

**ACHILLE
CANNAVALE**

APPUNTI INFORMATION THEORY 2023

**CIAO! QUESTI APPUNTI SONO
FRUTTO DEL MIO STUDIO E
DELLA MIA INTERPRETAZIONE,
QUINDI POTREBBERO
CONTENERE ERRORI, SVISTE O
COSE MIGLIORABILI. BUONO
STUDIO 📖 ✎**

ENTROPY IS NOT NEGATIVE

$$H(x) = \mathbb{E} \left[\log_2 \left(\frac{1}{p(x)} \right) \right] = \sum_{x \in \mathcal{A}_x} p(x) \cdot \log_2 \left(\frac{1}{p(x)} \right)$$

≥ 0 (for $p(x)$)
 ≥ 0 (for \log_2)
 ≥ 0 (for $\frac{1}{p(x)}$)

IMPORTANTE

Ciao! Questi appunti sono frutto del mio studio e della mia interpretazione, quindi potrebbero contenere errori, sviste o cose migliorabili. Buono studio 📖✍️

ENTROPY CHAIN RULE

$$H(x, y) = H(x|y) + H(y)$$

PROOF

$$\begin{aligned} H(x, y) &= \mathbb{E} \left[\log_2 \left(\frac{1}{p(x, y)} \right) \right] = \mathbb{E} \left[\log_2 \left(\frac{1}{p(x|y) \cdot p(y)} \right) \right] = \mathbb{E} \left[\log_2 \left(\frac{1}{p(x|y)} \right) + \log_2 \left(\frac{1}{p(y)} \right) \right] = \\ &= H(x|y) + H(y) \quad \square \end{aligned}$$

FANO'S INEQUALITY

$$H(x|y) \leq H(p_e) + p_e \log_2 (M-1), \quad M = \text{CARD}(\mathcal{A}_x)$$

LET BE:

$$E = \begin{cases} 0 & \text{IF } \hat{x} = x \\ 1 & \text{, OTHERWISE} \end{cases} \quad \text{SO } E \sim \mathcal{B}(p_e)$$

PROOF

$$H(E, x|y) = \begin{cases} H(E|x, y) + H(x|y) \\ H(x|E, y) + H(E|y) \end{cases}$$

\rightarrow IT'S 0 SINCE x KNOWN $\Rightarrow E$ KNOWN
 y KNOWN

$$\Rightarrow H(x|y) = H(x|E, y) + H(E|y) \leq H(E) = H(p_e)$$

$$= p_e H(x|y, E=1) + (1-p_e) H(x|y, E=0)$$

\rightarrow IT'S 0 SINCE $x = \hat{x} = y$

$$\leq \log_2 (M-1) \cdot \text{IT'S THE LARGEST ENTROPY}$$

$$\Rightarrow H(x|y) \leq H(p_e) + p_e \log_2 (M-1) \quad \square$$

DATA PROCESSING INEQUALITY

LET. $x \rightarrow y \rightarrow z$ BE A MARKOV'S CHAIN, x, y, z R.V.s

$$I(x; z) \leq I(x; y)$$

PROOF

$$I(x; y, z) = \begin{cases} I(x; y|z) + I(x; z) \\ I(x; z|y) + I(x; y) \end{cases}$$

0. SINCE x AND $z|y$ ARE INDEPENDENT

$$\Rightarrow I(x; y) = \underbrace{I(x; y|z)}_{\geq 0} + I(x|z) \Rightarrow I(x; z) \leq I(x; y) \quad \square$$

STATIONARY PROCESS

$\{x_i\}_{i \in \mathbb{Z}}$ IS A STATIONARY PROCESS IFF:

$$P((x_m, \dots, x_n) \in A) = P((x_{m-\Delta}, \dots, x_{n-\Delta}) \in A)$$

$$\forall m \in \mathbb{Z}, \forall m \geq n, m \geq \mathbb{Z}, \forall A, \forall \Delta \in \mathbb{N}$$

