

Ex 1.3 What is the minimum value of $H(p_1, \dots, p_n) = H(\mathbf{p})$ as \mathbf{p} ranges over the set of n -dimensional probability vectors? Find all \mathbf{p} 's which achieve this minimum.

$$\{ (p_1, \dots, p_n) \in [0, 1]^n \mid \sum_{i=1}^n p_i = 1, H(\mathbf{p}) \text{ minimal} \}$$

$$= \{ (1, 0, 0, \dots, 0), (0, 1, 0, \dots, 0), \dots, (0, 0, 1, \dots, 1) \}$$

$$H(\mathbf{p}) = - \sum_{i=1}^n p_i \log_2(p_i) \stackrel{\mathbf{p} = (1, 0, \dots, 0)}{=} - (p_1 \log_2(p_1) + \sum_{i=2}^n p_i \log_2(p_i))$$

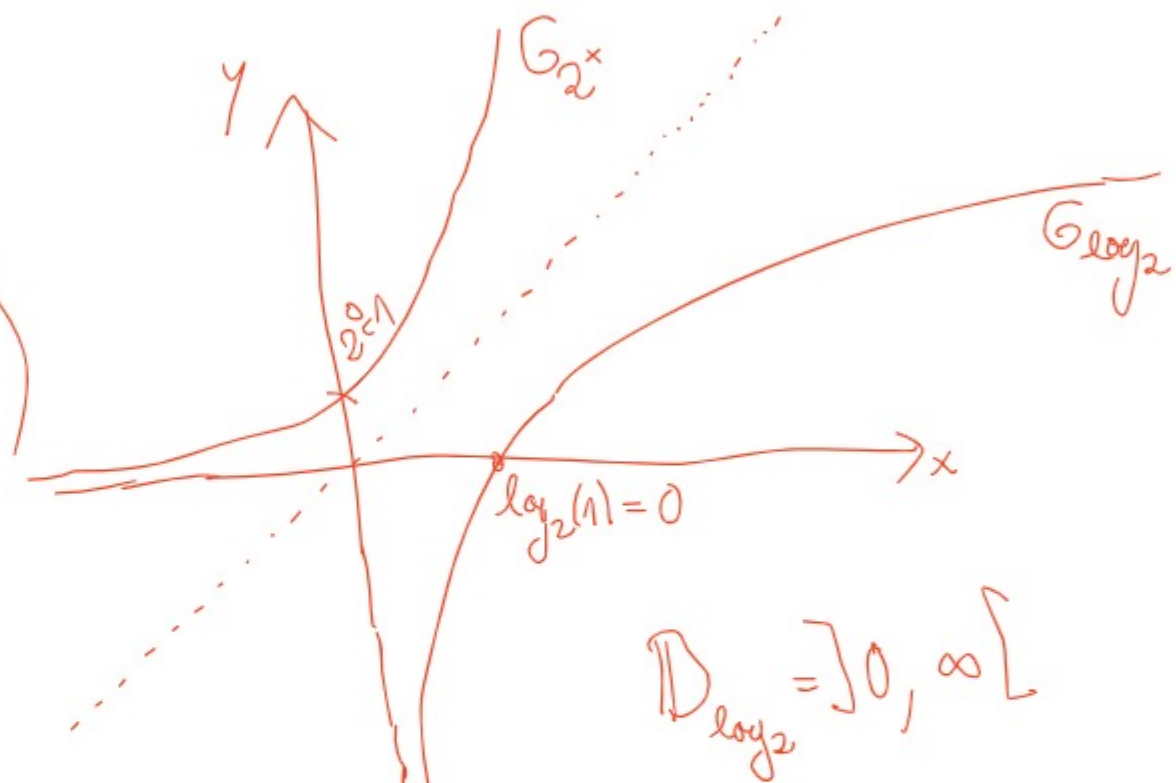
$$= - (1 \cdot \log_2(1) + (n-1) \cdot \lim_{p \rightarrow 0} p \cdot \log_2(p))$$

$$= 0$$

$\log_2(2^x) = x$
 $\log_2(1) = \log_2(2^0) = 0$

$\lim_{x \rightarrow 0} \frac{1}{x^2} = \lim_{x \rightarrow 0} \frac{1}{x} = \infty$

$$\lim_{x \rightarrow 0} x \cdot \log_2(x) = \lim_{x \rightarrow 0} \frac{\log_2(x)}{\frac{1}{x}} = \lim_{x \rightarrow 0} \frac{(\log_2(x))'}{(\frac{1}{x})'} = \lim_{x \rightarrow 0} \frac{\frac{1}{x \cdot \ln(2)}}{(-1) \cdot \frac{1}{x^2}} = \lim_{x \rightarrow 0} \frac{1}{\ln(2)} \cdot \frac{x}{-1} = \lim_{x \rightarrow 0} -\frac{x}{\ln(2)} = 0$$



$$(x^{10})' = 10x^9$$

$$(x^n)' = n \cdot x^{n-1}$$

$$(e^x)' = e^x$$

$$(\ln x)' = \frac{1}{x}$$

$$(a^x)' = a^x \cdot \ln(a)$$

$$(\log_a(x))' = \frac{1}{x \cdot \ln(a)}$$

$$\left(\frac{1}{x}\right)' = (x^{-1})' = (-1) \cdot x^{-2}$$

independent, identically distributed

$$X, Y \text{ independent} \Leftrightarrow \forall A, B: P(X \in A, Y \in B) = P(X \in A) \cdot P(Y \in B)$$

proof+

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$$X, Y \text{ independent} \Leftrightarrow \forall A, B: P(X \in A, Y \in B) = P(X \in A) \cdot P(Y \in B)$$

