

$$\textcircled{a} \quad I(g(x); Y) \text{ vs } I(X; Y)$$

$H(Y|X) = H(Y|x, g(x)); \quad H(x|x) \geq H(g(x)|x)$

$$H(g(x)|x) = \emptyset \quad I(g(x); Y) = H(g(x)) - H(g(x)|Y) \leq H(X) - \underbrace{H(g(x)|Y)}_{?}$$

$$I(X, g(X); Y) = I(X; Y) + \underbrace{I(g(X); Y|X)}_{\emptyset} =$$

$$= I(g(X); Y) + I(X; Y|g(X))$$

$$I(g(X); Y|X) = \underbrace{H(g(X)|X)}_{\emptyset} - \underbrace{H(g(X)|X, Y)}_{\emptyset} = \emptyset$$

$$I(X; Y) = I(g(X); Y) + \underbrace{I(X; Y|g(X))}_{\geq 0} \Rightarrow$$

$$\boxed{I(X; Y) \geq I(g(X); Y)}$$

Dowiedziano!!!

- Mażko w plaszczeniu o DATA-LOCARING inequality know  
 $I(X; Z) \leq I(X; Y)$  i.e.  $I(X; Y) \geq I(X; Z)$

$$\textcircled{c} \quad H(x_0|x_{-1}) \text{ vs. } H(x_0|x_{-1}, x_1)$$

$$(x_1, x_0, x_1) \quad H(x_{-1}, x_0, x_1) = H(x_{-1}) + H(x_0|x_{-1}) + H(x_1|x_{-1}, x_0)$$

$$H(x_0|x_{-1}) \geq H(x_0|x_{-1}, x_1) \quad \begin{matrix} + \\ \text{CONDITIONING REDUCES} \\ \text{ENTROPY.} \end{matrix}$$

$$P(1|2) \geq 0 \quad \sum_{x \in X} p(x) \log \frac{p(x)}{P(x)} \geq 0$$

$$I(X; Y) = H(X) - H(X|Y) \geq 0 \quad \boxed{H(X) \geq H(X|Y)}$$

$$I(x_0|x_1|x_{-1}) = H(x_0|x_{-1}) - H(x_0|x_{-1}, x_1) \geq 0 \Rightarrow$$

$$\boxed{H(x_0|x_{-1}) \geq H(x_0|x_{-1}, x_1)} \quad \text{Dowiedziano!!!}$$

$$\textcircled{d} \quad \frac{H(X, Y)}{(H(X) + H(Y))} \text{ vs. } 1$$

$$H(X, Y) = H(X) + H(Y|X) \leq H(X) + H(Y)$$

conditioning reduces entropy

$$H(X, Y) \leq H(X) + H(Y)$$

$$\boxed{\frac{H(X, Y)}{H(X) + H(Y)} \leq 1}$$

Dowiedziano!!!

• HWLS SOLUTIONS: (ILIKEE REZULTATI NO HAKU TOLAZHENIEA)

(a)  $\tau = f(x) = 5x$ ,  $f$  is bijective function i.e.  
 $x = f^{-1}(\tau) = g(\tau)$

- ZNAEME DOKA:  $H(x) \geq H(f(x)) = H(5x)$

- OD DRUG SLOVAK:  $H(\tau) \geq H(g(\tau)) = H(x)$

$$H(5x) \geq H(x)$$

ZNAKI ERAS PODIVAJI DOKA  $H(x) \geq H(5x)$  + COKO  
 DOKA  $H(5x) \geq H(x) \Rightarrow [H(x) = H(5x)]$

(b)  $I(g(\tau); \tau)$  vs.  $I(x; \tau)$

HOZE PA ST FORMIRAT MARKOV CHARTA:

$$\tau \rightarrow x \rightarrow g(x)$$

$\rightarrow$  OVA SEZNOD  
 NE ZAVISI OD

T.E. JE PREDOM ZA  
 MARKOV LAVICE

- DATA PROCESSING INEQUALITY:

$$I(\tau; x) \geq I(\tau; g(x)) \Rightarrow I(x; \tau) \geq I(g(x); \tau)$$

$$p(x, z | \tau) = \frac{p(x, y, z)}{p(y)} = \frac{p(y) \cdot p(z | x, \tau)}{p(y)} = \underline{p(x | \tau) \cdot p(z | \tau)}$$

### 2.4.3 MUTUAL INFORMATION OF HEADS AND TAILS

(a) CONSIDER A FAIR COIN FLIP. WHAT IS THE MUTUAL INFORMATION BETWEEN TOP AND BOTTOM SIDES OF THE COIN?

(b) A SIX-SIDE FAIR DIE IS THROWN. WHAT IS THE MUTUAL INFORMATION BETWEEN THE TOP SIDE AND THE FRONT SIDE (THE SIDE MOST FACING YOU)?

①  $X \in \{\text{HEAD, TAIL}\} = \{\text{T, H}\}$   
 ~~$p(x) = \left\{ \begin{array}{ll} 1 & \text{if } x = \text{HEAD} \\ 1 & \text{if } x = \text{TAIL} \end{array} \right.$~~   
 $x \in \{\text{TAIL, NON-TAIL}\}$   $\tau \in \{\text{HEAD, NON-HEAD}\}$

$X$	H	NH	$p(x)$
T	0	$\frac{1}{2}$	$\frac{1}{2}$
NT	$\frac{1}{2}$	0	$\frac{1}{2}$
$p(\tau)$	$\frac{1}{2}$	$\frac{1}{2}$	

$$I(x; \tau) = H(x) - H(x|\tau)$$

$\circled{P(x|\tau)}$

$X$	H	NH
T	0	1
NT	1	0

$$P(x|\tau) = \frac{P(x, \tau)}{P(\tau)}$$

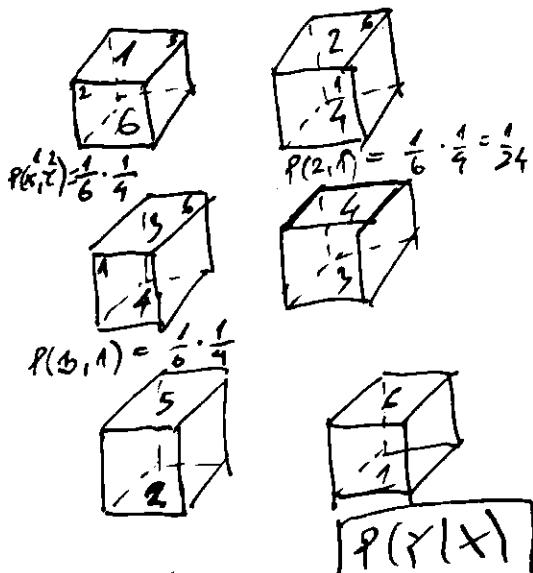
$$H(X) = \frac{1}{2} \log_2 + \frac{1}{2} \log_2 = \frac{1}{2} + \frac{1}{2} = 1$$

$$H(X|Y) = \sum_{x,y} p(x,y) \log \frac{1}{p(x|y)} = 0 \log 0 + \frac{1}{2} \log \frac{1}{2} + \frac{1}{2} \log \frac{1}{2} = 0$$

$$I(X;Y) = H(X) - H(X|Y) = 1 - 0 = 1 \text{ bit}$$

LOGIČNO je! Ako  $Y$  = SIGURNO GO ZNAČI,  $X$

(b)  $X$  = GORNA STRANA  $\in \{1, 2, 3, 4, 5, 6\}$   
 $Y$  = PREDNA STRANA  $\in \{1, 2, 3, 4, 5, 6\}$



$X \setminus Y$	1	2	3	4	5	6	$P(X)$
1	0	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	0	$\frac{1}{6}$
2	$\frac{1}{6}$	0	$\frac{1}{6}$	$\frac{1}{6}$	0	$\frac{1}{6}$	$\frac{1}{6}$
3	$\frac{1}{6}$	$\frac{1}{6}$	0	0	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$
4	$\frac{1}{6}$	$\frac{1}{6}$	0	0	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$
5	$\frac{1}{6}$	0	$\frac{1}{6}$	$\frac{1}{6}$	0	$\frac{1}{6}$	$\frac{1}{6}$
6	0	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	0	$\frac{1}{6}$
$P(Y)$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	0

$X \setminus Y$	1	2	3	4	5	6
1	0	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	0
2	$\frac{1}{4}$	0	$\frac{1}{4}$	$\frac{1}{4}$	0	$\frac{1}{4}$
3	$\frac{1}{4}$	$\frac{1}{4}$	0	0	$\frac{1}{4}$	$\frac{1}{4}$
4	$\frac{1}{4}$	$\frac{1}{4}$	0	0	$\frac{1}{4}$	$\frac{1}{4}$
5	$\frac{1}{4}$	0	$\frac{1}{4}$	$\frac{1}{4}$	0	$\frac{1}{4}$
6	0	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	0

VREDNOSTI

$$I(X;Y) = H(X) - H(X|Y)$$

$$H(X) = 6 \cdot \frac{1}{6} \log_2 6 = \log_2 6 = 2.58$$

$$H(X|Y) = - \sum_{x,y} P(x,y) \log P(x|y)$$

$$= 24 \cdot \frac{1}{24} \cdot \log_2 4 = \log_2 4 = 2$$

$$\boxed{I(X;Y) = \log_2 6 - \log_2 4 = \log_2 \frac{6}{4} = 0.585}$$

- POZNAVANJE NA GORNATA STRANA JA NAYAZVANA NEIZVES  
 NESTA NA VREDNOSTA NA PREDNATA STRANA ZA CELE  
 2 BITA !!!

- Isto go kriterij vo **H(X|Y)** so tda sto direktno odgovar go formulata za transformaciju (ANO GO ZAKI TOP)  $= H(X) - D = 2 \cdot \frac{1}{2} \log_2 2 = 1$
- $I(B;T) = H(B) - H(B|T) = (ANO GO ZAKI TOP) = H(B) - D = 2 \cdot \frac{1}{2} \log_2 2 = 1$
- $I(T;F) = H(T) - H(F|T) = 6 \cdot \frac{1}{6} \log_2 6 - 4 \cdot \frac{1}{4} \log_2 4 = \log_2 \frac{6}{4} = 0.585$

MO BREZ CEDOVNI  
 GORNA STRANA ZA PRED  
 NATA STRANA 3  
 INA 4 MOGLA SO COVAKI VEROJ

• PRIMEROT HA BEIJERETO VO HW2S TC DIREKCUA DA NE ODIS SO CORA FORMULA ZA H(X) TUKU SO SKLADENATA TE RIX SO TIEZIATI KOMO CORA SKLADENATA KOMENJVA ZA DA NE ODIS SO FORMULATA  $\sum p(x) \ln p(x)$ .

- SEPAR NOVO VJEZNE E MATEMATICKI POKAZOVAN A ICERETEO VO HW2S E POPUNTICHO, PONIDENTE-ZNO.

**2.44** PURE RANDOMNESS. WE WISH TO USE THREE-SIDES COIN TO GENERATE FAIR COIN TOSS. LET THE COIN HAVE PROBABILITY MASS FUNCTION  $X$

$$x = \begin{cases} A, & p_A \\ B, & p_B \\ C, & p_C \end{cases} \quad \boxed{p_C = 1 - p_A - p_B}$$

WHERE  $p_A, p_B, p_C$  ARE UNKNOWN.

(a) HOW WOULD YOU USE TWO INDEPENDENT BITS  $x_1, x_2$  TO GENERATE (IF POSSIBLE) BERNULLI ( $\frac{1}{2}$ ) RANDOM VARIABLE  $Z$ ?

(b) WHAT IS THE RESULTING MAXIMUM EXPECTED NUMBER OF FAIR BITS GENERATED?

(a) AA, AB, AC, BA, BB, BC, CC, CA, CB

$$H(X) = -p_A \ln p_A - p_B \ln p_B - p_C \ln p_C$$

$$X \in \{A, B, C\} \quad W \in \{W_1, W_2, \dots, W_{2^n}\}$$

$$\boxed{W_i = \{W_{i1}, W_{i2}\}_{n=2}}$$

$$\boxed{H(W) = n H(X)}$$

$$H(X) = -p_A \ln p_A - p_B \ln p_B - (1-p_A-p_B) \ln p_C =$$

$$= -p_A \ln p_A - p_B \ln p_B - \ln p_C + p_A \ln p_C + p_B \ln p_C$$

$$= p_A \ln \frac{p_C}{p_A} + p_B \ln \frac{p_C}{p_B} - \ln p_C = \ln \left( \frac{p_C}{p_A} \right)^{p_A} + \ln \left( \frac{p_C}{p_B} \right)^{p_B} - \ln p_C$$

$$= \ln \left[ \left( \frac{p_C}{p_A} \right)^{p_A} \cdot \left( \frac{p_C}{p_B} \right)^{p_B} \cdot \frac{1}{p_C} \right]$$

$$P(W) = \{p_A, p_A p_B, p_A p_C, p_B^2, p_B p_A, p_B p_C, p_C^2, p_C p_A, p_C p_B\}$$

$$\frac{p_A}{8} + \frac{p_B}{8} = \frac{1}{2} \quad p_C = \frac{1}{2} \quad \underbrace{\left\{ \frac{1}{8^2}, \frac{1}{8^3}, \frac{1}{8^4}, \dots \right\}}_{\text{NEMA}}$$

$$\{AA, AB, AC, BB, BA, BC, CC, CA, CB\}$$

$$H(x) \geq H(f(x))$$

$$\left\{ \underbrace{\overbrace{AA, AB, AC}^{\frac{1}{2}}, \overbrace{BA, BC}^{\frac{1}{2}}, \overbrace{BB, CC}^{\frac{1}{2}}, \overbrace{CA, CB}^{\frac{1}{2}} \right\}$$

$$\begin{aligned} p_A^2 + 2p_A p_B + 2p_A p_C &= \frac{1}{2} \\ p_B^2 + 2p_B p_C + p_C^2 &= \frac{1}{2} \\ p_A + p_B + p_C &= 1 \end{aligned}$$

$$\begin{aligned} x^2 + 2x\gamma + 2x\zeta &= \frac{1}{2} \\ \gamma^2 + 2\gamma\zeta + \zeta^2 &= \frac{1}{2} \\ x + \gamma + \zeta &= 1 \\ \zeta &= 1 - (x + \gamma) \end{aligned}$$

$$x^2 + 2x\gamma + 2x - 2x^2 - 2x\gamma = \frac{1}{2}$$

$$2x^2 + 4x - 4x^2 = 1$$

$$2x^2 - 4x + 1 = 0$$

$$-2x^2 + 4x - 1 = 0$$

$$x_{1,2} = \frac{4 \pm \sqrt{16 - 8}}{4}$$

$$x_{1,2} = 1 \pm \frac{1}{4}\sqrt{8} = 1 \pm \frac{1}{4}\sqrt{2} = 1 \pm \frac{\sqrt{2}}{2} = \frac{2 - \sqrt{2}}{2} = 0.293$$

$$\gamma^2 + 2\gamma(1-x-\gamma) + (1-(x+\gamma))^2 = \frac{1}{2} =$$

$$= \gamma^2 + 2\gamma - \frac{2 - \sqrt{2}}{2} \cdot 2\gamma - 2\gamma^2 + \left(1 - \frac{2 - \sqrt{2}}{2} - \gamma\right)^2 = \frac{1}{2}$$

$$= \gamma^2 + 2\gamma - 2\gamma + \sqrt{2}\gamma - 2\gamma^2 + \left(\frac{2 - \sqrt{2}}{2} - \gamma\right)^2 = -\gamma^2 + \sqrt{2}\gamma + \left(\frac{\sqrt{2}}{2} - \gamma\right)^2 =$$

$$= -\gamma^2 + \sqrt{2}\gamma + \left(\frac{2}{4} - \frac{2\sqrt{2}}{4}\gamma + \gamma^2\right) = -\gamma^2 + \sqrt{2}\gamma + \frac{1}{2} - \sqrt{2}\gamma + \gamma^2 =$$

$$\Rightarrow \boxed{\frac{1}{2} = \frac{1}{2}}$$

$$\zeta = 1 - x - \gamma = 1 - \frac{2 - \sqrt{2}}{2} - \gamma$$

$$\zeta = \frac{2 - 2 + \sqrt{2}}{2} - \gamma = \frac{\sqrt{2}}{2} - \gamma$$

$$\boxed{\zeta = \frac{\sqrt{2}}{2} - \gamma}$$

$$\text{IF: } \gamma = \frac{1}{4} \Rightarrow \zeta = \boxed{\frac{\sqrt{2}}{2}} \bullet \frac{1}{4} = \underline{0.707 - 0.25} = 0.457$$

$$\boxed{0.457 + 0.293 + 0.25 = 0.750 + 0.250 = 1}$$

$$\gamma^2 + 2\gamma\left(\frac{\sqrt{2}}{2} - \gamma\right) + \left(\frac{\sqrt{2}}{2} - \gamma\right)^2 = \gamma^2 + \gamma\sqrt{2} - 2\gamma^2 + \frac{1}{2} - \sqrt{2}\gamma + \gamma^2 = \frac{1}{2}$$

$$p(x) = \{p_A, p_B, p_C\} = \{0.293; 0.25; 0.457\}$$

$$H(x) = -0.293 \ln 0.293 - 0.25 \ln 0.25 - 0.457 \ln 0.457 = 1.535$$

$$H(W) = \sum_{i=1}^9 p_i(w) \ln p_i(w) = -p_1^2 \ln p_1^2 - p_2 p_3 \ln p_2 p_3 - \dots - p_9 p_9 \ln p_9 p_9$$

$$H(W) = 3.0704$$

$$H(W) = 2 \cdot H(X) = 2 \cdot 1.535 = 3.070$$

THE MAXIMUM NUMBER OF PURE RANDOM BITS THAT CAN BE OBTAINED FROM THREE SIDE COIN IS:

$$\ln H(X)$$

WE WANT SUCH:

$$p(x) = \left[ \frac{2-p_2}{2}, p_1, \frac{p_2}{2} - p_1 \right]$$

$$H(X) = -\frac{2-p_2}{2} \ln \frac{2-p_2}{2} - p_1 \ln p_1 - \left( \frac{p_2}{2} - p_1 \right) \ln \left( \frac{p_2}{2} - p_1 \right)$$

$$\frac{dH(X)}{dp_1} = 0 \quad \left[ p_1 \ln p_1 + \left( \frac{p_2}{2} - p_1 \right) \ln \left( \frac{p_2}{2} - p_1 \right) \right] = 0$$

MAXE:  
Multi-hop MIMO Cap. w.r.t.  $p_1$

$$p_1 = \frac{p_2}{4} \Rightarrow H(X) \text{ MAXIMUM}$$

$$H_{\max}(X) = 1.5795 \approx 1.580$$

$\Rightarrow$  BEST DISTRIBUTION FOR  $X$  IS:

$$p(x) = \{p_1, p_2, p_3\} = \left\{ \frac{2-p_2}{2}, \frac{p_2}{4}, \frac{p_2}{2} - \frac{p_2}{4} \right\}$$

$$p(x) = \left\{ \frac{2-p_2}{2}, \frac{p_2}{4}, \frac{p_2}{4} \right\}$$

$$H(W) = 2 \cdot 1.580 = 3.160 \Rightarrow \text{MAXIMUM NUMBER OF PURE RANDOM BITS.}$$

2.45 FINITE ENTROPY Show for a discrete random variable  $X \in \{1, 2, \dots\}$ , if  $E[\ln x] < \infty$  then  $H(X) < \infty$ .

JENSEN'S INEQUALITY:  $E[f(x)] \geq f(E[x])$   
IF  $f(t)$  IS CONVEX FUNCTION

$$H(X) = - \sum_{x \in X} p(x) \ln p(x) = -E[\ln p(x)] \geq -E[\ln E[p(x)]]$$

$$E[\ln p(x)] \leq \ln E[p(x)]$$

$$E[p(x)] = \sum_{x \in X} p^2(x)$$

$$-E[\ln p(x)] \geq -E[\ln E[p(x)]]$$

$$\begin{aligned} \ln(p(x)) &\Rightarrow \text{CONCAVE FUNCTION} \\ 2^{-H(X)} &= 2^{-E[\ln p(x)]} \leq 2^{-E[\ln E[p(x)]]} \\ &= E[2^{-p(x)}] = \sum_{x \in X} 2^{-p^2(x)} \\ 2^{-H(X)} &\leq \sum_{x \in X} 2^{-p^2(x)} \end{aligned}$$

$$H(x) \geq -\log \sum_{x \in X} p(x) = -\log \left[ \sum_{x \in X} p^*(x) \right] =$$

$$x \in \{1, 2, \dots\} \quad p(n) = \{p_1, p_2, \dots\}$$

$$E[x] = \sum_{x \in X} x \cdot p(x) = \sum_{i=1}^{\infty} i \cdot p_i \quad \sum_{i=1}^{\infty} i \cdot x^i = \frac{x}{(1-x)^2}$$

$$E[\log x] = \sum_{x \in X} p(x) \log(x) \leq \log \left[ \sum_{x \in X} x \cdot p(x) \right] =$$

$$E[\log x] \leq \log E[x]$$

$$\textcircled{A} \Rightarrow -H(x) \leq \log \sum_{x \in X} p^*(x) \quad H(x) \geq \log \sum_{x \in X} p(x)$$

• Tres da mesma forma  $H(x) \leq E[\log x]$

$$\sum_{x \in X} p(x) \log(x) = \sum_{x \in X} \log x \cdot p(x) = \log \left( \prod_{x \in X} x^{p(x)} \right)$$

$$x = \log 2^x \quad \sum_{x \in X} x \cdot p(x) = \sum_{x \in X} \log 2^x \cdot p(x) = E[\log 2^x]$$

$$\leq \log 2^{E[x]} = E[x]$$

$$H(x) = -E[\log p(x)] \geq -\log E[p(x)] = -\log \sum_{x \in X} p^*(x)$$

$$-H(x) \leq \log \sum_{x \in X} p^*(x) \quad \boxed{2^{-H(x)} \leq \sum_{x \in X} p^*(x)}$$

$$D(p||q) = \sum_{i=0}^{\infty} p_i \log \frac{p_i}{q_i} \geq 0 \quad \sum_{i=0}^{\infty} p_i \log p_i - \sum_{i=0}^{\infty} p_i \log q_i \geq 0$$

$$\sum_{i=0}^{\infty} p_i \log p_i \geq \sum_{i=0}^{\infty} p_i \log q_i - \sum_{i=0}^{\infty} p_i \log q_i \leq -\sum_{i=0}^{\infty} p_i \log q_i$$

$q_i = 2^{-i}$

$$-\sum_{i=0}^{\infty} p_i \log 2^{-i} \leq -\sum_{i=0}^{\infty} p_i \log 2^i = +\sum_{i=0}^{\infty} i \cdot p_i$$

$H(x) \leq E[x]$

$E[x] \leq \log 2 (x-1)$

$\log x \leq x-1$

$$f(x) = f(x_0) + \frac{f'(x_0)}{1!} (x-x_0) + \frac{f''(x_0)}{2!} (x-x_0)^2 + \dots$$

$$\begin{aligned} \text{Ed}[x] &\leq \frac{1}{\ln 2}(x-1) & \boxed{\text{Ed}(x) \leq x-1} & \text{Torre} \\ \mathbb{E}[\text{Ed}(x)] &\leq \text{Ed}[\mathbb{E}(x)] \leq \text{Ed}[H(x)+1] \leq \mathbb{E}[x] \\ \mathbb{E}[\text{Ed}[x]] &\leq \frac{1}{\ln 2}\mathbb{E}[x-1] = \frac{1}{\ln 2}[\mathbb{E}[x]-1] \geq \frac{1}{\ln 2}[H(x)-1] \\ \text{Ed}[x] &\leq \text{Ed}(x+1) \leq x & \cancel{\text{Ed}(x)} \end{aligned}$$

Znáčí:  
①  $\mathbb{E}[x] \geq \mathbb{E}[\text{Ed}(x)]$

②  $\mathbb{E}[x] \geq H(x)$

$\mathbb{E}[\text{Ed}(x)] < \infty$

NE maximálna dôsa  $H(x) < \infty$

$$\begin{aligned} H(x) &= -\sum_{x \in X} p(x) \text{Ed}[p(x)] \geq -\sum_{x \in X} p(x) \text{Ed}(p(x)+1) \geq \\ &\geq -\sum_{x \in X} p(x) \cdot p(x) = -\sum_{x \in X} p^2(x) \\ -H(x) &\leq \sum_{x \in X} p^2(x) \end{aligned}$$

$\text{Ed}[x] = \frac{\ln x}{\ln 2}$   
 $\ln x = \ln 2 \cdot \text{Ed}[x]$   
 $\text{Ed}[x] = \frac{1}{\ln 2} - \frac{1}{\ln 2 \cdot x}$

- kroko je konkav, teda Jeansens nequality

$$\begin{aligned} H(x) &= -\sum_{x \in X} p(x) \text{Ed}[p(x)] = -\mathbb{E}[\text{Ed}[p(x)]] \geq -\text{Ed}\{\mathbb{E}[p(x)]\} \geq \\ &\geq -\text{Ed}\{\mathbb{E}[p(x)]+1\} \geq -\mathbb{E}[p(x)] = -\sum_{x \in X} p^2(x) \end{aligned}$$

$\mathbb{E}[\text{Ed}(x)] < \infty$        $\mathbb{E}[\text{Ed}(x)] \leq \text{Ed}\{\mathbb{E}[x]\}$

$$\ln x \leq x-1 \quad \gamma = \frac{1}{x} \quad x = \frac{1}{\gamma} \quad -\ln \gamma \leq \frac{1}{\gamma} - 1$$

$$\boxed{\ln \gamma \geq 1 - \frac{1}{\gamma}} \quad \boxed{\gamma > 0} \quad \begin{aligned} &x \in \{1, 2, 3, \dots\} \\ &P(\gamma) = \{\gamma_1, \gamma_2, \dots\} \end{aligned}$$

$$\mathbb{E}[\ln \gamma] \geq \sum_{\gamma \in \Gamma} \left(1 - \frac{1}{\gamma}\right) P(\gamma) = 1 - \sum_{\gamma \in \Gamma} \frac{P(\gamma)}{\gamma} = 1 - \mathbb{E}\left[\frac{1}{\gamma}\right]$$

$$H(\gamma) = - \sum_{\gamma \in \Gamma} p(\gamma) \ln p(\gamma)$$

$$\ln p(\gamma) \leq \gamma(\gamma) - 1$$

$$H(\gamma) = - \sum_{\gamma \in \Gamma} p(\gamma) (\ln p(\gamma)) \geq - \sum_{\gamma \in \Gamma} p(\gamma) (1 - p(\gamma)) = 1 - \sum_{\gamma \in \Gamma} p^2(\gamma)$$

$$E[\ln \gamma] \geq 1 - E\left[\frac{1}{\gamma}\right]$$

$$H(\gamma) \geq 1 - \sum_{\gamma \in \Gamma} p^2(\gamma) \geq 1 - \frac{1}{2} H(\gamma)$$

$$\begin{aligned} \gamma(\gamma) + 2^{-H(\gamma)} &\geq 1 / \ln \\ \ln H(\gamma) &\geq H(\gamma) ? \end{aligned}$$

$$(\ln H(\gamma) - H(\gamma)) \geq 0 ?$$

$$H(\gamma) \geq 1 - \sum_{\gamma \in \Gamma} p^2(\gamma) \quad E[\ln \gamma] \geq 1 - E\left[\frac{1}{\gamma}\right]$$

$$E[\ln \gamma] \leq \ln E[\gamma]$$

$$H(\gamma) = -p_1 \ln p_1 - p_2 \ln p_2 - p_3 \ln p_3 - \dots$$

$$E[\ln \gamma] = p_1 \ln \gamma_1 + p_2 \ln \gamma_2 + p_3 \ln \gamma_3 + \dots$$

$$\gamma \in \{1, 2, 3, \dots\}$$

$$E[\ln \gamma] = p_2 \ln 2 + p_3 \ln 3 + p_4 \ln 4 + \dots$$

$$E[\ln \gamma] = \sum_{i=1}^{\infty} p_i \ln i = \sum_{i=1}^{\infty} p_i \ln \frac{1}{\left(\frac{1}{i}\right)}$$

$$H(\gamma) = \sum_{i=1}^{\infty} p_i \ln \frac{1}{p_i} = \sum_{i=1}^{\infty} p_i \ln \frac{1}{\left(\frac{1}{i}\right)} = - \sum_{i=1}^{\infty} p_i \ln p_i = - \sum_{i=1}^{\infty} p_i h_i$$

$$E[\ln \gamma] \geq \frac{1 - E\left[\frac{1}{\gamma}\right]}{H(2)} = \left[1 - \sum_{i=1}^{\infty} \frac{h(i)}{i}\right] \cdot \frac{1}{H(2)}$$

$$\ln p_i = \left| \begin{array}{l} \text{IF } \gamma_i = 2^i \\ \text{else } 0 \end{array} \right| = \ln 2^i = i$$

$$H(\gamma) = - \sum_{i=1}^{\infty} p_i \ln p_i \quad E[\ln \gamma] \geq \left[1 - \sum_{i=1}^{\infty} \frac{p_i}{i}\right] \cdot \frac{1}{H(2)}$$

$$H(\gamma) \geq 1 - \sum_{i=1}^{\infty} p_i^2$$

$$\text{Sind } \left[1 - \sum_{i=1}^{\infty} \frac{p_i}{i}\right] \cdot \frac{1}{H(2)} \geq 1 - \sum_{i=1}^{\infty} p_i^2 ?$$

$$\sum_{Y \in T} \frac{y(x)}{\gamma} \geq \sum_{Y \in T} y^2(x) \quad \text{GRENZV. SE 82}$$

V. 01 MAKE Multihop MIMO CARRIER. W 3.27.15

$$\frac{1}{L_2} \left[ 1 - \sum_{Y \in T} \frac{y(\gamma)}{\gamma} \right] \leq 1 - \sum_{Y \in T} y^2(\gamma) \quad \leq 1$$

[0, ... 1]      [0, ... 1]

FIRMAZ //

$$H(X) \geq E[\text{ld}(x)]$$

Das MUSAM DEVA e ODAZO je DEVA TWO  
 $H(X) < \infty$  TOGA  $E[\text{ld}(X)] \leq \infty$

$$\underbrace{E[\text{ld}(x)] \leq H(x) < \infty}_{\Rightarrow E[\text{ld}(x)] < \infty}$$

I RESENTE  
 TEST VREMOŠT  
 za  $\text{ld}(x)$ ,  
 $E[\text{ld}(x)]$  GO POKROV-  
 VAT OVA MAKE.

$$y = \alpha \cdot \beta^i \rightarrow \text{to generation } x \in \{i\} \text{ scučit: } i=1, 2, \dots$$

$$\frac{1}{L_2} - \frac{3\alpha\beta}{2L_2} \leq 1 - 2\alpha^2\beta^2$$

$$\frac{1}{L_2} - \frac{\gamma}{L_2} \leq 1 - \alpha^2$$

$$1 - \alpha \leq L_2 - L_2\alpha^2$$

$$L_2 \alpha^2 - \alpha + 1 - L_2 \leq 0$$

$$2 - 3\alpha\beta \leq 2L_2 - 4L_2\alpha^2\beta^2$$

$$4L_2\alpha^2\beta^2 - 3\alpha\beta + 2 - 2L_2 \leq 0$$

$$y = \alpha \cdot \beta^i \quad h(x) = \sum_{i=1}^{\infty} \alpha \cdot \beta^i \text{ld}(\alpha \cdot \beta^i) =$$

$$= \sum_{i=1}^{\infty} \alpha \cdot \beta^i \text{ld}(\beta^i) + \sum_{i=1}^{\infty} \alpha \cdot \beta^i \cdot i \text{ld}(\beta^i) = \alpha \cdot \text{ld}(\beta) \sum_{i=1}^{\infty} \beta^i +$$

$$+ \alpha \beta \sum_{i=1}^{\infty} (\alpha \cdot \beta^i) \cdot i = \text{ld}(\alpha) + \text{ld}(\beta) \cdot g(x)$$

$\in [x] \quad \in [x] \quad \in [x]$

OUT POMOČI SOD KOREK  
 (2.19) ED.2

$$H(x) = \text{ld}x + \text{ld}\beta \cdot E[x] \geq \text{ld}x + \text{ld}\beta \cdot E[\text{ld}(x)]$$

$$E[\text{ld}(x)] \leq \frac{H(x) - \text{ld}x}{\text{ld}\beta}$$

$$\underbrace{H(x)}_{\text{erzielbar durch Verwendung von } p=1-\beta} \geq -\text{ld}x - \text{ld}\beta \cdot E[\text{ld}(x)]$$

• OVA KENNEN PERTURBATIONEN VO GENERATEN SICHTEN ZU  
 $p = \alpha \cdot \beta^i$  ET  $\alpha, \beta \in [0..1]$   $H(x) \geq E[\text{ld}(x)]$

• AVO: LÖSUNG UND DAZIET MÖGLICH (2.30)  
 TOETS:

$$E[\text{ld}x] \leq \text{ld}\{\overline{E[x]}\} \leq E[x]$$

zu 5.10  $x \in \{1, 2, \dots\}$

$$E[x] = \sum_{i=1}^{\infty} i \cdot p(i)$$

AVO  $E[\text{ld}x] < \infty$   
 NE MÖGLICHE ZAHLEN  $E[x] < \infty$  !!!

