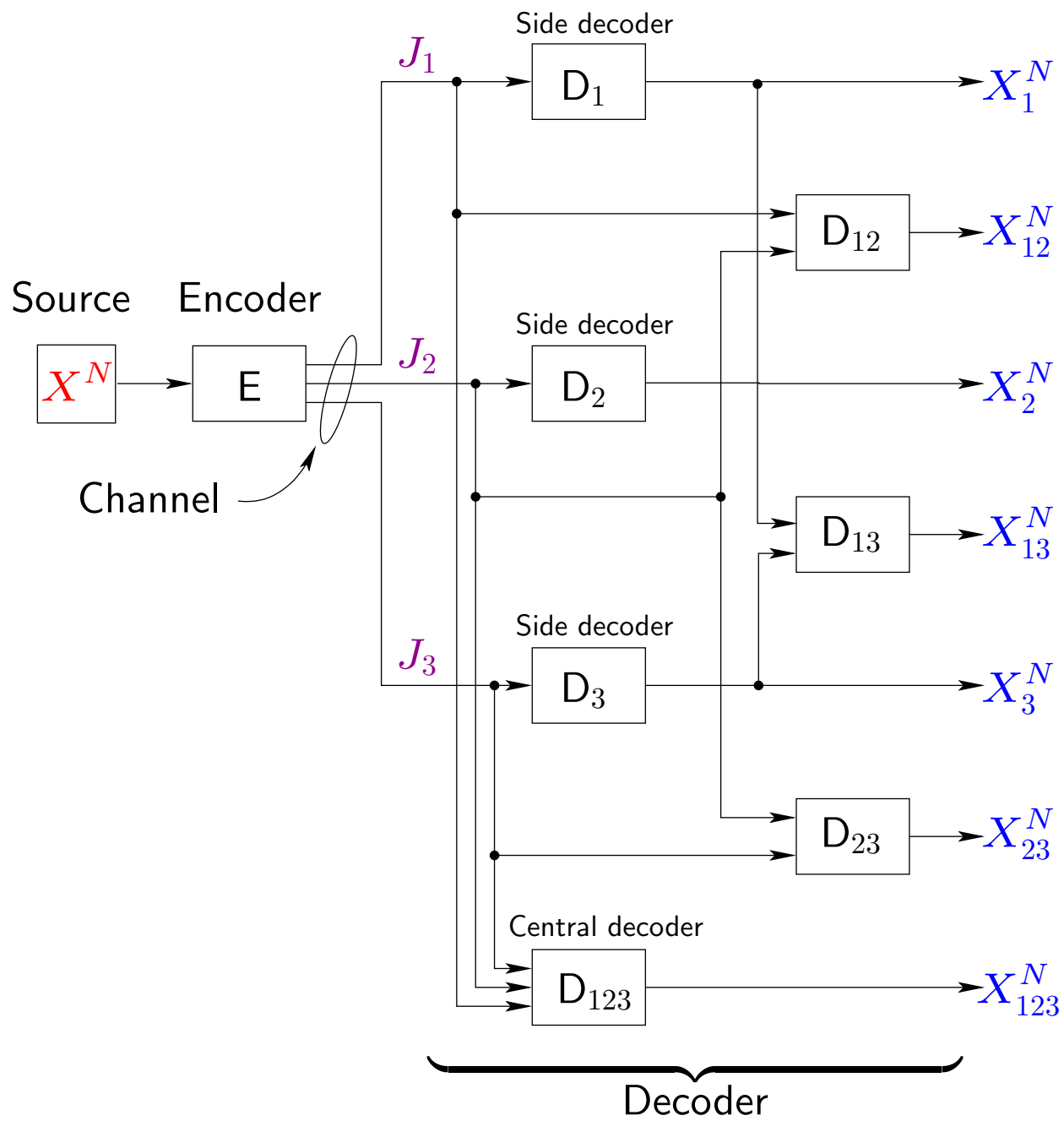
Decoder: Consists of $2^L - 1$ sub-decoders: one for each non-empty subset of the available descriptions.

Decoder Outputs: $X_{\mathcal{S}}^N = g_{\mathcal{S}}(\{J_l : l \in \mathcal{S}\})$ where $\mathcal{S} \subseteq \{1, \dots, L\}$, $\mathcal{S} \neq \emptyset$.

Example: Coding with 3 Descriptions



- The source is i.i.d. vector X^N (length N).
- The encoder outputs L descriptions J_1, J_2, J_3 of X^N .
- The decoder produces an output \hat{X}^N from the **available** descriptions.