Ex 1.8 Let X_1, X_2, \dots , be an i.i.d. sequence of discrete random variables with entropy $H(X)$. Let $C_n(t) := \{x^n \in X^n : p(x^n) \geq 2^{-nt}\}$ denote the subset of n -sequences with probabilities $\geq 2^{-nt}$.

a) Show $|C_n(t)| \leq 2^{nt} = 2^{nt}$

b) For what values of t does $P(X^n \in C_n(t)) \rightarrow 1$?

c) Denoted by $p(x)$ the probability mass of the random variable X compute the limit

$$\lim_{n \rightarrow \infty} (p(X_1, X_2, \dots, X_n))^{1/n}$$

$$C_n(t) = \left\{ (x_1, \dots, x_n) \in X^n \mid \underbrace{p(x_1, \dots, x_n)}_{= P(X_1=x_1, \dots, X_n=x_n)} \geq 2^{-nt} \right\}$$

1.8 a) Assume for contradiction $|C_n(t)| > 2^{nt}$.

$$P(C_n(t)) = \sum_{x^n \in C_n(t)} \underbrace{p(x^n)}_{\geq 2^{-nt}} \geq \sum_{x^n \in C_n(t)} 2^{-nt} = |C_n(t)| \cdot 2^{-nt} > 2^{nt} \cdot 2^{-nt} = 2^{nt-nt} = 2^0 = 1$$

$$a^n \cdot a^m = a^{n+m}$$

$\Rightarrow P(C_n(t)) > 1$
Contradiction \downarrow
 $\Rightarrow |C_n(t)| \leq 2^{nt}$

$$\sum_{i=3}^7 5 = |\{3, 4, 5, 6, 7\}| \cdot 5 = 5 \cdot 5 = 25$$

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• Show $|C_n(t)| \leq 2^{nt}$.

• For what values of t does $P(X^n \in C_n(t)) \rightarrow 1$?

• Denoted by $p(x)$ the probability mass of the random variable X compute the limit

$$\lim_{n \rightarrow \infty} (p(X_1, X_2, \dots, X_n))^{1/n}$$

$$\ln \left(\underbrace{P(X_1=x_1, \dots, X_n=x_n)}_{= P(X_1=x_1) \cdot \dots \cdot P(X_n=x_n)} \right)^{\frac{1}{n}} = \frac{1}{n} \ln \left(\prod_{i=1}^n P(X_i=x_i) \right) = \frac{1}{n} \sum_{i=1}^n \ln P(X_i=x_i)$$

$$X, Y \text{ independent} \Leftrightarrow \forall A, B: P(X \in A, Y \in B) = P(X \in A) \cdot P(Y \in B)$$

$$a^{n+m} = a^n \cdot a^m$$

$$\ln(a^{n+m}) = \ln(a^n) + \ln(a^m)$$

Ex 1.8 Let X_1, X_2, \dots be an i.i.d. sequence of discrete random variables with entropy $H(X)$. Let $C_n(t) = \{x^n \in X^n : p(x^n) \geq 2^{-nt}\}$ denote the subset of n -sequences with probabilities $\geq 2^{-nt}$.

- Show $|C_n(t)| \leq 2^{-nt}$.
- For what values of t does $P(X^n \in C_n(t)) \rightarrow 1$?
- Denoted by $p(x)$ the probability mass of the random variable X compute the limit

X, Y independent \Leftrightarrow
 $\forall A, B : P(X \in A, Y \in B) = P(X \in A) \cdot P(Y \in B)$

X_1, X_2, X_3, \dots i.i.d., $E[X_1] < \infty$
 Strong law of large numbers
 $\frac{1}{n} \sum_{i=1}^n X_i \xrightarrow[n \rightarrow \infty]{\text{almost surely}} E[X_1]$

$\lim_{n \rightarrow \infty} (p(X_1, X_2, \dots, X_n))^{1/n} = \lim_{n \rightarrow \infty} (P(X_1=x_1) \cdot \dots \cdot P(X_n=x_n))^{1/n}$
 $= \lim_{n \rightarrow \infty} (P(X_1=x_1) \cdot \dots \cdot P(X_n=x_n))^{1/n} = \lim_{n \rightarrow \infty} (P(X_1=x_1) \cdot \dots \cdot P(X_n=x_n))^{1/n}$
 $= \frac{1}{n} \cdot \ln(P(X_1=x_1) \cdot \dots \cdot P(X_n=x_n))$
 $= \frac{1}{n} \cdot (\ln(P(X_1=x_1)) + \dots + \ln(P(X_n=x_n)))$
 $a^{n+m} = a^n \cdot a^m$
 $\log_a(x \cdot y) = \log_a(x) + \log_a(y)$
 $\ln(a^b) = b \cdot \ln(a)$

$P_{X_n} = P_X$

$p = P(x_1, \dots, x_n)$

$p(x_1, \dots, x_n) = P_{X_1}(x_1) \cdot \dots \cdot P_{X_n}(x_n)$
 $= P_X(x_1) \cdot \dots \cdot P_X(x_n)$

$\ln(p(x_1, x_2, \dots, x_n))^{1/n}$

$= \frac{1}{n} (\ln(P_X(x_1)) + \dots + \ln(P_X(x_n)))$

$p(x_1, \dots, x_n)^{1/n} \xrightarrow[n \rightarrow \infty]{\text{a.s.}}$

$\xrightarrow[n \rightarrow \infty]{\text{d.s. (almost surely)}} E[\ln(P_X(x_1))] = -H(X_1)$
 $E[\log_2(P_X(x_1))] = 2$