ENTROPY RATE

$$\text{IF } \exists \lim_{n \rightarrow \infty} \frac{1}{n} H(x_1, \dots, x_n),$$

$$\text{THEN } H_\infty = \lim_{n \rightarrow \infty} \frac{1}{n} H(x_1, \dots, x_n) \left[\frac{\text{bits}}{\text{symbol}} \right]$$

KRAFT'S INEQUALITY

1. C. IS AN INSTANTANEOUS D -ARY SOURCE CODE WITH CODEWORDS LENGTH $\{l_i\}_{i=1}^m$, $m < \infty$

$$\Rightarrow \sum_{i=1}^m D^{-l_i} \leq 1$$

2. IF $\{l_i\}_{i=1}^m$, $m < \infty$ ARE SUCH THAT $l_i \in \mathbb{N}$, $\forall i$ AND $\sum_{i=1}^m D^{-l_i} \leq 1$

$\Rightarrow \exists$ A D -ARY INSTANTANEOUS CODEWORD WITH CODEWORD LENGTH: $\{l_i\}_{i=1}^m$

PROOF. 1

Every D-ary code with maximum codeword l_{\max} can be represented with a d-ary tree.

If the code is prefix and a node is used for a codeword, then all the descending nodes cannot be used, otherwise the corresponding codewords would violate the prefix condition.

\Rightarrow A codeword of length l_i eliminates $D^{l_{\max} - l_i}$ nodes at depth l_{\max}

$$\underbrace{\sum_{i=1}^m D^{l_{\max} - l_i}}_{\text{* ELIMINATED TERMINAL NODES}} \leq \underbrace{D^{l_{\max}}}_{\text{* TERMINAL NODES}} \Rightarrow \sum_{i=1}^m D^{-l_i} \leq 1 \quad \square$$

PROOF. 2

IF $l_i \in \mathbb{N} \cdot \Delta$ AND $\sum_{i=1}^m D^{-l_i} \Rightarrow$

1) Build a d-ary tree of depth $l_{\max} = \max\{l_1, \dots, l_m\}$

2) Place a codeword on a node of depth l_i

3) Remove all descending nodes

4) Iterate until there is no l_i available

5) Assign a label to every branch of the tree

6) Assign a codeword to every node by reading the labels from the root to the node



STRONG LAW LARGE NUMBER

$\{x_i\}_{i \in \mathbb{N}}$ i.i.d. r.v.

$$\mathbb{E}[|x_i|] < \infty \quad \forall i \in \mathbb{N}$$

LET $\bar{x}_m = \frac{1}{m} \sum_{i=1}^m x_i$, THEN $\bar{x}_m \xrightarrow{m \rightarrow \infty} \mathbb{E}[x_i]$ ALMOST SURELY AND IN MEAN SQUARE.

PROOF (WEAK LAW): $\text{VAR}(x_i) = \sigma^2 < \infty$, CONV. IN. P

$$\forall \varepsilon > 0 \quad \mathbb{P}(|\bar{x}_m - \mu| > \varepsilon) \xrightarrow{m \rightarrow \infty} 0$$

$$\mathbb{P}(|\bar{x}_m - \mu| > \varepsilon) = \mathbb{P}(|\bar{x}_m - \mu|^2 > \varepsilon^2) \leq \frac{\mathbb{E}[(\bar{x}_m - \mu)^2]}{\varepsilon^2} = \frac{\text{VAR}(\bar{x}_m)}{\varepsilon^2} = *$$

$$\begin{aligned} \mathbb{E}[\bar{x}_m] &= \mathbb{E}\left[\frac{1}{m} \sum_{i=1}^m x_i\right] = \\ &= \frac{1}{m} \sum_{i=1}^m \mathbb{E}[x_i] = \mu \end{aligned}$$

MARKOV'S INEQUALITY

$$\text{VAR}(\bar{x}_m) = \text{VAR}\left(\frac{1}{m} \sum_{i=1}^m x_i\right) = \frac{1}{m} \left(\sum_{i=1}^m \text{VAR}(x_i) + \sum_{i=1}^m \sum_{\substack{j=1 \\ j \neq i}}^m \text{VAR}(x_i, x_j) \right)$$

$$* = \frac{\sigma^2}{m \varepsilon^2} \xrightarrow{m \rightarrow \infty} 0 \quad \square$$

