



SLEPIAN WOLF ENCODING

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DISTRIBUTED SOURCE CODING

- An important problem in information theory and communication.
- DSC problems regard the compression of multiple correlated information sources that do not communicate with each other.
- By modeling the correlation between multiple sources at the decoder side together with channel codes, DSC is able to shift computation complexity from encoder side to the decoder side.
- Provides appropriate frameworks for video/multimedia compression and sensor networks.

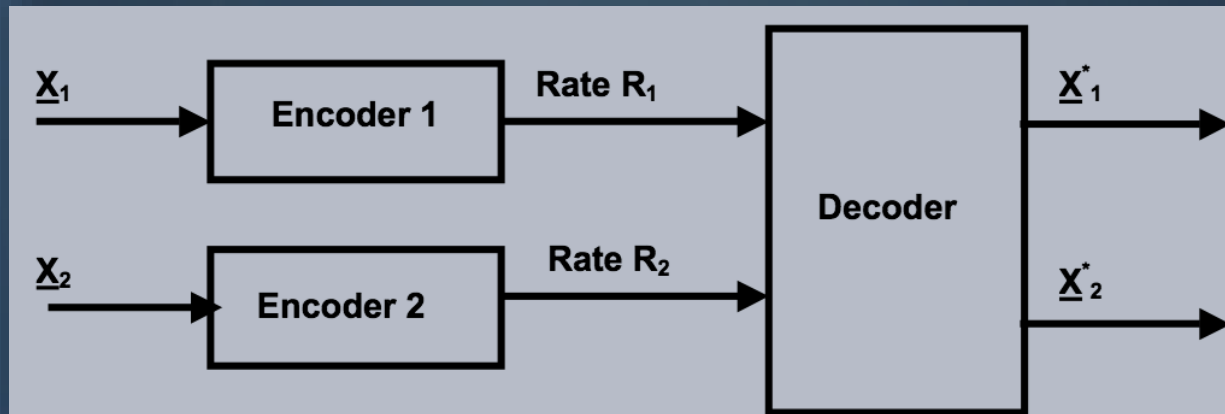
SLEPIAN-WOLF BOUND

- Suppose Alice has X , and Bob has Y .
- Alice wants to compress data using entropy $H(X)$
- Bob wants to compress data using entropy $H(Y)$
- They use one decoder and the rate of compression is $R(X)+R(Y) = H(X)+H(Y)$.
- But, Slepian Wolf encoding states that even if there is no link between Alice and Bob, you can do with $H(X,Y)$ compression rate, knowing the correlation between X and Y .

AN EXAMPLE

- Suppose, weather in 2 neighboring towns-Janestown and Thomasville, is correlated.
- Each town has a weather “good” or “bad” with equal probabilities, and the probability that the 2 towns have different weather on a given day is p .
- Jane, a resident in Janestown wants to send last year’s weather to Tom in Thomasville.
 - If Jane knows the weather in Thomasville, she can just send the difference string, which will have $365p$ bits as “1” and $365(1-p)$ as a “0”. The difference string can be compressed to $365h_2(p)$ length h_2 being the entropy of difference string.
 - Even if Jane is unaware of the weather in Thomasville, she can send the same length signal.
This follows from **Slepian Wolf encoding**.

FORMULATION



Let X and Y be two discrete sources, with probabilities $p(x)$ and $p(y)$, which are formed from N -independent drawings from a joint distribution $p(x,y)$.

An encoding function f_1 is a map from X^n (n length drawing from X) to 2^{nR_1} of a codeword,