Problem Statement

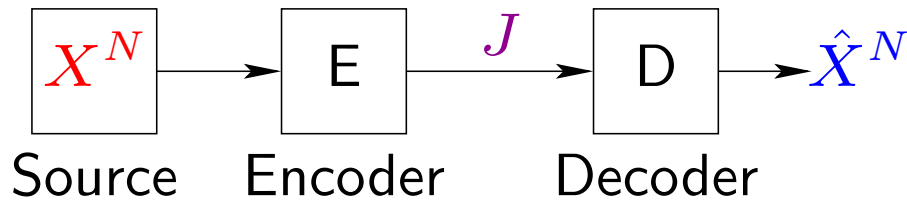
- **Problem:** What is the Rate-Distortion (R-D) region?
- The **rates** (L parameters) are R_1, \dots, R_L .
- The **distortions** ($(2^L - 1)$ parameters) are

$$D_{\mathcal{S}} = \frac{1}{N} \mathbb{E} d(\mathbf{X}^N, \mathbf{X}_{\mathcal{S}}^N), \quad \mathcal{S} \subseteq \{1, \dots, L\}, \mathcal{S} \neq \emptyset$$

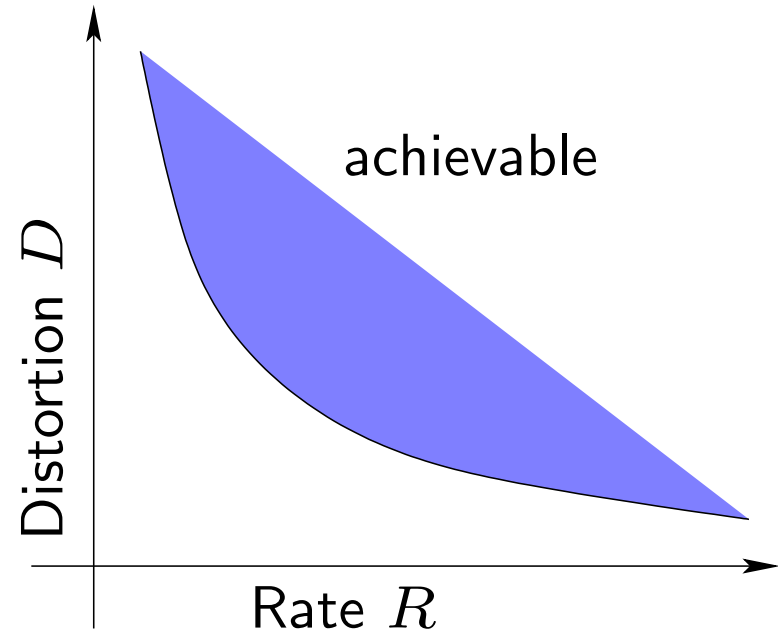
- The R-D region is $(L + 2^L - 1)$ -dimensional.

- **Why do we need the R-D region?**
 - Provides complete knowledge about tradeoff between packet redundancy and fidelity of reconstruction.
 - Helps design good MD coders.
- For $L = 1$, it is Shannon's R-D region.

Rate-Distortion Theory



$$\frac{1}{N} H(J) \leq R$$
$$\frac{1}{N} d(X^N, \hat{X}^N) \leq D$$



The R-D region is the set of achievable (R, D) as $N \rightarrow \infty$.

Theorem 1. [Shannon] *The R-D region is the convex set $R \geq R(D)$ where*

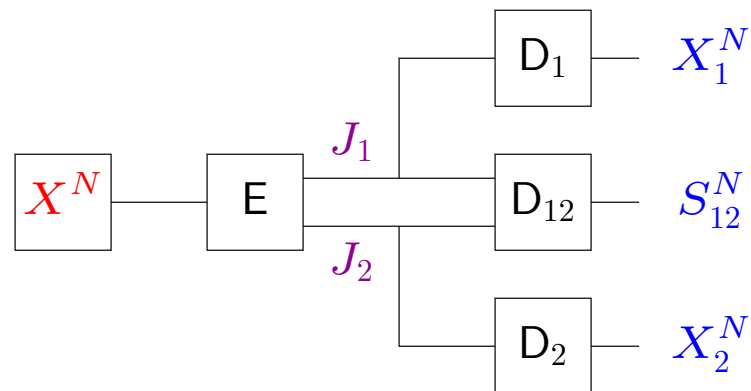
$$R(D) = \min_{\hat{X}} I(\mathbf{X}; \hat{X}) \quad \text{s.t.} \quad \mathbb{E}d(\mathbf{X}, \hat{X}) \leq D$$

minimized over all \hat{X} jointly distributed with \mathbf{X} .

For the **Gaussian Source**: $\mathbf{X} \sim N(0, 1)$, the **rate-distortion** function is

$$R(D) = \frac{1}{2} \log \left(\frac{1}{D} \right)$$

Two-Description Coding



The rates and distortions are:

$$R_1 = \frac{1}{N}H(J_1)$$

$$R_2 = \frac{1}{N}H(J_2)$$

$$D_1 = \frac{1}{N}Ed(X^N, X_1^N)$$

$$D_2 = \frac{1}{N}Ed(X^N, X_2^N)$$

$$D_{12} = \frac{1}{N}Ed(X^N, X_{12}^N)$$

Two-Description Coding

El Gamal and Cover (1982) found an **achievable region** for $L = 2$:

$$R_1 \geq I(\mathbf{X}; X_1)$$

$$R_2 \geq I(\mathbf{X}; X_2)$$

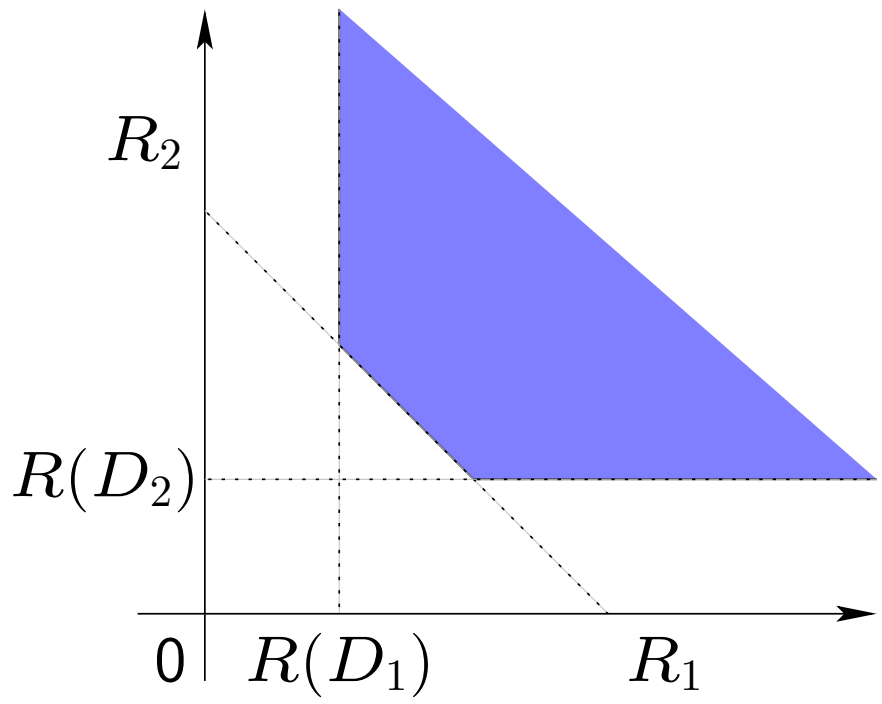
$$R_1 + R_2 \geq I(\mathbf{X}; X_1 X_2 X_{12}) + I(X_1; X_2)$$

$$D_S \geq \text{Ed}(\mathbf{X}, X_S), \quad S = 1, 2, 12$$

where X_1, X_2, X_{12} are any r.v.'s jointly distributed with the source \mathbf{X} .

- The convex hull of this region is achievable by time-sharing.
- This is *an* achievable region, **not necessarily *the* R-D region.**
- **Ozarow (1980)** computed the R-D region for the Gaussian Source.

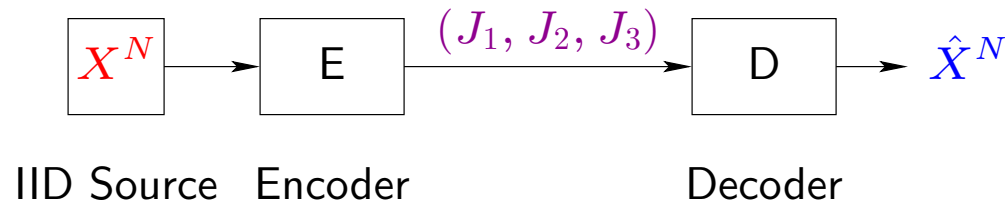
The Gaussian Source



The Gaussian rate region

- **Ahlsvede (1987)** showed tightness of the inner bound by El Gamal and Cover in the “no excess rate” case ($R_1 + R_2 = R(D_{12})$) for the joint description output: X_{12} .
- **Zhang and Berger (1987)** provided a stronger achievable result than El Gamal and Cover for $L = 2$. For the **binary symmetric source** with Hamming distortion measure, their result provides a **strict improvement**.

Progressive Data Transmission



The decoding is **progressive**:

$$(J_1) \rightarrow D_1 \geq D(R_1)$$

$$(J_1, J_2) \rightarrow D_{12} \geq D(R_1 + R_2)$$

$$(J_1, J_2, J_3) \rightarrow D_{123} \geq D(R_1 + R_2 + R_3)$$

- **Question:** Can we attain all the minimum distortions simultaneously?
- **Answer:** Not always. X is **successively refinable** if the Shannon bounds are achievable with equality:

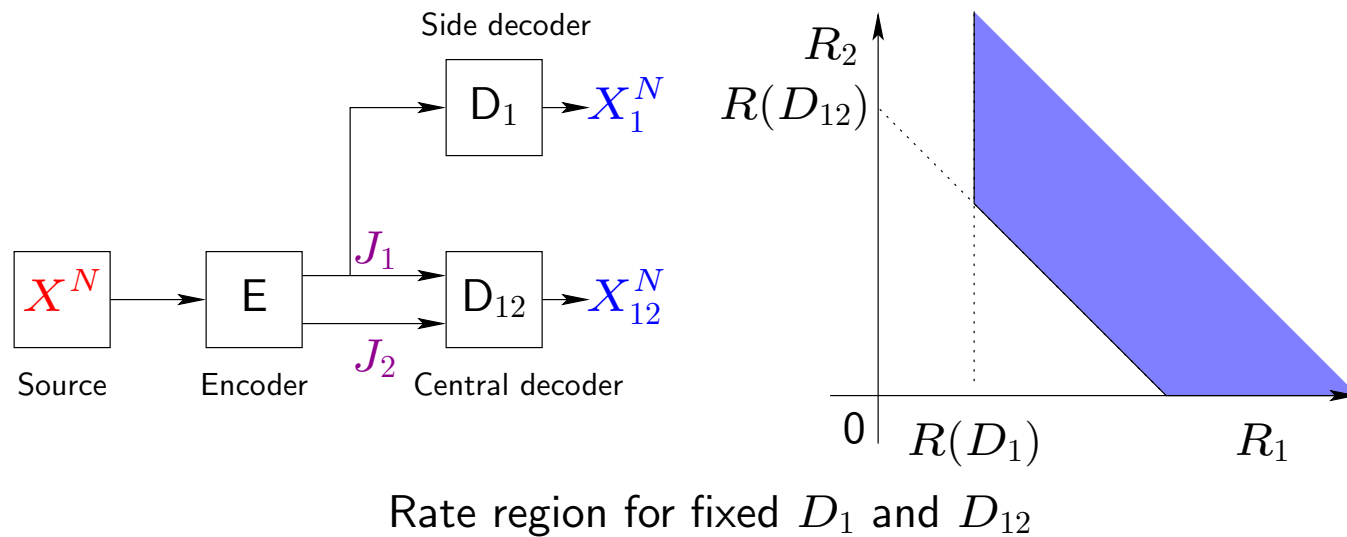
$$D_1 = D(R_1)$$

$$D_{12} = D(R_1 + R_2)$$

$$D_{123} = D(R_1 + R_2 + R_3)$$

Successive Refinement

- The Gaussian source is successively refinable:



R-D region: $R_1 \geq R(D_1)$, $R_1 + R_2 \geq R(D_{12})$.

- This structure is a **special case of MD coding** with X_2 absent.

Summary of New Results

- Achievable region for MD coding of an arbitrary source ($L \geq 2$).
- R-D region bounds for the Gaussian source.
- Special cases of Gaussian MD coding.
- Examples.

Achievable Region for $L \geq 2$

Theorem 2. *The R-D region contains the rates and distortions satisfying*

$$\begin{aligned} \sum_{l \in \mathcal{S}} R_l &\geq (|\mathcal{S}| - 1)I(\mathbf{X}; \mathbf{X}_\emptyset) - H(\mathbf{X}_\mathcal{U} : \mathcal{U} \in 2^{\mathcal{S}} | \mathbf{X}) \\ &\quad + \sum_{\mathcal{T} \subseteq \mathcal{S}} H(\mathbf{X}_\mathcal{T} | \mathbf{X}_\mathcal{U} : \mathcal{U} \in 2^{\mathcal{T}} - \mathcal{T}) \\ D_{\mathcal{S}} &\geq \text{Ed}_{\mathcal{S}}(\mathbf{X}, \mathbf{X}_{\mathcal{S}}) \end{aligned}$$

for every $\emptyset \neq \mathcal{S} \subseteq \mathcal{L} = \{1, \dots, L\}$ and some joint distribution between outputs $\{\mathbf{X}_{\mathcal{S}}\}$ and the source \mathbf{X} .

- Best region is obtained by **optimizing over the joint p.d.f** of $(X, \{X_S : S \in 2^{\mathcal{L}}\})$.
- For a given p.d.f, the rate-region is a polyhedron:
 - Every constraint is active (polymatroid structure).
- Region is an inner bound on the R-D region.
- Proof uses random coding.
- For $L = 2$, this produces the results of **El Gamal and Cover**, and of **Zhang and Berger**.

Outer Bound on the R-D Region

- Outer bounds: difficult to find for **arbitrary sources**.
 - Proof depends strongly on properties of source.
- We focus on
 - **Gaussian source**: $X \sim N(0, 1)$.
 - **Squared-error distortion**: $d(x, y) = |x - y|^2$.