ASYMPTOTIC EQUIPARTITION PROPERTY

$\{x_i\}_{i \in \mathbb{N}}$ i.i.d.

$$H(x_i) = H(x) < \infty$$

$$\frac{1}{m} \log_2 \left(\frac{1}{P(x_1, \dots, x_m)} \right) \xrightarrow{m \rightarrow \infty} H(x) \quad \text{ALMOST SURELY AND IN MEAN SQUARE.}$$

PROOF

$$\frac{1}{m} \log_2 \left(\frac{1}{P(x_1, \dots, x_m)} \right) = \frac{1}{m} \sum_{i=1}^m \log_2 \left(\frac{1}{P(x_i)} \right)$$

$$\text{IF IT SATISFIES STRONG LAW LARGE NUMBER.} \Rightarrow \xrightarrow{m \rightarrow \infty} \mathbb{E}[Y_i] = \mathbb{E}\left[\log_2 \left(\frac{1}{P(x_i)} \right)\right] = H(x)$$

- i.i.d.? YES, CAUSE Y_i IS A FUNCTION OF x_i THAT IS i.i.d.
- $\mathbb{E}[|Y_i|] < \infty$? YES, CAUSE $\log_2(\cdot)$ IS POSITIVE. \square

TYPICAL SET

LET $\{x_i\}_{i \in \mathbb{N}}$ i.i.d, $H(x) < \infty$

$$A_\epsilon^{(m)} = \left\{ \underline{x} = (x_1, \dots, x_m) \in \mathcal{A}_x^m : \left| \frac{1}{m} \log_2 \left(\frac{1}{P(\underline{x})} \right) - H(x) \right| < \epsilon \right\}$$

PROP. 1

$$\mathbb{P} \left((x_1, \dots, x_m) \in A_\epsilon^{(m)} \right) > 1 - \epsilon \quad \forall \epsilon > 0, \text{ FOR } m \cdot \text{SUFF. LARGE.}$$

PROOF. 1

$$\text{SINCE } \mathbb{P} \left((x_1, \dots, x_m) \in A_\epsilon^{(m)} \right) \xrightarrow{m \rightarrow \infty} 1 \Rightarrow \mathbb{P} \left((x_1, \dots, x_m) \in A_\epsilon^{(m)} \right) > 1 - \epsilon$$

$\forall \epsilon > 0, \text{ FOR } m \cdot \text{SUFF. LARGE.}$

PROP. 2

$$\forall \epsilon > 0 \quad \text{CARD} \left(A_\epsilon^{(m)} \right) \leq 2^{m(H(x) + \epsilon)}$$

PROOF. 2

FROM THE DEF. OF TYPICAL SET:

$$-\frac{1}{m} \log_2 (P(\underline{x})) - H(\underline{x}) < \epsilon \Rightarrow P(\underline{x}) \geq 2^{-m(H(x) + \epsilon)}$$

$$-\left(-\frac{1}{m} \log_2 (P(\underline{x})) - H(\underline{x})\right) < \epsilon \Rightarrow P(\underline{x}) \leq 2^{-m(H(x) - \epsilon)}$$

$$\begin{aligned} \text{NOW WE CAN WRITE: } 1 &= \sum_{\underline{x} \in \mathcal{A}_x^m} P(\underline{x}) \geq \sum_{\underline{x} \in A_\epsilon^{(m)}} P(\underline{x}) \geq \sum_{\underline{x} \in A_\epsilon^{(m)}} 2^{-m(H(x) + \epsilon)} = \\ &= \text{CARD} \left(A_\epsilon^{(m)} \right) \cdot 2^{-m(H(x) + \epsilon)} \Rightarrow 1 \geq \text{CARD} \left(A_\epsilon^{(m)} \right) \cdot 2^{-m(H(x) + \epsilon)} \Rightarrow \\ &\Rightarrow 2^{m(H(x) + \epsilon)} \geq \text{CARD} \left(A_\epsilon^{(m)} \right) \end{aligned}$$