$$f_1: X^n \rightarrow \{1,2,\dots,2^{nR_1}\}$$

Similarly for Y ,

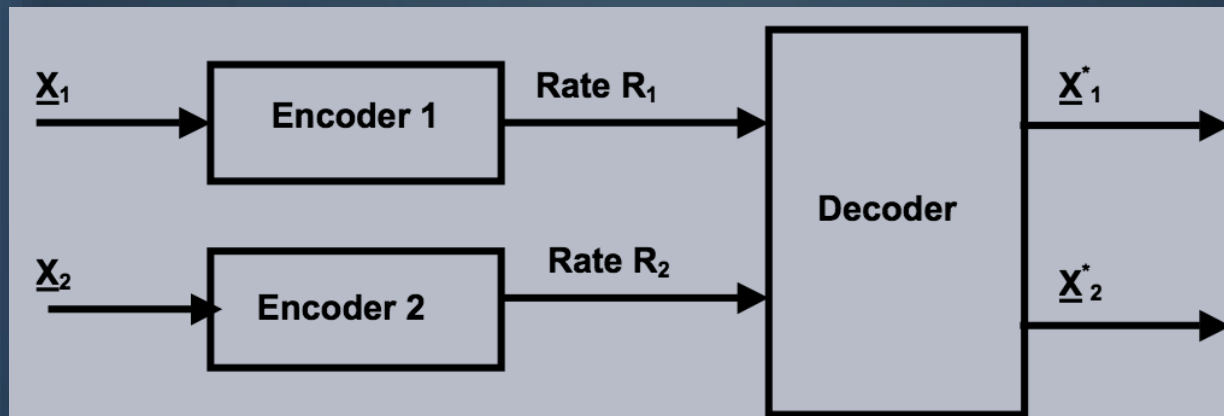
$$f_2: Y^n \rightarrow \{1,2,\dots,2^{nR_2}\}$$

Then there is a single decoding function g , which is:

$$g: \{1,2,\dots,2^{nR_1}\} \times \{1,2,\dots,2^{nR_2}\} \rightarrow X^n \times Y^n$$

This gives us back the sequences. In this process the error probability $\Pr(g(f_1(x^n), f_2(y^n)) \neq (x^n, y^n)) = P_e^{(n)}$. (R_1, R_2) are achievable if \exists encoding and decoding functions, such that $P_e^{(n)} \rightarrow 0$, and $R_1 \geq H(X)$ and $R_2 \geq H(Y)$.

ADMISSIBLE RATE PAIRS



The systems of interest are those for which the probability that \underline{X}_1^* doesn't equal \underline{X}_1 and \underline{X}_2^* doesn't equal \underline{X}_2 can be made as small as possible by choosing sufficiently large n .

- ❖ Such systems are said to be *ADMISSIBLE SYSTEMS*.
- ❖ The rate pair (R_1, R_2) for an admissible system is called an *ADMISSIBLE RATE PAIR*.
- ❖ The closure of all such admissible pairs is called the *ADMISSIBLE RATE REGION*.

SLEPIAN WOLF THEOREM

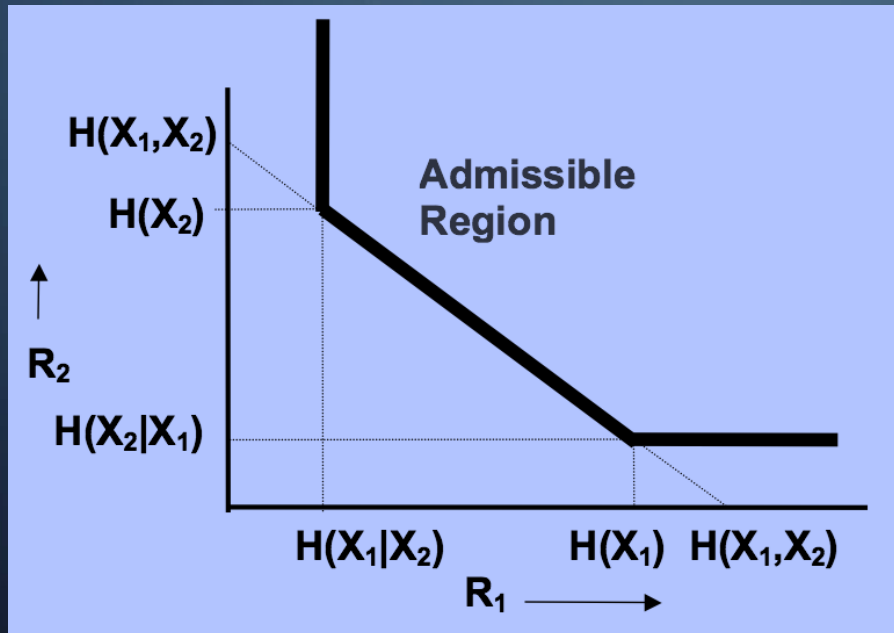
states that:

The admissible rate region for the pair of rates (R_1, R_2) is the set of points that satisfy the three inequalities:

$$R_1 \geq H(X_1 | X_2)$$

$$R_2 \geq H(X_2 | X_1)$$

$$R_1 + R_2 \geq H(X_1, X_2)$$



SIGNIFICANCE OF THE THEOREM

- The significance of the theorem is seen by comparing it to the entropy bound for the compression rates of single sources.
- Separate encoders which ignore the source correlation can achieve rates only upto:

$$R_1 + R_2 \geq H(X_1) + H(X_2).$$

- However for Slepian-Wolf coding the encoders exploit the knowledge of correlation to achieve the same rates as an optimal joint encoder, namely

$$R_1 + R_2 \geq H(X_1, X_2).$$

Note that: $H(X_1) + H(X_2) \geq H(X_1, X_2)$

PROOF

- Necessity:

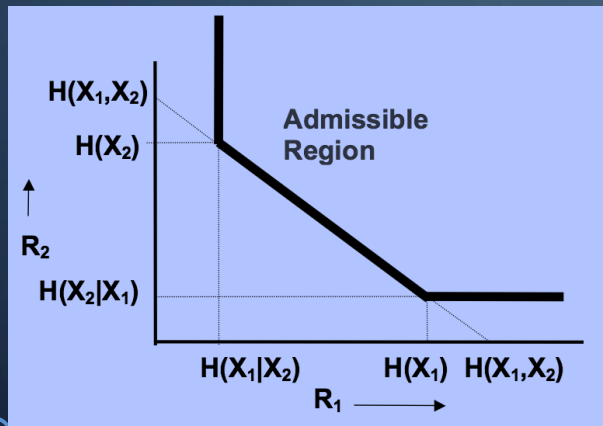
- The necessity of the 3 inequalities follows by considering a modified system in which the pair of source sequences, X_1 and X_2 are input to a single encoder.
 - The output of this single encoder must have a rate at least $H(X_1, X_2)$. Thus, $R_1 + R_2 \geq H(X_1, X_2)$
 - Furthermore if the decoder somehow knows X_2 , the single encoder would require a code whose rate is at least $H(X_1 | X_2)$. Thus, $R_1 \geq H(X_1 | X_2)$
 - By symmetry, $R_2 \geq H(X_2 | X_1)$

PROOF (CONTD.)

SUFFICIENCY:

- Claim: Proving the admissibility at $R_1=H(X_1 | X_2)$, $R_2=H(X_2)$ is established, then we can say that the claimed region is admissible.

- Proof:

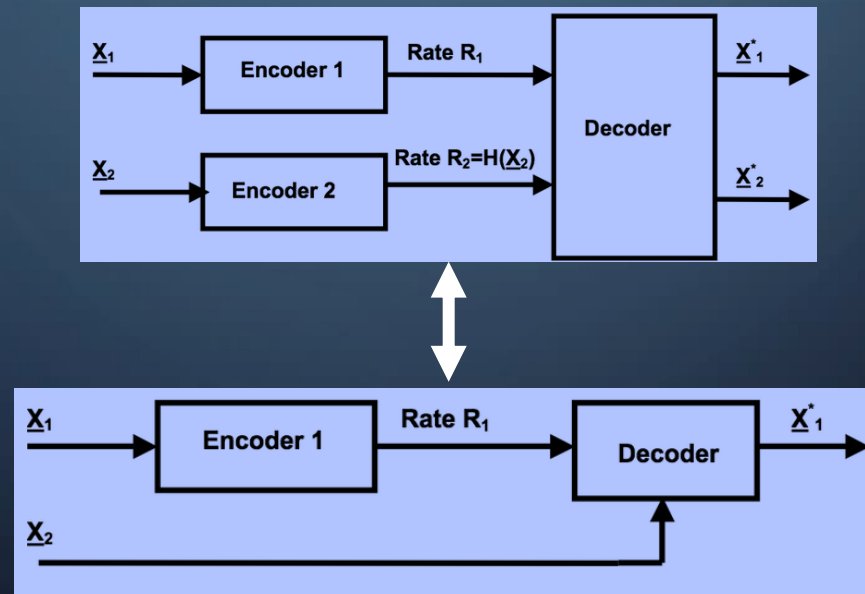


- the admissibility of the rate pair $R_1=H(X_1)$, $R_2=H(X_2 | X_1)$ follows from symmetry;
- the admissibility of all rate pairs on the straight line connecting the rate pair $R_1=H(X_1 | X_2)$, $R_2=H(X_2)$ with the rate pair $R_1=H(X_1)$, $R_2=H(X_2 | X_1)$ follows from time-sharing; and
- the admissibility of all rates within the admissibility region but not on the boundaries follows from wasting bits.

PROOF

- Sufficiency:

- Consider the particular rate pair $R_1 = H(X_1 | X_2)$ and $R_2 = H(X_2)$.
- Note that if $R_1 = H(X_1 | X_2)$, $R_2 = H(X_2)$, then the output of Encoder-2 suffices to reconstruct X_2 . So, the system becomes equivalent to the below system.



PROOF (CONTD.)

- The original construction of an admissible system at the rate point $R_1=H(X_1 | X_2)$, $R_2=H(X_2)$ was found for a particular model of the statistics of the correlated source pair called the **twin binary symmetric source**.
 - A twin binary symmetric source is a memoryless source whose outputs X_1 and X_2 are binary random variables (taking on the values 0 and 1) described by
 - $P(X_1=0)=P(X_1=1)$
 - $P(X_2=0 | X_1=1)=P(X_2=1 | X_1=0)=p$
 - $P(X_2=0 | X_1=0)=P(X_2=1 | X_1=1)=1-p,$ $p, (1-p)$ where p is a parameter satisfying $0 \leq p \leq 1$. Note that $P(X_2=0)=P(X_2=1)=1/2$.
 - Define $h_2(p)=-[p \log_2(p) + (1-p) \log_2(1-p)]$. For the twin binary symmetric source:
 - $H(X_1)=1$
 - $H(X_2)=1$
 - $H(X_2 | X_1)=H(X_1 | X_2)=h_2(p)$
 - $H(X_1, X_2)=1+h_2(p).$

For the twin binary symmetric source, the rate point of interest has $R_1=H(X_1 | X_2)=h_2(p)$ and $R_2=H(X_2)=1$.

PROOF (CONTD.)

- Compressing X_1 :

- The problem of compressing X_1 can be thought of as the problem where we transmit X_1 and X_2 is received at the other end of a Binary Symmetric channel. What we need is a parity check code!
- It is known that a parity check code exists for a BSC with approx. $2^{n(1-h_2(p))}$ codewords such that when X_2 is seen on the channel, we can reliably tell which codeword was actually sent (i.e. X_1).
- But there are 2^n binary strings that can be X_1 .
- So, we do a construction called co-set decomposition.
- Each co-set has a length $2^{n(1-h_2(p))}$, therefore the number of cosets are: $2^n / 2^{n(1-h_2(p))} = 2^{nh_2(p)}$
- Now the encoder of X_1 can send the information about X_1 in $nh_2(p)$ bits.
- Decoder uses this information and the already present information X_2 to construct X_1 reliably.

PROOF (CONCLUSION)

- Applying this construction to the problem at hand, X_1 must be in one of the cosets of the group code.
- If the source encoder transmits to the decoder the identity of the coset containing X_1 , the decoder can determine X_1 from this knowledge and its knowledge of X_2 by using a decoder for the coset code that operated on the "received" word X_2 .
- Since there are $2^{nh_2(p)}$ cosets, the encoder must transmit $nh_2(p)$ binary digits.

The rate of transmission is $h_2(p) = H(X_1 | X_2)$. This establishes the admissibility of the rate point $R_1 = H(X_1 | X_2) = h_2(p)$ and $R_2 = H(X_2) = 1$.

The entire admissible rate region then follows from symmetry, time-sharing and wasting bits.

REMARKS

- The Slepian-Wolf theorem applies to a much wider class of problems than that described here.
- The statistical description of the source sequences can be of a much more general nature,
 - there can be more than two correlated source sequences, and
 - the configuration of the encoders and decoder can be of many different types.