NO SICHTEN MUß  $E[\text{ld}x] = \text{const} \leq \infty$  ICH

$$D(g|g) \geq 0 \quad \sum_{i=0}^{\infty} p_i \text{ld} \frac{p_i}{g_i} \geq 0$$

$$+ \sum_{i=0}^{\infty} p_i \text{ld} p_i - \sum_{i=0}^{\infty} p_i \text{ld} g_i \geq 0 \quad - \sum_{i=0}^{\infty} p_i \text{ld} p_i \leq - \sum_{i=0}^{\infty} p_i \text{ld} g_i =$$

$$= |2i = \alpha \cdot \beta^i| = - \sum_{i=0}^{\infty} p_i \text{ld}(\alpha \cdot \beta^i) = - \sum_{i=0}^{\infty} p_i \text{ld}\alpha - \sum_{i=0}^{\infty} p_i \text{ld}\beta^i =$$

$$= - \text{ld}\alpha - \text{ld}\beta \sum_{i=0}^{\infty} i \cdot p_i = - \text{ld}\alpha - \text{ld}\beta E[x]$$

$$\alpha, \beta \in [0..1] \Rightarrow H(x) \leq -\text{ld}\alpha - \text{ld}\beta E[x]$$

$$H(x) \leq \text{ld}x + \text{ld}\beta E[x] = \begin{cases} \alpha = 1 \\ \beta = 1/2 \end{cases} = 0 + E[x]$$

$$H(x) \leq E[x] \quad \text{PARK SE DORTA MÖGLICH !!}$$

$$\sum_{i=0}^{\infty} \alpha \cdot \beta^i = \alpha \frac{1}{1-\beta} = 1 \quad \alpha \sum_{i=0}^{\infty} i \cdot \beta^i = \frac{\alpha \cdot \beta}{(1-\beta)^2} = A$$

$$\frac{1}{1-\beta} = A \quad \frac{i}{1-\beta} = A \cdot \frac{1}{1-\beta} = A \quad \boxed{\frac{1}{1-\beta} = 1} \quad \boxed{\alpha = 1-\beta}$$

$$\frac{1-a}{1-A+a} = A \quad 1-\alpha = \alpha \cdot A \quad \alpha = \frac{a}{A+1}$$

$$\alpha(A+a) = a \quad \alpha = \frac{a}{A+a}$$

$$\beta = 1 - \frac{1}{A+1} = \frac{A+1-1}{A+1} = \frac{A}{A+1} //$$

$$\boxed{\beta = \frac{1}{A+1} \left( \frac{A}{A+1} \right)^{-1}}$$

Aho TEGNES so omsæsen

$$E[X] = \sum_{i=0}^{\infty} i p(i) = A$$

$$E[\ln d(X)] \leq \ln d E[X] \quad \rightarrow E[\ln d(X)] < \infty$$

$$E[X] = A < \infty$$

$$-H(x) = E[\ln d(p(x))] \leq \ln d \underbrace{E[p(x)]}_{E[0..1]} = \ln d \sum_{i=0}^A p^2(i) \leq E[\gamma(x)] - 1$$

MACLOREN

(L1):

$$-H(x) = E[\ln d(\gamma(x))] \leq \ln d \underbrace{\{E[p(x)]\}}_{E[0..1]} \leq \ln d \{E[\gamma(x)] + 1\} \leq \underline{E[\gamma(x)]}$$

N13.12  $\Rightarrow +H(x) \leq -\ln d - \ln d \beta \cdot E[X] =$

$$= -\ln \frac{1}{1+A} - \ln \frac{A}{1+A} E[X] = \ln(1+A) + \ln\left(\frac{A+1}{A}\right) E[X]$$

$$+H(x) \leq \ln(1+A) + \ln(1+A) \cdot E[X] - \ln d(A) E[X] =$$

$$= \ln(1+A) [1 + E[X]] - \ln d(A) E[X] = \frac{(1+A)\ln(1+A) - A\ln d(A)}{A}$$

$$\leq (1+A) \cdot \left(1 - \frac{1}{A}\right) - A \left(1 - \frac{1}{A}\right) = (1+A) \frac{A-1}{A} - \frac{A-1}{A} \cdot A$$

$$= \frac{A-1}{A} \quad \boxed{+H(x) \leq \frac{A-1}{A}}$$

• VD OVS SLEJER EN DØ KAKVO,  $A = p(x)$   
 STENGER I E PONVERSENTE FOR KURVA (VIDI MÅLTET  
 MULTITRØY M/MOCSENSE. MW (3.27.25)

$$+H(x) = E[\ln d(p(x))] \leq \ln d E[p(x)] \leq \ln d [E[\gamma(x)] + 1] \leq E[\gamma(x)]$$

$$\leq p(E[X]) - 1$$

$$+H(x) = \sum_{i=0}^{\infty} p(i) \ln d \frac{1}{p(i)} = - \sum_{i=0}^{\infty} \underbrace{\frac{1}{A+1} \left(\frac{A}{A+1}\right)^i}_{p(i)} \ln \frac{1}{A+1} \left(\frac{A}{A+1}\right)^i$$

for  $A \rightarrow 1$   $\boxed{p(x) = \left(\frac{1}{2}\right)^x}$

$$+H(x) = -E[\ln d(p(x))] = -E[\ln d\left(\frac{1}{2}\right)^x] = E[\ln 2^x] = E[X]$$

$$H(x) \leq \frac{(1+A) \cdot \text{ld}(1+A) - A \cdot \text{ld}A}{\text{maximum entropy pr. 2.30}} = \text{ld}(1+A) + \cancel{A \cdot \text{ld}(A)} - A \cdot \cancel{\text{ld}A}$$

$$H(x) \leq \text{ld}(1+A) \leq A$$

$$\text{ld}(1+A) \leq A \quad \left. \begin{array}{l} \text{ld}(1+A) \leq x \\ \text{ld}(1+A) \leq A \end{array} \right\} \text{MACLOEDOV RED}$$

ZNAČI VONEČNO, ALE JE ZELENÍ PDF-OT DA  
BIDEZI:  $\text{E}[X] \leq A$

$$\boxed{H(x) \leq A = E[X]} \quad (z_4 A \geq 4)$$

$$\text{BIDEZI: } E[\text{ld}(x)] \leq \text{ld} E[x] \leq \text{ld} A$$

USCOVOT  $E[\text{ld}(x)] < \infty$  DA  $H(x) < \infty$   
JE SIGURNO (SLOVAKO) ZDÍTO VÍTĚ ROVĚ  
 $E[X] < \infty$ .

→ OVA SEMOGA VÁŽI ALE ZELENÍ  $P(X) = 2^{-x}$   
POVÁZOT JE TO RELEVANTNÉ ENTROPIA  $(P(X=2) \geq 0)$  KOM  
NE PFT.  
SUSTINISUJOT POHÁZ  $\infty$ .

**2.46 Axiomatic definition of entropy.** If we assume certain axioms for our measure of information we will be forced to use a logarithmic measure such as entropy. Shannon used this to justify his initial definition of entropy. In this book we focus more on the other properties of entropy rather than its axiomatic derivation to justify its use. Axiomatic derivation to justify its use. If the sequence of symmetric functions  $H_m(\gamma_1, \gamma_2, \dots, \gamma_m)$  satisfies the following properties:

- Normalization:  $H_2\left(\frac{1}{2}, \frac{1}{2}\right) = 1$

- Continuity:  $H_2(\gamma_1, 1-\gamma_1)$  is continuous function of "p"

- Grouping:  $H_m(\gamma_1, \gamma_2, \dots, \gamma_m) = H_{m-1}(\gamma_1, \gamma_2, \dots, \gamma_{m-1})$

PROVE TOTAT  $H_m$  MUST BE OF THE FORM:

$$H_m(\gamma_1, \gamma_2, \dots, \gamma_m) = - \sum_{i=1}^m \gamma_i \log \gamma_i \quad \alpha = 2, 3, \dots$$

## GROWING PROPERTY i.e. AXIOM:

$$H_m(p_1, p_2, \dots, p_m) = H_{m-1}(p_1 + p_2, p_3, \dots, p_m) +$$

$$(p_1 + p_2) H_2\left(\frac{p_1}{p_1 + p_2}, \frac{p_2}{p_1 + p_2}\right)$$

\*

$$H_m = \sum_{i=1}^m p_i \ln \frac{1}{p_i} = p_1 \ln p_1^{-1} + p_2 \ln p_2^{-1} + \dots + p_m \ln p_m^{-1}$$

$$+ (p_1 + p_2) \ln \frac{1}{p_1 + p_2} - (p_1 + p_2) \ln \frac{1}{p_1 + p_2}$$

$$H_{m-1}(p_1 + p_2, p_3, \dots, p_m) = (p_1 + p_2) \ln \frac{1}{p_1 + p_2} + p_3 \ln \frac{1}{p_3} + \dots$$

$$\textcircled{3} = (p_1 + p_2) \left[ \frac{p_1}{p_1 + p_2} \ln \frac{p_1 + p_2}{p_1} + \frac{p_2}{p_1 + p_2} \ln \frac{p_1 + p_2}{p_2} \right] =$$

$$= p_1 \ln \left( \frac{p_1 + p_2}{p_1} \right) + p_2 \ln \left( \frac{p_1 + p_2}{p_2} \right)$$

$$\underline{p_1 \ln \frac{1}{p_1}} + \underline{p_2 \ln \frac{1}{p_2}} - \underline{p_1 \ln \frac{1}{p_1 + p_2}} - \underline{p_2 \ln \frac{1}{p_1 + p_2}} =$$

$$= p_1 \ln \frac{p_1 + p_2}{p_1} + p_2 \ln \frac{p_1 + p_2}{p_2} = (p_1 + p_2) H\left(\frac{p_1}{p_1 + p_2}, \frac{p_2}{p_1 + p_2}\right)$$

∴ PROOF:

QED !!

$$H_m = (p_1 + p_2) H_2\left(\frac{p_1}{p_1 + p_2}, \frac{p_2}{p_1 + p_2}\right) + H_{m-1}(p_1 + p_2, p_3, \dots, p_m)$$

④ SOLUTION 1 SOLUTION:

- EXPAND THE GROWING AXIOM BY INDUCTION AND PROVE THAT:

$$H_n(p_1, p_2, \dots, p_n) = H_{n-k}(p_1 + p_2 + \dots + p_k, p_{k+1}, \dots, p_n) +$$

$$+ (p_1 + p_2 + \dots + p_k) H_k\left(\frac{p_1}{p_1 + p_2 + \dots + p_k}, \dots, \frac{p_k}{p_1 + p_2 + \dots + p_k}\right)$$

LET  $f(n)$  BE FACTOR OF UNIFORM DISTRIBUTION OF  $n^n$  SYMBOLS i.e.:

$$f(n) = H_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right)$$

- For any two integers  $r$  and  $s$ :

$$f(r,s) = f(r) + f(s)$$

- For rationals:  $\gamma = \frac{r}{s}$

$$H_2(\gamma, 1-\gamma) = -\gamma \ln \gamma - (1-\gamma) \ln(1-\gamma)$$

$$f(n) = H_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right) = -\sum_{i=1}^n \gamma_i \ln \gamma_i =$$

$$= \sum_{i=1}^n \frac{1}{n} \ln \frac{1}{n} = n \cdot \frac{1}{n} \ln n = \ln n$$

$$\boxed{n = r \cdot s} \Rightarrow f(r,s) = \ln(r,s) = \ln r + \ln s$$

DA AHA ALSO  
DA ZNAČE  
DEFINICIJA NA  
ENTROPIJU, A  
ZAPRIMENJU ČAK  
NE ŠTA ZNAČI.

$$H_2(\gamma, 1-\gamma) = \frac{\gamma}{s} \ln \frac{s}{\gamma} + \left(1 - \frac{\gamma}{s}\right) \ln \frac{1}{1-\frac{\gamma}{s}} =$$

$$= \frac{\gamma}{s} \ln \frac{s}{\gamma} + \ln \frac{s}{s-\gamma} - \frac{\gamma}{s} \ln \frac{s}{s-\gamma} =$$

$$= \frac{\gamma}{s} \ln \frac{s}{\gamma} + \ln \frac{s}{s-\gamma} = \frac{\gamma}{s} \ln \left(\frac{s-\gamma}{s}\right) + \ln \frac{s}{s-\gamma}$$

$$= \left(1 - \frac{\gamma}{s}\right) \ln \frac{s}{s-\gamma} = \frac{s-\gamma}{s} \ln \frac{s}{s-\gamma}$$

$$h\left(\frac{y_2}{s_2}\right) = H\left(1 - \frac{y_2}{s_2}, \frac{y_2}{s_2}\right) \\ = H\left(\frac{y_1 y_2 - y_1}{s_2}, \frac{y_2}{s_2}\right) = H\left(\frac{y_1}{s_2}, \frac{y_2}{s_2}\right)$$

$$S_k = \sum_{i=1}^k \gamma_i$$

$$h(I) = H_2(1, 1-2)$$

$$S_3 = S_2 + y_3$$

$$H_m(y_1, \dots, y_m) = H_{m-1}(S_2, y_2, \dots, y_m) + S_2 h\left(\frac{y_1}{S_2}\right)$$

$$= H_{m-2}(S_3, y_4, \dots, y_m) + S_3 h\left(\frac{y_3}{S_3}\right) + S_2 h\left(\frac{y_1}{S_2}\right)$$

$$= H_{m-(k-1)}(S_k, y_{k+1}, \dots, y_m) + \sum_{i=2}^k S_i h\left(\frac{y_i}{S_i}\right) \quad \text{④}$$

- And so on, representing recursive algorithm

$$N_A: H_K(y_1/S_K, y_2/S_K, \dots, y_K/S_K)$$

$$H_K\left(\frac{y_1}{S_K}, \frac{y_2}{S_K}, \dots, \frac{y_K}{S_K}\right) = H_2\left(\frac{S_{K-1}}{S_K}, \frac{y_K}{S_K}\right) + \sum_{i=2}^{K-1} \frac{S_i}{S_K} h\left(\frac{y_i}{S_i}\right)$$

$$= H_2\left(\frac{S_K}{S_K}\right) + \sum_{i=2}^K \frac{S_i}{S_K} h\left(\frac{y_i}{S_i}\right) = \sum_{i=2}^K h\left(\frac{y_i}{S_i}\right)$$

$$\text{ZNAČÍ: } H_k\left(\frac{p_1}{s_k}, \frac{p_2}{s_k}, \dots, \frac{p_k}{s_k}\right) = \frac{1}{s_k} \sum_{i=1}^k \sin\left(\frac{p_i}{s_k}\right)$$

Ačo je záverečné vo  $\textcircled{A}$  je doslova

$$H(p_1, p_2, \dots, p_m) = H_{m-k+1}(s_k, p_{k+1}, \dots, p_m) + \underbrace{s_k H_k\left(\frac{p_1}{s_k}, \frac{p_2}{s_k}, \dots, \frac{p_k}{s_k}\right)}$$

- So všetkými dôkazovými postupmi je možné na každú kôtu

- CONSIDER:  $f(m, n) = H_m\left(\frac{1}{m}, \frac{1}{m}, \dots, \frac{1}{m}\right) =$

$$= H_{m-n+1}\left(\frac{1}{m}, \frac{1}{m}, \dots, \frac{1}{m}\right) + \underbrace{\frac{1}{m} H_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right)}_{s_k}$$

$$= H_{m-n+1}\left(\frac{1}{m}, \frac{1}{m}, \dots, \frac{1}{m}\right) + \underbrace{\frac{1}{m} H_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right)}_{s_k}$$

$$= H_{m-n+1}\left(\frac{1}{m}, \frac{1}{m}, \frac{1}{m}, \dots, \frac{1}{m}\right) + \underbrace{\frac{2}{m} H_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right)}_{s_k}$$

$$= H_m\left(\frac{1}{m}, \frac{1}{m}, \dots, \frac{1}{m}\right) + H_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right)$$

$$= f(n) + f(m) \quad \text{DOKAZANÉ !!!}$$

- Obrana automacie scénu:  $f(m^k) = k \cdot f(m)$

UŽIEĽNO INEASNO VYVODOVATEĽNÉ JE  
GLEDAJ AČO ZNAČÍ: DOKAŽE VYVODOVATEĽNÉ  
KÔTA COGAM RÁZOM:

$$H(m^k) = \sum_{i=1}^{m^k} \frac{1}{m^k} \ln m^k = \underbrace{m^k \cdot \frac{1}{m^k} \ln m^k}_{H(m)} = k \cdot \underline{\ln m}$$

- $H_2(1, 0) = G(1) = \emptyset$   $\boxed{p_1 + p_2 = 1}$

$$\begin{aligned} H_3(p_1, p_2, 0) &= H_2(p_1, p_2) + p_2 H_2(1, 0) = \\ &= H_2(p_1 + p_2, 0) + (p_1 + p_2) H_2\left(\frac{p_1}{p_1 + p_2}, \frac{p_2}{p_1 + p_2}\right) = \\ &= H_2(1, 0) + H_2(p_1, p_2) \end{aligned}$$

$$\Rightarrow p_2 H_2(1, 0) = H_2(1, 0) \Rightarrow p_2 = 0 \Rightarrow H_2(1, 0) = 0$$

①  $f(n+1) - f(n) \rightarrow 0$  as  $n \rightarrow \infty$  (DA SE DOKAWE  $\frac{1}{\infty}$ )

$$\begin{aligned}
 f(n+1) &= h\left(\frac{1}{n+1}, \frac{1}{n+1}, \dots, \frac{1}{n+1}\right) = h_{\underbrace{n+1-n+1}_2}\left(\frac{n}{n+1}, \frac{1}{n+1}\right) \\
 &+ \frac{n}{n+1} h_n\left(\underbrace{\frac{n}{n+1}}_{\frac{n}{n+1}}, \underbrace{\frac{n+1}{n+1}}_{\frac{n}{n+1}}, \dots, \underbrace{\frac{n+1}{n+1}}_{\frac{n}{n+1}}\right) = \\
 &= h_2\left(\frac{n}{n+1}, \frac{1}{n+1}\right) = h_2\left(1 - \frac{1}{n+1}, \frac{1}{n+1}\right) = h\left(\frac{1}{n+1}\right) = \\
 &= h\left(\frac{1}{n+1}\right) + \frac{n}{n+1} h_n\left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right)
 \end{aligned}$$

$$\boxed{f(n+1) = h\left(\frac{1}{n+1}\right) + \frac{n}{n+1} f(n)} \quad \boxed{f(n+1) f(n) = h\left(\frac{1}{n+1}\right) f(n)}$$

$$\boxed{f(n+1) - \frac{n}{n+1} f(n) = h\left(\frac{1}{n+1}\right) f(n)}$$

$$\lim_{n \rightarrow \infty} \frac{(n+1)f(n+1) - nf(n)}{n+1} = \lim h\left(\frac{1}{n+1}\right) f(n)$$

$$\lim_{n \rightarrow \infty} f(n+1) - \frac{n}{n+1} f(n) = f(n) - f(n) = 0 \Rightarrow h\left(\frac{1}{n+1}\right) f(n) = 0$$

• Let's define:

$$a_{n+1} = f(n+1) - f(n)$$

$$b_n = h\left(\frac{1}{n}\right) f(n)$$

$$\begin{aligned}
 a_{n+1} &= h\left(\frac{1}{n+1}\right) + \frac{n}{n+1} f(n) - f(n) = h\left(\frac{1}{n+1}\right) + \frac{n f(n) - f(n)}{n+1} \\
 &= h\left(\frac{1}{n+1}\right) + \frac{n f(n) - n f(n) - f(n)}{n+1} = h\left(\frac{1}{n+1}\right) - \frac{f(n)}{n+1}
 \end{aligned}$$

$$\boxed{a_{n+1} = -\frac{1}{n+1} f(n) + b_{n+1}}$$

$$a_{n+1} = -\frac{1}{n+1} \sum_{i=2}^n a_i + b_{n+1}$$

$$a_2 = f(2) - f(1) \quad a_3 = f(3) - f(2) \quad a_4 = f(4) - f(3)$$

$$a_n = f(n) - f(n-1) \quad \boxed{a_2 + a_3 + a_4 + \dots + a_n = f(2) - f(1) + f(3) - f(2) + f(4) - f(3) = -f(1) + f(n) = f(n) \quad 17}$$

$$(n+1) \cdot a_{n+1} = - \sum_{i=2}^n q_i + (b_{n+1}) \cdot (n+1) \quad (n+1) b_{n+1} = (n+1) a_{n+1} + \sum_{i=2}^n q_i$$

• Summing over  $n^{th}$

$$\sum_{n=2}^N n b_n = \sum_{n=2}^N n a_n + \sum_{n=2}^N \sum_{i=2}^n q_i = \\ = \sum_{n=2}^N (n a_n + q_{n-1} + q_{n-2} + \dots + q_2)$$

$$\boxed{n+1 = m} \quad m b_m = m a_m + \sum_{i=2}^{m-1} q_i \quad \sum_{n=2}^N$$

$$\sum_{m=2}^N m b_m = \sum_{m=2}^N m a_m + \sum_{m=2}^N \sum_{i=2}^{m-1} q_i =$$

$$= \sum_{m=2}^N m a_m + q_{m-1} + q_{m-2} + \dots + q_2 \Rightarrow$$

$$\boxed{\sum_{n=2}^N n b_n = \sum_{n=2}^N (n a_n + q_{n-1} + q_{n-2} + \dots + q_2)}$$

$$\sum_{q=2}^N q a_q + \sum_{n=2}^N q_{n-1} + \sum_{n=2}^N q_{n-2} + \dots + \underbrace{\sum_{n=2}^N q_2}_{(N-1) q_2}$$

$$\sum_{n=2}^N q_{n-2} = \begin{vmatrix} q_1 = 4-2 & n = n-2 \\ q = 2 & n = 0 \\ n = N & n = N-2 \end{vmatrix} = \sum_{n=0}^{N-2} q_n$$

$$\sum_{n=2}^N q_{n-1} = \begin{vmatrix} q = n-1 & n = 1 \\ q = 2 & n = 1 \\ n = N & n = N-1 \end{vmatrix} = \sum_{n=1}^{N-1} q_n$$

$$\sum_{n=2}^4 (n a_n + q_{n-1} + q_{n-2} + \dots + q_2) = 2 \underline{q_2} + 3 \underline{q_3} + \underline{q_2} + \\ 4 \underline{q_4} + \underline{q_3} + \underline{q_2} =$$

$$\sum_{n=2}^5 (n a_n + q_{n-1} + q_{n-2} + \dots + q_2) = 2 \underline{q_2} + 3 \underline{q_3} + \underline{q_2} + 4 \underline{q_4} + \underline{q_3} + \underline{q_2} + \\ 5 \underline{q_5} + \underline{q_4} + \underline{q_3} + \underline{q_2} = 5 \underline{q_2} + 5 \underline{q_3} + 5 \underline{q_4} + 5 \underline{q_5}$$

$$\sum_{n=2}^N (a_2 + a_{n-1} + a_{n-2} + \dots + a_1) = N \cdot \sum_{i=2}^N a_i \quad / \sum_{n=1}^N n = \frac{N(N+1)}{2}$$

$$\begin{aligned} S &= 1 + 2 + \dots + (n-1) + n \\ S &= n + n-1 + \dots + 2 + 1 \\ 2S &= (n+1) + (n+1) + \dots + (n+1) \end{aligned} \quad | \quad \boxed{S = \frac{n \cdot (n+1)}{2}}$$

$$\sum_{n=2}^N n b_n = N \sum_{i=2}^N a_i \quad / \quad \sum_{n=1}^N n = \frac{(n+1)n}{2}$$

$$\begin{aligned} S &= 2 + 3 + \dots + n \\ S &= n + n-1 + \dots + 2 \\ 2S &= (n+2) + (n+2) + \dots + (n+2) \end{aligned} \quad | \quad \boxed{S = \frac{(n-1) \cdot (n+2)}{2}}$$

$$\frac{\sum_{n=2}^N n b_n}{\sum_{n=1}^N n} = \frac{2}{N+1} \sum_{i=2}^N a_i$$

$$\sum_{n=1}^N n = \frac{(N-1)(N+2)}{2}$$

$$b_n = h\left(\frac{1}{n}\right)$$

$$\begin{aligned} h\left(\frac{1}{n}\right) &= \left(1 - \frac{1}{n}\right) \ln\left(1 - \frac{1}{n}\right) + \frac{1}{n} \ln\left(\frac{1}{n}\right)^{-1} = \\ &= \frac{n-1}{n} \ln\left(\frac{n-1}{n}\right) + \frac{1}{n} \ln n = \frac{n-1}{n} \ln \frac{n}{n-1} + \frac{1}{n} \ln n \\ &= \frac{n-1}{n} \ln n - \frac{n-1}{n} \ln(n-1) + \frac{1}{n} \ln n = \left(\frac{n-1}{n} + \frac{1}{n}\right) \ln n - \frac{n-1}{n} \ln(n-1) \\ &= \ln n - \frac{n-1}{n} \ln(n-1) = 0 \quad \text{cogitare e vero!} \quad \text{MASSIMA TENSIONE!} \\ \Rightarrow b_n &\rightarrow 0 \quad \text{IF} \quad n \rightarrow \infty \quad \Rightarrow \quad \frac{\sum_{n=2}^N n b_n}{\sum_{n=1}^N n} \rightarrow 0 \\ \Rightarrow a_{n+1} &= - \underbrace{\frac{1}{n+1} \sum_{i=2}^n a_i}_{\rightarrow 0} + b_{n+1} \quad \Rightarrow a_{n+1} \rightarrow 0 \Rightarrow \\ &\quad a_{n+1} = f(n+1) - f(n) = 0 \quad \text{domani!!!} \end{aligned}$$

LEMMA 2.0.1 Let the function  $f(n)$  satisfy the following assumptions:

- $f(mn) = f(m) + f(n)$
- $\lim_{n \rightarrow \infty} [f(n+1) - f(n)] = 0$
- $f(2) = 1$

Then the function  $f(n) = \log_2 n$ .

PROOF: Let  $p$  be smallest prime number and let:

$$g(n) = f(n) - \frac{f(p) \cdot \log n}{\log p}$$

$$g(m \cdot n) = f(m \cdot n) - \frac{f(p) \cdot \log(m \cdot n)}{\log p} =$$

$$= f(m \cdot n) - \underbrace{\frac{f(p) \log n}{\log p}}_{= f(p) \log(n/p)} + \frac{f(p) \log m}{\log p} =$$

$$f(m \cdot n) = f(n) + f(m)$$

$$= f(n) - \frac{f(p) \log n}{\log p} + f(n) - \frac{f(p) \log(n/p)}{\log p} =$$

$$= g(n) + g(n)$$

Znací,  $g(n)$  je konstantní funkce

očekáváme.

$$g(p) = f(p) - \frac{f(p) \cdot \log p}{\log p} = f(p) - f(p) = 0$$

$$\alpha_n = g(n+1) - g(n) = f(n+1) - \frac{f(p) \cdot \log(n+1)}{\log p} - f(n) + \frac{f(p) \log n}{\log p}$$

$$= f(n+1) - f(n) - \frac{f(p)}{\log p} \log \frac{n+1}{n}$$

$$\lim_{n \rightarrow \infty} \left[ f(n+1) - f(n) \right] - \lim_{n \rightarrow \infty} \left[ \frac{f(p)}{\log p} \log \frac{n+1}{n} \right] = 0$$

$$\Rightarrow \boxed{\lim_{n \rightarrow \infty} (\alpha_n) = 0}$$

$$\lim_{n \rightarrow \infty} (\ln x) = \ln \lim_{n \rightarrow \infty} x^n = 0$$

FOR AN INTEGER  $n$ , DEFINE

$$y^{(1)} = \left\lfloor \frac{y}{P} \right\rfloor$$

$$y^{(1)} < \frac{y}{P}$$

$$y = P \cdot y^{(1)} + l$$

$$0 \leq l \leq P$$

$$g(P) = \emptyset \quad g(P \cdot y^{(1)}) = g(y^{(1)})$$

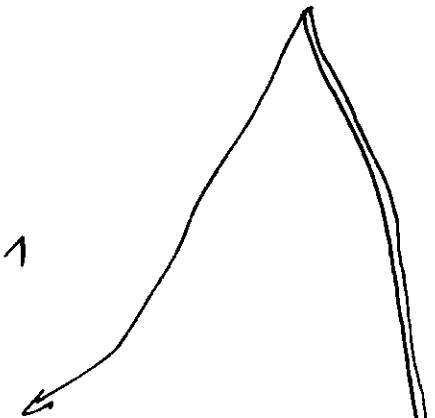
$$g(P \cdot y^{(1)}) = g(P) + g(y^{(1)}) = g(y^{(1)})$$

$$g(y) = \underbrace{g(y^{(1)})}_{\text{0}} - \underbrace{g(P \cdot y^{(1)})}_{\sum_{i=1}^{n-1} a_i} + \underbrace{g(y)}_{\sum_{i=P \cdot y^{(1)}}^{n-1} a_i} = \underbrace{g(y^{(1)})}_{\text{0}} + \sum_{i=P \cdot y^{(1)}}^{n-1} a_i$$

$$g(y) = g(y^k) + \sum_{i=1}^k \left( \sum_{i=P \cdot y^k}^{n-1} a_i \right)$$

$$y^{(k)} \leq \frac{y}{P^k}$$

$$k = \left\lceil \frac{\log y}{\log P} \right\rceil + 1$$



$$a_i = g(i+1) - g(i)$$

$$\sum_{i=P \cdot y^{(1)}}^{n-1} a_i = a_{P \cdot y^{(1)}} + \dots + a_{n-1} = g(P \cdot y^{(1)} + 1) - g(P \cdot y^{(1)}) + g(P \cdot y^{(1)}) - g(P \cdot y^{(1)} - 1) + \dots + g(n) - g(n-1) = g(P \cdot y^{(1)} + 1) - g(n-1)$$

• ACO DEFINICJA

$$a_i = g(i) - g(i-1)$$

$$\sum_{i=P \cdot y^{(1)}}^{n-1} a_i = g(P \cdot y^{(1)}) - g(P \cdot y^{(1)} - 1) + g(P \cdot y^{(1)} + 1) - g(P \cdot y^{(1)}) + g(n-1) - g(n-2) = g(P \cdot y^{(1)} - 1) - g(n-1)$$

• SIEOK NIE DA SO OWAJ DEFINICJA

$$\sum_{i=P \cdot y^{(1)}}^{n-1} a_i = g(P \cdot y^{(1)} + 1) - g(P \cdot y^{(1)}) + g(P \cdot y^{(1)} + 2) - g(P \cdot y^{(1)} + 1) + g(P \cdot y^{(1)} + 3) - g(P \cdot y^{(1)} + 2) + \dots + g(n) - g(n-1)$$

$$\sum_{i=P \cdot y^{(1)}}^{n-1} a_i = g(n) - g(P \cdot y^{(1)})$$

DZIĘKUJĘ!!!