PROP. 3

$$\forall \epsilon > 0 \quad \text{CARD} \left(A_\epsilon^{(m)} \right) > (1 - \epsilon) \cdot 2^{m(H(x) - \epsilon)}, \text{ FOR } m \cdot \text{SUFF. LARGE}$$

PROOF. 3

FOR $m \cdot \text{SUFF. LARGE}$ $\mathbb{P}(\underline{x} \in A_\epsilon^{(m)}) > 1 - \epsilon$, SO:

$$\begin{aligned} 1 - \epsilon < \mathbb{P}(\underline{x} \in A_\epsilon^{(m)}) &\stackrel{②}{\leq} \sum_{\underline{x} \in A_\epsilon^{(m)}} 2^{-m(H(x) - \epsilon)} = \text{CARD} \left(A_\epsilon^{(m)} \right) \cdot 2^{-m(H(x) - \epsilon)} \Rightarrow \\ &\Rightarrow \text{CARD} \left(A_\epsilon^{(m)} \right) > (1 - \epsilon) 2^{m(H(x) - \epsilon)}, \quad m \rightarrow \infty \end{aligned}$$

SOURCE CODE

A SOURCE CODE FOR A R.V. X IS AN APPLICATION:

$$C: \mathcal{A}_X \longrightarrow \mathbb{D}^*$$

\mathcal{A}_X ALPHABET OF X

SET OF ALL FINITE LENGTH STRING OF SYMBOLS TAKEN FROM THE D -ARY ALPHABET:

$$\mathbb{D} = \{1, \dots, D-1\}$$

NON-SINGULAR

A CODE IS NON-SINGULAR IF C IS INJECTIVE:

$$\text{IF } \forall x_1 \neq x_2 \Rightarrow C(x_1) \neq C(x_2)$$

EXTENDED CODE

THE EXTENSION C^* OF A CODE C IS THE APPLICATION:

$$C^*: \mathcal{A}_X^* \longrightarrow \mathbb{D}^* = C^*(x_1, \dots, x_m) = C(x_1) C(x_2) \dots C(x_m)$$

SET OF ALL FINITE LENGTH SEQUENCES OF SYMBOLS TAKEN FROM \mathcal{A}_X .

UNIQUELY DECODABLE

A CODE C IS UNIQUELY DECODABLE IF C^* IS NON-SINGULAR.

PREFIX CODE

C IS A PREFIX CODE IF IT SATISFIES THE PREFIX CONDITION:

NO CODEWORD IS PREFIX
OF ANY CODEWORD

PROVE THAT EXPECTED LENGTH OF ANY PREFIX CODE IS ALWAYS GREATER THAN ENTROPY

C-15. A D-ARY PREFIX CODE FOR X R.V.

$$L = \mathbb{E}[l(x)] \geq H(x)$$

= IFF $P(x)$ IS D-ADIC

PROOF

$$L - H(x) = \mathbb{E}[l(x)] - \mathbb{E}\left[\log_D\left(\frac{1}{P(x)}\right)\right] = \mathbb{E}\left[l(x) - \log_D\left(\frac{1}{P(x)}\right)\right] =$$

$$= \mathbb{E}\left[-\log_D(D^{-l(x)}) - \log_D\left(\frac{1}{P(x)}\right)\right] = \mathbb{E}\left[\log_D\left(\frac{P(x)}{D^{-l(x)}}\right)\right] =$$

$$= \mathbb{E}\left[\log_D\left(\frac{P(x)}{\sum_{y \in A_x} D^{-l(y)}}\right)\right] + \mathbb{E}\left[\log_D\left(\frac{1}{\sum_{y \in A_x} D^{-l(y)}}\right)\right] =$$

IT'S JUST A NUMBER

$$= \mathbb{E}_P\left[\log_D\left(\frac{P(x)}{q(x)}\right)\right] + \log_D\left(\frac{1}{\sum_{y \in A_x} D^{-l(y)}}\right) =$$

$$= \underbrace{D_P(P(x) // q(x))}_{\geq 0} + \underbrace{\log_D\left(\frac{1}{\sum_{y \in A_x} D^{-l(y)}}\right)}_{\geq 0}$$