- An **outer bound** on the R-D region:

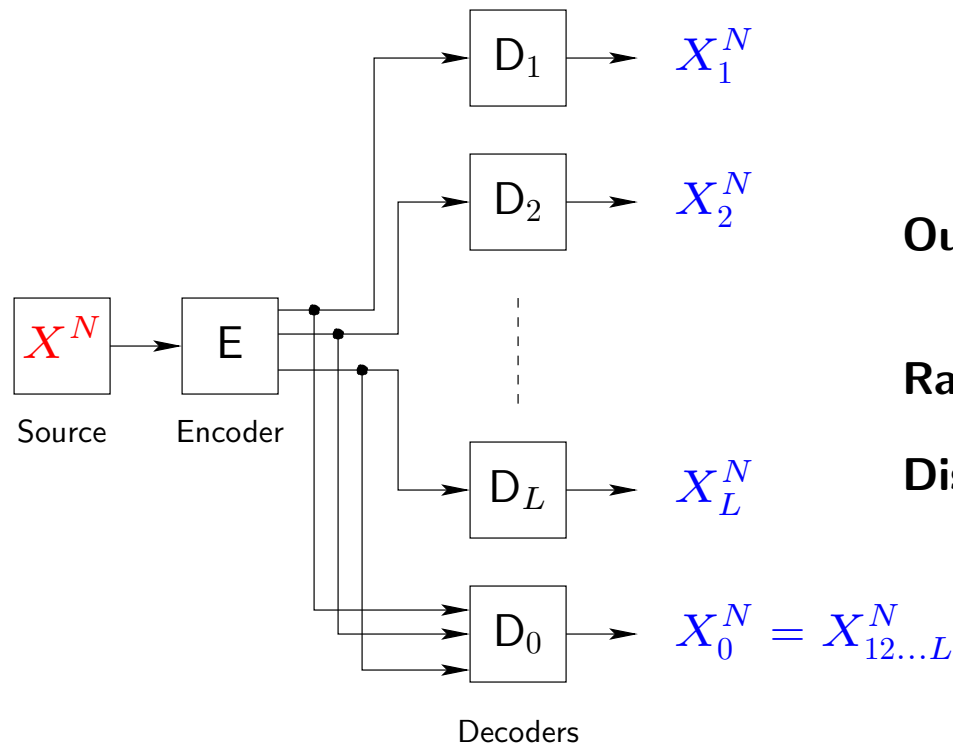
Theorem 3. *The R-D region is contained in*

$$\sum_{k \in \mathcal{K}} R_k \geq -\frac{1}{2} \log \left[\min_{\{\mathcal{K}_m\}_{m=1}^M} \inf_{\lambda \geq 0} \left(D_{\mathcal{K}} \frac{\prod_{m=1}^M (D_{\mathcal{K}_m} + \lambda)}{(D_{\mathcal{K}} + \lambda)(1 + \lambda)^{M-1}} \right) \right], \quad \forall \mathcal{K} \in 2^{\mathcal{L}}$$

minimized over all partitions $\{\mathcal{K}_m\}$ of \mathcal{K} .

- The outer bound is a polyhedron.
- We also have a stronger outer bound.

L Channels and $L + 1$ Decoders



Keep only **side** and **central** decoders. Ignore all other decoders.