- Vano sio go definicione  $u^{(k)}$  oo  $u$  et definito  
 $u^{(2)} \text{ oo } u^{(1)}$   $\sum_{i=P_{u^{(2)}}}^{n^{(1)}-1} a_i$

$$g(u^{(4)}) = g(u^{(2)}) + \sum_{i=P_{u^{(2)}}}^{n^{(1)}-1} a_i$$

$$g(u) = g(u^{(1)}) + \sum_{i=P_{u^{(1)}}}^{n^{(1)}-1} a_i = g(u^{(2)}) + \sum_{i=P_{u^{(2)}}}^{n^{(1)}-1} a_i + \sum_{i=P_{u^{(1)}}}^{n^{(1)}-1} t_i$$

$$g(u) = g(u^*) + \sum_{i=P_{u^*}}^{n^{(1)}-1} a_i + \dots + \sum_{i=P_{u^{(k)}}}^{n^{(k)}-1} a_i + \sum_{i=P_{u^{(k)}}}^{n^{(k)}-1} q_i =$$

$$= g(u^*) + \sum_{j=1}^k \sum_{i=P_{u^*}+j}^{n^{(k)}-1} q_i$$

$$u^{(k)} = \left\lfloor \frac{u}{P^k} \right\rfloor \quad u^{(k)} \leq \frac{u}{P^k} \quad u = P^k \cdot u^{(k)} + c$$

$$0 \leq c \leq P^k$$

$$k = \left\lceil \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} \right\rceil + 1 \quad \left\lceil \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} \right\rceil = k-1$$

$$k-1 = \left\lceil \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} \right\rceil \leq \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} \quad \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} \geq k-1 \quad \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} \geq k-1 \quad \boxed{u \geq P^{k-1}}$$

$$u^{(k)} = \left\lfloor \frac{P^{k-1}}{P^k} \right\rfloor = \left\lfloor \frac{1}{P} \right\rfloor = |P > 1| = \emptyset$$

$$u^{(k)} = \emptyset \Rightarrow g(u^{(k)}) = g(\emptyset) = \emptyset$$

$$g(u) = \sum_{i=1}^k q_i \quad \lfloor u \rfloor \leq P \cdot \left[ \frac{\lfloor u \rfloor}{\lfloor u \rfloor P} + 1 \right]$$

Since  $a_n \rightarrow 0 \Rightarrow \frac{g(u)}{\lfloor u \rfloor} \rightarrow 0$   $g(u)$  has at least

$\Theta(\log_2 u)$  terms  $a_i$

$$\lim_{u \rightarrow \infty} \frac{f(u)}{\lfloor u \rfloor} = \frac{f(P)}{\lfloor u \rfloor}$$

$$g(u) = f(u) - \frac{f(P) \lfloor u \rfloor}{\lfloor u \rfloor}$$

$$\frac{g(u)}{\lfloor u \rfloor} = \frac{f(u)}{\lfloor u \rfloor} - \frac{f(P)}{\lfloor u \rfloor}$$

• THIRD TERM

$$\frac{f(P)}{\lfloor u \rfloor} = c$$

$$\lim_{u \rightarrow \infty} \frac{f(u)}{\lfloor u \rfloor} = \frac{f(P)}{\lfloor u \rfloor}$$

$$\frac{f(2)}{\lfloor u \rfloor} = 1 \Rightarrow \boxed{f(P) = \lfloor u \rfloor}$$

$$N = p_1 p_2 \cdots p_L \quad f(N) = \sum_i f(p_i) = \sum_i \log p_i = \log N$$

SOME OF THE DNA PRIME NUMBERS NOTED  
 THAT ONE IS ALSO GOOD INTEGERS POSITIVE BESIDES.  
 (FUNDAMENTAL THEOREM OF ARITHMETIC)

- SO OVERLEMMA IS PROVEN.

- PROVE TWO SEPARATELY THAT  $f(n)$  IS MONOTONIC.
- ONLY FUNCTION  $f(m)$  SUCH THAT  $f(mn) = f(m) + f(n)$  FOR ALL INTEGERS  $m, n$  IS OF THE FORM  $f(n) = \log_b(n)$  FOR SOME BASE  $b > 1$ .

$$\text{LET } c = f(2). \quad f(4) = f(2) + f(2) = f(2 \cdot 2) = 2c$$

$$f(2^k) = k \cdot f(2) = k \cdot c = c \cdot \log_2 2^k$$

- EXTENSION FOR INTEGERS THAT ARE NOT POWERS OF 2.

FOR ANY INTEGER  $n$ , let  $r > 0$ , BE ANOTHER INTEGER  
 AND LET  $2^k \leq n^r < 2^{k+1}$

$$kc \leq r f(n) < (k+1)c \quad \frac{c \frac{k}{r}}{c} \leq f(n) < c \frac{k+1}{r}$$

$$\log 2^k \leq r \log n \leq (k+1) \quad \frac{k}{r} \leq \log n < \frac{k+1}{r}$$

$$\frac{c \frac{k}{r}}{c} \leq c \log n \leq c \frac{k+1}{r}$$

$$|x| = \begin{cases} x & x > 0 \\ -x & x < 0 \end{cases}$$

$$\frac{k}{r} \leq \frac{f(n)}{c} \leq \frac{k+1}{r}$$

$$\frac{k}{r \cdot c} \leq \frac{\log n}{c} \leq \frac{k+1}{r \cdot c}$$

$$\left| f(n) - \frac{\log n}{c} \right| = \begin{cases} f(n) - \frac{\log n}{c} & f(n) - \frac{\log n}{c} > 0 \\ \frac{\log(n) - f(n)}{c} & f(n) - \frac{\log n}{c} \leq 0 \end{cases}$$

$$\frac{k}{r} - \frac{k}{rc} \leq f(n) - \frac{\log n}{c} \leq \frac{k+1}{r} - \frac{k+1}{rc}$$

$$L = \frac{kc - k}{rc} \quad R = \frac{rc + c - k - 1}{rc} = \frac{c(k+1) - (k+1)}{rc} =$$

$$= \frac{(c-1)(k+1)}{rc}$$

$$\left| f(n) - \frac{\log n}{c} \right| < \frac{1}{r}$$

$\Rightarrow$  MOREOVER  $f(n) = \frac{\log n}{c}$

$$f(2) = \frac{\log(2)}{c} > 1 \quad c = f(2) = 1$$

- DOWARZEME DENS:

$$f(u) = \text{H}_m\left(\frac{1}{u}, \frac{1}{u_1}, \dots, \frac{1}{u_m}\right) = \text{ld} u$$

- DA DOWARZEME DENS:

$$\boxed{\text{H}_2(q, 1-q) = -q \text{ld} q - (1-q) \text{ld}(1-q)} \quad \#$$

- BEZ EXTENDED GROUPING ATION!

$$\boxed{\text{H}(p_1, p_2, \dots, p_n) = \text{H}_{n-k-n}(s_k, q_{n+k}, \dots, p_n) + s_k t_k \left( \frac{p_1}{s_k} \frac{p_2}{s_k} \dots \frac{p_k}{s_k} \right)}$$

$$f(s) = \text{H}_3\left(\frac{1}{s}, \dots, \frac{1}{s}\right) = \text{H}\left(\underbrace{\frac{1}{s}, \frac{1}{s}, \dots, \frac{1}{s}}_{s-r}, \frac{s-r}{s}\right) + \frac{s-r}{s} f(s-r) = \#$$

$s_k = (s-r) \cdot \frac{1}{s}$

$$f(s-r) = \text{H}\left(\frac{\frac{1}{s}}{\frac{s-r}{s}}, \frac{\frac{1}{s}}{\frac{s-r}{s}}, \dots, \frac{\frac{1}{s}}{\frac{s-r}{s}}\right)$$

$$= \text{H}_{s-r}\left(\frac{1}{s-r}, \frac{1}{s-r}, \dots, \frac{1}{s-r}\right)$$

$$\# = \text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) + \frac{r}{s} \text{H}_r\left(\frac{\frac{1}{s}}{\frac{r}{s}}, \frac{\frac{1}{s}}{\frac{r}{s}}, \dots, \frac{\frac{1}{s}}{\frac{r}{s}}\right) + \frac{s-r}{s} f(s-r)$$

$$= \text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) + \frac{r}{s} \text{H}_r\left(\frac{1}{r}, \frac{1}{r}, \dots, \frac{1}{r}\right) + \frac{s-r}{s} f(s-r)$$

$$= \text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) + \frac{r}{s} f(r) + \frac{s-r}{s} f(s-r)$$

P2S  $\Rightarrow$  #

$$f(s) = \text{ld} s \Rightarrow \text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) = \text{ld} s - \frac{r}{s} f(r) - \frac{s-r}{s} f(s-r) =$$

$$= \text{ld} s - \frac{r}{s} \text{ld} r - \frac{s-r}{s} \text{ld}(s-r) = \text{ld} s - \frac{r}{s} \text{ld} r - \text{ld}(s-r) + \frac{r}{s} \text{ld}(sr)$$

$\# = \text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) + \frac{r}{s} f(r) + \frac{s-r}{s} f(s-r)$ 
 $\text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) = f(s) - \frac{s-r}{s} f(s-r) \xrightarrow{\text{H}_2(s-r) + \text{ld} s - \frac{r}{s} \text{ld} r = \text{ld} s}$ 
 $\text{H}_2\left(\frac{r}{s}, \frac{s-r}{s}\right) f(s-r) = \left(1 - \frac{r}{s}\right) \text{ld} s - \left(1 - \frac{r}{s}\right) \frac{s-r}{s} f(s-r)$ 

TAKA BEZGUNA!! NG-HOZE KOLIGATOR PA GO POTCAAS!!

 $= \text{ld} s - \frac{r}{s} \text{ld} s + \frac{r}{s} \text{ld} s - \frac{r}{s} \text{ld} r - \left(1 - \frac{r}{s}\right) \text{ld}(s-r) =$ 
 $= \text{ld} s - \frac{r}{s} \text{ld} s - \frac{r}{s} \text{ld} s - \left(1 - \frac{r}{s}\right) \text{ld}(s-r) =$ 
 $^{24} = -\frac{r}{s} \text{ld} \frac{r}{s} + \left(1 - \frac{r}{s}\right) \text{ld} s - \quad \left(1 - \frac{r}{s}\right) \text{ld}(s-r) =$

$$= -\frac{r}{s} \ln \frac{r}{s} - \left(1 - \frac{r}{s}\right) \ln \frac{1}{s} - \left(1 - \frac{r}{s}\right) \ln(s-r) = ①$$

$$-\left(1 - \frac{r}{s}\right) = \left(\frac{s-r}{r} - 1\right) = \cancel{\frac{s-r}{r}} \quad -\left(1 - \frac{r}{s}\right) = \left(\frac{r}{s} - 1\right) = \frac{r-s}{s}$$

$$② = -\frac{r}{s} \ln \frac{r}{s} + \frac{s-r}{r} \ln \frac{1}{s} + \frac{r-s}{s} \ln(s-r) =$$

$$= -\frac{r}{s} \ln \frac{r}{s} + \frac{r-s}{r} \ln s + \frac{r-s}{s} \ln(s-r) =$$

$$= -\frac{r}{s} \ln \frac{r}{s} + \left(1 - \frac{r}{s}\right) \ln s + \left(\frac{r}{s} - 1\right) \ln(s-r)$$

$$p = \frac{r}{s} \quad \left(1 - \frac{r}{s}\right) \ln s + \left(\frac{r}{s} - 1\right) \ln(s-r)$$

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~~$\cancel{\ln s \ln s + (1-r) \ln(s-r)}$~~

~~$\cancel{\ln s \ln s + (1-r) \ln(s-r) + r \ln s + (s-r) \ln(s-r)}$~~

~~$\cancel{\ln s \ln s + (1-r) \ln(s-r) + (1-r) \ln(s-r)}$~~

~~$\cancel{\ln s \ln s + (1-r) \ln(s-r) =}$~~

~~$\cancel{\ln s \ln s + (1-r) \ln(s-r) + (1-r) \ln(s-r)}$~~

~~$\cancel{\ln s \ln s + (1-r) \ln(s-r) + (1-r) \ln(s-r)}$~~

~~$\cancel{\ln s \ln s + (1-r) \ln(s-r) = \ln s \left(\frac{1-r^2}{s} + r^2\right)} \quad ③$~~

$$H_2\left(\frac{v}{s}, \frac{s-v}{s}\right) = \ln s - \underbrace{\frac{v}{s} \ln r}_{\text{Faktor!!!}} - \frac{s-v}{s} f(s-r) = ④$$

$$= \ln s - \frac{v}{s} \ln r + \frac{v}{s} \ln s - \frac{v}{s} \ln s - \frac{s-v}{s} \ln(s-r) =$$

$$= \ln s - \frac{v}{s} \ln r - \frac{v}{s} \ln \left(\frac{1}{s}\right) - \frac{v}{s} \ln s - \frac{s-v}{s} \ln(s-r) =$$

$$= \ln s \left(1 - \frac{r}{s}\right) - \frac{v}{s} \ln \frac{r}{s} - \frac{s-v}{s} \ln(s-r) =$$

$$= -\frac{r}{s} \ln \frac{r}{s} - \left(1 - \frac{r}{s}\right) \ln \frac{1}{s} - \left(1 - \frac{r}{s}\right) \ln(s-r) =$$

$$= -\frac{r}{s} \ln \frac{r}{s} - \left(1 - \frac{r}{s}\right) \ln \frac{s-v}{s} = -\frac{r}{s} \ln \frac{r}{s} - \left(1 - \frac{r}{s}\right) \ln \left(1 - \frac{v}{s}\right)$$

DOKAEND

• Assume that N26.④ holds for  $n=4-1$

$$H_7(q_1, q_2, \dots, q_n) = \underline{H_{n-1}(q_1+q_2, q_3, \dots, q_n)} + (q_1+q_2) H_2\left(\frac{q_1}{q_1+q_2}, \frac{q_2}{q_1+q_2}\right)$$

$$= - \sum_{i=3}^n q_i \text{ld} q_i + (q_1+q_2) \text{ld} \frac{1}{q_1+q_2} + \cancel{(q_1+q_2)} \cdot \frac{q_1}{q_1+q_2} \text{ld} \frac{q_1+q_2}{q_1} +$$

$$+ \cancel{(q_1+q_2)} \frac{q_2}{q_1+q_2} \text{ld} \frac{q_1+q_2}{q_2} = - \sum_{i=2}^n q_i \text{ld} q_i + (q_1+q_2) \text{ld} \frac{1}{q_1+q_2}$$

$$+ \underline{q_1 \text{ld} (q_1+q_2)} - \underline{q_1 \text{ld} q_1} + \underline{q_2 \text{ld} (q_1+q_2)} - \underline{q_2 \text{ld} q_2}$$

$$= - \sum_{i=3}^n q_i \text{ld} q_i - (q_1+q_2) \text{ld} (q_1+q_2) + \cancel{(q_1+q_2) \text{ld} (q_1+q_2)} + \sum_{i=1}^2 q_i \text{ld} q_i$$

$$= \sum_{i=1}^n q_i \text{ld} q_i$$

DOKAZANOV  $\Rightarrow$  DOKAZANIE  $\Rightarrow$  DOKAZANIE

DOKAZATE ZELENKI  $\Rightarrow$  DOKAZANIE SE ISOLCEV  
 1.) DOKAZANIE  $\Rightarrow$  VLASTNOST  
 2.) ADO VLAST

16.06.

$$\begin{array}{r|rr|l} & n=4-1 & \text{VLAST} & \text{1 do } n \\ \hline 2772997 & | 3141776 & \div 9 & \end{array}$$

08.08.2012

2.47

NUMBER OF MISORTED PCK. A DECK OF  $n^2$

CARDS IN ORDER  $1, 2, \dots, n$  IS PROVIDED.  
 ONE CARD IS REMOVED AT RANDOM, AND  
 THEN REARRANGED AT RANDOM. WHAT IS THE  
 NUMBER OF MISORTED CARDS?

$$H(n) = \sum_{i=1}^n \frac{1}{n} \text{ld} n = n! \text{ld} n = \underline{\underline{n}}$$

$\underbrace{(n-1) \cdot \text{ld} n}_{\cdot 1 \cdot 2 \cdot \dots \cdot n}$

PLAVICHE  
CARTI NA N27

1234	1234	1234	1234
2134*	2134*	3214	4231
3214	1324*	1324*	1432
4231	1432	1243*	1243*

ADRESA  
 6 MISORTED KOMARACI  
 KOJ SE PODER ALKO VELIKI  
 $4 \cdot 3 = 12$  MISORTED KOMARACI  
 CI. ADRESANT  $2(n-1) = 6$

$n \cdot (n-1)$  BILO VOD OD N-TE BROJVI MOZE DA SE SVAKI  
NA  $(n-1)$  MESTA

$2(n-1) \Rightarrow$  OD NAPOTE  $n(n-1)$  MA  $\frac{1}{2}(n-1)$  SREDNJI  
PAZNI I SVAKI SREDJEN KES MA L DULJEVA  
TI.

$$P(x) = \begin{cases} \frac{1}{4^2} & \text{Case I: } x(i)=j, x(j)=i, x(k)=k \quad |i-j|>1 \\ \frac{2(n-1)}{4^2} & \text{Case II: } x(i)=i, x(j)=i, x(k)=k \quad |i-j|=2 \\ \frac{1}{4} & \text{Case III: } x(k)=k \text{ for all } k \end{cases}$$

$$\bullet n(n-1) - 2(n-1) = n^2 - n - 2n + 1 = n^2 - 3n + 1 = \\ \downarrow = (n-2)(n-1)$$

INSTANCES OF  
CASE I

DULJEVATE SI BROJIS EDASI

•  $(n-1)$  INSTANCES OF CASE II

- • (1) INSTANCE OF CASE III (THE NUMBERS TREAT ITS OWN STAVCI, 1 = STAVCI TA 1, VIDI'S 2) (TRENT PARE, I.E. NO MISORDERED FIGURES)
- Operacija III C' VODI ZA 20! TO KOMADA MOZE DA VRATI VO NEZVONO KRETANJE MESTO
- NE SE DOBIM PRIMERITE NA P.26 (NE SE ANO IZVELEI'S 2 KETI, I MESTA ZAKRIVI'S MESTOVA)

1 2 3		
2 1 3	2 1 3	3 1 2
2 3 1	1 2 2	1 3 2

1 2 3 4	/ / / /	/ / / /	/ / / /
2 1 3 4	* 2 1 3 4	3 1 2 4	4 1 2 3
2 3 1 4	1 3 2 4	* 1 3 2 4	1 4 2 3
2 3 4 1	1 3 4 2	1 2 3 4	* 1 2 4 3

④ PRESENT CASES

$$(n-2)(n-1)$$

$$H(x) = \sum_{i=1}^{n-1} \frac{1}{4^2} \ln \frac{n^2}{4^2} + \sum_{k=1}^{(n-1)} \frac{2}{4^2} \ln \frac{n^2}{2^2} + \frac{1}{4} \ln n$$

$$= \frac{(n-2)(n-1)}{4^2} \ln \frac{n^2}{4^2} + \frac{2(n-1)}{4^2} \ln \frac{n^2}{2^2}$$

$$(n-1) + (n-2)(n-1) + 1 = n - 1 + n^2 - 3n + 1 = n^2 - 2n + 1 = (n-1)^2$$

$$P(x=x_1) = \frac{(n-2)(n-1)}{(n-1)^2} = \frac{n-2}{n-1}$$

$$P(x=x_2) = \frac{2}{(n-1)^2}$$

$$\begin{aligned}
 P(X=x_1) &= \frac{n-2}{n-1} && \text{veličina je } x_1, \text{ misorting case I} \\
 P(X=x_2) &= \frac{1}{n-1} && \text{veličina je } x_2, \text{ misorting case II} \\
 P(X=x_3) &= \frac{n}{(n-1)^2} && \text{veličina je } x_3, \text{ misorting case III} \\
 \bullet \text{ samo za even misorting od orden tip} \\
 \gamma_1 &= \frac{P(X=x_1)}{(n-2)(n-1)} = \frac{1}{(n-1)^2} && ? \\
 \gamma_2 &= P(X=x_2)
 \end{aligned}$$

Vidimo da je:  $x_1, x_2, \dots, x_{n-1}, x_n$

$$\begin{aligned}
 n(n-1) + n &= h^2 - h + h = h^2 \\
 P(X \in x_1) &= \frac{(n-2)(n-1)}{h^2} & P(X \in x_2) &= \frac{2(n-1)}{h^2} \\
 P(X \in x_3) &= \frac{n}{h^2} & (n-2)(n-1) + 2(n-1)
 \end{aligned}$$

$$\bullet \text{ samo za even misorting od orden tip} \\
 \gamma_1 = P(X=x_1) = \frac{P(X \in x_1)}{(n-2)(n-1)} = \frac{1}{h^2} \quad \gamma_2 = P(X=x_2) = \frac{P(X \in x_2)}{(n-1)}$$

$$\gamma_3 = P(X \in x_3) = \frac{1}{h} \quad \rightarrow \text{DUPLICATIJE} \\
 \text{za záležitost} \quad \text{veličina}$$

$$\boxed{\gamma_1 = \frac{1}{h^2} ; \gamma_2 = \frac{2}{h^2} ; \gamma_3 = \frac{1}{h}}$$

$$H(X) = (n-1)(n-2) \cdot \frac{1}{h^2} \ln h^2 + \underbrace{(n-1) \cdot \frac{2}{h^2}}_{\text{bez duplikatov}} \cdot \ln \frac{h^2}{2} + \underbrace{\frac{1}{h} \ln h}_{\text{bez duplikatov}}$$

$$\begin{aligned}
 &= \frac{2(n-1)(n-2)}{h^2} \ln h + \frac{4(n-1) \ln h}{h^2} - \frac{2(n-1) \ln 2}{h^2} + \frac{1}{h} \ln h = \\
 &= \ln(h) \left[ \frac{2(n-1)(n-2) + 4(n-1) + 4}{h^2} \right] - \frac{n-1}{h^2} \ln 4
 \end{aligned}$$

$$2(n-1)(n-2) + 4(n-1)+n = 2(n^2 - 2n - n + 2) + 4n - 4 + n = \\ = 2n^2 - 6n + 4 + 5n - 4 = 2n^2 - n$$

$$H(x) = \frac{n(2n-1)}{n^2} \text{ld} n - \frac{n-1}{n^2} \text{ld} 4$$

$$H(x) = \frac{2n-1}{n} \text{ld} n - \frac{n-1}{n^2} \text{ld} 4$$

MOREZ ZAVACOVÁ MESA DA SE VIKA: EXAMPLE OF TAKING OUT AND RETURNING BACK THE FILE FROM ARCHIVE.

**2.48** SEQUENCE LENGTH; HOW MUCH INFORMATION DOES THE LENGTH OF THE SEQUENCE GIVE ABOUT THE CONTENT OF THE SEQUENCE? Suppose

WE CONSIDER A BEINULLI ( $\frac{1}{2}$ ) PROCESS  $\{X_i\}$ . STOP THE PROCESS WHEN THE FIRST 1 APPEARS. LET "N" DESIGNATE THIS STOPPING TIME. THIS  $X_N$  IS AN ELEMENT OF THE SET OF ALL FINITE-LENGTH BINARY SEQUENCES  $\{0,1\}^* = \{0,1,00,01,10,11,000, \dots\}$ .

(a) FIND  $I(N; X^N)$ . (b) FIND  $H(X^N/N)$  (c) FIND  $H(X^N)$   
LET'S NOW CONSIDER A DIFFERENT STOPPING TIME. FOR THIS PART, AGAIN ASSUME THAT  $X_i$  IN BEINULLI ( $\frac{1}{2}$ ) BUT STOP AT TIME  $N=6$ , WITH PROBABILITY  $1/3$  AND STOP AT TIME  $N=12$  WITH PROBABILITY  $2/3$ .  
LET THIS STOPPING TIME BE INDEPENDENT OF THE SEQUENCE  $X_1, X_2, \dots, X_{12}$ .

(d) FIND  $I(N; X^N)$ . (e) FIND  $H(X^N/N)$ . (f) FIND  $H(X^N)$ .

$$(a) P(N) = \underbrace{\frac{1}{2} \cdot \frac{1}{2} \cdots \frac{1}{2}}_{N-1} \left(1 - \frac{1}{2}\right) = \left(\frac{1}{2}\right)^N$$

$$H(N) = \sum_{N=1}^{\infty} \left(\frac{1}{2}\right)^N \text{ld} 2^N = \text{ld} 2 \sum_{N=1}^{\infty} N \left(\frac{1}{2}\right)^N = \text{ld} 2 \frac{\frac{1}{2}}{\left(1 - \frac{1}{2}\right)^2}$$

$$H(N) = \text{ld} 2 \frac{\frac{1}{2}}{\frac{1}{4}} = (\text{ld} 2) \cdot 2 = \underline{2 \text{ld} 2} = 2$$

$$I(N; X^N) = H(N) - \underbrace{H(N|X^N)}_{\text{NO SO ZAHEJ } X^N \text{ SO ENAES } "N"} = H(N) = 2 \text{ld} 2 = 2$$

$$(6) \quad H(X^N|N) = ? \quad I(N; X^N) = H(N) \quad H(\underbrace{N|X^N}_{\emptyset}) = H(N)$$

$$H(X^N|N) = H(X_1|N) + H(X_2|N, X_1) + \dots + H(X_N|N, X_1, \dots, X_{N-1})$$

$$X^N \in \{x_1, x_2, \dots, x_N\}$$

$$H(X^N|N) = \sum_{i=1}^N P(i) \cdot H(X^N|N_i)$$

$$I(N; X^N) = [H(X^N) - H(\cancel{X^N}|N)] = N - H(X^N|N)$$

$$H(X^N) = N \cdot H(X_i) = N \cdot [-q \ln q - (1-q) \ln(1-q)] = N$$

$$H(X^N|N) = N - I(N; X^N) = N - 2 \cancel{\ln 2}$$

$$X^N \in \{0, 1, 00, 01, 10, 11, 000, 001, 010, 011, 100, 101, 110, 111\}$$

$$N=1 \Rightarrow X^N \in \{1, 10, 11, 100, 101, 110, 111\}$$

$$X^N \in \{0, 1\}$$

$$\boxed{N=2} \quad X^N \in \{0, 1, 00, 01, 10, 11\}$$

$$H(X^N|N=2) = \emptyset \quad \begin{array}{l} \text{ako } X^N \text{ e ograniceno na } N=2 \\ \text{2. } X^N \text{ je ograniceno na } N=2 \\ \text{1. nema nevernost.} \end{array}$$

$$\boxed{N=3} \quad X^N \in \{0, 1, 00, 01, 10, 11, 000, 001, 010, 011, 100, 101, 110, 111\}$$

$$\rightarrow H(X^N|N=2) = P(00) \ln \frac{1}{P(00)} + P(01) \ln \frac{1}{P(01)} + P(10) \ln \frac{1}{P(10)} + P(11) \ln \frac{1}{P(11)} = \left(\frac{1}{4} \cdot \ln 4\right) \cdot 4 = \frac{1}{2} \cdot 4 = 2$$

$$H(X^N|N=4) = \left(\frac{1}{2^4} \ln 2^4\right) \cdot 2^4 = 4 \ln 2 = 4$$

$$H(X^N|N) = \sum_{n=1}^N p(n) H(X^N|N=n) = \sum_{n=1}^N \frac{1}{2^n} n$$

$$S = \sum_{n=1}^N n x^n \quad \int x^n dx = \frac{x^{n+1}}{n+1} \quad \int x^{n-1} dx = \frac{x^n}{n}$$

$$\frac{S}{x} = \sum_{n=1}^N n x^{n-1} / \int \Rightarrow \int \frac{S}{x} dx = \sum_{n=1}^N n \frac{x^n}{n} =$$

$$\int \frac{S dx}{x} = \sum_{n=1}^N x^n$$

$$S = x + x^2 + \dots + x^N$$

$$\frac{S}{x} = 1 + x + \dots + x^{N-1}$$

$$S - \frac{S}{x} = x^N - 1$$

$$S \left(1 - \frac{1}{x}\right) = x^N - 1$$

$$S = \frac{x^N - 1}{x - 1} = \frac{x^{N+1} - x}{x - 1}$$

$$\int \frac{S dx}{x} = \frac{x^{N+1} - x}{x - 1}$$

$$\frac{S}{x} = \left( \frac{x^{N+1} - x}{x - 1} \right)' = \frac{[(N+1)x^N - 1](x-1) - x^{N+1} + x}{(x-1)^2}$$

$$S = x \cdot \frac{(N+1)x^N(x-1) - x^{N+1} + x^{N+1} - x}{(x-1)^2} = x \frac{(N+1)x^{N+1} - (N+1)x^{N+1} + x}{(x-1)^2}$$

$$S = x \frac{Nx^{N+1} - Nx^{N+1} - Nx^N + x^{N+1} - x^{N+1}}{(x-1)^2} = x \frac{Nx^{N+1} - (N+1)x^{N+1} + x}{(x-1)^2}$$

$$S = x \cdot \frac{x^N(Nx - N - 1) + 1}{(x-1)^2}$$

$$x = \frac{1}{2}$$

$$S = \frac{1}{2} \frac{\frac{1}{2^N} \left(\frac{N}{2} - N - 1\right) + 1}{\frac{1}{4}} = 2 \left[ \frac{1}{2^N} \left(-\frac{N}{2} - 1\right) + 1 \right]$$

$$S = 2 - \frac{1}{2^N} \left(1 + \frac{N}{2}\right)$$

$$S = 2 - \frac{N+2}{2^N}$$

$$S = 2 - N \cdot 2^{-N} - 2^{1-N}$$

$$H(x^N|N) = 2 - \frac{N+2}{2^N}$$

$$N \rightarrow \infty \quad H(x^N|N) = 2$$

$$(c) I(N; x^N) = H(x^N) - H(x^N|N) = 2 \Rightarrow$$

$$H(x^N) = 2 + H(x^N|N) = 4 - \frac{N+2}{2^N}$$

$$\Rightarrow (N=1) \quad H(x^N|1) = 2 - \frac{3}{2} = 0.5$$

$$H(x^N|1) = P(N=1) \cdot H(x^N|N=1) = \frac{1}{2} \cdot \left[ \frac{1}{2} \log_2 + \frac{1}{2} \log_2 \right] = \frac{1}{2}$$

$$H(x^N) = 4 - \frac{N+2}{2^N}$$

$$(d) P(N=6) = \frac{1}{3} \quad P(N=12) = \frac{2}{3}$$

$$H(N) = P(N=6) \cdot \text{ld} \frac{1}{P(N=6)} + P(N=12) \cdot \text{ld} \frac{1}{P(N=12)} = \\ = \frac{1}{3} \text{ld} 3 + \frac{2}{3} \text{ld} \frac{1}{2} = \frac{1}{3} \text{ld} 3 + \frac{2}{3} \text{ld} 2 - \frac{2}{3} \text{ld} 2 \\ = \text{ld} 3 - \frac{2}{3} \text{ld} 2 \quad H(N) = \text{ld} 3 - \frac{2}{3} \text{ld} 2 = \text{ld} 3 - \frac{2}{3}$$

$$\boxed{I(N; X^N) = H(N) - H(N|X^N) = H(N) = \text{ld} 3 - \frac{2}{3} = 0,92}$$

$$(e) H(X^N|N) = ? \quad I(N; X^N) = H(X^N) - H(X^N|N)$$

$$H(X^N|N=6) = -P(000000) \text{ld} P(000000) - P(000001) \text{ld} P(000001) - \\ - P(000010) \text{ld} P(000010) - P(000011) \text{ld} P(000011) - \dots - P(111111) \text{ld} P(111111)$$

$$H(X^N|N=6) = \left( \frac{1}{2^6} \text{ld} 2^6 \right) \cdot 2^6 = \text{ld} 2^6 = 6$$

$$H(X^N|N=12) = \left( \frac{1}{2^{12}} \text{ld} 2^{12} \right) \cdot 2^{12} = 12$$

$$H(X^N|N) = P(N=6) \cdot H(X^N|N=6) + P(N=12) H(X^N|N=12) = \\ = \frac{1}{3} \cdot 6 + \frac{2}{3} \cdot 12 = 2 + 8 = \underline{\underline{10}}$$

$$H(X^N|N) = 10$$

$$(d) H(X^N) = ? \quad H(X^N) = I(N; X^N) + H(X^N|N) \Rightarrow$$

$$H(X^N) = \text{ld} 3 - \frac{2}{3} + 10 = \text{ld} 3 + \frac{30-2}{3} = \text{ld} 3 + \frac{28}{3}$$

$$\boxed{H(X^N) = 10,92}$$

### CHAPTER 3 ASYMPTOTIC EQUIVALENTS

$$\frac{1}{n} \text{ld} \frac{1}{P(x_1, x_2, \dots, x_n)} \rightarrow H \quad x_1, x_2, \dots, x_n \\ P(x_1, x_2, \dots, x_n) \quad \quad \quad \quad \quad P(x_1, x_2, \dots, x_n)$$

$$\left( P(x_1, x_2, \dots, x_n) \right)^{\frac{1}{n}} \rightarrow 2^H \quad \left[ P(x_1, x_2, \dots, x_n) \right]^{\frac{1}{n}} = 2^{-H} \quad ( )^n$$

$$\boxed{P(x_1, x_2, \dots, x_n) \rightarrow 2^{-H}}$$

## SE NARRATIVE ON PROBLEM 2.48

$$H(x^n) = 4 - \frac{n+2}{2^n}$$

$$P(0,000000000001) = \frac{1}{2^{12}} \quad 2^{-4H} = \frac{4^H + \frac{n+2}{2^n}}{2}$$

• TYPICAL SET WHERE SAMPLE OUTPUT IS CLOSE TO THE TRUE OUTPUT.