PREFIX \Rightarrow KRAFT $\Rightarrow \leq 1$

$$\Rightarrow L - H(x) \geq 0 \quad \square$$

BOUNDS TO THE AVERAGE LENGTH

C-15. AN OPTIMAL CODE $\Rightarrow H(x) \leq L^* \leq H(x) + 1$

PROOF

LET C BE THE SHANNON CODE:

$$\Rightarrow l(x) = \left\lceil \log_D\left(\frac{1}{P(x)}\right) \right\rceil \quad \text{SINCE } 0 \leq \lceil x \rceil \leq x + 1$$

$$\Rightarrow \log_D\left(\frac{1}{P(x)}\right) \leq l(x) \leq \log_D\left(\frac{1}{P(x)}\right) + 1$$

$$\xrightarrow{\mathbb{E}[\cdot]} H(x) \leq L \leq H(x) + 1 \quad \square$$

CHANNEL CODE

A (M, m) CHANNEL CODE IS THE MAPPING:

$$\underline{x}_m: \{1, \dots, M\} \rightarrow \mathcal{A}_x^m$$

$$\{\underline{x}_m(1), \dots, \underline{x}_m(M)\} = \text{CODEBOOK}$$

└─ CODEWORDS

DECODING RULE

A DECODING RULE IS THE MAPPING:

$$g: \mathcal{A}_y^m \longrightarrow \{1, \dots, M\}$$

$$\hat{w} = g(\underline{y}_m) = \text{ESTIMATE OF } w$$

RATE

THE RATE R OF A (M, m) CHANNEL CODE IS:

$$R = \frac{\log_2(M)}{m} \left[\frac{\text{bits}}{\text{CHANNEL USES}} \right]$$

ACHIEVABLE RATE

THE RATE R IS ACHIEVABLE IF \exists A SEQUENCE OF $(\lceil 2^{mR} \rceil, m)$ CHANNEL CODES:

$$P_n^{\max}(e) \xrightarrow{m \rightarrow \infty} 0$$

CHANNEL CODING THEOREM

1 $R < C \Rightarrow \exists$ A SEQUENCE OF $(\lceil 2^{mR} \rceil, m)$ CHANNEL CODES: $P_m^{\max}(e) \xrightarrow{m \rightarrow \infty} 0$

2 IF FOR A SEQUENCE OF $(\lceil 2^{mR} \rceil, m)$ CHANNEL CODES: $P_m^{\max}(e) \xrightarrow{m \rightarrow \infty} 0 \Rightarrow R \leq C$

PROOF

LET $M = \lceil 2^{mR} \rceil \approx 2^{mR}$, THEN

1) RANDOM CODING

THE (M, m) CHANNEL CODE IS GENERATED AS FOLLOWS:

$$\begin{pmatrix} x_1(1) & \dots & x_m(1) \\ x_1(2) & \dots & x_m(2) \\ \vdots & & \vdots \\ x_1(m) & \dots & x_m(m) \end{pmatrix} = \begin{pmatrix} \underline{x}_m(1) \\ \underline{x}_m(2) \\ \vdots \\ \underline{x}_m(m) \end{pmatrix} = \text{CODEBOOK}$$

THE ENTRIES OF THIS MATRIX ARE i.i.d. R.V. DRAWN FROM THE DISTRIBUTION $P^*(x) = \underset{p(x)}{\text{ARGMAX}} I(x; Y)$

THE CODE IS SHARED WITH THE DESTINATION.