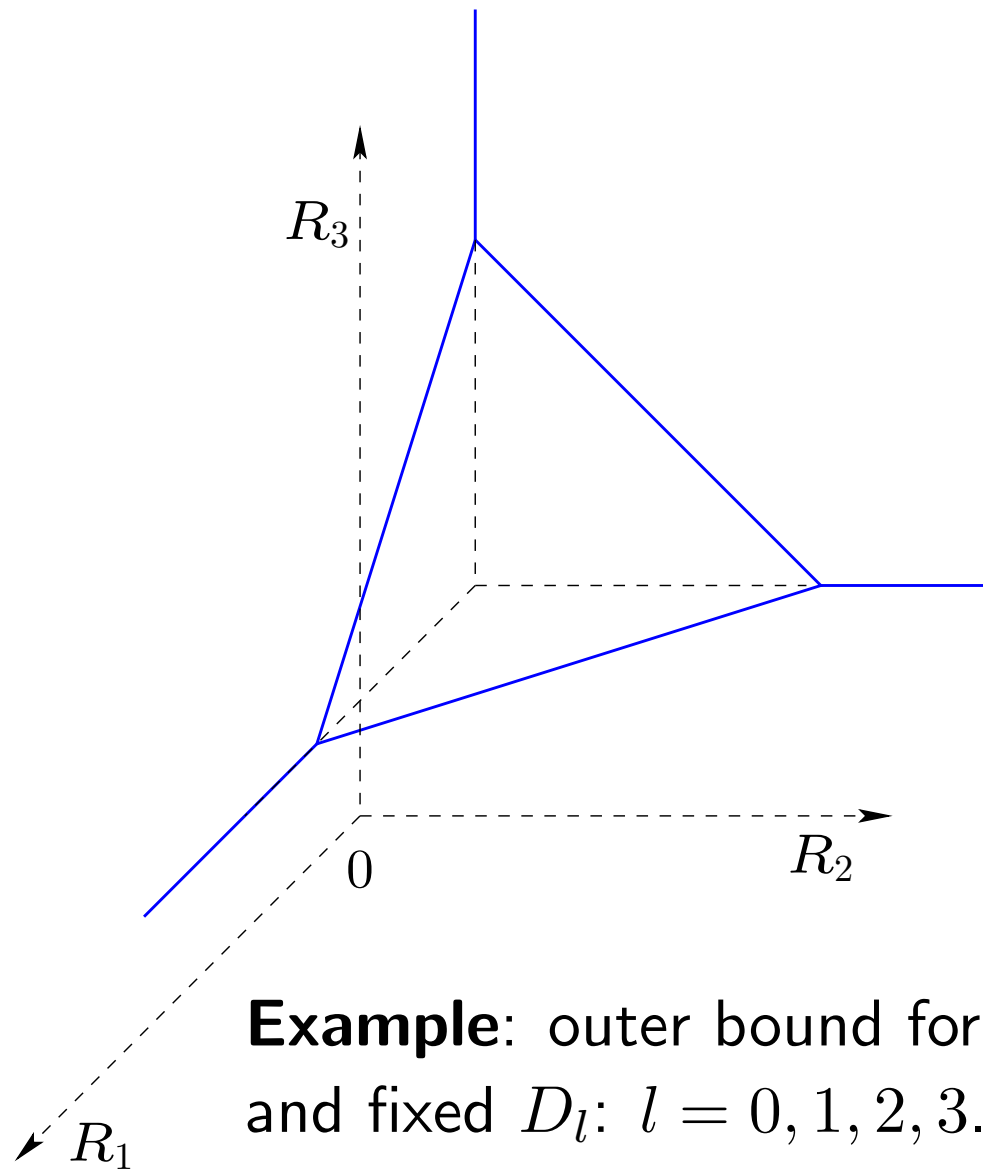
Bounds on the R-D region

- **Outer Bound:** For the L -channel $(L + 1)$ -receiver problem, the *outer bound* reduces to

$$R_l \geq -\frac{1}{2} \log D_l, \quad (l = 1, \dots, L)$$

$$R_1 + \dots + R_L \geq -\frac{1}{2} \log \varphi_L(D_0, \dots, D_L)$$

$$\varphi_L(D_0, \dots, D_L) = \inf_{\lambda \geq 0} \left[D_0 \frac{\prod (D_l + \lambda)}{(D_0 + \lambda)(1 + \lambda)^{L-1}} \right]$$



Inner Bound

Pick a **joint Gaussian p.d.f.** for $(\mathbf{X}, \{X_{\mathcal{S}} : \mathcal{S} \in 2^{\mathcal{L}}\})$:

Let (W_1, \dots, W_L) be jointly Gaussian with covariance matrix \mathbf{K} .

$$X_l = \mathbb{E}(\mathbf{X} | \mathbf{X} + W_l) \equiv \alpha_l(\mathbf{X} + W_l), \quad (l = 1, \dots, L)$$

$$X_0 = \mathbb{E}(\mathbf{X} | X_1, \dots, X_L) \equiv \sum_{l=1}^L \beta_l X_l$$

for appropriate α_l, β_l .

- For given distortions D_1, \dots, D_L and D_0 , we optimize over \mathbf{K} to find the best rate region.
- The optimal solution takes the form:

$$\mathbf{K} = \begin{pmatrix} \sigma_1^2 & u & u & \dots & u \\ u & \sigma_2^2 & u & \dots & u \\ u & u & \sigma_3^2 & \dots & u \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ u & u & u & \dots & \sigma_L^2 \end{pmatrix}$$

Tightness of Bounds

The inner and outer bounds meet for some rates and distortions.

Theorem 4. *For any $L \geq 2$ and distortions D_0, D_1, \dots, D_L , the best achievable sum rate meets the outer bound:*

$$\sum_{l=1}^L R_l = 1 \frac{1}{2} \log \varphi_L(D_0, \dots, D_L).$$

Example: 3-Channel 4-Decoder Problem

Take $L = 3$, $D_1 = D_2 = D_3 = 1/2$ and $D_0 = 1/16$.

Outer Bound:

$$R_l \geq 0.5, \quad l = 1, 2, 3$$

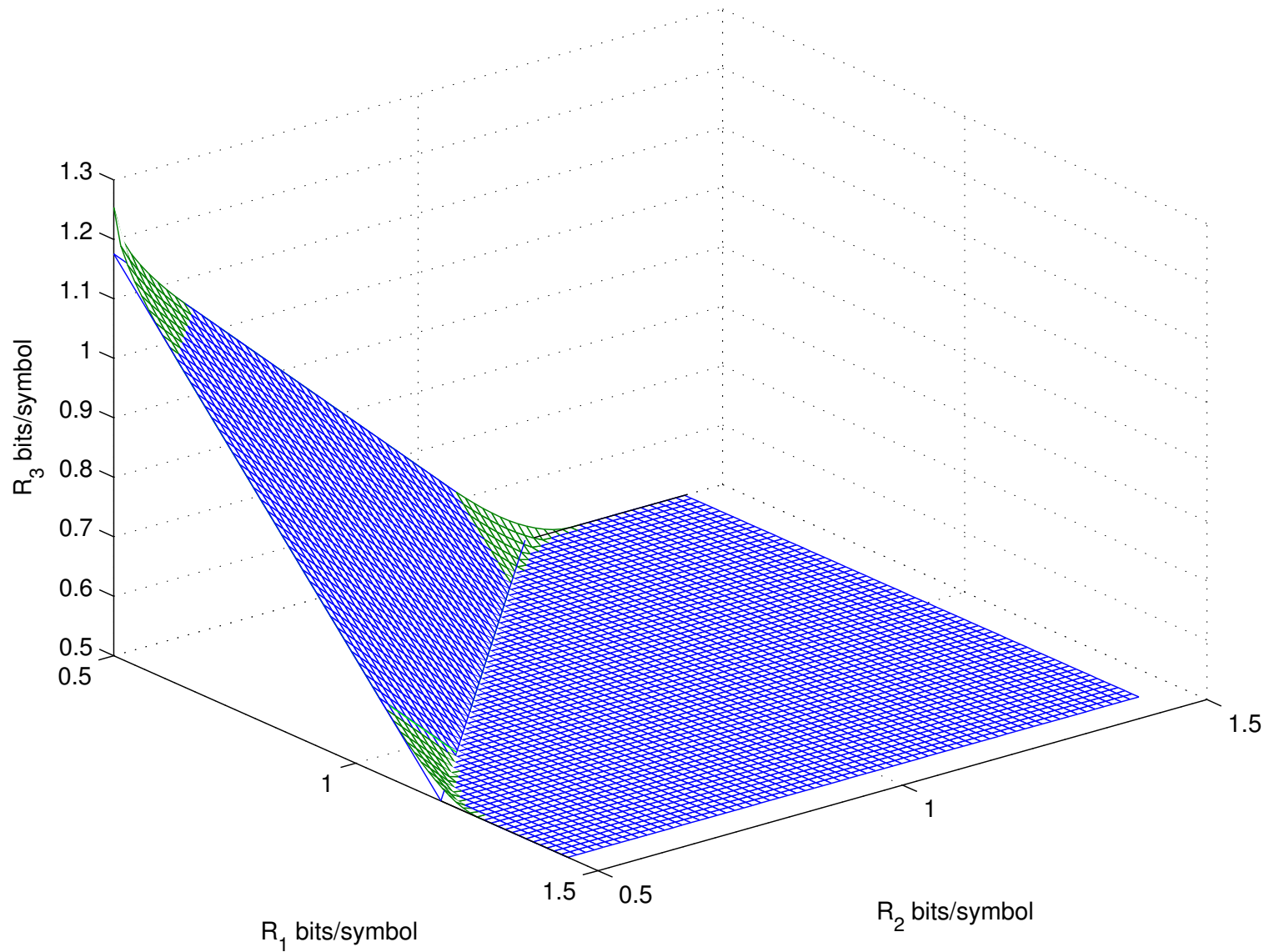
$$R_1 + R_2 + R_3 \geq 2.1755$$

Achievable Rates:

$$R_l \geq 0.5, \quad l = 1, 2, 3$$

$$R_1 + R_2 + R_3 = 2.1755$$

$$R_l + R_m \geq 1.1258, \quad l < m$$



Blue Region: Inner and outer bounds meet on a hexagon.

Green Region: Inner bound (does not meet outer bound).