• Example. Let random variables  $x_i \in \{0, 1\}$  have probabilistic mass function  $\gamma(1) = p, \gamma(0) = q$ . If  $x_1, x_2, \dots, x_n$  are i.i.d. probability of  $x_1, x_2, \dots, x_n$  is  $\prod p(x_i)$ .

• e.g.:  $(1, 0, 1, 1, 0, 1) \xrightarrow{\text{by def}} P(X) = p^4 \cdot q^2 = p^{\sum x_i} q^{n - \sum x_i}$   
NOT ALL SEQUENCES OF LENGTH  $n=6$  HAS SAME PROBABILISTICITY.

$$\Pr\{x_1, x_2, \dots, x_n : \gamma(x_1, x_2, \dots, x_n) = 2^{-n(H+\epsilon)}\} = 1$$

$$\gamma(x_1, x_2, \dots, x_n) = \prod p^{\sum x_i} q^{n - \sum x_i}$$

NUMBER OF 1'S IN SEQUENCE IS CLOSE TO  $n \cdot p$  AND ALL SUCH SEQUENCES HAVE RUGHTLY THE SAME PROBABILITIY

$$x_1 x_2 x_3 x_4 x_5 \quad \underline{p = 0.6} \quad \underline{q = 0.4}$$

$$n = 5$$

$$\boxed{n \cdot p = 0.6 \cdot 5 = 3}$$

$$\boxed{4 \cdot 2 = 5 \cdot 0.4 = 2}$$

$$p^3 \cdot q^2 = (0.6)^3 \cdot (0.4)^2 = 0.03456 \quad 2^{-5 \cdot H(p)} = 0.03575$$

DEFINITION: CONVERGENCE OF RANDOM VARIABLES Given A SEQUENCE OF RANDOM VARIABLES  $x_1, x_2, \dots$ , WE SAY THAT SEQUENCE  $x_1, x_2, \dots$  CONVERGES TO RANDOM VARIABLE  $X$ :

1. In probability, if  $\forall \epsilon > 0, \Pr\{|x_n - X| > \epsilon\} \rightarrow 0$ .
2. In mean square, if  $E[(x_n - X)^2] \rightarrow 0$ .
3. (With probability 1) (also called almost surely) if  $\Pr\{\lim_{n \rightarrow \infty} x_n = X\} = 1$

### 3.1 Asymptotic Equitability Property Theory

Theorem 3.1.1 (AEP) If  $x_1, x_2, \dots$  are i.i.d.  $\sim p(x)$

THEN :  $-L \ln p(x_1, x_2, \dots, x_n) \xrightarrow{a.s.} H(x)$

PROBABILITY OBNEZAO GLEBOS A1 ISKEDVANITA VO YAKO MNOGOCYATIYE

$$-L \ln p(x_1, x_2, \dots, x_n) = -\frac{1}{n} [ \ln p(x_1) + \ln p(x_2) + \dots + \ln p(x_n) ] =$$

$$= -\frac{1}{n} \sum_{i=1}^n \ln p(x_i) \xrightarrow{\text{in probability}} -E[\ln p(x_i)] = H(x)$$

SAMPLE AVERAGE

EXPECTED VALUE OF  
 $\ln p(x_i)$

LAW OF LARGE NUMBERS

$$\Rightarrow \ln p(x_1, x_2, \dots, x_n) \xrightarrow{a.s.} H(x) \quad p(x_1, x_2, \dots, x_n) = 2^{-n \cdot H(x)}$$

$$p(x_1, x_2, \dots, x_n) = 2^{-n \cdot H(x)} \quad (*)$$

PROBABILITAT VARIETAT SEQUENCE SO YOP  
BUD NA EDIFICI

DEFINITION: The typical set  $A_\epsilon^{(n)}$  with respect to  $p(x)$  is the set of sequences  $(x_1, x_2, \dots, x_n)$  such that  $H(x) + \epsilon \leq \ln p(x_1, x_2, \dots, x_n) \leq H(x) - \epsilon$

THE TYPICAL SET IS SO WHERE SAMPLE ENTROPY CLOSE TO REAL ENTROPY.

Theorem 3.1.2 As a consequence of AEP, we can show that the set  $A_\epsilon^{(n)}$  has the following properties.

(1) If  $(x_1, x_2, \dots, x_n) \in A_\epsilon^{(n)}$ , then  $H(x) - \epsilon \leq -\frac{1}{n} \ln p(x_1, x_2, \dots, x_n) \leq H(x) + \epsilon$

(2)  $\Pr\{A_\epsilon^{(n)}\} \geq 1 - \epsilon$  for  $n$  sufficiently large

(3)  $|A_\epsilon^{(n)}| \leq 2^{n(H(x)+\epsilon)}$  where  $|A|$  denotes the number of elements in the set  $A$ .

(4)  $|A_\epsilon^{(n)}| \geq (1-\epsilon) 2^{n(H(x)-\epsilon)}$

Thus, the typical set has probability nearly 1.

All elements of typical set are nearly equiprobable, and number of element is nearly  $2^{nH(x)}$

$$\text{PROOF: (1)} - u(h(t)+\epsilon) \leq \text{ld} p(x_1, x_2, \dots, x_n) \leq -u(h(t)-\epsilon)$$

$$u(h(t)+\epsilon) \geq -\text{ld} p(x_1, x_2, \dots, x_n) \geq u(h(t)-\epsilon) \Rightarrow$$

$$h(t)-\epsilon \leq -\frac{1}{n} \text{ld} p(x_1, x_2, \dots, x_n) \leq h(t)+\epsilon$$

$$(2) \oplus \Rightarrow p(x_1, x_2, \dots, x_n) = 2^{-u(h(t))}$$

$$\lim_{n \rightarrow \infty} p(x_1, x_2, \dots, x_n) = \lim_{n \rightarrow \infty} 2^{-u(h(t))} = \underline{\frac{1}{2^{u(h(t))}}}$$

• FOR ANY  $\delta > 0$  THERE EXIST  $n_0$  SUCH THAT  
 $u \geq n_0$  WE HAVE:

$$\Pr \left\{ \left| -\frac{1}{n} \text{ld} p(x_1, x_2, \dots, x_n) - h(t) \right| < \epsilon \right\} > 1 - \delta.$$

$$000, \underbrace{001}_a, \underbrace{010}_b, \underbrace{011}_c, \underbrace{100}_1, \underbrace{101}_2, \underbrace{110}_3, 111$$

$$\begin{cases} r = 0.7 \\ g = 0.3 \end{cases}$$

$$0.7^2 \cdot 0.3 = \underline{0.49 \cdot 0.3}$$

$$\oplus (0.7)^2 \cdot 0.7 = \underline{0.49 \cdot 0.7} = \underline{0.063}$$

$$(3) 1 = \sum_{x \in X^n} p(x) \geq \sum_{x \in A_E^{(n)}} p(x) \Rightarrow \sum_{x \in A_E^{(n)}} 2^{-u(h(t)+\epsilon)} =$$

$$= |A_E^{(n)}| \cdot 2^{-u(h(t)+\epsilon)}$$

⇒  $2^{-u(h(t)+\epsilon)} \geq |A_E^{(n)}|$   
 i.e.  $|A_E^{(n)}| \leq 2^{-u(h(t)+\epsilon)}$

$$(4) (1-\epsilon) \leq \Pr \{ A_E^{(n)} \} \leq \sum_{x \in A_E^{(n)}} 2^{-u(h(t)-\epsilon)} = |A_E^{(n)}| \cdot 2^{-u(h(t)-\epsilon)}$$

$$|A_E^{(n)}| \geq (1-\epsilon) 2^{-u(h(t)-\epsilon)}$$

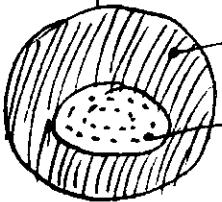
Dоказано!!!

### 3.2 CONSEQUENCES OF THE AEP: DATA COMPRESSION

In  $A_E^{(n)}$  there are  $\leq 2^{u(h(t))}$  sequences which can be indexed by  $u(h(t)+\epsilon)+1$  bits. Prefix  $\oplus$  for each sequence  $\Rightarrow$  LENGTH OF INDEXING NUMBER  $\leq u(h(t)+\epsilon)+2$  bits

NON-TERMINAL SET:  $u(\text{ld}(t)) + 2$  bits

TERMINAL SET:  $u(h(t)+\epsilon) + 2$  bits



- $x^n$  denotes a sequence  $x_1, x_2, \dots, x_n$
- $l(x^n)$  - length of the codeword corresponding to  $x^n$
- If  $n$  is sufficiently large so that  $P(A_E^{(n)}) > 1 - \epsilon$  the expected length of codeword is:

$$E[l(x^n)] = \sum_{x^n} p(x^n) \cdot l(x^n)$$

$$\begin{aligned} E[l(x^n)] &= \sum_{x^n} p(x^n) l(x^n) = \sum_{x^n \in A_E^{(n)}} p(x^n) l(x^n) + \sum_{x^n \in A_E^{(n)c}} p(x^n) l(x^n) \\ &\leq \sum_{x^n \in A_E^{(n)}} p(x^n) (n(H+\epsilon) + 2) + \sum_{x^n \in A_E^{(n)c}} p(x^n) (n(Hd|x|) + 2) \\ &= P_V\{A_E^{(n)}\} (n(H+\epsilon) + 2) + P_V\{A_E^{(n)c}\} (n(Hd|x|) + 2) \leq \\ &n(H+\epsilon) + \epsilon n(Hd|x|) + 2 = n(H+\epsilon) \\ \epsilon' &= \epsilon \cdot Hd|x| + \frac{2}{n} + \epsilon \end{aligned}$$

**Theorem 3.2.1** Let  $x^n$  be i.i.d  $\sim \gamma(x)$ . Let  $\epsilon > 0$   
 Then there exists a code that maps  $x^n$  of length  $n$  into smaller strings such that the mapping is one-to-one (and therefore invertible) and:

$$E\left[\frac{1}{n} l(x^n)\right] \leq H(x) + \epsilon$$

Hence we can represent sequence  $x^n$  using  $nH(x)$  bits on average.

### 3.3 HIGH PROBABILITY SETS AND TYPICAL SET

From the definition of  $A_E^{(n)}$  it is clear that  $A_E^{(n)}$  is ~~FAIRLY~~ small set that contains most of the probability. ~~INCIDENTALLY~~ it is the smallest set that contains the same number of elements as the largest set, to pass off  $\epsilon$  in statement.

DEFINITION: For each  $n = 1, 2, \dots$  let  $B_\delta^{(n)} \subset X^n$  be the smallest set with:

$$P\{B_\delta^{(n)}\} \geq 1 - \delta$$

We argue that  $B_\delta^{(n)}$  must have significant intersection with  $A_E^{(n)}$  and therefore must have about

AS MANY ELEMENTS. IN THEOREM 3.11 WE OBTAIN  
THE PROOF OF FOLLOWING THEOREM:

**Theorem 3.3.1.** Let  $x_1, x_2, \dots, x_n$  be i.i.d.  $\sim f(t)$ ,  
for  $\delta < \frac{1}{2}$  and  $\delta' \geq 0$ , if  $\Pr\{B_{\delta'}^{(n)}\} > 1 - \delta$  then  
 $\frac{1}{n} \ln(B_{\delta'}^{(n)}) > H - \delta'$  for  $n =$  SUFFICIENTLY LARGE.

Thus,  $B_{\delta'}^{(n)}$  must have at least  $\boxed{2^H}$  elements  
to FIRST ORDER in the exponent.  $A_{\delta'}^{(n)}$  has  
 $2^{n(H(\lambda) + \epsilon)}$  elements.

DEFINITION: The notation  $a_n \doteq b_n$  means

$$\lim_{n \rightarrow \infty} \left[ \frac{1}{n} \ln \frac{a_n}{b_n} \right] = 0$$

This implies that  $a_n$  &  $b_n$  are EQUAL TO  
THE FIRST ORDER IN THE EXPONENT.

• Therefore, if  $\delta_n \rightarrow 0$  &  $\epsilon_n \rightarrow 0$  then  
 $|B_{\delta_n}^{(n)}| \doteq |A_{\epsilon_n}^{(n)}| = 2^{nH}$  071309700

### 3.1 Markov Inequality & Chebyshev Inequality.

(a) (Markov's inequality) For ANY NONNEGATIVE RANDOM VARIABLE  $X$ , AND ANY  $t > 0$ , SHOW THAT EXIST RANDOM VARIABLE THAT ACHIEVES THIS INEQUALITY WITH EQUALITY

$$\Pr\{X \geq t\} \leq \frac{E[X]}{t}$$

(b) (Chebyshev's inequality) Let  $\gamma$  be a random variable WITH MEAN  $\mu$  AND VARIANCE  $\sigma^2$ . By letting  $X = (\gamma - \mu)^2$  SHOW THAT FOR ANY  $\epsilon > 0$

$$(a) \Pr\{X \geq t\} = 1 - \Pr\{X \leq t\} \quad \Pr\{|\gamma - \mu| > \epsilon\} \leq \frac{\sigma^2}{\epsilon^2}$$

$$E[X] = \sum_{i=1}^n p(x_i) \cdot x_i$$

$$x \in \{1, 2, \dots, n\}$$

$$p(x) = \{p_1, p_2, \dots, p_n\}$$

$$\Pr(X > t) = 1 - \Pr(X \leq t) = 1 - \sum_{i=1}^t p_i \cdot \gamma_i$$

$$\frac{E[X]}{t} = \frac{1}{t} \sum_{i=1}^t x_i p_i \geq 0$$

$$\frac{E[X]}{t} \geq \Pr(X > t)$$

$$\frac{1}{t} \sum_{i=1}^t x_i p_i \geq 1 - \sum_{i=1}^t x_i \gamma_i = \sum_{i=t+1}^n x_i \gamma_i$$

$$\sum_{i=1}^t x_i \gamma_i \geq + \sum_{i=t+1}^n x_i \gamma_i$$

$$\sum_{i=t+1}^n x_i \gamma_i - \sum_{i=t+1}^n t \gamma_i + \sum_{i=1}^t x_i \gamma_i$$

$$\sum_{i=t+1}^n (1-t) x_i \gamma_i + \sum_{i=1}^t x_i \gamma_i \geq 0 ?$$

$$\sum_{i=1}^t x_i \gamma_i \geq + - + \sum_{i=t+1}^n x_i \gamma_i$$

$$\sum_{i=1}^t x_i \gamma_i + \sum_{i=t+1}^n x_i \gamma_i + t \sum_{i=1}^t x_i \gamma_i \geq t.$$

$$(1+t) \sum_{i=1}^t x_i \gamma_i + \sum_{i=t+1}^n x_i \gamma_i \geq t$$

$$\frac{1}{t} \sum_{i=1}^t x_i \gamma_i = \frac{1}{t} \left[ \underbrace{\sum_{i=1}^t x_i \gamma_i}_{\gamma_1, \gamma_2, \dots, \gamma_t, \dots, \gamma_n} + \underbrace{\sum_{i=t+1}^n x_i \gamma_i}_{\gamma_{t+1}, \dots, \gamma_n} \right]$$

$$x \in \{1, 2, \dots, t, \dots, n\}$$

$$\gamma = \{\gamma_1, \gamma_2, \dots, \gamma_t, \dots, \gamma_n\}$$

$$E[X] = \sum_{i=1}^n i \cdot \gamma_i = p_1 + 2p_2 + 3p_3 + \dots + tp_t + \dots + np_n$$

$$\frac{E[X]}{t} = \frac{p_1}{t} + \frac{2p_2}{t} + \dots + \frac{tp_t}{t} + \dots + \frac{n p_n}{t} \geq$$

$$p_1 + p_2 + \dots + p_t + \dots + \frac{n}{t} p_n$$

$$1 - \sum_{i=1}^t \gamma_i = 1 - \gamma_1 - \gamma_2 - \gamma_3 - \dots - \gamma_n$$

$$\Pr(X > t) = \sum_{i=t+1}^n p_i = \gamma_{t+1} + \gamma_{t+2} + \dots + \gamma_n$$

$$\frac{E[X]}{t} = \frac{\gamma_1}{t} + \frac{2\gamma_2}{t} + \dots + \gamma_t + \frac{(t+1)\gamma_{t+1}}{t} + \dots + \frac{n}{t}\gamma_n.$$

$$\leq \frac{\gamma_1}{t} + \frac{2\gamma_2}{t} + \dots + \gamma_t + \underbrace{(\gamma_{t+1} + \dots + \gamma_{n-1})}_{\Pr(X > t)}$$

$$x \in [t, 2t, 3t, \dots, nt]$$

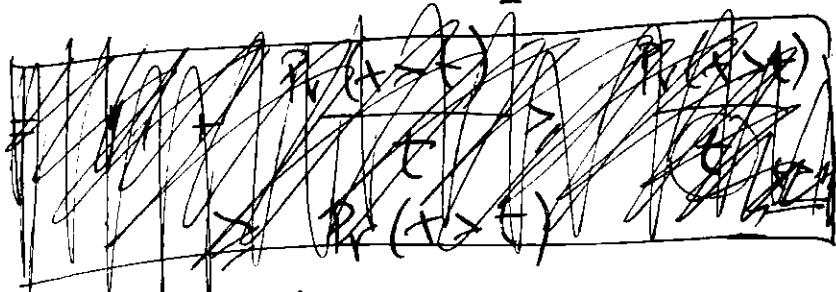
$$\frac{E[X]}{t} = \sum_{i=1}^n i p_i$$

$$\Pr(X > t) = \sum_{i=2}^n i t p_i$$

$$\frac{E[X]}{t} = \sum_{i=1}^n i p_i$$

$$\Pr(X > t) = t + \sum_{i=2}^n i p_i$$

$$\frac{E[X]}{t} = \gamma_1 + \sum_{i=2}^n i p_i$$



$$\frac{E[X]}{t} = \frac{\gamma_1}{t} + \frac{2\gamma_2}{t} + \dots + \gamma_t + \frac{(t+1)\gamma_{t+1}}{t} + \dots + \frac{n-1}{t}\gamma_{n-1}$$

$$\geq \underbrace{\gamma_1 + \gamma_2 + \dots + \gamma_t}_{1 - \Pr(X > t)} + \frac{(t+1)\gamma_{t+1}}{t} + \dots + \frac{n-1}{t}\gamma_{n-1}$$

$$\frac{E[X]}{t} = \frac{1}{t} \cdot \sum_{i=1}^n i = \frac{1}{t} \cdot \frac{t \cdot n+1}{2} = \boxed{\frac{n+1}{2}}$$

$$\frac{E[X]}{t} = \frac{1}{t} \cdot \frac{n+1}{2}$$

$$x \in [1, 2, \dots, t, \dots, n] \quad \Pr(X > t) = \sum_{i=t+1}^n p_i$$

$$\Pr(X > t) = \gamma_{t+1} + \gamma_{t+2} + \dots + \gamma_{(n)}$$

$$\frac{E[X]}{t} = \frac{\gamma_1}{t} + \frac{2\gamma_2}{t} + \dots + \frac{(t+1)\gamma_{t+1}}{t} + \dots + \frac{n}{t}\gamma_n \leq \frac{\gamma_1}{t} + \frac{2\gamma_2}{t} + \dots + \gamma_{t+1} + \dots + \gamma_n$$

• FORMA UNIFORME

$$x \in \{1, 2, 3, \dots, n\}$$

$$\gamma(x) = \left\{\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right\}$$

$$\Pr\{x > t\} = (n-t) \cdot \frac{1}{n}$$

$$\begin{aligned} E[x] &= \frac{1}{n} \sum_{i=1}^n i \cdot \frac{n(n+1)}{2} \\ E[x] &= \frac{n+1}{2} \end{aligned}$$

$$\frac{E[x]}{t} = \frac{n+1}{2t} \quad \text{④}$$

$$\frac{n+1}{2t} \geq \frac{(n-t)}{n}$$

$$n^2 + n \geq 2nt - 2t^2$$

$$n^2 + n + 2t^2 - 2nt \geq 0$$

$$f(t) = 2t^2 - 2nt + n(n+1) \quad \text{JPSSDP}$$

VO OVO  
SUCATNE  
VAG  
NEGATIVNO.

• PODEMOS GEOMETRICA

$$x \in \{1, 2, 3, \dots, n\}$$

$$\gamma(x) = \left\{\frac{1}{2}, \frac{1}{2}, \dots, \frac{1}{2^{n-1}}\right\}$$

$$S = 1 + 2 + 2^2 + \dots + 2^n$$

$$2S = 2^2 + 2^3 + \dots + 2^{n+1}$$

$$S(1-2) = 2 - 2^{n+1}$$

$$S = \frac{2(1 - \frac{1}{2^n})}{1 - 2}$$

$$S = \left(1 - \frac{1}{2^n}\right)$$

$$S = \left(1 - \frac{1}{2^{n-1}}\right) + \frac{1}{2^{n-1}} = 1$$

$$\Pr(x > t) = 1 - \left(1 - \frac{1}{2^n}\right) = \frac{1}{2^n}$$

MATE!!

$$E[x] = \sum_{i=1}^{n-1} i \cdot \gamma(i) = \sum_{i=1}^{n-1} i \cdot \frac{1}{2^i} + \frac{n}{2^{n-1}} = \frac{n}{2^{n-1}} - \frac{1}{2^{n-1}} + 2 + \frac{n}{2^{n-1}}$$

$$E[x] > 2 - \frac{1}{2^{n-1}}$$

~~$$\frac{1}{2^{n-1}} = \frac{2}{2^n}(2^{-n}) + \frac{2}{2^{n-1}}(2^{-1})$$~~

$$\mathbb{E}[X] = \frac{1}{2} + \frac{1}{2^{n+1}} \sum_{i=1}^n i$$

$$\frac{\mathbb{E}[X]}{t} = \frac{2}{t} - \frac{1}{t \cdot 2^{n+1}}$$

$$P_V(X \geq t) = \frac{1}{2} t$$

$$\frac{2}{t} - \frac{1}{t \cdot 2^{n+1}} \geq \frac{1}{2} t$$

$$2 \cdot 2^t - 2^{t-n+1} \geq t$$

$$\frac{2 \cdot 2^t}{t} - \frac{2^t}{t \cdot 2^{n+1}} \geq 1$$

$$\left[ 2^{t-1} - 2^{t-n+1} - t \geq 0 \right]$$

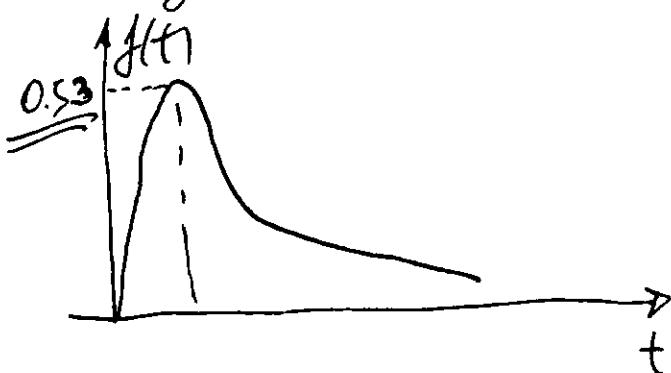
$$\frac{2^t}{t} \left[ 2 - \frac{1}{2^{n+1}} \right] \geq 1$$

$$\left( 2 - \frac{1}{2^{n+1}} \right) \geq \left[ \frac{t}{2^t} \right]$$

$$\left( 2 - \frac{1}{2^{n+1}} \right) \geq t \cdot 2^{-t}$$

$g(n)$

$f(t)$



$$f(0) = 0$$

$$f(1) = \frac{1}{2}$$

$$f(2) = \frac{2}{2^2} = \frac{1}{4}$$

$$g(1) = g(2) = 1$$

$$g(3) = 2 - \frac{1}{8} = \frac{7}{8}$$

$$g(4) = 2 - \frac{1}{16} = \frac{15}{16}$$

VAS RETTANGOLARE!!!

• SE CARATTERI DI UNIFORMITÀ 4 N13.40

$$P\{X \geq t\} = \frac{(n-t+1)}{n}$$

$$\frac{n+1}{2t} \geq \frac{n-t+1}{n}$$

$$n+t > 2t \cdot n - 2t^2 + 2t$$

$$2t^2 - 2(n+1)t + n(n+1) \geq 0$$

(STOJO SE DOPPIA VERSO 1 ESSO E'  $\Pr\{X \geq t\}$ )

• SEGA DA GRANDEZA EA:

$$\mathbb{E}[X] = \sum_{i=1}^n x_i \cdot p(i) = \sum_{i=1}^{n-1} \frac{1}{2^{n-i}} \cdot \frac{1}{2^n} \cdot \frac{1}{2^{n-i}} =$$

$$x = \left\{ \frac{1}{2^{n-i}} : i = 1 \dots n \right\}$$

$$P_i = \left\{ \frac{1}{2^i} : i = 1 \dots n-1 \right\}, \frac{1}{2^n}$$

$$\sum_{i=1}^{n-1} \frac{1}{2^i} = \sum_{i=1}^n \frac{1}{2^i} = \frac{n-1}{2^n} + \frac{1}{2^n} = \frac{n}{2^n}$$

$$\boxed{\mathbb{E}[X] = \frac{n}{2^n}}$$

$$\frac{E[X]}{t} = \frac{n+1}{t \cdot 2^n}$$

$P_r[X > t] = ?$     $P_r[X > t] = \frac{1}{2^t}$

$X = \left[ \frac{1}{2^{n-1}}, \frac{1}{2^{n-2}}, \dots, \frac{1}{2^2}, \frac{1}{2^1}, \frac{1}{2^0} \right]$

$P(X) = \left[ \frac{1}{2^1}, \frac{1}{2^2}, \dots, \frac{1}{2^{n-2}}, \frac{1}{2^{n-1}}, \frac{1}{2^n} \right]$

$t = \frac{1}{2^{n-m}}$     $t = \frac{1}{2^n} \cdot 2^m$

$2^m = t \cdot 2^n$

$$2^m = \frac{1}{256} \cdot 2^{10} = 4$$

$$m = \log_2 4 = 2$$

$$\boxed{P_r[X > t] = \frac{1}{t \cdot 2^n}}$$

$$\frac{n+1}{t \cdot 2^n} \geq \frac{1}{t \cdot 2^m} \quad n+1 \geq 1 \quad \boxed{m \geq 0}$$

1 no over success after reassignment.

$$X = [1, 2, \dots, t, \dots, n] \quad \frac{E[X]}{t} = \frac{1}{t} \frac{\sum_{i=1}^n i}{\sum_{i=1}^n 1}$$

$$\boxed{\frac{E[X]}{t} = \frac{n+1}{2t}}$$

$\underbrace{\frac{1}{t}, \frac{2}{t}, \dots, \frac{n}{t}}_{\{ \} \text{ OPT.1}}$

$$q = [q_1, q_2, \dots, q_t, \dots, q_n]$$

$$\boxed{\sum_{i=1}^n p_i = \underbrace{q_1 + q_2 + \dots + q_t}_{\frac{n+1}{2} \{ \} \text{ OPT.1}} = \frac{n+1}{2 \cdot t}}.$$

$$q = \left[ \frac{1}{n}, \frac{1}{n-1}, \dots, \frac{1}{t}, \dots, \frac{1}{2} \right]$$

$$\boxed{P_r[X \geq t] \leq \frac{E[X]}{t}}$$

$$\sum_{i=t}^n p_i \leq \frac{1}{t} \sum_{i=1}^n p_i \cdot x_i$$

$$t \cdot \sum_{i=t}^n \gamma_i \leq \sum_{i=1}^{t-1} \gamma_i x_i + \sum_{i=t}^n \gamma_i x_i$$

$$\sum_{i=t}^n (\gamma_i - \gamma_t) x_i \leq \boxed{\sum_{i=1}^{t-1} \gamma_i x_i}$$

$$\boxed{\gamma_1 x_1 + \gamma_2 x_2 + \dots + \gamma_{t-1} x_{t-1} + (\gamma_t - \gamma_t) x_t + (\gamma_t - \gamma_{t+1}) x_{t+1} + \dots + (\gamma_t - \gamma_n) x_n} > 0$$

$$x \in [x_1, x_2, \dots, x_t, \dots, x_n] = [x_1, x_2, \dots, \cancel{x_t}, \dots, \cancel{x_n}]$$

$$\gamma \in [\gamma_1, \gamma_2, \dots, \gamma_t, \dots, \gamma_n] = [\gamma_1, \gamma_2, \dots, \gamma_t, \dots, \gamma_n]$$

$$\frac{E[x]}{t} = \frac{1}{t} \sum_{i=1}^{t-1} \gamma_i x_i + \sum_{i=t}^n \gamma_i x_i = \sum_{i=1}^{t-1} \gamma_i x_i + \gamma_t \sum_{i=t}^n x_i$$

### EDITION 1 SOLUTION

(a) IF  $X$  has distribution  $F(x)$

$$E[X] = \int_0^\infty x dF = \int_0^\delta x dF + \int_\delta^\infty x dF \geq \int_\delta^\infty x dF$$

$$\geq \int_\delta^\infty F dF = F \cdot \Pr[x \geq \delta]$$

$$\Pr[x \geq \delta] \leq \frac{E[X]}{\delta}$$

Proved!!