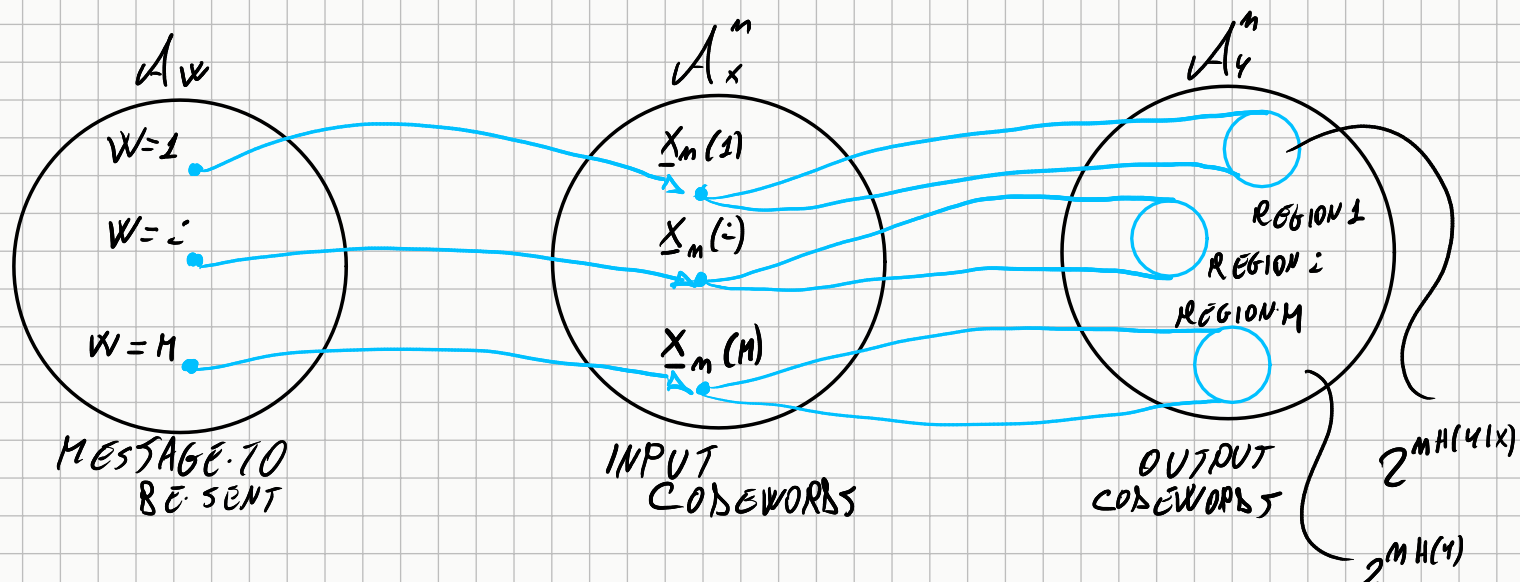
2) JOINTLY TYPICAL DECODING RULE

$\hat{W} = g(\underline{Y}_m) = i \Leftrightarrow \underline{x}_m(i)$ (THE CODEWORD ASSOCIATED TO $W = i$) IS THE ONLY SEQUENCE

JOINTLY TYPICAL WITH \underline{Y}_m , i.e.

$$\begin{cases} (\underline{x}_m(i), \underline{Y}_m) \in A_\epsilon^{(m)} \\ (\underline{x}_m(j), \underline{Y}_m) \notin A_\epsilon^{(m)} \quad \forall j \neq i \end{cases}$$

3) SPHERE



If n is large, the received codewords lie in well defined regions, and, if these regions do not overlap, there is no error in decoding; this is possible only if the rate is sufficiently small.

$$(R \text{ SMALL} \Rightarrow M \text{ SMALL} \Rightarrow \text{FEW REGIONS})$$

BUT: •) THE TOTAL NUMBER OF TYPICAL (RECEIVED) SEQUENCES IS $2^{nH(Y)}$

•) $\forall \underline{x}_m$ TYPICAL, THERE ARE $2^{nH(Y|X)}$ \underline{y}_m TYPICAL.

•) THE NUMBER OF REGIONS IS M .

IN ORDER TO HAVE NON-OVERLAPPING REGIONS, WE MUST HAVE:

$$\underbrace{M}_{\text{\# OF REGIONS}} \underbrace{2^{nH(Y|X)}}_{\text{\# OF CODEWORDS IN EACH REGION}} \leq \underbrace{2^{nH(Y)}}_{\text{TOTAL \# OF CODEWORDS}}$$