Theorem 5. *Suppose*

$$D_L + \left(1 - L + \sum_{l=1}^{L-1} D_l^{-1}\right)^{-1} \geq 1 + D_0.$$

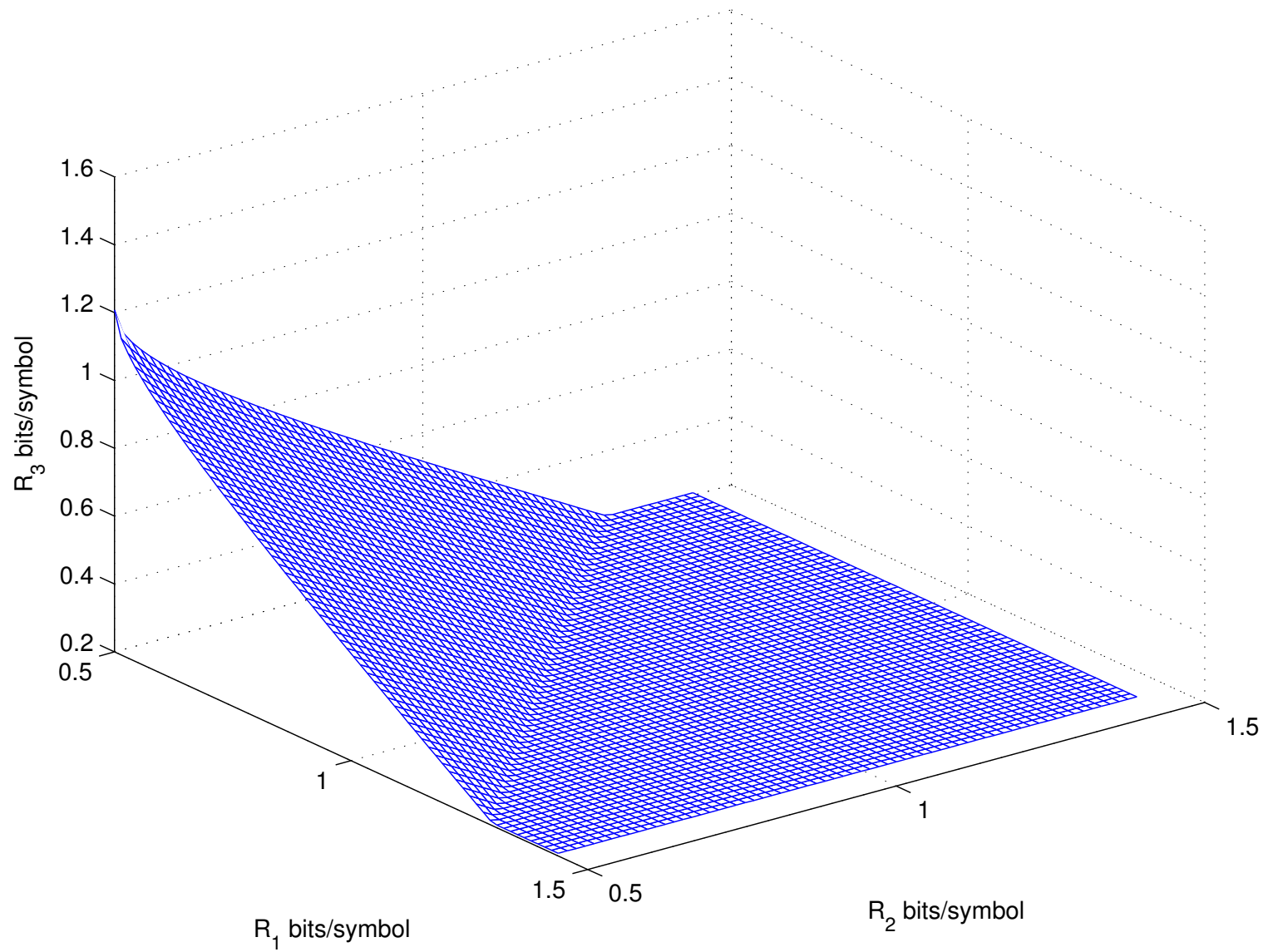
then the outer and inner bounds meet everywhere.

Example: $L = 3$, $D_1 = D_2 = 1/2$, $D_3 = 3/4$, and $D_0 = 1/16$:

The above inequality

$$D_3 + \frac{D_1 D_2}{D_1 + D_2 - D_1 D_2} \geq 1 + D_0$$

is satisfied.



The R-D Region

Successive Refinement on Chains

$$\begin{array}{c} X \xrightarrow{R_1} X_1 \xrightarrow{R_2} X_{12} \xrightarrow{R_3} X_{123} \dots \\ \text{Gaussian} \end{array}$$

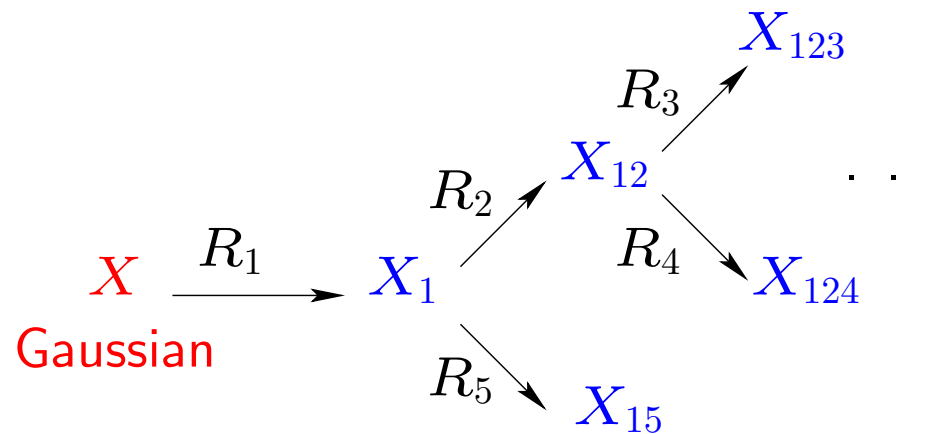
$$D_1 \geq D(R_1)$$

$$D_{12} \geq D(R_1 + R_2)$$

$$D_{123} \geq D(R_1 + R_2 + R_3)$$

Proved using the achievability result.

Successive Refinement on Trees



$$D_1 \geq D(R_1)$$

$$D_{12} \geq D(R_1 + R_2)$$

$$D_{123} \geq D(R_1 + R_2 + R_3)$$

$$D_{124} \geq D(R_1 + R_2 + R_4)$$

$$D_{15} \geq D(R_1 + R_5)$$

Summary

- **Multiple Description Coding:**
 - Discussed applications to real-time transmission over underwater acoustic networks.
 - Reviewed known MD coding results.
- **Our new results:**
 - Achievable rate region for MDC with *many* channels.
 - Outer bound on the R-D region for the Gaussian source.
 - Demonstrated tightness of bounds.