$$E[X] = E[X | X \leq \delta] + E[X | X > \delta] \cdot \Pr[X > \delta]$$

$$E[X] = E[X | X \geq \delta] \cdot \Pr[X \geq \delta] + E[X | X < \delta] \cdot \Pr[X < \delta]$$

$$E[X | Y=\gamma] = \sum_{x \in X} x \cdot P(X=x | Y=\gamma) = \sum_{x \in X} x \cdot \frac{P(X=x, Y=\gamma)}{P(Y=\gamma)}$$

$$E[X | X \leq \delta] = \sum_{x \in X} x \cdot \underline{\gamma(x | X \leq \delta)} = \sum_{x \in X} x \cdot \frac{\gamma(x, X \leq \delta)}{\gamma(X \leq \delta)}$$

$$E[X] = \sum_{x \in X} x \cdot \gamma(x) = \sum_{x \in X} x \cdot \sum_n \gamma(x | n)$$

$$\rightarrow E[X] \geq E[X | X \geq \delta] \cdot \Pr[X \geq \delta] \geq \delta \cdot \Pr[X \geq \delta] \text{ JAKO!!!}$$

## CONDITIONAL EXPECTATION AND TOTAL EXPECTATION

- If  $R$  is random variable and " $E$ " is event then conditional expectation is defined by:

$$Ex(R|E) = \sum_{\omega \in E} R(\omega) \cdot Pr(\omega|E)$$

- For example, let  $R$  be the number that comes up on a roll of a fair die, and let " $E$ " be the event that number is even. Let's compute  $Ex(R|E)$ , the expected value of a die roll, given that the result is even.

$$Ex(R|E) = \sum_{\omega \in E} R(\omega) \cdot Pr(\omega|E) = 1 \cdot 0 + 2 \cdot \frac{1}{3} + 3 \cdot 0 + 4 \cdot \frac{1}{3} + 5 \cdot 0 + 6 \cdot \frac{1}{3} = \\ = \frac{2}{3} + \frac{4}{3} + \frac{6}{3} = \frac{12}{3} = 4$$

## Theorem (Total Expectation)

$$Ex(R) = Ex(R|E_1) \cdot Pr(E_1) + Ex(R|E_2) \cdot Pr(E_2) + \dots + Ex(R|E_n) \cdot Pr(E_n)$$

ANALOGOUS TO TOTAL PROBABILITY THEOREM

$$P(B) = P(B|A_1) \cdot P(A_1) + P(B|A_2) \cdot P(A_2) + \dots + P(B|A_n) \cdot P(A_n)$$

- For example, let  $R$  be the number that comes up on fair die, and " $E$ " is the event that the result is even. Then the event  $\bar{E}$  is the event that the result is odd. So total expectation theory says:

$$Ex(R) = \underbrace{Ex(R|E)}_{1/2} \cdot \underbrace{Pr(E)}_{1/2} + \underbrace{Ex(R|\bar{E})}_{?} \cdot \underbrace{Pr(\bar{E})}_{1/2}$$

$\begin{matrix} 1+1 \\ 3+1 \\ 5+2 \end{matrix}$

$$Ex(R|\bar{E}) = 1 \cdot \frac{1}{3} + 2 \cdot 0 + 3 \cdot \frac{1}{3} + 4 \cdot 0 + 5 \cdot \frac{1}{3} + 6 \cdot 0$$

$$Ex(R|\bar{E}) = \frac{1+3+5}{3} = \frac{9}{3} = 3$$

$$\boxed{Ex(R) = 4 \cdot \frac{1}{2} + 3 \cdot \frac{1}{2} = 2 + \frac{3}{2} = \frac{4+3}{2} = \frac{7}{2}} \quad 4$$

$E R$	1	2	3	4	5	6	7	8	9	10	$P(E)$
$E$	0	$1/4$	0	$1/16$	0	$1/64$	0	$1/256$	0	$1/1024$	$\frac{1+1}{1024}$
$\bar{E}$	$1/2$	0	$1/8$	0	$1/32$	0	$1/128$	0	$1/512$	0	$\frac{3+1}{512}$
$P(R)$	$1/2$	$1/4$	$1/8$	$1/16$	$1/32$	$1/64$	$1/128$	$1/256$	$1/512$	$1/1024$	1

$P(R|E)$

14016  
6400

604

77.28.101.27

E	1	2	3	4	5	6	7	8	9	10
E	0	$\frac{512}{1715} \cdot \frac{1}{4}$	0	$\frac{512}{1715} \cdot \frac{1}{16}$	0	$\frac{512}{1715} \cdot \frac{1}{64}$	0	$\frac{512}{1715} \cdot \frac{1}{256}$	0	$\frac{512}{1715} \cdot \frac{1}{1024}$
E	$\frac{512}{1715} \cdot \frac{1}{2}$	0	$\frac{512}{1715} \cdot \frac{1}{8}$	0	$\frac{512}{1715} \cdot \frac{1}{32}$	0	$\frac{512}{1715} \cdot \frac{1}{128}$	0	$\frac{512}{1715} \cdot \frac{1}{512}$	0

$$E_x(R) = \underbrace{E_x(R|E)}_{?} \cdot \underbrace{P_R(E)}_{\frac{171}{512}} + \underbrace{E_x(R|\bar{E})}_{?} \cdot \underbrace{P_R(\bar{E})}_{\frac{341}{512}}$$

[CONTINUE FROM PP. 43]

- For given  $\delta$ , the distribution achieving:

$$Pr\{X \geq \delta\} = \frac{E[X]}{\delta} \quad \text{i.e.: } X = \begin{cases} \delta & \text{with prob. } \frac{1}{\delta} \\ 0 & \text{with prob. } 1 - \frac{1}{\delta} \end{cases}$$

where  $\mu \leq \delta$

$$E[X] = \delta \cdot \frac{1}{\delta} + 0 \cdot (1 - \frac{1}{\delta}) \quad \boxed{E[X] = \mu}$$

$$Pr\{X \geq \delta\} = \frac{\mu}{\delta} \quad \boxed{\frac{\mu}{\delta} = \frac{1}{\delta}}$$

- Likewise with ex. rate probability distro:

$$\delta: \quad X = \begin{cases} \delta & \text{with probability } \mu \\ 0 & \text{with probability } 1 - \mu \end{cases}$$

$$E[X] = \delta \cdot \mu + 0 \cdot (1 - \mu) = \delta \mu$$

$$Pr\{X \geq \delta\} = \mu \quad \Rightarrow \quad \mu = \frac{\delta \mu}{\delta} \quad \boxed{\mu = \mu}$$

(b) Chebyshev inequality: Let  $\tau$  be a random variable with mean  $\mu$  and variance  $\delta^2$ . By letting  $X = (\tau - \mu)^2$ , show that for any  $\epsilon > 0$

$$Pr\{|\tau - \mu| > \epsilon\} \leq \frac{\delta^2}{\epsilon^2}$$

$$\boxed{Pr\{X \geq \delta\} \leq \frac{E[X]}{\delta}}$$

$$\rightarrow Pr\{[\tau - \mu]^2 > \epsilon^2\} \leq Pr\{(\tau - \mu)^2 \geq \epsilon^2\} \leq \frac{E[(\tau - \mu)^2]}{\epsilon^2}$$

$$E[(\tau - \mu)^2] = \delta^2 \rightarrow Pr\{[\tau - \mu]^2 > \epsilon^2\} \leq \frac{\delta^2}{\epsilon^2} \quad \text{using Chebyshev}$$

$$Pr\{(\tau - \mu)^2 > \epsilon^2\} = Pr\{(\tau - \mu) > \epsilon\} \rightarrow Pr\{(\tau - \mu) > \epsilon\} \leq \frac{\delta^2}{\epsilon^2}$$

PROOFED!!!

(c) (SEEOR IN EDITION 1) PROVE THE WEAK LAW OF LARGE NUMBERS. Let  $Z_1, Z_2, \dots, Z_n$  be a sequence of i.i.d. random variables with mean  $\mu$  and variance  $\sigma^2$ . Let  $\bar{Z}_n = \frac{1}{n} \sum_{i=1}^n Z_i$  be a sample mean.

SHOW THAT:

$$P_r\{| \bar{Z}_n - \mu | > \varepsilon\} \leq \frac{\sigma^2}{n\varepsilon^2}$$

Thus  $P_r\{| \bar{Z}_n - \mu | > \varepsilon\} \rightarrow 0$  as  $n \rightarrow \infty$ . This is known as THE WEAK LAW OF LARGE NUMBERS.

$$E[\bar{Z}_n] = \frac{1}{n} E\left[\sum_{i=1}^n Z_i\right] = \frac{1}{n} \sum_{i=1}^n E[Z_i] = \frac{1}{n} : n \cdot \mu = \mu$$

$$\begin{aligned} \text{Var}[\bar{Z}_n] &= E[(\bar{Z}_n - \mu)^2] = E[\bar{Z}_n^2 - 2\mu \bar{Z}_n + \mu^2] = \\ &= E[\bar{Z}_n^2] - 2\mu E[\bar{Z}_n] + \mu^2 = \underline{E[\bar{Z}_n^2]} - \mu^2 \end{aligned}$$

$$\begin{aligned} E[\bar{Z}^2] &\stackrel{1}{=} E[(z_1 + z_2) \cdot (z_1 + z_2)] = \frac{1}{4} E[z_1^2 + 2z_1 z_2 + z_2^2] \\ &= \frac{1}{4} \{E[z_1^2] + 2\underbrace{E[z_1 z_2]}_{\text{ANALOGO}} + E[z_2^2]\} = \frac{1}{4} \underbrace{E[z_1^2]}_{\text{NE STATISTISCHE}} = \frac{1}{2} (\sigma^2 + \mu^2) \end{aligned}$$

$$\text{FOR } n = \underline{E[\bar{Z}^2]} = \frac{1}{n^2} \cdot (\sigma^2 + \mu^2) = \frac{1}{n} (\sigma^2 + \mu^2)$$

$$\text{Var}[\bar{Z}] = \frac{1}{n} (\sigma^2 + \mu^2) - \mu^2$$

$$\begin{aligned} \sigma^2 &= E[(z_1 - \mu)^2] = E[z_1^2] - 2\mu E[z_1] + \mu^2 = \underline{E[z_1^2]} - \mu^2 \\ \bullet \text{Var}[\bar{Z}_n] &= E\left[\left(\frac{1}{n} \sum_{i=1}^n z_i - \mu\right)^2\right] = \frac{1}{n^2} E\left[\sum_{i=1}^n z_i^2\right] - 2\mu E\left(\frac{1}{n} \sum_{i=1}^n z_i\right) + \mu^2 \end{aligned}$$

$$\boxed{\text{Var}[\bar{Z}_n] = \frac{1}{n^2} E\left[\sum_{i=1}^n z_i^2\right] - \mu^2}$$

VIELMehr  $\leftarrow$  POSSIBEL!!!

$$E\left[\left(\sum_{i=1}^2 z_i\right)^2\right] = E[(z_1 + z_2)(z_1 + z_2)] = E[z_1^2] + 2\underbrace{E[z_1 z_2] + E[z_2^2]}_{\text{ANALOGO}}$$

$$E\left[\left(\sum_{i=1}^2 z_i\right)^2\right] = 2 \cdot (\sigma^2 + \mu^2) \quad \text{ANALOGO: } E\left[\left(\sum_{i=1}^n z_i\right)^2\right] = n \cdot (\sigma^2 + \mu^2)$$

$$\text{Var}\left[\bar{Z}_n\right] = \frac{1}{n^2} n \cdot (\sigma^2 + \mu^2) - \mu^2 = \frac{\sigma^2}{n} + \frac{\mu^2}{n} - \mu^2$$

$$\bar{z}_4 = \frac{1}{4} \sum_{i=1}^4 z_i$$

$$E[\bar{z}_4] = \mu \quad E[(z_i - \mu)^2] = \sigma^2 \quad \text{(*)}$$

$$E[\bar{z}_4] = \frac{1}{4} \cdot \sum_{i=1}^4 E[z_i] = \frac{1}{4} \cdot 4 \cdot \mu = \mu$$

~~$$E[(\bar{z}_4 - \mu)^2] = E[\bar{z}_4^2] - 2\mu E[\bar{z}_4] + \mu^2 = E[\bar{z}_4^2] - \mu^2$$~~

~~$$E[(\bar{z}_4 - \mu)^2] = E\left[\frac{1}{4^2} \sum_{i=1}^4 (z_i - \mu)^2\right] = \frac{1}{4^2} \cdot 4 \cdot E[z_i^2]$$~~

~~$$= E\left[\left(\frac{1}{4} \sum_{i=1}^4 z_i - \mu\right)^2\right] = E\left[\left(\frac{1}{4} \sum_{i=1}^4 z_i - \frac{1}{4} \sum_{i=1}^4 \mu\right)^2\right] =$$~~

~~$$= E\left[\frac{1}{4} \sum_{i=1}^4 (z_i - \mu)\right]^2 = \frac{1}{4^2} \sum_{i=1}^4 E[(z_i - \mu)^2] = \frac{1}{4^2} \cdot 4 \cdot \sigma^2 = \frac{5}{4}$$~~

OM È DOPO TREPUNZE!!!



MMV

$$[(z_1 - \mu) + (z_2 - \mu)] [(z_1 - \mu) + (z_2 - \mu)] =$$

$$(z_1 - \mu)^2 + 2(z_1 - \mu) \cdot (z_2 - \mu) + (z_2 - \mu)^2$$

$$E[(z_1 - \mu) \cdot (z_2 - \mu)] = \underbrace{E[z_1 z_2]}_{\text{STATISTIČKI NEZAVISNI}} - \mu \underbrace{E[z_1]}_{\mu} - \mu \underbrace{E[z_2]}_{\mu} + \mu^2 = -\mu^2$$

- OVA ŠOKTE KORAKA NEĆE DOGO NO NISCA ODA  
POZNO I TO DA MU NA STATISTIČKI NEZAVISNI  
STOČNIČKI ZVRŠI POZAVATE LI ODZVONJATE ALE VODI-  
KUĆA NE ODZVONJATE STATISTIČKI REZULTATI TA  
ZBOGDA:

$$E[(z_1 - \mu) \cdot (z_2 - \mu)] = 0$$

$$Z = X + Y : \quad E[X] = \mu_X \quad E[Y] = \mu_Y \quad E[(X - \mu_X)^2] = \sigma_X^2 \quad \sigma_X^2 = E[X^2] - \mu_X^2$$

$$E[(Y - \mu_Y)^2] = \sigma_Y^2$$

$$E[Z] = E[X + Y] = E[X] + E[Y] = \mu_X + \mu_Y$$

$$E[(Z - \mu_Z)^2] = E[Z^2] - 2\mu_Z \cdot E[Z] + \mu_Z^2 = E[Z^2] - \mu_Z^2$$

$$E[Z^2] = E[X^2 + 2XY + Y^2] = E[X^2] + E[Y^2] = \sigma_X^2 + \mu_X^2 + \sigma_Y^2 + \mu_Y^2$$

$$E[(Z - \mu_Z)^2] = \sigma_X^2 + \mu_X^2 + \sigma_Y^2 + \mu_Y^2 - (\mu_X + \mu_Y)^2 =$$

$$\sigma_Z^2 = \sigma_X^2 + \sigma_Y^2 - 2\mu_X \mu_Y$$

IZVODJENJE  
ZAKON JEDNOSTAVNO  
DODAJE SE  $\sigma_Z^2 = \sigma_X^2 + \sigma_Y^2$   
DODAJE SE  $\mu_Z = \mu_X + \mu_Y$   
JO KAZNEĆE  
PRUĆA UNI MIKRO.

CREASING MEASURE:

$$\Pr\{|Y-\mu| > \varepsilon\} \leq \frac{6^2}{\varepsilon^2}$$

$$\bar{Z}_n = \frac{1}{n} \sum_{i=1}^n Z_i$$

$$E[\bar{Z}_n] = \mu \quad \text{Var}\{\bar{Z}_n\} = \frac{6^2}{n}$$

$$\Pr\{| \bar{Z}_n - \mu | > \varepsilon\} \leq \frac{5^2}{n \cdot \varepsilon^2}$$

BY LETTING  $n \rightarrow \infty \Rightarrow \Pr\{| \bar{Z}_n - \mu | > \varepsilon\} \rightarrow 0$   
PROVED!!!

**3.2 AEP AND NOISE INFORMATION.** Let  $(x_i, z_i)$

be i.i.d.  $\sim \gamma(\theta, \gamma)$ . We form the log likelihood ratio of the hypothesis that  $X$  and  $Z$  are independent vs. hypothesis that  $X$  and  $Z$  are dependent. What is the limit of:

$$I = \frac{1}{n} \ln \frac{p(x^n) \gamma(z^n)}{p(x^n, z^n)}$$

$$- \frac{1}{n} \ln p(x_1, x_2, \dots, x_n) = - \underbrace{\frac{1}{n} \sum_{i=1}^n \ln p(x_i)}_{\text{SAMPLE NOISE}} \xrightarrow{\text{LD}} -E[\ln p(x_i)] = H(X)$$

$$I = \frac{1}{n} \ln p(x^n) + \frac{1}{n} \ln \gamma(z^n) - \frac{1}{n} \ln p(x^n, z^n) \xrightarrow{n \rightarrow \infty}$$

$$\begin{aligned} & \rightarrow -H(X) - H(Z) + H(X, Z) = H(X) - H(Z) + (H(Z) + H(X|Z)) \\ & = -H(X) + H(X|Z) = -\underline{I(X, Z)} \end{aligned}$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \ln \frac{p(x^n) \cdot \gamma(z^n)}{p(x^n, z^n)} = I(X, Z)$$

Concordant or discordant condition

$$\frac{1}{n} \ln \frac{p(x^n) \cdot \gamma(z^n)}{p(x^n, z^n)} = \frac{1}{n} \ln \prod_{i=1}^n \frac{p(x_i) \cdot \gamma(z_i)}{p(x_i, z_i)} =$$

$$\begin{aligned}
 &= \frac{1}{4} \sum_{i=1}^n \text{lo} \frac{f(x_i) \cdot g(y_i)}{f(x_i, y_i)} \cdot \lim_{n \rightarrow \infty} \frac{1}{n} \text{lo} \frac{f(x_i) \cdot g(y_i)}{g(x_i, y_i)} = \\
 &= \sum_{i=1}^n \underbrace{\lim_{n \rightarrow \infty} \frac{1}{n} \text{lo} \frac{f(x_i) \cdot g(y_i)}{f(x_i, y_i)}}_{I(x_i, y_i)} = \sum_{i=1}^n I(x_i, y_i) \quad \text{VIA } \underline{075318796}
 \end{aligned}$$

$$\frac{1}{4} \sum_{i=1}^n \text{lo} \frac{f(x_i) \cdot g(y_i)}{g(x_i, y_i)} \rightarrow E \left[ \text{lo} \frac{f(x_i) \cdot g(y_i)}{g(x_i, y_i)} \right] = -I(x, y) \quad \text{TROVED}$$

3.3 Piece of Cake: A cake is sliced roughly in half, the other pieces discarded. We will assume that a random cut creates pieces of proportion

$$P = \begin{cases} \left(\frac{2}{3}, \frac{1}{3}\right) & \text{WITH PROBABILITY } \frac{3}{4} = p \\ \left(\frac{2}{5}, \frac{3}{5}\right) & \text{WITH PROBABILITY } \frac{1}{4} = q \end{cases}$$

True for example, the first cut (and chance of largest piece) that result in piece size  $\frac{3}{5}$ . Cutting and choosing from this piece might reduce it to size  $\left(\frac{2}{5}\right) \cdot \left(\frac{2}{3}\right)$  at the 2, and so on. How large to first piece in general, is the piece of cake after  $n$  cuts.

$$1) \frac{1}{4} \cdot \frac{1}{2} = P\left(\frac{3}{5}\right) = \frac{1}{8} \quad P\left(\frac{2}{5}\right) = \frac{1}{4} \cdot \frac{1}{2} = \frac{1}{8}$$

$$2) P\left(\frac{3}{5} \cdot \frac{2}{3}\right) = P\left(\frac{3}{5}\right) \cdot P\left(\frac{2}{3}\right) = \frac{1}{8} \cdot \frac{3}{4} \cdot \frac{1}{2} = \frac{1}{8} \cdot \frac{3}{8} = \frac{3}{64}$$

$$P(2) = P\left[\frac{1}{2} P\left(\frac{2}{3}\right) + \frac{1}{2} P\left(\frac{1}{3}\right)\right]$$

$$P\left(\frac{3}{15}\right) = P\left(\frac{3}{5} \cdot \frac{1}{3}\right) = P\left(\frac{3}{5}\right) \cdot P\left(\frac{1}{3}\right) = \frac{1}{8} \cdot \frac{3}{4} \cdot \frac{1}{2} = \frac{3}{64} = P\left(\frac{3}{5} \cdot \frac{2}{3}\right) = P\left(\frac{6}{15}\right)$$

$$P\left(\frac{4}{15}\right) = P\left(\frac{2}{5} \cdot \frac{2}{3}\right) = P\left(\frac{2}{5}\right) \cdot P\left(\frac{2}{3}\right) = \frac{1}{8} \cdot \frac{3}{4} \cdot \frac{1}{2} = \frac{3}{64} = P\left(\frac{2}{5} \cdot \frac{1}{3}\right) = P\left(\frac{2}{15}\right)$$

- and so on

$$P\left(\frac{2}{3}\right) = \frac{3}{4} \cdot \frac{1}{2} = \frac{3}{8} \quad P\left(\frac{1}{3}\right) = \frac{1}{4} \cdot \frac{1}{2} = \frac{1}{8}$$

$$P\left(\frac{4}{15}\right) = P\left(\frac{2}{3} \cdot \frac{2}{5}\right) = P\left(\frac{2}{3}\right) \cdot P\left(\frac{2}{5}\right) = \frac{3}{8} \cdot \frac{1}{8} = \frac{3}{64} = P\left(\frac{1}{3} \cdot \frac{2}{5}\right) = P\left(\frac{2}{15}\right)$$

$$P\left(\frac{1}{15}\right) = P\left(\frac{2}{3} \cdot \frac{3}{5}\right) = P\left(\frac{2}{3}\right) \cdot P\left(\frac{3}{5}\right) = \frac{3}{8} \cdot \frac{3}{8} = P\left(\frac{1}{3} \cdot \frac{3}{5}\right) = P\left(\frac{3}{15}\right)$$

Ako se mogući go iznosi: 1040LENOO paric  
 procent st potrošenja:  $P = \begin{cases} \left(\frac{2}{3}, \frac{1}{3}\right) & \text{with monopolic} \\ \left(\frac{2}{5}, \frac{3}{5}\right) & \text{with + -} \end{cases}$   $\begin{cases} \left(\frac{3}{4}\right), p^* \\ \left(\frac{1}{4}\right), 2^* \end{cases}$

Ako se tkoje od  $(2/3)$   
 ①  $P\left(\frac{2}{3}\right) = P\left(\frac{3}{4}\right) \cdot \frac{1}{2} = \frac{3}{4} \cdot \frac{1}{2} = \frac{3}{8} = \frac{1}{2}$   $\begin{cases} \text{sekor} \\ \text{go dicas} \\ \text{1040LENOO} \\ \text{i.e } 2/3 \end{cases}$   
 $q\left(\frac{1}{3}\right) = \left(\frac{1}{4}\right) \cdot \frac{1}{2} = \frac{3}{8} = \frac{1}{2}$

②  $\underline{q\left(\frac{2}{3} \cdot \frac{3}{5}\right)} = q\left(\frac{2}{3}\right) \cdot P\left(\frac{3}{5}\right) = \frac{3}{8} \cdot \frac{1}{4} \cdot \frac{1}{2} = \frac{3}{8} \cdot \frac{1}{8} = \frac{3}{64}$   $\begin{cases} \text{GO} \\ \text{BITAS} \\ \text{1040LENOO} \\ \text{3/5} \end{cases}$   
 $q\left(\frac{2}{3} \cdot \frac{2}{5}\right) = 1\left(\frac{2}{3}\right) \cdot q\left(\frac{2}{5}\right) = \frac{3}{8} \cdot \frac{1}{8} = \frac{3}{64}$   
 $P\left(\frac{2}{3} \cdot \frac{2}{3}\right) = 1\left(\frac{2}{3}\right) \cdot 1\left(\frac{2}{3}\right) = \frac{3}{8} \cdot \frac{3}{8} = \frac{9}{64}$   
 $q\left(\frac{2}{3} \cdot \frac{1}{5}\right) = 1\left(\frac{2}{3}\right) \cdot 1\left(\frac{1}{5}\right) = \frac{3}{8} \cdot \frac{3}{8} = \frac{9}{64}$   $\begin{cases} \text{GO} \\ \text{BITAS} \\ \text{1040LENOO} \\ 2/3 \end{cases}$

$P\left(\frac{2}{3} \cdot \frac{2}{5}\right) = P\left(\frac{2}{3} \cdot \frac{2}{3}\right) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{2}$   $\begin{cases} \text{VO VTORIOT} \\ \text{ZENOK GO} \\ \text{IMAS IC} \\ P\left(\frac{6}{15}\right) \text{ LI } P\left(\frac{4}{9}\right) \end{cases}$   
 $P\left(\frac{2}{3} \cdot \frac{2}{3}\right) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{2}$

- Ako se tkoje od  $(3/5)$  vo vtori zavor  
 Otkrite se:  
 $q\left(\frac{2}{5} \cdot \frac{2}{3}\right) = \frac{1 \cdot 2}{2^2}$   $P\left(\frac{2}{3} \cdot \frac{3}{5}\right) = \frac{9}{2}$   $\begin{cases} \text{MOZETI NE} \\ \text{TREBA DA} \\ \text{MOZIS GO} \\ 1/2 ZATO \\ PREDICO E DA \\ GO ZEMEI \\ 1040LENOO \\ PARIC !!! \end{cases}$   
 ENTRONIZACIJA za vtori zavor e:  
 $H = - \sum_{k=0}^{2} \binom{4}{k} p^k \cdot 2^{4-k} \ln \binom{4}{k} p^4 \cdot 2^{4-k} =$   
 $\frac{\binom{2}{1}}{1! \cdot 1!} = 2$   $\begin{cases} \text{ZAVOD ODRS} \\ \text{AKTIV. } 2^2 \end{cases}$   
 $= - \left[ \binom{4}{0} p^0 \cdot 2^2 \ln \binom{4}{0} p^0 \cdot 2^2 + \binom{4}{1} p^1 \cdot 2^1 \ln \binom{4}{1} p^2 \cdot 2^1 + \binom{4}{2} p^2 \cdot 2^0 \ln \binom{4}{2} p^2 \cdot 2^0 \right]$   
 $= - \left[ p^2 \ln 2^2 + 2 p^2 \ln 12 + p^2 \ln p^2 \right]$   
 - Ako se generiraju skoci entronizacija je Broe:  
 $H(S) = - \sum_{k=0}^{4} \binom{4}{k} p^k \cdot 2^{4-k} \ln \binom{4}{k} p^4 \cdot 2^{4-k}$   $S = \{x_1, x_2, x_3\}$   
 $\text{ENTROPIJA NA NIZA TAKO GO SLUZAVI PLAN.}$

$$\textcircled{30} \quad P\left(\frac{2}{3} \cdot \frac{3}{5} \cdot \frac{2}{3}\right) = \begin{cases} \text{ODA M REZ} \\ \text{PERCENTE SO} \\ \text{ZAKLADJENI ZBOZ} \\ \text{NA PROSOCNITE VYKONA} \end{cases} = P \cdot 2 \cdot 1 = \underline{\underline{P^2}} \\ P\left(\frac{2}{3} \cdot \frac{2}{3} \cdot \frac{2}{3}\right) = \underline{\underline{P^3}}; \quad P\left(\frac{2}{3} \cdot \frac{3}{5} \cdot \frac{2}{5}\right) = \underline{\underline{P \cdot P^2}}; \\ P\left(\frac{2}{3} \cdot \frac{2}{3} \cdot \frac{3}{5}\right) = \underline{\underline{P^2 \cdot P}}; \quad P\left(\frac{3}{5}, \frac{2}{3}, \frac{2}{3}\right) = \underline{\underline{P \cdot P^2}}; \\ P\left(\frac{2}{5} \cdot \frac{2}{3} \cdot \frac{3}{5}\right) = \underline{\underline{P^2 \cdot P}}; \quad P\left(\frac{2}{5} \cdot \frac{3}{5} \cdot \frac{2}{3}\right) = \underline{\underline{P^2 \cdot P}}; \quad P\left(\frac{3}{5} \cdot \frac{3}{5} \cdot \frac{2}{3}\right) = \underline{\underline{P^3}}$$

MAIS VYKRO:  $1+P^3, 3+P^2P, 3+P^2P^2, 1+P^2$   
HOZMI ISTOPI.

$$P\left(\frac{9}{64}\right); \quad 3 \cdot P\left(\frac{9}{16} \cdot \frac{1}{4}\right) = 3 \cdot P\left(\frac{9}{64}\right); \quad 3 P\left(\frac{3}{4} \cdot \frac{1}{16}\right) = 3 \cdot P\left(\frac{3}{64}\right); \\ ; \quad P\left(\left(\frac{1}{4}\right)^3\right) = P\left(\frac{1}{64}\right)$$

$$x_i \in \begin{cases} \frac{2}{3} & \text{PROBABILITY } P \\ \frac{3}{5} & \text{PROBABILITY } Q = 1 - P \end{cases}$$

$$P(x_1, x_2, \dots, x_i, \dots, x_n) \rightarrow 2^{-H(X)} = 2^{-n \cdot (P \lg \frac{1}{P} + Q \lg \frac{1}{Q})}$$

$$H(X) = P \lg \frac{1}{P} + Q \lg \frac{1}{Q} = \frac{3}{4} \lg \frac{4}{3} + \frac{1}{4} \lg 4 = \underline{\underline{0.8113}}$$

• TIPERIOT SET JEKA

$$- n \cdot \frac{3}{4} \quad 2/3 - K_1 \quad \text{OTSEČOČI} \\ (\text{DVOZLOZENI})$$

$$- n \cdot \frac{1}{4} \quad 3/5 - K_1 \quad \text{VETOCI}$$

zakladnye svedeni  $\therefore n = \underline{\underline{P \vee \{A_E^{(n)}\}}} \geq 1 - \epsilon$

$$\begin{cases} L_{\alpha}^{X_i} = x \\ L_{\alpha+1}^{X_i} = x+1 \end{cases}$$

PRIMER:  $n = 4 \Rightarrow$  NADVEROVATNOST GOCEMINT NA VYKOTO JE ALE

$$\left(\frac{2}{3}\right)^{\frac{3}{4}} \cdot \left(\frac{3}{5}\right)^{\frac{1}{4}} = \left(\frac{2}{3}\right)^3 \cdot \left(\frac{3}{5}\right)^1 = \frac{8}{27} \cdot \frac{3}{5} = \frac{24}{135} = \frac{8}{45}$$

$$P\left(\left(\frac{2}{3}\right)^3 \cdot \left(\frac{3}{5}\right)^1\right) \div 2^{-H(X)} = 2^{-4 \cdot 0.8113}$$

$$\text{DOPRJE DVEK} \\ P \cdot P \cdot Q \cdot Q = 2^{-H(X)}$$

$$2^{-4 \left( \frac{3}{4} \lg \frac{4}{3} + \frac{1}{4} \lg 4 \right)} = 2^{-4 \left( \frac{3}{4} \lg \frac{16}{9} + \frac{1}{2} \right)} = 2^{-3 \lg \frac{4}{3} - 2 - 2 - 3 \lg \frac{4}{3}} \\ = \frac{1}{3} \cdot 2^{\lg \left(\frac{4}{3}\right)^3} = \frac{1}{3} \cdot \left(\frac{4}{3}\right)^3 = 2$$

• Scores nelle diverse suddivisioni del governo di fatto:  
 totale 100% "y = scelta di 100%"

$$L(y) = \left(\frac{2}{3}\right)^{\frac{3}{4}y} \cdot \left(\frac{3}{5}\right)^{\frac{1}{4}(100-y)} \quad L(4) = \frac{8}{45} = 0,178 \quad L(10) = 0,0133$$

$$2^{n-H} = \binom{n}{y} = 2^{4 \cdot 0.8} = 9.48 \quad H(4) = 0.8113$$

$$\binom{4}{1} = \frac{4!}{3!} = \frac{4 \cdot 3 \cdot 2 \cdot 1}{3 \cdot 2 \cdot 1} = 4 \quad \text{MAXIMUM BLDAT mit Elementen von TOTALE MAIORIO}$$

3.4 ACP. Let  $x_i$  be iid  $\sim \gamma(\lambda)$ ,  $i \in \{1, 2, \dots, n\}$

Let  $\mu = E[x]$  and  $H = -\sum \gamma(\lambda) \ln \gamma(\lambda)$ .