SINCE $M = 2^{nR}$

$$2^{nR} 2^{nH(Y|X)} \leq 2^{nH(Y)}$$

$$2^{nR} \leq 2^{n(H(Y) - H(Y|X))} = 2^{nI(X;Y)} = 2^{nC}$$

SINCE $P^*(X) = \underset{P(X)}{\text{ARGMAX}} I(X;Y)$

$$\Rightarrow R \leq C \quad \square$$

CAPACITY OF A GAUSSIAN CHANNEL WITH POWER CONSTRAINT

$$C = \frac{1}{2} \log_2 \left(1 + \frac{P}{N} \right)$$

AND CAN BE ACHIEVED WITH: $x \sim \mathcal{N}(0, P)$

PROOF

$$I(x; y) = h(y) - \underbrace{h(y|x)}_{h(x+z|x) = h(z|x) = h(z)} = \frac{1}{2} \log_2 (2\pi e N)$$

z AND x ARE INDEP.
z ~ N(0, N)

$$= h(y) - \frac{1}{2} \log_2 (2\pi e N)$$

$$\text{VAR}(y) = \text{VAR}(x+z) = \text{VAR}(x) + \text{VAR}(z) = \underbrace{\mathbb{E}[x^2]}_{\leq P} - \underbrace{\left(\mathbb{E}[x] \right)^2}_{\geq 0} + N \leq P + N$$

x, z INDEP

$$\Rightarrow h(y) \leq \frac{1}{2} \log_2 (2\pi e (P+N))$$

"=" IF $y \sim \mathcal{N}(0, (P+N))$
IF $x \sim \mathcal{N}(0, P)$

$$\begin{aligned} \Rightarrow I(x; y) &= h(y) - \frac{1}{2} \log_2 (2\pi e N) \leq \frac{1}{2} \log_2 (2\pi e (P+N)) - \frac{1}{2} \log_2 (2\pi e N) \\ &= \frac{1}{2} \log_2 \left(\frac{P+N}{N} \right) = \frac{1}{2} \log_2 \left(1 + \frac{P}{N} \right) \end{aligned}$$

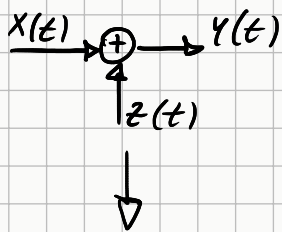
$$\Rightarrow C = \max_{P_x \leq P} \{I(x; y)\} = \frac{1}{2} \log_2 \left(1 + \frac{P}{N} \right) \quad \text{FOR } x \sim \mathcal{N}(0, P) \quad \square$$

BAND-LIMITED GAUSSIAN CHANNEL

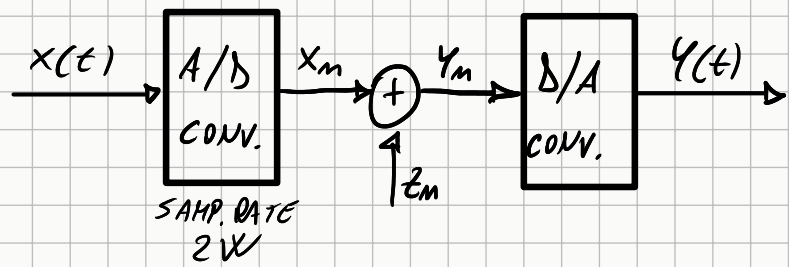
$$C = W \log_2 \left(1 + \frac{P}{N_0 W} \right) \left[\frac{\text{bit}}{\text{s}} \right]$$

PROOF

THE CHANNEL:



HAS THE SAME
CAPACITY
AS



$$C = \frac{1}{2} \log_2 \left(1 + \frac{P}{N_0 W} \right) \left[\frac{\text{bit}}{\text{CH. USE.}} \right]$$

SINCE IT IS USED 2W TIMES PER SECOND
WE HAVE:

$$\left[\frac{\text{CH. USE.}}{\text{s}} \right] \cdot \underbrace{\frac{1}{2} \log_2 \left(1 + \frac{P}{N_0 W} \right)}_{\left[\frac{\text{bit}}{\text{CH. USE.}} \right]} \Rightarrow$$

$$\Rightarrow W \cdot \log_2 \left(1 + \frac{P}{N_0 W} \right) \left[\frac{\text{bit}}{\text{s}} \right] \quad \square$$