Let  $A^H = \{x^H \in X^H : \left| -\frac{1}{n} \sum_{i=1}^n \ln \gamma(x_i) - H \right| \leq \epsilon \}$ .

Let  $B^H = \{x^H \in X^H : \left| \frac{1}{n} \sum_{i=1}^n x_i - \mu \right| \leq \epsilon \}$ . LAW OF  
LARGE  
NUMBERS  
weak law

(a) Does  $\Pr\{x^H \in A^H\} \rightarrow 1$ . Yes

(b) Does  $\Pr\{x^H \in A^H \cap B^H\} \rightarrow 1$ . Yes

(a.)  $A^H$  è piano monotono, è vero che  
 $\Pr\{A^{(H)}\} \leq 1 - \epsilon$  Theorem 3.1.2.2

(b)  $B^H$  - sottoset di  $\Pr\{B^{(H)}\} \geq 1 - \delta$

(c) Show that:  $|A^H \cap B^H| \leq 2^{n(H+\epsilon)}$  for  $H$

(d) Show that:  $|A^H \cap B^H| \geq \left(\frac{1}{2}\right) \cdot 2^{n(H-\epsilon)}$  for  
n sufficiently large

(c)  $|A^{(H)}| \leq 2^{n(H+\epsilon)}$   $A_n$  è tipica set

$$2 |A^{(H)} \cap B^{(H)}| \leq |A^{(H)}| \quad |A^{(H)}| \geq |A^{(H)} \cap B^{(H)}|$$

$$2^{n(H+\epsilon)} \geq |A^{(H)}| \geq |A^{(H)} \cap B^{(H)}| \quad |A^{(H)} \cap B^{(H)}| \leq 2^{n(H+\epsilon)}$$

(d)  $|A^H| \geq (1 - \epsilon) 2^{n(H-\epsilon)}$

$$1 = \sum_{x \in X^H} \gamma(x) \geq \sum_{x \in A^H} \gamma(x) = \begin{cases} \gamma(x) \geq 2^{n(H(x)-\epsilon)} & \\ 2^{-n(H(x)+\epsilon)} \leq \gamma(x) \leq 2^{-n(H(x)-\epsilon)} & \end{cases} = \textcircled{1}$$

$$\textcircled{1} = \sum_{x^i \in A^i} \gamma(x^i) \geq \sum_{x^i \in A^i} 2^{-n(H(i)+\varepsilon)} = |A_i| 2^{-n(H(i)+\varepsilon)}$$

$$|A_i| \leq 2^{-n(H(i)-\varepsilon)} \leq |X^i| \cdot 2^{-n(H(i)-\varepsilon)}$$

$$|X^i| \geq \sum_{x^i \in X^i} 2^{-n(H(i)-\varepsilon)} \leq |X^i| \cdot 2^{-n(H(i)-\varepsilon)}$$

$$\frac{|A^i|}{2^{-n(H(i)-\varepsilon)}} \leq |X^i| + \varepsilon \cdot 2^{-n(H(i)-\varepsilon)}$$

$$\frac{1}{2^{-n(H(i)-\varepsilon)}} \geq \frac{1}{|A^i| + \varepsilon 2^{-n(H(i)-\varepsilon)}}$$

$$\gamma(x^i) \geq \frac{1}{|A^i| - \varepsilon 2^{-n(H(i)-\varepsilon)}}$$

$$2^{-n(H(i)-\varepsilon)} \geq \frac{1}{|A^i| + \varepsilon 2^{-n(H(i)-\varepsilon)}}$$

$$|A^i| 2^{-n(H(i)-\varepsilon)} + \varepsilon \geq 1 \quad |A^i| \geq (1-\varepsilon) 2^{-n(H(i)-\varepsilon)}$$

is it correct??

$$(1+\varepsilon) \geq \Pr\{X^{(i)}\} \geq \Pr\{A^{(i)}\} = \sum_{x^i \in A^i} \gamma(x^i) \geq \sum_{x^i \in A^i} 2^{-n(H(i)+\varepsilon)}$$

$$= |A^i| 2^{-n(H(i)+\varepsilon)}$$

$$(1+\varepsilon) \geq |A^i| 2^{-n(H(i)+\varepsilon)}$$

$$|A^i| \leq (1+\varepsilon) 2^{n(H(i)+\varepsilon)}$$

$$|A^i| \leq |A^i| 2^{n(H(i)+\varepsilon)}$$

$$(1+\varepsilon) 2^{n(H(i)+\varepsilon)} \geq (1-\varepsilon) 2^{n(H(i)-\varepsilon)} \quad (1+\varepsilon) 2^{n\varepsilon} \geq (1-\varepsilon) 2^{-n\varepsilon}$$

$$(1+\varepsilon) 2^{n\varepsilon} \geq 1-\varepsilon$$

$$2^{n\varepsilon} \geq \frac{1-\varepsilon}{1+\varepsilon}$$

$$2^{n\varepsilon} \geq \sqrt{\left(\frac{1-\varepsilon}{1+\varepsilon}\right)}$$

$$2^{-n\varepsilon} \leq \sqrt{\frac{1+\varepsilon}{1-\varepsilon}}$$

$$|A^i| \geq (1-\varepsilon) 2^{nH} \cdot 2^{-n\varepsilon}$$

$$(1-\varepsilon) \leq (1+\varepsilon) 2^{n\varepsilon}$$

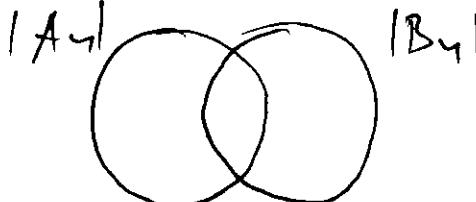
$(1+\varepsilon) \geq -(1+\varepsilon) 2^{n\varepsilon} \rightarrow$  STEP OF PROOF 3.1.2.2  
 UNIVERSITY OF TORONTO SOLUTIONS:

$$\Pr\{X^i \in A^i \cap B^i\} = \Pr\{X^i \in A^i\} + \Pr\{X^i \in B^i\} - \Pr\{X^i \in A^i \cup B^i\}$$

$$> (1-\varepsilon) + (1-\varepsilon_1) - \Pr\{X^i \in A^i \cup B^i\} \geq 1-\varepsilon-\varepsilon_1 \Rightarrow \Pr\{X^i \in A^i \cap B^i\} \rightarrow 1$$

(c) (ε-εrosione στον ουρανό, το νηπίο στέκεται κατεύθυνση)

$$|A^{\gamma} \cap B^{\gamma}| \leq 2^{\gamma(H+\epsilon)} \quad \text{πάλι σύγκριση μεταξύ}$$

$$|A^{\gamma} \cap B^{\gamma}| = |A^{\gamma}| + |B^{\gamma}| - |A^{\gamma} \cup B^{\gamma}| \quad |A^{\gamma}| \geq |A^{\gamma} \cap B^{\gamma}|$$


$$|A^{\gamma}| + |B^{\gamma}| = |A^{\gamma} \cup B^{\gamma}| + |A^{\gamma} \cap B^{\gamma}|$$

$$|A^{\gamma} \cap B^{\gamma}| = |A^{\gamma}| + |B^{\gamma}| - |A^{\gamma} \cup B^{\gamma}| \quad \cancel{|A^{\gamma}| + |B^{\gamma}| > |A^{\gamma} \cap B^{\gamma}|}$$

$$|A^{\gamma}| + |B^{\gamma}| \geq |A^{\gamma} \cap B^{\gamma}| \quad |A^{\gamma} \cap B^{\gamma}| \leq |A^{\gamma}| + |B^{\gamma}| \leq$$

$$(A^{\gamma} \cap B^{\gamma}) = |A^{\gamma}| + |B^{\gamma}| - |A^{\gamma} \cup B^{\gamma}| = \begin{cases} |A^{\gamma} \cup B^{\gamma}| > |B^{\gamma}| \\ |B^{\gamma}| \leq |A^{\gamma} \cup B^{\gamma}| \end{cases}$$

$$\leq |A^{\gamma}| + |A^{\gamma} \cup B^{\gamma}| - |A^{\gamma} \cup B^{\gamma}| = |A^{\gamma}|$$

$$|A^{\gamma} \cap B^{\gamma}| \leq |A^{\gamma}| \quad \text{δουλειά για νομιμοποίηση!!!}$$

Σε αυτήν την περιπτώση δουλειά. Σε σχέση με 3.1.2.3

$$|A^{\gamma}| \leq 2^{\gamma(H+\epsilon)} \rightarrow |A^{\gamma} \cap B^{\gamma}| \leq |A^{\gamma}| \leq 2^{\gamma(H+\epsilon)}$$

$$|A^{\gamma} \cap B^{\gamma}| \leq 2^{\gamma(H+\epsilon)} \quad |A^{\gamma}| \geq (1-\epsilon) 2^{\gamma(H-\epsilon)}$$

$$(d) \quad |A^{\gamma} \cap B^{\gamma}| \geq \frac{1}{2} 2^{\gamma(H-\epsilon)} \quad |A^{\gamma}| \geq (1-\epsilon) 2^{\gamma(H-\epsilon)}$$

$$|A^{\gamma} \cap B^{\gamma}| = |A^{\gamma}| + |B^{\gamma}| - |A^{\gamma} \cup B^{\gamma}| \geq \begin{cases} |A^{\gamma} \cup B^{\gamma}| \leq 2^{\gamma H} \\ |A^{\gamma}| \geq |B^{\gamma}| \end{cases}$$

$$\geq |A^{\gamma}| + |B^{\gamma}| - 2|B^{\gamma}| = |A^{\gamma}| - |B^{\gamma}|$$

$$|A^{\gamma}| \geq (1-\epsilon) 2^{\gamma(H-\epsilon)} \quad |\Pr\{A^{\gamma}\} \geq (1-\epsilon)|$$

$$|A^{\gamma}| \gg$$

(c) UNIVERSITY OF TORONTO SOLUTION

$$1 \geq \sum_{x^n \in A^{n+\epsilon}} \gamma(x^n) \geq \sum_{x^n \in A^{n+\epsilon}} 2^{-n(H+\epsilon)} = |A^{n+\epsilon}| \cdot 2^{-n(H+\epsilon)}$$

$$\Rightarrow |A^{n+\epsilon}| \leq 2^{-n(H+\epsilon)} \quad \text{proven!!!}$$

(d) UNIVERSITY OF TORONTO SOLUTION

QD. (c) equivalent PERA TWO  $|A^{n+\epsilon}| \leq 2^{-n(H+\epsilon)}$

$$|C^n| \leq 2^{-n(H+\epsilon)} \Rightarrow C^n \in \text{TYPICAL PROBES}$$

$$\rightarrow \Pr\{C^n\} \rightarrow 1 \Rightarrow \Pr\{A^{n+\epsilon}\} \rightarrow 1$$

$$\frac{1}{2} \leq \Pr\{A^{n+\epsilon}\} = \sum_{x^n \in A^{n+\epsilon}} \gamma(x^n) \leq \sum_{x^n \in A^{n+\epsilon}} 2^{-n(H-\epsilon)} =$$

$$= |A^{n+\epsilon}| 2^{-n(H-\epsilon)} \quad \boxed{|A^{n+\epsilon}| \geq \frac{1}{2} 2^{-n(H-\epsilon)}} \quad \boxed{\text{PROVED!!!}}$$

**3.5** SETS DEFINED BY NORMALITIES. Let  $x_1, x_2, \dots$  be an i.i.d sequence of discrete random variables with entropy  $H(X)$ . Let

$$C_n(t) = \{x^n \in \mathcal{X}^n : \gamma(x^n) \geq 2^{-n+t}\}$$

denote the subset of  $n$ -sequences with prob. abilities  $\geq 2^{-n+t}$ .

(a) SHOW THAT  $|C_n(t)| \leq 2^{nt}$

(b) FOR WHAT VALUES OF "t" DOES  $\Pr\{x^n \in C_n(t)\} \rightarrow 1$

(a)  $\underline{2^{-n(H(x))+\epsilon}} \leq \gamma(x^n) \leq \overline{2^{-n(H(x)-\epsilon)}} \quad / \text{ld}$

$$1 = \sum_{x^n \in \mathcal{X}^n} \gamma(x^n) \geq \sum_{x^n \in \mathcal{X}^n} 2^{-nt} \geq |C_n(t)| \cdot 2^{-nt} \Rightarrow$$

$$\boxed{|C_n(t)| \leq 2^{nt}}$$

(b)  $t = H(X) + \epsilon \Rightarrow |C_n(t)| \leq 2^{n(H(X)+\epsilon)}$

$$-n(H(X)+\epsilon) \leq \underline{\gamma(x^n)} \leq \overline{\gamma(x^n)} \leq -n(H(X)-\epsilon) \quad (\leftarrow 1)$$

$$n(H(X)+\epsilon) \geq -\underline{\gamma(x^n)} \geq n[H-\epsilon] \Rightarrow$$

$$n(h(x) - \varepsilon) \leq -\frac{1}{4} \text{ld}(g(x)) \leq n(h(D) + \varepsilon)$$

$$1 - \varepsilon = \sum_{x^n \in X^n} p(x^n) \geq \sum_{x^n \in X^n} 2^{-nt} = |C_n(t)| \cdot 2^{-nt}$$

$$2^{-nt} \leq \frac{1-\varepsilon}{|C_n(t)|} ; -nt \leq \text{ld} \frac{1-\varepsilon}{|C_n(t)|} ;$$

$$t \geq \frac{1}{n} \text{ld} \frac{|C_n(t)|}{1-\varepsilon}$$

• UNIVERSITY OF TORONTO SOLUTIONS

$$\frac{1}{n} \text{ld} |\beta_\delta^{(n)}| > h - \delta \quad (\beta_\delta^{(n)})^{\frac{1}{n}} > 2^{-(h-\delta)}$$

$$|\beta_\delta^{(n)}| > 2^{n(h-\delta)} \Rightarrow |C_n(t)| > 2^{n(h-\delta)}$$

- FROM PART (i)  $|a_n(n)| \leq 2^{-nt} \Rightarrow$
- $t \geq h - \varepsilon$  is necessary condition such  $\Pr\{x^n \in C_n(t)\} \rightarrow 1$
- $t \geq h + \varepsilon$  is sufficient condition ( $\varepsilon > 0$ ) such  $A_\delta^{(n)} \subseteq C_n(t)$  where  $\Pr\{x^n \in C_n(t)\} \rightarrow 1$
- WHETHER  $t = h$  is sufficient condition (IT DEPENDS TO THE DISTRIBUTION).
  - E.G.  $\Pr\{x=0\} = \Pr\{x=1\} = \frac{1}{2} \rightarrow t = 1$

$$h = \frac{1}{2} \text{ld} 2 + \frac{1}{2} \text{ld} 2 = 1$$

ALL SEQUENCES HAVE PROBABILITY  $2^{-n}$

$$\begin{cases} x_1 & 00 \\ x_2 & 01 \\ x_3 & 10 \\ x_4 & 11 \end{cases} \quad P(x_i) = \frac{1}{2^n} = \frac{1}{4} \quad i = 1, 2, 3, 4$$

□ AND  $\Pr\{x=0\} = p \quad \Pr\{x=1\} = 1-p \quad 0 < p < 1$

$t = h$  is NOT SUFFICIENT !!!

- NO TIME  $\Rightarrow \gamma = 2$

$$\left. \begin{array}{l} x_1 = 00 \\ x_2 = 01 \\ x_3 = 10 \\ x_4 = 11 \end{array} \right\} \quad \begin{array}{l} p(t_1) = 1^2 \cdot \gamma(t_2) = \gamma(1-\gamma) \\ \gamma(t_3) = \gamma(1-\gamma) \quad p(t_4) = (1-\gamma)^2 \end{array}$$

E.G.  $\gamma = 0.6 \quad 1-\gamma = 0.4 \quad \gamma(t_1) = (0.6)^2 = 0.36$

$$\gamma(x_2) = 0.6 \cdot 0.4 = 0.24 \quad \gamma(t_3) = 0.24 \quad \gamma(t_4) = 0.16$$

$$H = -0.6 \log 0.6 - 0.4 \log 0.4 = 0.971$$

$$2^{-nH} = 2^{-2H} = 2^{-1.942} = 0.26$$

$$\boxed{\gamma(t_1) > 0.26}$$

$$\boxed{\gamma(t_2 \dots t_4) < 0.26}$$

- Zusätzl. nechsten Schritt  $\Pr\{X_n \in C_i(t)\} \rightarrow 1 \in [t > t - \epsilon]$

**36 AEP-like limit.** Let  $x_1, x_2, \dots$  be i.i.d drawn according to probability mass function  $\gamma(t)$ . Find:  $\lim_{n \rightarrow \infty} \left[ \gamma(x_1, x_2, \dots, x_n) \right]^{\frac{1}{n}}$ .

$$2^{-n(H+\epsilon)} \leq \gamma(x_1, x_2, \dots, x_n) \leq 2^{-n(H(x)-\epsilon)}$$

$$-\frac{1}{n} \log \gamma(x_1, x_2, \dots, x_n) \xrightarrow{n \rightarrow \infty} H(t)$$

$$\frac{1}{n} \log \gamma(x_1, x_2, \dots, x_n) \rightarrow -H(t)$$

$$\lim_{n \rightarrow \infty} \gamma(x_1, x_2, \dots, x_n) = 2^{-H(x)}$$

$$\begin{aligned} 2^{-n(H+\epsilon)} &\leq \gamma(x_1, x_2, \dots, x_n) \\ &\leq 2^{-n(H(x)-\epsilon)} \end{aligned}$$

• Trinity College Dublin section:

$$\lim_{n \rightarrow \infty} \gamma(x_1, x_2, \dots, x_n) = \lim_{n \rightarrow \infty} 2^{\frac{1}{n} \log \gamma(x_1, x_2, \dots, x_n)} =$$

$$= \lim_{n \rightarrow \infty} 2^{\frac{1}{n} \sum_{i=1}^n \log \gamma(x_i)} = \left| \begin{array}{l} \text{STRONG LAW OF LARGE NUMBERS} \\ \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \gamma(x_i) = E[\gamma] \end{array} \right| =$$

$$= \lim_{n \rightarrow \infty} 2^{E[\log \gamma]} = 2^{-H(x)} \quad \left| \begin{array}{l} \text{STRONG LAW: ALMOST SURELY} \\ \frac{1}{n} \sum_{i=1}^n \gamma(x_i) \xrightarrow{A.S.} E[\gamma] \end{array} \right|$$

**5.7 ACP AND SOURCE CODING:** A DISCRETE MEMORYLESS SOURCE EMITS A SEQUENCE OF STATISTICALLY INDEPENDENT BINARY DIGITS WITH PROBABILITIES  $P(1) = 0.005$  AND  $P(0) = 0.995$ . THE DIGITS ARE TAKEN 100 AT TIME AND BINARY CODEWORD IS PROVIDED FOR EACH SEQUENCE OF 100 DIGITS CONTAINING THREE OR FEWER 1'S.

(a) ASSUMING THAT THE CODEWORDS ARE THE SAME LENGTH, FIND THE MINIMUM LENGTH REQUIRED TO PROVIDE CODEWORDS FOR THE SEQUENCES WITH THREE OR FEWER 1'S.

(b) CALCULATE THE POSSIBILITY OF OBSERVING A SOURCE SEQUENCE FOR WHICH NO CODEWORD HAS BEEN ASSIGNED.

(c) USE CHERTSHOV'S INEQUALITY TO FIND THE PROBABILITY OF OBSERVING A SOURCE SEQUENCE FOR WHICH NO CODEWORD HAS BEEN ASSIGNED. COMPARE THE BOUND WITH ACTUAL PROBABILITY COMPUTED IN PART (b).

Q)				MAX. LENGTH
- 2	SOURCE	NO	1000	SEQUENCE I
- 9	SOURCE	NO	5000	SEQUENCE II
- 18	SOURCE	NO	10000	SEQUENCE III
- 85	SOURCE	NO	50000	SEQUENCE IV
- VERIFATRIST DA SE GENERATA SEQUENCA SO > 3	e:			
601111				5000 4991 1

$$q = 1 - P(S_3 + S_2 + S_1 + S_0)$$

$S_i \rightarrow$  SEQUENCE WITH  $i =$  ONES.

$$P(S_0 + S_1 + S_2 + S_3) = \binom{100}{3} P^{97} q^3 + \binom{100}{2} P^{98} q^2 + \binom{100}{1} P^{99} q + \binom{100}{0} P^{100}$$

000 101  
001 110  
010 111  
011 111  
100 111

$$\binom{3}{2} = \frac{6}{2 \cdot 1!} = 3$$

$$\binom{3}{1} = \frac{6}{2! \cdot 1!} = 3$$

$$N(q) = 0.115$$

$$P(1) =$$

$$H(p) = \left[ -100 \cdot 0.005 \log 0.005 + 99 \cdot 0.995 \log 0.995 \right] = 0.0454$$

$$H(X) = 100 \cdot 0.0454 = 4.54525 \text{ BITS}$$

PEROZATRIST DA SE GENERATA SEQUENCA SO RAZJEZOPISI E ISTA SO VERIFATRIST DA SE GENERATA IZVETRA SEQUENCA ZA KOTAKNE E GENERIRAN KOPEN ZDOR

$$P = 1 - \left( \binom{100}{3} P^{97} q^3 + \binom{100}{2} P^{98} q^2 + \binom{100}{1} P^{99} q + \binom{100}{0} P^{100} \right) = 0.0017$$

$$(a) \text{ Minimálna výberová súčieta na ktorú je zárukou} \in$$

$$n \cdot H(q) = 100 \cdot \left( q \ln \frac{q}{q} + (1-q) \ln \frac{1}{(1-q)} \right) \quad q = 0.995$$

$$n \cdot H(q) = 100 \cdot 0.0454 = 4,5415 \doteq 5 \quad \underline{\underline{\text{BTS}}}$$

$$(c) \boxed{\mu = p \cdot \varnothing + 1 \cdot 1 = 0.005}$$

$$\sigma^2 = \sum_{x=0}^1 (x-\mu)^2 p(x) = (1-0.005)^2 \cdot 0.005 + (0-0.005)^2 \cdot 0.995$$

$$\boxed{\sigma^2 = (0.995)^2 \cdot 0.005 + (0.005)^2 \cdot 0.995 = 0.005 = (0.995+0.005) \cdot \frac{0.005}{0.995} = 0.005 \cdot 0.995}$$

$$P\{|x-\mu| > \varepsilon\} \leq \frac{\sigma^2}{\varepsilon^2} \quad \underline{\underline{\text{Vezo}}} \quad \text{DRUGA} \quad \boxed{x \xrightarrow{a.s.} \mu}$$

$$P\{|x-\mu| > \varepsilon\} \leq \frac{0.005}{\varepsilon^2}$$

$$P\{|x-0.005| > \varepsilon\} \leq \frac{0.005}{\varepsilon^2}$$

ova je ekvivalentne na  $P\{x=1\} = 0.005$

Najlepšia výberová súčieta za  $\varepsilon$   $\in \varepsilon = 0.995$

$$P\{|x-0.005| > \varepsilon\} \leq \frac{0.005}{0.995^2} = 0.0051$$

Znali užívateľského  $\varepsilon$ : 0.005.

### • Solučia 1st Edition

$$(a) \binom{100}{0} + \binom{100}{1} + \binom{100}{2} + \binom{100}{3} = 166751$$

$$f_0(166751) = 17.74 \doteq \underbrace{17}_{\text{MINIMUM WEIGHT}}$$

$$n \cdot H(q) = 4.5 \doteq 5 \quad M_1 \Rightarrow \underbrace{\text{MINIMUM WEIGHT}}_{(\text{ZADANÝ TERAZ})} \quad \begin{array}{l} 18 \text{ IS} \\ \text{QUITE LARGE} \\ \text{THEN } n \cdot H(q) \end{array}$$

$$(b) P = 1 - \sum_{i=0}^{100} \binom{100}{i} \underbrace{p^{100-i} \cdot q^i}_{\substack{0 \\ 1 \\ \dots \\ 100}} = 0.00167$$

$$(c) S_n - konštantná "výberová" RIAK \cdot \Omega \text{ sum of } \overset{n}{\underset{i=1}{\text{i.i.d.}}} \text{ čísiek} \quad \text{možnosť je: } P\{|S_n - \mu| > \varepsilon\} \leq \frac{4 \cdot 52}{\varepsilon^2}$$



$$y=100 \quad \mu = 1 \cdot 0.005 + 0 \cdot 0.995 = 0.005$$

$$G^2 = |\rho_{\mu, S}| = 0.005 \cdot 0.995$$

$S_{100} \geq 4$  IF AND ONLY IF:  $|S_{100} - 100 \cdot 0.005| \geq 3.5$

$$|S_{100} - 0.5| \geq 3.5 \Rightarrow \boxed{\epsilon = 3.5}$$

$$\Pr\{|S_n - 4 \cdot \mu| \geq \epsilon\} \leq \frac{G^2}{\epsilon^2}$$

OVAA DEFRICI DA 90 VGRADUVA VZLOVOT  
DEJA ZEOLIVE DA SERVENCI ZA VOI RE  
DEFINIRAN KOPEN ZELOV.

→ 90 DA OVA JE BIJE VEROPAT OT. DA  
SERVENCI ZA IMA 4 IC, TOVE JE DIOJ.

$$\Pr\{\text{NEKA KOPEN ZELOV}\} \leq \frac{100 \cdot (0.005 \cdot 0.995)}{3.5^2} = \frac{0.5 \cdot 0.995}{3.5^2} = 0.04061$$

$0.04061 > 0.0167$  } LOS ZAIND DATA CREDISLEVATA  
NEENHOVOST.

### 3.8 Products

LET

$$X = \begin{cases} 1, & \text{WITH PROBABILITY } 1/2 \\ 2, & \text{WITH PROBABILITY } 1/4 \\ 3, & \text{WITH PROBABILITY } 1/4 \end{cases}$$

MMV KARAGU  
JAVA SVODA  
CDA VO  
MATE  
MULTIPLYING Law  
MMV

LET  $X_1, X_2, \dots, X_n$  BE DRAWN  
TO THE BEHAVIOR OF THE PRODUCT  
 $(X_1 \cdot X_2 \cdot \dots \cdot X_n)^{1/n}$ .

1.1. D ACCORDING  
THE LIMITING

BY LAW OF  
LARGE NUM.

$$(X_1 \cdot X_2 \cdot \dots \cdot X_n)^{1/n} = \sqrt[n]{\log(X_1 \cdot X_2 \cdot \dots \cdot X_n)} = \sqrt[n]{\sum_{i=1}^n \log(X_i)}$$

$$\text{a.s. } E[\log(X)] = \sqrt[n]{(1 + \log 3)} = \underline{1.56508}$$

$$H(X) = \frac{1}{2} \log 2 + \frac{1}{4} \log 4 + \frac{1}{4} \log 4 = \frac{1}{2} + \frac{1}{2} + \frac{1}{2} = \frac{3}{2}$$

$$E[\log(X)] = \log(1) \cdot \frac{1}{2} + \log 2 \cdot \frac{1}{4} + \log 3 \cdot \frac{1}{4} = \frac{1}{4}(1 + \log 3) = \frac{1}{4} \log 6$$

3.9 AEP. Let  $x_1, x_2, \dots$  be independent, identically distributed random variables given according to the probability mass function  $p(x)$ ,  $x \in \{1, 2, \dots, m\}$ . Thus,  $p(x_1, x_2, \dots, x_n) = \prod_{i=1}^n p(x_i)$ . We know that  $-\frac{1}{n} \log p(x_1, x_2, \dots, x_n) \rightarrow H(x)$  in probability.

Let  $g(x_1, x_2, \dots, x_n) = \prod_{i=1}^n g(x_i)$ , where  $g$  is another probability mass function on  $\{1, 2, \dots, m\}$ .

- Evaluate  $\lim(-\frac{1}{n} \log g(x_1, x_2, \dots, x_n))$  where  $x_1, x_2, \dots, x_n$  are i.i.d  $\sim p(x)$
- Now evaluate the limit of the log-likelihood ratio.  $\frac{1}{n} \log \frac{g(x_1, x_2, \dots, x_n)}{p(x_1, x_2, \dots, x_n)}$  where  $x_1, x_2, \dots, x_n$  are i.i.d  $\sim p(x)$ . Thus, the odds favoring  $g$  are small when  $p$  is true.

$$\begin{aligned}
 (a) -\frac{1}{n} \log g(x_1, x_2, \dots, x_n) &= -\frac{1}{n} \sum_{i=1}^n \log g(x_i) \xrightarrow{\text{def}} -E[\log g] \\
 &= -\sum_{i=1}^n g(x_i) \log g(x_i) = \underline{H_g(x)} \\
 -\frac{1}{n} \log \frac{g(x_1, x_2, \dots, x_n)}{p(x_1, x_2, \dots, x_n)} &= -\frac{1}{n} \left( \sum_{i=1}^n \log g(x_i) - \sum_{i=1}^n \log p(x_i) + \sum_{i=1}^n \log \frac{p(x_i)}{g(x_i)} \right) \\
 &= -\frac{1}{n} \sum_{i=1}^n \log \frac{g(x_i)}{p(x_i)} - \frac{1}{n} \sum_{i=1}^n \log p(x_i) \xrightarrow{\text{def}} -E\left[\log \frac{g(x_i)}{p(x_i)}\right] - E[\log p(x)] \\
 &= E\left[\log \frac{p(x_i)}{g(x_i)}\right] + H(x) = \underline{D(p||g) + H(x)}
 \end{aligned}$$

$$(b) \frac{1}{n} \log \frac{g(x_1, x_2, \dots, x_n)}{p(x_1, x_2, \dots, x_n)} \xrightarrow{\text{def}} -E_p\left[\log \frac{p(x_1, \dots, x_n)}{g(x_1, \dots, x_n)}\right] = -D(p||g)$$

PROOF OF RESULT IN SECTION 1 SO THAT WE GET

$$\begin{aligned}
 -\frac{1}{n} \sum_{i=1}^n \log g(x_1, x_2, \dots, x_n) &= -\lim_{n \rightarrow \infty} \sum_{i=1}^n \log g(x_i) \text{ w.p. 1} = -\sum p(x) \log g(x) \\
 &= +\sum p(x) \log \frac{p(x)}{g(x)} \quad \blacksquare \quad \sum p(x) \log p(x) = \underline{D(p||g) + H(p)}
 \end{aligned}$$

$$(b) \lim_{n \rightarrow \infty} \frac{1}{n} \ln \frac{\gamma(x_1, x_2, \dots, x_n)}{\gamma(x_1, x_2, \dots, x_n)} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \ln \frac{\gamma(x_i)}{\gamma(x_i)} \text{ w.r.t. } \mathbb{E} \left[ \ln \frac{\gamma(x_i)}{\gamma(x_i)} \right] \\ = - D(\gamma \| \varphi)$$

(080700865)

3.10 Random box size An  $n$ -dimensional rectangular box with sides  $x_1, x_2, \dots, x_n$  is to be constructed. The volume is  $V_n = \prod x_i$ . The edge length  $\ell$  of a  $n$ -cube with the same volume as a random box is  $\ell = V_n^{1/n}$ . Let  $x_1, x_2, \dots, x_n$  be uniform random variables over unit interval  $[0, 1]$ . Find  $\lim_{n \rightarrow \infty} V_n^{1/n}$  and compare to  $(\mathbb{E}[V_n])^{1/n}$ .

Clearly the expected edge length does not capture the idea of the volume of the box. The geometric mean, rather than the arithmetic mean, characterizes the behavior of the products.

$$H(x) = - \int \gamma(x) \ln \gamma(x) dx = \int_0^1 \gamma(x) \ln \frac{1}{\gamma(x)} dx = 0$$

$$\gamma(x) = \left[ \frac{1}{2}, \frac{1}{2} \right] \quad H(x) = \frac{1}{2} \ln 2 + \frac{1}{2} \ln 2 = 1$$

$$\gamma(x) = \left[ \frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4} \right] \quad H(x) = \left( \frac{1}{4} \cdot \ln 4 \right) 4 = \underline{\underline{\left( \frac{1}{2} \right) \cdot 4 = 2}}$$

$$\gamma(x) = \left[ \frac{1}{8}, \frac{1}{8}, \dots, \frac{1}{8} \right] \quad H(x) = \left( \frac{1}{8} \cdot 3 \right) 8 = \frac{24}{8} = 3$$

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

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

$x_i$  zamszci g1 urolo da se PDF  $\Rightarrow$  to g1 so  
so ukojost vo rang  $[0..1]$  (tak moga da bude so)  
tak moga da bude so)

$$\text{④} \lim_{n \rightarrow \infty} (x_1 \cdot x_2 \cdots x_n)^{\frac{1}{n}} = e^{\lim_{n \rightarrow \infty} \ln(x_1 \cdot x_2 \cdots x_n)} = e^{\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \ln x_i}$$

$$\xrightarrow{\text{A.S.}} e^{\mathbb{E}[\ln x_i]} = \underbrace{e^{-1}}_{\text{PP.63}} = \frac{1}{e}$$

$e^x$  e kontinuirat funkcia je zadaa  
lim odi vo cao-  
nevtot. (stewart continuity theorem)

$$\in [V_n]^{\frac{1}{n}} = \in [x_1 x_2 \dots x_n]^{\frac{1}{n}}$$

- 1ST REZON DA JE VJEZBE VOJNA SREDINA NA PROSTREDNJA HA  
SAMO  $\in$  MI E GREBO T.J. NEKA SREDINA NA PROSTREDNJA HA.

$$E[\ln(x)] = \int_0^1 \ln(x) dx = \begin{cases} M = \ln x \\ dM = \frac{1}{x} dx \\ U = \int 1 dx = x \end{cases} = x \cdot \ln x - \int x \cdot \frac{dx}{x} = x \ln x - x$$

$$E[\ln(x)] = + \left( x \ln x - x \right) \Big|_0^1 = -1$$

TEOREMA ZA  
SREDNA VREDNOST  
 $\int_{\Omega} f(x) \cdot h(x) dx$   
OD VEROJATNOST

$$\lim_{n \rightarrow \infty} V_n^{\frac{1}{n}} = e^{\lim_{n \rightarrow \infty} \frac{1}{n} \ln V_n} = \frac{1}{e} < \frac{1}{2}$$

AKT. ④:  $\lim_{n \rightarrow \infty} \ln(V_n)^{\frac{1}{n}} = \lim_{n \rightarrow \infty} \frac{1}{n} \ln(x_1 x_2 \dots x_n) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \ln(x_i)$

$\xrightarrow{n \rightarrow \infty} E[\ln(x)] = -1$

DO OVA SREDI SREDNA VREDNOST POZIDA  
NA OVA TEZO  $E: e^{-1}$

$$E[x_1 \cdot x_2 \dots x_n] = \prod_{i=1}^n E[x_i] = \left(\frac{1}{2}\right)^n$$

$$\iiint_{x_1 x_2 \dots x_n} x_1 \cdot x_2 \dots x_n \underbrace{f(x_1, \dots, x_n) dx_1 dx_2 \dots dx_n}_{= f(t_1) \cdot f(t_2) \dots f(t_n)}$$

$$E[x_i] = \frac{1}{2} = \frac{0+1}{2} \quad \text{ARITHMETIC MEAN}$$

$$\textcircled{1} \Rightarrow \underbrace{(x_1 \cdot x_2 \dots x_n)^{\frac{1}{n}}}_{\text{DEFINITION FOR GEOMETRIC MEAN}} = 1/e < \left(\frac{1}{2}\right)$$

DEFINITION FOR GEOMETRIC MEAN

$$\left(\frac{1}{e}\right)^n < \left(\frac{1}{2}\right)^n \Rightarrow \text{SREDEN VREDNOST NA VREDNOSTI GEOMETRISKO TEZO}$$

- SAMO DA VJEZBE OBLIKUJUZI (SLOVAKA)  
POZIDA PODJEDNA VREDNOST GEOMETRISKE SREDINE  
G PONIKEK OZ. ONA ZE DOBRA KAKO STVORETIC  
SLOVAKA.

3.11

# PROOF OF THEOREM 3.3.1.

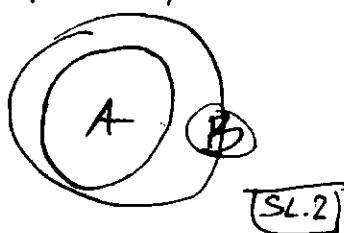
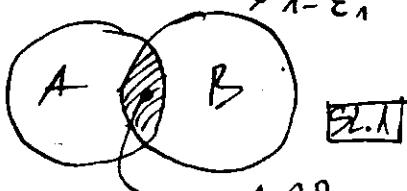
1075426898

GOL

This process shows that the size of the smallest "modular" set is about  $2^{nH}$ . Let  $x_1, x_2, \dots, x_n$  be i.i.d.  $\sim p(x)$ . Let  $B_{\delta}^{(n)}$   $\subset X^n$  such that  $Pr(B_{\delta}^{(n)}) > 1 - \delta$ . Fix  $\epsilon < \frac{\delta}{2}$ .

(a) Given the two sets  $A, B$  such as  $Pr(B_{\delta}^{(n)}) > 1 - \epsilon_1$ , and  $Pr(B) > 1 - \epsilon_2$ , show that  $Pr(A \cap B) > 1 - \epsilon_1 - \epsilon_2$ . Hence  $Pr(A_{\epsilon}^{(n)} \cap B_{\delta}^{(n)}) \geq 1 - \epsilon - \delta$

$$Pr(A + B) = \underbrace{Pr(A)}_{\geq 1 - \epsilon_1} + \underbrace{Pr(B)}_{\geq 1 - \epsilon_2} - Pr(A \cap B) > 1 - \epsilon_1 + 1 - \epsilon_2 - Pr(A \cap B)$$



$$Pr(A \cap B) = Pr(A) + Pr(B) - \underbrace{Pr(A + B)}_{\leq 1} \geq 1 - \epsilon_1 + 1 - \epsilon_2 - Pr(A + B)$$

$\geq 1 - \epsilon_1 + 1 - \epsilon_2 - \delta \geq 1 - \epsilon_1 - \epsilon_2$  proved!!!

(b) Justify the steps in the chain of inequalities:

$$1 - \epsilon - \delta \leq Pr(A_{\epsilon}^{(n)} \cap B_{\delta}^{(n)}) = \sum_{A_{\epsilon}^{(n)} \cap B_{\delta}^{(n)}} p(x^n) \leq \sum_{A_{\epsilon}^{(n)} \cap B_{\delta}^{(n)}} e^{-n(H-\epsilon)}$$

$$\boxed{2^{-n(H+\epsilon)} \leq p(x^n) \leq 2^{-n(H-\epsilon)}}$$

ASYMPTOTIC  
EQUALITY  
THEOREM

$$= |A_{\epsilon}^{(n)} \cap B_{\delta}^{(n)}| e^{-n(H-\epsilon)} \leq |A_{\epsilon}^{(n)}| e^{-n(H-\epsilon)} \leq |B_{\delta}^{(n)}| e^{-n(H-\epsilon)}$$

$$|B_{\delta}^{(n)}| \geq (1 - \epsilon_1 - \epsilon_2) \cdot e^{n(H-\epsilon)}$$

↑ presesnor + point Vermanov  
of somero prozessno

(c) Conclude the proof of the theorem

$$(1 - \epsilon - \delta) \leq |B_{\delta}^{(n)}| e^{-n(H-\epsilon)} / \ln$$

$$\ln(1 - \epsilon - \delta) \leq \ln |B_{\delta}^{(n)}| - n(H-\epsilon)$$

$$\ln |B_{\delta}^{(n)}| \geq \ln(1-\varepsilon-\delta) + \gamma(n-\varepsilon) \geq \ln\left(1-\frac{1}{2}-\frac{\delta}{\varepsilon}\right) + \gamma(n-\varepsilon)$$

$$= n\gamma - \gamma\varepsilon + \ln\left(\frac{1}{2}-\delta\right) \geq n\gamma - \frac{\gamma}{2} + \ln\left(\frac{1}{2}-\delta\right)$$

$$\frac{1}{\gamma} \ln |B_{\delta}^{(n)}| \geq n + \underbrace{\frac{1}{\gamma} \ln\left(\frac{1}{2}-\delta\right)}_{\leq 0} - \frac{1}{2} = n - \delta'$$

$$\boxed{\delta' = \frac{1}{2} + \frac{1}{\gamma} \ln\left(\frac{1}{2}-\delta\right)} \quad \delta' > 0$$

• EDITION 1 SOLUTION (WITH CORRECTIONS)

- AC KOMPLEMENT OF A (ZLENOVI OD ZE NAO  
NE PREDSTAVLJAT NA A).

$$P(A^c \cup B^c) = P(A^c) + P(B^c) - P(A^c \cap B^c)$$

$$\leq P(A^c) + P(B^c)$$

$$P(A) \geq 1 - \varepsilon_1 \Rightarrow \boxed{\begin{aligned} P(A^c) &= 1 - P(A) \leq 1 - 1 + \varepsilon_1 \\ P(A^c) &\leq \varepsilon_1 \end{aligned}} \Rightarrow$$

$$P(B) \geq 1 - \varepsilon_2 \Rightarrow P(B^c) = 1 - P(B) \leq 1 - 1 + \varepsilon_2 = \varepsilon_2$$

$$\boxed{P(B^c) \leq \varepsilon_2}$$

$$P(A \cap B) = 1 - P(A^c \cup B^c) \geq 1 - P(A^c) - P(B^c) \geq 1 - \varepsilon_1 - \varepsilon_2$$

$$X_n \in \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$$

$$A \in \{1, 2, 3, 4, 5, 6\}$$

$$B \in \{1, 2, 3, 4, 5, 6, 7\}$$

$$A^c \in \{7, 8, 9, 10\}$$

$$B^c \in \{8, 9, 10\}$$

$$\boxed{A^c \cup B^c = A^c + B^c - A^c \cap B^c = \{7, 8, 9, 10\}}$$

$$A \cap B = \{1, 2, 3, 4, 5, 6\}$$

$$\rightarrow \boxed{P(A \cap B) = 1 - P(A^c \cup B^c)}$$

ERMI PISU  
DOKAZANO !!!

$$\boxed{N.4. \star \quad (A_{\varepsilon}^{(n)} \cap B_{\delta}^{(n)}) / e^{-\gamma(H-\varepsilon)} \leq |B_{\delta}^{(n)}| e^{-\gamma(H-\varepsilon)}}$$

SLIČNI OP FAKTOR 350 :

$$\boxed{A_{\varepsilon}^{(n)} \cap B_{\delta}^{(n)} \subseteq B_{\delta}^{(n)}}$$

3.12

MONOTONIC CONVERGENCE OF EMPIRICAL DISTRIBUTION. Let  $\hat{P}_n$  denote empirical probability mass function corresponding to  $x_1, x_2, \dots, x_n$  i.e.  $\hat{P}_n(y) = \frac{1}{n} \sum_{i=1}^n I(x_i = y)$ , specifically,

is the proportion of times  $x_i = y$  in the first  $n$  samples, where  $I$  is indicator function.

(a) Show for  $X$  discrete that

$$E[D(\hat{P}_n || \gamma)] \leq E[D(\hat{P}_m || \gamma)] \rightarrow$$

EMPIRICAL DISTRIBUTION IS MORE MULTIMODAL

thus termed negative entropy "distance" from the empirical distribution to the true distribution decreases with sample size.

(Hint: Write  $\hat{P}_n = \frac{1}{n} \hat{P}_n + \frac{n-1}{n} \hat{P}_{n-1}$  and use the convexity of  $D$ .)

(b) Show for an arbitrary discrete  $X$  that

$$E[D(\hat{P}_n || \gamma)] \leq E[D(\hat{P}_{n-1} || \gamma)]$$

(Hint: Write  $\hat{P}_n$  as the average of  $n$  empirical mass functions with each of the  $n$  samples selected in turn.)

$$E[D(\hat{P}_n || \gamma)] \rightarrow \frac{1}{n} D(\hat{P}_{n-1} || \gamma) \quad n \rightarrow \text{BOTH MAJORANT}$$

$$D(\hat{P}_n || \gamma) = \sum_{i=1}^n p_n(i) \log \frac{p_n(i)}{p(i)}$$

$$\hat{P}_{n-1} = [\gamma_n, p_{-}]$$

•  $\Rightarrow$  convexity follows via Jensen's inequality.

$$E[D(\hat{P}_n || \gamma)] \geq D(E[\hat{P}_{n-1} || \gamma])$$

$$\begin{aligned} D(\hat{P} || \gamma) &= \hat{P}_1 \cdot \log \frac{\hat{P}_1}{\gamma_1} + \hat{P}_2 \cdot \log \frac{\hat{P}_2}{\gamma_2} \\ &= E_{\hat{P}} \left[ \log \frac{\hat{P}}{\gamma} \right] \end{aligned}$$

$$D(\hat{p}_1 || \gamma) = \frac{1}{n} \sum_{i=1}^n I(x_i=1) \cdot \text{ld} \left( \frac{\frac{1}{n} \sum_{i=1}^n I(x_i=1)}{p_1} \right) + \\ + \frac{1}{n} \sum_{i=1}^n I(x_i=2) \text{ld} \left( \frac{\frac{1}{n} \sum_{i=1}^n I(x_i=2)}{p_2} \right)$$

$$D(\hat{\gamma}_{2n} || \gamma) = \frac{1}{2n} \sum_{i=1}^{2n} I(x_i=1) \cdot \text{ld} \left( \frac{\frac{1}{2n} \sum_{i=1}^{2n} I(x_i=1)}{p_1} \right) + \\ + \frac{1}{2n} \sum_{i=1}^{2n} I(x_i=2) \text{ld} \left( \frac{\frac{1}{2n} \sum_{i=1}^{2n} I(x_i=2)}{p_2} \right) - \\ = \frac{1}{2} \left[ \frac{1}{n} \sum_{i=1}^n I(x_i=1) \text{ld} \frac{\frac{1}{2n} \sum_{i=1}^{2n} I(x_i=1)}{p_1} + \right. \\ \left. + \frac{1}{n} \sum_{i=1}^n I(x_i=2) \text{ld} \frac{\frac{1}{2n} \sum_{i=1}^{2n} I(x_i=2)}{p_2} \right] + \\ + \frac{1}{2n} \sum_{i=n+1}^{2n} I(x_i=1) \text{ld} \frac{\frac{1}{2n} \sum_{i=1}^{2n} I(x_i>1)}{p_2} + \\ + \frac{1}{2n} \sum_{i=n+1}^{2n} I(x_i=2) \text{ld} \frac{\frac{1}{2n} \sum_{i=1}^{2n} I(x_i=2)}{p_2}$$

$$\hat{p}_{2n} = \frac{1}{2n} \sum_{i=1}^{2n} I(x_i=x) = \frac{1}{2} \underbrace{\frac{1}{n} \sum_{i=1}^n I(x_i=x)}_{\in [p_1]} + \frac{1}{2} \underbrace{\frac{1}{n} \sum_{i=n+1}^{2n} I(x_i>x)}_{\in [p_2]}$$

$$\hat{\gamma}_{2n} = \frac{1}{2} p_1 + \frac{1}{2} p_2 \quad \left| \quad \hat{\gamma}_{2n} \rightarrow \in [\gamma_1] \quad p_n \quad \circledast \quad p_2 \right.$$

$$D(p_{2n} || \gamma_1) = \underline{p_{2n}(1)} \cdot \text{ld} \frac{p_{2n}(1)}{\gamma(1)} + \underline{p_{2n}(0)} \cdot \text{ld} \frac{p_{2n}(0)}{\gamma(0)} \rightarrow \\ = \left( \frac{1}{2} p_{2n}(1) + \frac{1}{2} p_{2n}(0) \right) \text{ld} \left[ \frac{\frac{1}{2} p_{2n}(1) + \frac{1}{2} p_{2n}(0)}{\gamma(1)} \right] + p_{2n}(0) \text{ld} \frac{p_{2n}(0)}{\gamma(0)}$$

$$D(\gamma_{2n} || \gamma) = \frac{1}{2} (\gamma_1(1) + \gamma_1'(1)) \text{ld} \frac{\gamma_1(1) + \gamma_1'(1)}{2\gamma(1)} + \frac{1}{2} (\gamma_1(0) + \gamma_1'(0)) \text{ld} \frac{\gamma_1(0) + \gamma_1'(0)}{2\gamma(0)}$$

$$= \frac{1}{2} (\gamma_1(1) + \gamma_1'(1)) \text{ld} \frac{\gamma_1(1) + \gamma_1'(1)}{2\gamma(1)} + \frac{1}{2} (1 - \gamma_1(1) + 1 - \gamma_1'(1)) \cdot \text{ld} \left[ \frac{2 - \gamma_1(1) - \gamma_1'(1)}{2(1 - \gamma(1))} \right]$$

$$= \frac{1}{2} (1 + \gamma_1') \text{ld} \frac{\gamma_1 + \gamma_1'}{2\gamma} + \frac{1}{2} (2 - \gamma_1 - \gamma_1') \text{ld} \frac{2 - \gamma_1 - \gamma_1'}{2(1 - \gamma)} =$$

$$= \frac{1}{2} (\gamma_1 + \gamma_1') \text{ld} \frac{\gamma_1 + \gamma_1'}{2\gamma} - \frac{1}{2} (\gamma_1 + \gamma_1') \text{ld} \frac{2 - \gamma_1 - \gamma_1'}{2(1 - \gamma)} + \text{ld} \frac{2 - \gamma_1 - \gamma_1'}{2(1 - \gamma)}$$

$$= \frac{1}{2} (\gamma_1 + \gamma_1') \text{ld} \frac{(\gamma_1 + \gamma_1') 2(1 - \gamma)}{2\gamma (2 - \gamma_1 - \gamma_1')} + \text{ld} \frac{2 - \gamma_1 - \gamma_1'}{2(1 - \gamma)}$$

$$D(\gamma_1 || \gamma) = \gamma_1(1) \text{ld} \frac{\gamma_1(1)}{\gamma} + \gamma_1(0) \text{ld} \frac{\gamma_1(0)}{1 - \gamma}$$

$$D(\gamma_1 || \gamma) = \gamma_1 \text{ld} \frac{\gamma_1}{\gamma} + (1 - \gamma_1) \text{ld} \frac{1 - \gamma_1}{1 - \gamma}$$

zu  $\gamma \rightarrow \infty$   $\boxed{\gamma_1 = \gamma_1'}$

$$D(\gamma_{2n} || \gamma) = \frac{1}{2} (\gamma_1 + \gamma_1') \underbrace{\text{ld} \frac{(\gamma_1 + \gamma_1')(1 - \gamma)}{\gamma (2 - \gamma_1 - \gamma_1')}}_{\gamma (2 - \gamma_1 - \gamma_1')} + \text{ld} \frac{2 - \gamma_1 - \gamma_1'}{2(1 - \gamma)}$$

IF:  $\gamma_1 = \gamma_1'$   $D(\gamma_{2n} || \gamma) = \gamma_1 \text{ld} \frac{2\gamma_1(1 - \gamma)}{2\gamma(1 - \gamma_1)} + \text{ld} \frac{1 - \gamma_1}{1 - \gamma}$

$$D(\gamma_{2n} || \gamma) = \gamma_1 \text{ld} \frac{\gamma_1}{\gamma} + (1 - \gamma_1) \text{ld} \frac{1 - \gamma_1}{1 - \gamma} = D(\gamma_1 || \gamma)$$

$$D(\hat{P}_{un} || \gamma) = \frac{1}{m} \left[ \hat{P}_{u1}(1) + \hat{P}_{u1}(1) + \dots + \hat{P}_{un}(1) \right] \text{ld} \frac{1}{m} \left[ \hat{P}_{u1}(1) + \dots + \hat{P}_{un}(1) \right]$$

$$+ \frac{1}{m} \left[ \hat{P}_{u1}(0) + \hat{P}_{u2}(0) + \dots + \hat{P}_{un}(0) \right] \text{ld} \frac{\frac{1}{m} \left[ \hat{P}_{u1}(0) + \dots + \hat{P}_{un}(0) \right]}{\gamma(0)}$$

$$\rightarrow \mathbb{E}[P_{ui}(1)] \text{ld} \frac{\mathbb{E}[P_{ui}(1)]}{\gamma(1)} + \dots + \mathbb{E}[P_{ui}(0)] \text{ld} \frac{\mathbb{E}[P_{ui}(0)]}{\gamma(0)}$$

(b) ARBITRARY DISCRETE "X"  $x \in \{x_1, x_2, \dots, x_n\}$   
 $\in [D(\hat{P}_n || \gamma)] \leq \in [D(P_{n-n} || \gamma)]$

$$D(\hat{P}_n || \gamma) = \hat{P}_n(x_1) \log \frac{\hat{P}_n(x_1)}{\gamma(x_1)} + \hat{P}_n(x_2) \log \frac{\hat{P}_n(x_2)}{\gamma(x_2)} + \dots$$

$$+ \hat{P}_n(x_n) \log \frac{\hat{P}_n(x_n)}{\gamma(x_n)} = \mathbb{E}_{\hat{P}_n} \left[ \log \left( \frac{\hat{P}_n(x)}{\gamma(x)} \right) \right]$$


---

- RETURN TO P.67\*

$$D(\hat{P}_{2n} || \gamma) = \hat{P}_{2n}(1) \log \frac{\hat{P}_{2n}(1)}{\gamma(1)} + \hat{P}_{2n}(0) \log \frac{\hat{P}_{2n}(0)}{\gamma(0)}$$

$$\hat{P}_{2n}(1) = \frac{1}{2} P_n(1) + \frac{1}{2} P_n(1) \xrightarrow{\text{AS}} \in [P_n(1)] = \gamma(1)$$

$$P_n(1) = \sum_{i=1}^n I(X_i=1) \quad \left. \begin{array}{l} \text{sono coste } \gamma = \text{tasse} \\ \text{e così, se } \gamma \text{ è fissa,} \\ \gamma(1) \text{ è solo una} \\ \text{parte dei costi.} \end{array} \right\}$$

SOGNATO SAREBBERO SÌ GOLEM, BROVI, PARCE  
 NEGOI GOLEM SÌ BROVI PARCE  
 POPOLI SÌ POLIZIA SÌ PRIMEROSI C'È  
 BIS DI ORGANIZZARSI NA TERRITORIO  
 NGE NA CITTÀ.

$$D(\hat{P}_n || \gamma) \xrightarrow{n \rightarrow \infty} 0 \Rightarrow D(\hat{P}_{2n} || \gamma) < D(\hat{P}_n || \gamma)$$

$$D(\lambda \gamma_1 + (1-\lambda) \gamma_2 || \gamma) \geq \lambda D(\gamma_1 || \gamma) + (1-\lambda) D(\gamma_2 || \gamma) \quad \boxed{0 \leq \lambda \leq 1}$$

→ CONVEXITY OF REACTIVE ENTROPY

solv-2-229 SOLUZIONE

$$P_n'(x) = \frac{1}{n} \sum_{i=1}^n I(X_{n+i} = x) \quad \text{FOR } n \in \mathbb{Z}$$

- N.67\*  $\Rightarrow \hat{P}_{2n}' = (\hat{P}_n' + \hat{P}_n') \cdot \frac{1}{2}$   
 → CONVEXITY OF REACTIVE ENTROPY

$$D(P_{2n} || \gamma) = D\left(\frac{1}{2} \hat{P}_n' + \frac{1}{2} \hat{P}_n' || \frac{1}{2} \gamma + \frac{1}{2} \gamma\right) \leq \frac{1}{2} D(\hat{P}_n' || \gamma) +$$

$$+ \frac{1}{2} D(\hat{P}_n' || \gamma) / \in [\dots]$$

$$E[D(\gamma_{2n} || \gamma)] = \frac{1}{2} E[D(\gamma_1 || \gamma)] + \frac{1}{2} E[D(\gamma_2 || \gamma)] = E[D(\tilde{\gamma}_n || \gamma)]$$

$\hat{p}_{2n}, \hat{p}_n, \tilde{\gamma}_n$  are random variables taking values in the space of probability distributions of  $X$ .

(b) For  $1 \leq i \leq n$ , let  $\gamma_n^{(i)}$  denote the empirical distribution of  $(n-i)$  random variables  $x_j$   $1 \leq j \leq n$ ,  $j \neq i$ , i.e.

$$\hat{p}_n^{(i)}(x) = \frac{1}{n-i} \sum_{j=1, j \neq i}^n \mathbb{1}(x_j = x) \text{ for } x \in \mathcal{X}$$

$$\hat{p}_n = \frac{1}{n} \sum_{i=1}^n \hat{p}_n^{(i)}$$

- By using convexity of relative entropy function

$$D(\tilde{\gamma}_n || \gamma) \leq \frac{1}{n} \sum_{i=1}^n D(\gamma_n^{(i)} || \gamma) \quad \begin{array}{l} \text{LEH LOGKA} \\ \text{KAKO BA KES} \\ \text{SOT, } \end{array}$$

$$E[D(\tilde{\gamma}_n || \gamma)] \leq \frac{1}{n} \sum_{i=1}^n E[D(\gamma_n^{(i)} || \gamma)] = E[D(\hat{p}_{n-1} || \gamma)]$$

$$E[D(\hat{p}_n^{(i)} || \gamma)] = E[D(\hat{p}_{n-1} || \gamma)] \text{ FOR } \forall i$$

• RECALL: Log-sum inequality

$$\sum_{i=1}^n a_i \ln \frac{a_i}{b_i} \geq \left( \sum_{i=1}^n a_i \right) \ln \frac{\sum_{i=1}^n a_i}{\sum_{i=1}^n b_i}$$

• Theorem 2.7.2 (Convexity of relative entropy)

$$D(\lambda \gamma_1 + (1-\lambda) \gamma_2 || \lambda \gamma_1 + (1-\lambda) \gamma_2) \leq \lambda D(\gamma_1 || \gamma_1) + (1-\lambda) D(\gamma_2 || \gamma_2)$$

$$(\lambda \gamma_1 + (1-\lambda) \gamma_2) \ln \frac{\lambda \gamma_1 + (1-\lambda) \gamma_2}{\lambda \gamma_1 + (1-\lambda) \gamma_2} \leq \lambda \gamma_1 \ln \frac{\gamma_1}{\gamma_1} + (1-\lambda) \gamma_2 \ln \frac{\gamma_2}{\gamma_2}$$

$(\gamma_1, \gamma_1)$  &  $(\gamma_2, \gamma_2)$  are two pairs of probability mass functions.

**3.13** CALCULATION OF TYPICAL SET: To clarify the notion of typical set  $A_{\epsilon}^{(n)}$  and the smalleset set of high probability  $A_{\epsilon}^{(n)}$ , we will calculate the set for a simple example. Consider a sequence of 1-dimensional binary random variables,  $x_1, x_2, \dots, x_n$  where the probability that  $x_i = 1$  is 0.6 ( $\gamma = 0.6, 1 - \gamma = 0.4$ ).

(a) Calculate  $H(x)$

(b) With  $n = 25$  and  $\epsilon = 0.1$ , which sequences fall in the typical set  $A_{\epsilon}^{(n)}$ ? What is the probability of the typical set? How many elements are there in the typical set? (This involves computation of the table of probabilities for sequences with  $k$  1's,  $0 \leq k \leq 25$ , and finding those sequences that are in the typical set.)

(c) How many elements are there in the smalleset set that has probability 0.9?

(d) How many elements are there in the intersection of the sets in (b) and (c)? What is the probability of this intersection.

$$2^{-n(H+\epsilon)} \leq p(x^n) \leq 2^{-n(H-\epsilon)}$$

$$\Pr\{A_{\epsilon}^{(n)}\} \geq 1 - \epsilon \quad |A_{\epsilon}^{(n)}| \leq 2^{n(H+\epsilon)} \quad |A_{\epsilon}^{(n)}| \geq (1-\epsilon)2^{n(H-\epsilon)}$$

$$(H-\epsilon) \leq -\frac{1}{n} \text{ld } p(x^n) \leq (H+\epsilon)$$

all sequences for which  $\epsilon$  is this  $\epsilon$  typical set

(a)  $H(x) = \gamma \text{ld } \frac{1}{\gamma} + (1-\gamma) \text{ld } \frac{1}{1-\gamma} = 0.6 \text{ld } \frac{1}{0.6} + 0.4 \text{ld } \frac{1}{0.4} = 0.970951$

(b)  $n = 25 \quad 1 \leq k \leq 25$

TOTAL NUMBER OF SEQUENCES WITH GIVEN "K" IS

$\binom{n}{k}$  e.g.  $\binom{25}{1} = \frac{25!}{24!} = 25$   $\rightarrow$  there are 25 sequences with 1's one

PROBABILITY OF THOSE SEQUENCES IS

$$\binom{n}{k} \cdot \gamma^k \cdot (1-\gamma)^{n-k} \quad \text{e.g. } 25 \cdot (0.6)^1 \cdot (0.4)^{24} = 4.22212 \cdot 10^{-9}$$

$$-\frac{1}{n} \text{ld } ((0.6)^1 \cdot (0.4)^{24}) = \underline{1.29853}$$

•  $A_{\epsilon}^{(n)}$  VO OVA PROZESTVO MISCAM DENTA RADANT SITE SEKVENCIJ SO  $25 \cdot 0.6 = 15,0$  EDINICI NO

$$0.970951 \cdot 0.1 = 0.870951 \leq -\frac{1}{25} \text{ld } ((0.6)^{15} \cdot (0.4)^{10}) = 0.970951 \leq \frac{0.970951 + 0.1}{1.070951} = 0.970951$$

$$H - \epsilon \leq -\frac{1}{n} \ln p(x^n) \leq H + \epsilon \quad -\epsilon \leq -\frac{1}{n} \ln p(x^n) - H \leq \epsilon.$$

$$\left| 1 - \frac{1}{n} \ln p(x^n) - H \right| \leq \epsilon$$

SKETCHED FORMULA  
USING LOGARITHM  
METHOD

$$\Pr\{A_{\epsilon}^{(n)}\} \geq 1 - \epsilon \geq 0.9$$

ON A SO 15-POLYHEDRAL  
SITE SURVEY SO:  
11, 12, ..., 13 CORRECT

MVN

- PROBABILITY OF TYPICAL SET IS:

$$P_A = \sum_{i=11}^{19} \binom{25}{i} p^i 1^{25-i} = \underline{0.936246}$$

$$P_A = 0.936246 \geq 0.9 = 1 - \epsilon$$

ON A 6-GRADUATE HOME-  
WORK. IS THIS A  
PROBABLY TO THE UNIVERSITY  
OF TORONTO.

- THE NUMBER OF ELEMENTS IN REPLICATE SET IS:

$$N_A = \sum_{i=11}^{19} \binom{25}{i} = \underline{26.366.510} \text{ ELEMENTS}$$

$$|A_{\epsilon}^{(n)}| \leq 2^{n(H+1+\epsilon)} = 2^{25(1.070951)} = 114.737.700$$

$$\frac{26.366.510}{8+7+6+4+3+2} = 15+10+7 = 30$$

$$P_B = 0.89284 \doteq 0.9$$

- MINIMUM SET  $[12, 19]$

$$N_B = \sum_{i=12}^{19} \binom{25}{i} = 21.909.110$$

$$P_B = 0.9$$

- NUMBER OF ELEMENTS IN THE INTERSECTION

$$|A_{\epsilon}^{(n)} \cap B_{\epsilon}^{(n)}| = |B_{\epsilon}^{(n)}| = 21.909.110 \text{ ELEMENTS}$$

$$\textcircled{1} \quad \frac{1}{n} \ln |B_{\epsilon}^{(n)}| > H - \delta'$$

$$H = 0.970951$$

$$\frac{1}{25} \ln (21.909.110) = 0.97540$$

$$\delta' = \frac{1}{2} + \frac{1}{n} \left| \ln \left( \frac{1}{2} - \delta \right) \right| = \frac{1}{2} + \frac{1}{25} \left| \ln (0.4) \right| = 0.53665$$

$$H - \delta' = 0.970951 - 0.53665 = 0.43430$$

?

SUM REMAINING SIDE =  $\frac{1}{2} + \frac{1}{25} \ln (0.4)$

?

$$25+24+23+\dots+2 = \frac{4 \cdot (n+1)}{2} - 1$$

Fasaderex Pi

$$S = 1 + 2 + 3 + \dots + n$$

$$S = n + n-1 + n-2 + \dots + 1$$

$$2S = n+1 + n+1 + n+1 + \dots + (n+1)$$

$$2S = (n+1) \cdot n$$

$$S = \frac{(n+1) \cdot n}{2}$$

- Vez orov na simulacie vo male obdov deva minimozem set e so súčinnat sekvencie zo tohto miestu mnoho:

12, 13, 14, ..., 20 coinc

Súčet a reprezentuje na one sekvenci je  $\geq 0.9$ , vtedy totéž množestvo zo DMZ bude na ešte zto 40 (8102) množov.

$$\sum_{i=12}^{20} \binom{25}{i} \cdot 2^{25-i} = 0.9127 > \sum_{i=12}^{20} \binom{25}{i} = 21.962.240$$

$$|A_{\epsilon}^{(n)} \cap B_{\epsilon}^{(n)}| = \sum_{i=12}^{19} \binom{25}{i} = 21.909.110 \quad \text{①}$$

Nájsť a pres eksp e dgára

- V<sub>0</sub> # scúčajte sum na mobil / statické veciost

a ne sum báz množestvo da má minimozem množ v elementi.

WEIGHT COLLEGE SOLUTION

$$(c) N \cdot p^2 \cdot q^{13} \leq 0.9 - \sum_{j=13}^{25} \binom{25}{j} \cdot j \cdot 2^{25-j}$$

Základné skúčajte so 12 coinc za da veciost a veciost bude 0.3

$$N \leq \frac{0.9 - \sum_{j=13}^{25} \binom{25}{j} \cdot j \cdot 2^{25-j}}{1^{12} \cdot 2^{13}}$$

$$N = 3.68067 \cdot 10^6$$

- Znači veciost a množ v elementi vo minimozem.

$$P = N \cdot p^2 \cdot q^{13} + \sum_{j=13}^{25} \binom{25}{j} \cdot j \cdot 2^{25-j}$$

$$|B_{\delta}^{(n)}| = N + \sum_{j=13}^{25} \binom{25}{j} = 2.04573 \cdot 10^7$$

$$|\beta_8^{(n)}| = |A_e^{(n)}| = 2^{n+1} = 2^{25 \cdot 0.970951} = \underline{2.0283 \cdot 10^7}$$

- Avo se storedi nötöök tresserat (LÄÄRNUK AS MAALI), määratud on kultuur kolleegi, se glööva deka (① vs ② N.72):

$$\textcircled{1} = 21,90911 \cdot 10^6$$

$$\textcircled{2} = 20,4579 \cdot 10^6$$

$$\boxed{2^{n+1} = 20,287 \cdot 10^6}$$

$$2^{n+1 + e} = 114,7377 \cdot 10^6$$

$$n+1 = 25 \cdot 0,97 =$$

DEFINITION:  $\textcircled{2}$  on üksik e koe ja ne so vooliv vo mäice 20370 fügleeduvad ohi gruppi ra servenci, a ne olec on servenci. See parbott ja optimizatsioon probleem.

$$(d) |A_e^{(n)} \cap \beta_8^{(n)}| = |A_{0..1}^{(n)} \cap \beta_{0..1}^{(n)}| =$$

$$A_e^{(n)} \in \{11, 12, \dots, 19\}$$

$$\beta_8^{(n)} \in \{\text{part of } 12, 13, \dots, 25\}$$

$$|A_{0..1}^{(n)} \cap \beta_{0..1}^{(n)}| \Rightarrow k \in \{\text{part of } 12, 13, \dots, 19\}$$

$$|A_{0..1}^{(n)} \cap \beta_{0..1}^{(n)}| = N + \sum_{j=12}^{19} \binom{25}{j} = 3,68067 \cdot 10^6 + 16,708210 \\ = 2.03895 \cdot 10^7$$

$$P_{A \cap B} = N \cdot 2^{12} 2^{13} + \sum_{j=12}^{19} \binom{25}{j} q^j 2^{25-j} = \underline{0.87064}$$

Süstema 20370  $\beta_{0..1}^{(n)}$ :  $k \in \{\text{part of } 12, 13, \dots, 25\}$

C ZADOKA 380  $q > 0$  PA 200000 ZADOKU VK  
OD K=25 TA SUMMA VADOKU - TAKA  
102400 (ZA lõimal BAG RA EERUVI) SE SITGA + 90  
VADOKA VEDOTATUST 0.9. **MMV**

## CHAPTER 4 Empirical rates of ergodic processes

AEP ESTABLISH THAT  $H \cdot h(x)$  BITS SUFFICE ON AVERAGE TO DESCRIBE  $\underline{H \text{ INDEPENDENT AND IDENTICALLY DISTRIBUTED RANDOM VARIABLES}}$ .

- If  $X_1, X_2, \dots, X_n$  are dependent we will show that the ex.  $\text{ENTROPY } H(X_1, X_2, \dots, X_n)$  grows (asymptotically) linearly with  $n$  at rate  $H(x)$ , which we will call THE ENTROPY RATE of the process.

### 4.1 MARKOV CHAINS

A stochastic process  $\{X_i\}$  is an indexed sequence of random variables. In general there can be arbitrary dependence among the random variables. Process is characterized by the joint probability mass function:

$$P_r \{ (X_1, X_2, \dots, X_n) = (x_1, x_2, \dots, x_n) \} = \gamma(x_1, x_2, \dots, x_n), \\ (x_1, x_2, \dots, x_n) \in \mathcal{X}^n \text{ for } n = 1, 2, \dots. \quad \text{MMV}$$

#### DEFINITION

A stochastic process is said to be STATIONARY if the joint distribution of any subset of the sequence of random variables is invariant with respect to shifts in time index, that is:

$$P_r \{ X_1 = x_1, X_2 = x_2, \dots, X_n = x_n \} = P_r \{ X_{1+l} = x_1, X_{2+l} = x_2, \dots, X_{n+l} = x_n \} \quad \checkmark$$

FOR every  $n$  AND every shift " $l$ ", AND FOR ALL  $x_1, x_2, \dots, x_n \in \mathcal{X}$

**DEFINITION** A discrete stochastic process  $X_1, X_2, \dots$  is said to be a MARCOV CHAIN or MARCOV PROCESS if for  $n = 1, 2, \dots$

$$P_r (X_{n+m} = x_{n+m} | X_n = x_n, X_{n-1} = x_{n-1}, \dots, X_1 = x_1) = P_r (X_{n+m} = x_{n+m} | X_n = x_n) \quad \text{for all } x_1, x_2, \dots, x_n, x_{n+1} \in \mathcal{X}$$

$$P(X_1, X_2, \dots, X_n) = \gamma(x_1) \cdot \gamma(x_2 | x_1) \cdot \gamma(x_3 | x_2) \cdots \gamma(x_n | x_{n-1})$$

$$\begin{aligned} \gamma(x_1, x_2, \dots, x_n) &= \gamma(x_1) \cdot \gamma(x_2 | x_1) \cdot \gamma(x_3 | x_2, x_1) \cdots \gamma(x_n | x_{n-1}, \dots, x_1) \\ &= \gamma(x_1) \cdot \underbrace{\gamma(x_2 | x_1) \cdot \gamma(x_3 | x_1, x_2) \cdots \gamma(x_n | x_1, \dots, x_{n-1})}_{P(X_2, \dots, X_n | X_1)} = \gamma(x_1) \cdot \gamma(x_2 | x_1) \\ &\cdot \gamma(x_3 | x_2) \cdot \gamma(x_4 | x_1, x_2, x_3) \cdots \gamma(x_n | x_1, \dots, x_{n-1}) \end{aligned}$$

FOR TIME INVARIANT MARKOV CHAIN:

$$\Pr(X_{n+1}=6|X_n=9) = \Pr\{X_1=6|X_0=9\} \text{ for all } n \in \mathbb{N}$$

- If  $\{X_n\}$  is a Markov Chain,  $X_n$  is called THE STATE AT TIME  $n$ . A TIME INVARIANT MARKOV CHAIN IS CHARACTERIZED BY ITS INITIAL STATE AND PROBABILITIES TRANSITION MATRIX.

$$P = [P_{ij}], i, j \in \{1, 2, \dots, m\} \quad P_{ij} = \Pr\{X_{n+1}=j|X_n=i\}$$

IF IT IS POSSIBLE TO GO WITH POSITIVE PROBABILITY FROM ANY STATE OF MARKOV CHAIN TO ANY OTHER STATE IN THE FINITE NUMBER OF STEPS, THE MARKOV CHAIN IS SAID TO BE IRREDUCIBLE.

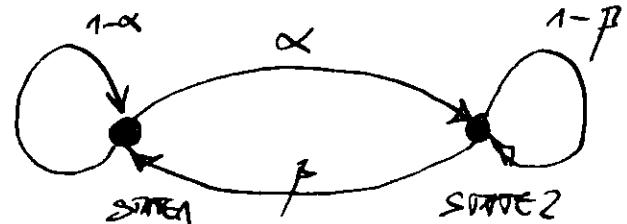
- IF THE PROBABILITY MASS FUNCTION OF RANDOM VARIABLE AT TIME  $n$  IS  $f(x_n)$  THE PROBABILITY MASS FUNCTION AT TIME  $(n+1)$  IS:

$$f(x_{n+1}) = \sum_{x_n} f(x_n) \cdot P_{x_n x_{n+1}}$$

A DISTRIBUTION OF STATES SUCH THAT DISTRIBUTION AT TIME  $n+1$  IS THE SAME AS THE DISTRIBUTION AT  $n$  IS CALLED STATIONARY DISTRIBUTION.

**EXAMPLE 4.1.1** CONSIDER A TWO STATE MARKOV CHAIN WITH PROBABILITIES TRANSITION MATRIX

$$\checkmark P = \begin{bmatrix} 1-\alpha & \alpha \\ \beta & 1-\beta \end{bmatrix}$$



$\mu$ -VECTOR REPRESENTING STATIONARY PROBABILITIES OF STATES 1 & 2, RESPECTIVELY.

$$\boxed{\mu \cdot P = \mu} \quad \text{FIND STATIONARY PROBABILITIES BY SOLVING THIS EQUATION.}$$

$$[\mu_1 \mu_2] \begin{bmatrix} 1-\alpha & \alpha \\ \beta & 1-\beta \end{bmatrix} = [\mu_1(1-\alpha) + \mu_2 \cdot \beta, \alpha \mu_1 + (1-\beta) \mu_2]$$

$$[\mu_1 \mu_2] = [\mu_1, \mu_2]$$

$$\mu_1(1-\alpha) + \mu_2 \beta = \mu_1 \quad -\alpha \mu_1 + \mu_2 \beta = 0 \quad \alpha \mu_1 = \mu_2 \beta$$

$$\alpha \mu_1 + (1-\beta) \mu_2 = \mu_2 \quad \alpha \mu_1 - \beta \mu_2 = 0 \quad (\alpha \mu_1 = \beta \mu_2)$$

$$\boxed{\mu_1 + \mu_2 = 1} \quad \mu_1 = 1 - \mu_2 \quad \alpha(1 - \mu_2) = \beta \mu_2$$

$$\alpha - \alpha p_2 = p_1 p_2 \quad (\alpha + p)p_2 = \alpha$$

$$p_2 = \frac{\alpha}{\alpha + p}$$

$$M_1 = \frac{p}{\alpha} \cdot p_2 = \frac{p^2}{\alpha + p}$$

- If process is stationary the entropy of the state  $x_n$  at time  $n$  is:

$$H(x_n) = H\left(\frac{p}{\alpha+p}, \frac{\alpha}{\alpha+p}\right) = \frac{p}{\alpha+p} \log \frac{\alpha+p}{p} + \frac{\alpha}{\alpha+p} \log \frac{\alpha}{\alpha+p}$$

$$H(X|X) = \sum_{x \in X} p(x) H(X|x=x)$$

$$H(X|X) = \sum_{x,y} p(x,y) \log \frac{1}{p(y|x)}$$

## 4.2 Entropy Rate

If we have a sequence of random variables, a natural question to ask is: How does the entropy of the sequence grow with  $n$ ? We define the entropy rate as this rate of growth as follows.

Definition: The entropy of stochastic process

$\{x_i\}$  is defined by

$$H(X) = \lim_{n \rightarrow \infty} \frac{1}{n} H(x_1, x_2, \dots, x_n)$$

when the limit exists.

EXAMPLES:

1.) Typewriter:  $w$  - equally likely output letters  
 $w^n$  - sequences of length  $n$ , all equally likely.

$$H(x_1, x_2, \dots, x_n) = \sum_{i=1}^n \frac{1}{m} \cdot \log m = n \cdot \frac{1}{m} \cdot \log m$$

$H(x_1, x_2, \dots, x_n) = \log m$  the entropy rate is:

$$H(X) = \lim_{n \rightarrow \infty} \frac{1}{n} H(x_1, x_2, \dots, x_n) = \frac{1}{m} \log m = \underline{(1/m)}$$

2.)  $x_1, x_2, \dots, x_n$  are i.i.d random variables. Then

$$H(X) = \lim_n \frac{H(x_1, x_2, \dots, x_n)}{n} = \lim_n \frac{n H(x_1)}{n} = H(x_1)$$

3.) Sequence of independent but not identically distributed random variables

$$H(x_1, x_2, \dots, x_n) = \sum_{i=1}^n H(x_i)$$

E.G. SEQUENCE OF PISOR PARTITION SUCH IS:

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n H(x_i) \text{ DOES NOT EXIST}$$

$P_i = P(X_i=1)$  IS CONSTANT BUT FUNCTION OF  $i =$

$$P_i = \begin{cases} 0.5 & \text{IF } 2k \leq i \leq 2k+1 \\ 0 & \text{IF } 2k+1 \leq i \leq 2k+2 \end{cases} \text{ FOR } k = 0, 1, 2, \dots$$

$H(x_i)$  WILL OSCILLATE BETWEEN  $0.5 \text{ AND } 1$ , AND WILL NOT HAVE THE CRIMIT.

• QUANTITY OF ENTROPY RATE: (Alternative Definition)

$$H'(X) = \lim_{n \rightarrow \infty} H(X_1 | X_{n-1}, X_{n-2}, \dots, X_1)$$

$H(X) \sim$  PER SYMBOL ENTROPY OF  $4^2$  POSSIBLE VARIANCES.

$H'(X) \sim$  CONDITIONAL ENTROPY OF LAST SYMBOL VARIABLE GIVEN THE REST.

**Theorem 4.2.1** For STATIONARY STOCHASTIC PROCESS THE LIMITS  $H(X)$  AND  $H'(X)$  EXIST AND ARE EQUIV:

$$H(X) = H'(X)$$

$$H(X_1, X_2, \dots, X_n) = H(X_1) + H(X_2 | X_1) + H(X_3 | X_2, X_1) \dots$$

$$H(X_1, X_2, \dots, X_n) = \sum_{i=1}^n H(X_i | X_{i-1}, \dots, X_1)$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} H(X_1, X_2, \dots, X_n) = \sum_{i=1}^n \underbrace{\lim_{n \rightarrow \infty} \frac{1}{n} H(X_i | X_{i-1}, \dots, X_1)}_{\text{IF EXISTS}}$$

THIS CRIT EXISTS ALSO

**Theorem 4.2.2** For STATIONARY STOCHASTIC PROCESS  $H(X_n | X_{n-1}, \dots, X_1)$  IS NONINCREASING IN  $4^2$  AND HAS LIMIT  $H'(X)$ .

PROOF:  $H(X_{n+1} | X_1, X_2, \dots, X_n) \leq H(X_{n+1} | X_n, \dots, X_1) = H(X_n | X_{n-1}, \dots, X_1)$

SEE TO DEFINITION OF STATIONARY STOCHASTIC PROCESS.

**Theorem 4.2.3** (Converse mean) If  $a_i \rightarrow a$  THEN  $\sum a_i \rightarrow a$

PROOF: Let  $\epsilon > 0$ . Since  $a_n \rightarrow a$  there exist numbers such that  $|a_i - a| < \epsilon$  for all  $i \geq N(\epsilon)$ . Hence,

$$|b_n - a| = \left| \frac{1}{n} \sum_{i=1}^n (a_i - a) \right| \leq \frac{1}{n} \sum_{i=1}^n |a_i - a|$$

$$\leq \frac{1}{n} \sum_{i=1}^{N(\epsilon)} |a_i - a| + \frac{n - N(\epsilon)}{n} \cdot \epsilon \leq \frac{1}{n} \sum_{i=1}^{N(\epsilon)} |a_i - a| + \epsilon$$

FIRST TERM GOES TO 0 AS  $n \rightarrow \infty$ , we can make  $|b_n - a| \leq 2\epsilon$  by taking  $n$  large enough. Hence,  $b_n \rightarrow a$  as  $n \rightarrow \infty$ .

PROOF OF THEOREM 4.2.1 By chain rule

$$H(X_1, X_2, \dots, X_n) = \sum_{i=1}^n H(X_i | X_{i-1}, \dots, X_1)$$

$$\frac{H(X_1, X_2, \dots, X_n)}{n} = \frac{1}{n} \sum_{i=1}^n H(X_i | X_{i-1}, \dots, X_1)$$

ENTROPY RATE IS AVERAGE OF CONDITIONAL ENTROPIES.

T4.2.3

$$\frac{1}{n} \sum_{i=1}^n H(X_i | X_{i-1}, \dots, X_1) = H(X_n | X_{n-1}, \dots, X_1)$$

$$\rightarrow H(X) = \lim_{n \rightarrow \infty} \frac{1}{n} H(X_1, X_2, \dots, X_n) = \lim_{n \rightarrow \infty} H(X_n | X_{n-1}, \dots, X_1) = H'(X)$$

T4.2.1

(PROOF IN 168)

AEP FOR STATIONARY ERGODIC PROCESS

For ANY STATIONARY ERGODIC PROCESS:

$$-\frac{1}{n} \log \left[ \frac{1}{n} H(X_1, X_2, \dots, X_n) \right] \rightarrow H(X)$$

**Markov Chains** For STATIONARY MARKOV CHAIN, THE ENTROPY RATE IS GIVEN BY:

$$H(X) = H'(X) = \lim H(X_n | X_{n-1}, \dots, X_1) = \lim H(X_n | X_1)$$

CONDITIONAL ENTROPY  $\rightarrow$  CALCULATED USING THE GIVEN STATIONARY DISTRIBUTION. STATIONARY DISTRIBUTION  $\pi$  IS SOLUTION OF THE EQUATIONS

$$\pi_j = \sum_i \pi_i p_{ij} \text{ for all } j =$$

$$[\mu_1, \mu_2] = [\mu_1, \mu_0] \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix} = [\mu_1] \cdot [p_{ij}]$$

$$\mu_i = \sum_{j=1}^2 \mu_j \cdot p_{ij}$$

(MORA INTEGRALA ZA SUMA NA VEROATNOSTI TE-TRANSICIJA KOI OD BILO KOTA SODRGA "1" VEDR VO SISTEMU  $i, j = 2$ )

$$\mu_1 = \mu_1(1-\alpha) + \mu_2 \cdot \beta = \sum_{i=1}^2 \mu_i \cdot p_{i1}$$

MMV

Theorem 4.2.4

Let  $\{x_i\}$  be a stationary Markov chain with stationary probability matrix  $P$ .

FOR  $x \sim \pi$  AND TRANSITION  
LET  $(x_i \sim \pi)$  THEN THE

MMV

$$H(x) = - \sum_{i,j} \mu_i \cdot p_{ij} \ln p_{ij}$$

DA E KLASIKA:  
PROCESOT PRODA OT C VD SISTEMO,  $i \in \{1, 2, \dots, n\}$   
I PREGAZ VD SISTEMA  
NE SO VEROVATNOST  $p_{ij}$   
ANO VICE VD SISTEMA  $\pi_i$

PROOF:  $H(x) = H(x_2 | x_1) = \sum_i \mu_i \sum_j p_{ij} \ln \frac{1}{p_{ij}}$

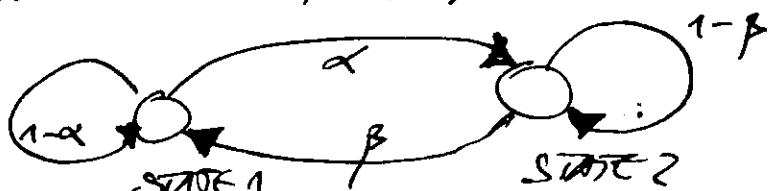
$$H(Y | x=x) = - \sum_{y \in Y} p(y | x=x) \ln p(y | x)$$

$$H(Y | x) = \overline{H(Y | x=x)} = - \sum_{x \in X} p(x) p(y|x) \ln p(y|x)$$

$$= \sum_x \sum_y p(x, y) \ln p(y|x) = \sum_x p(x) H(Y | x=x)$$

Example 4.2.1 (Two State Markov Chain) The  
entropy rate of two state Markov chain

is:



$$H(x) = - \sum_{i,j} \mu_i \cdot p_{ij} \ln p_{ij}$$

$$p_{ij} = \begin{bmatrix} 1-\alpha & \alpha \\ \beta & 1-\beta \end{bmatrix}$$

$$= \sum_i \mu_i \sum_j p_{ij} \ln \frac{1}{p_{ij}} = \mu_1 \sum_j p_{1j} \ln \frac{1}{p_{1j}} +$$

$$\mu_2 \sum_j p_{2j} \ln \frac{1}{p_{2j}} = \mu_1 [\alpha \ln \frac{1}{\alpha} + (1-\alpha) \ln \frac{1}{1-\alpha}] + \mu_2 [\beta \ln \frac{1}{\beta} + (1-\beta) \ln \frac{1}{1-\beta}]$$

$$H(x) = \underbrace{\frac{p}{\alpha+p} H(\alpha)}_{\mu_1} + \underbrace{\frac{1-p}{\beta+\gamma} H(\beta)}_{\mu_2}$$

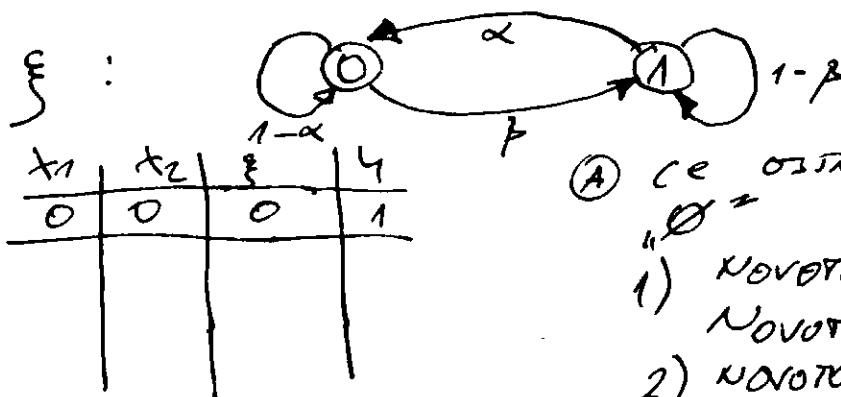
Mod 1 binärer Kreis zu Matrix Werte 0  
 P(X=x)  $\rightarrow$   $\begin{cases} x_1 \in (0,1) & P(x_1) = \left[ \begin{array}{c} 0 \\ 1 \end{array} \right] \\ x_2 \in (0,1) & P(x_2) = \left[ \begin{array}{c} 1 \\ 0 \end{array} \right] \end{cases}$

- Prozessor  $\leftarrow$ :  $\xi = \underline{x_1 \oplus x_2} \mod 2$

$$\xi \in [0,1]$$

$x_1$	$x_2$	$\xi$
0	0	0
0	1	1
1	0	1
1	1	0

?



- (A) ce este valoarea "0" de sortat  
 "0" = nu:  
 1) nu este 1 rezultat  $x_1 = 0$   
 2) nu este 1 rezultat  $x_2 = 0$   
 3) nu este 1 rezultat  $x_1 = 1$   
 4) nu este 1 rezultat  $x_2 = 1$

VERUZATNOSTA DA PROZESSOR  $\xi$  NE JA SUNEI  
 SOSROT PADA VO "0" E:

$$P_0 = \frac{1}{2} \cdot \frac{1}{3} + \frac{1}{2} \cdot \frac{2}{3} = \frac{1}{6} + \frac{2}{6} = \frac{3}{6} = \frac{1}{2} = \mu_1$$

(B)  $\xi$  ce BIDE VO SOSROT PADA, "1" = nu:

$$1) \underline{x_1=0, x_2=1} \quad \text{I.e. } x_1=1, x_2=0$$

$$P_1 = \frac{1}{2} \cdot \frac{2}{3} + \frac{1}{2} \cdot \frac{1}{3} = \frac{2}{6} + \frac{1}{6} = \frac{3}{6} = \frac{1}{2} = \mu_2$$

(C)  $\xi$  ce REPREZINE VO SOSROT PADA 1 DE SOSROTADA

$x_1=0$	$x_2=0$	$n$	$\gamma+1$
0	0		1
0	1		
1	0		
1	1	0	1

$$\begin{aligned}
 P(1|0) &= P(00) \left[ P(x_1=0) + P(x_2=1) \right] + \\
 &\quad P(00) \left[ P(x_1=1) + P(x_2=0) \right] + \\
 &\quad P(11) \left[ P(x_1=0) + P(x_2=1) \right] + P(11) \left[ P(x_1=1) + P(x_2=0) \right] \\
 &= (P_{11} + P_{00}) \left[ P(x_1=0) + P(x_2=1) \right] + (P_{11} + P_{00}) \left[ P(x_1=1) + P(x_2=0) \right]
 \end{aligned}$$

$$= \left( \frac{2}{6} + \frac{1}{6} \right) \left[ \frac{1}{2} + \frac{2}{3} \right] + \left( \frac{2}{6} + \frac{1}{6} \right) \left[ \frac{1}{2} + \frac{1}{3} \right] = \frac{1}{2} \cdot \frac{3+4}{6} + \frac{1}{2} \cdot \frac{3+2}{6} = \frac{12}{12} = 1$$

	$x_1$	$x_2$	$P(x_1=0, x_2=0)$
	0	0	0
	1	0	0
	0	1	0
	1	1	0

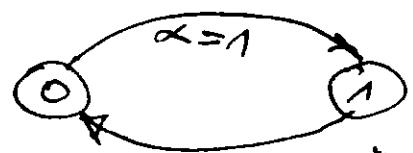
$$P(1|0) = P(01)[P(x_1=1) + P(x_2=1)] + \\ P(00)[P(x_1=0) + P(x_2=0)] + P(10) \cdot [P(x_1=0) \\ + P(x_2=0)] + P(11) \cdot [P(x_1=1) + P(x_2=1)]$$

$$P(1|0) = [P(01) + P(10)] \cdot [P(x_1=1) + P(x_2=1) + [P(01) + P(10)]].$$

$$\cdot [P(x_1=0) + P(x_2=0)] = \left(\frac{2}{6} + \frac{1}{6}\right)\left(\frac{1}{2} + \frac{2}{3}\right) + \left(\frac{2}{6} + \frac{1}{6}\right) \cdot$$

$$\left(\frac{1}{2} + \frac{1}{3}\right) = \frac{1}{2} \frac{3+4}{6} + \frac{1}{2} \frac{1+2}{6} = \frac{12}{12}$$

$$[\alpha=1, \beta=1]$$



$$P_1 = \frac{1}{2}, P_2 = \frac{1}{2}$$

AND GO  
MOZILLA M  
VO MATCH  
BC THEODO  
DA PODIUM  
NATURAL  
SEQUENCE

4.5 Example: EXITING STATE OF RANDOM WALK ON THE WEIGHTED SHIFT.

Consider graph with  $n^2$  nodes labeled  $\{1, 2, \dots, n\}$ . WEIGHT

WEIGHT  $w_{ij} \geq 0$  on the edge joining  $i$  to node  $j$   
 $w_{ii} = w_{ji}$  } UNDIRECTED

$w_{ij} = 0$  if there is no edge connecting  $i$  &  $j$

A PARTICLE WALKS RANDOMLY FROM NODE TO NODE.  
THE RANDOM WALK  $\{x_n\}, x_n \in \{1, 2, \dots, n\}$

Given  $x_0 = i$ , THE next vertex  $j$  is chosen from the nodes connected to node  $i$  with PROBABILITY PROPORTIONAL TO THE WEIGHT OF THE EDGE CONNECTING  $i$  TO  $j$

$$P(ij) = \frac{w_{ij}}{\sum w_{ik}}$$

THE STATIONARY DISTRIBUTION FOR THIS MARKOV CHAIN ASSIGNS PROBABILITY TO NODE  $i$ . LET

$w_i = \sum w_{ij}$  BE TOTAL WEIGHT OF EDGES EXITING FROM NODE  $i$  AND LET  $w = \sum_{i,j} w_{ij}$  BE THE SUM OF WEIGHTS OF ALL EDGES. THEN:  $\sum w_i = 2w$   
 - Now we see that stationary distribution is:

$$M_i = \frac{w_i}{2w}$$

WE CHECK THIS IS ERATORER DISTRIBUTION BY  
CHECKING  $\sum_i \mu_i p_{ij} = m$

$$\sum_i \mu_i p_{ij} = \sum_i \frac{w_i}{2w} \frac{w_{ij}}{w_i} = \frac{w_j}{2w} = m_j$$

THUS, THE STATIONARY PROBABILITY OF STATE  $i^{\circ}$   
IS PROPORTIONAL TO THE WEIGHT OF THE EDGES  
EXTRADUCING FROM NODE  $i^{\circ}$ .

• ENTROPY RATE IS:

$$H(x) = H(x_2 | x_1) = - \sum_i \mu_i \sum_j p_{ij} \ln p_{ij} =$$

$$= - \sum_i \frac{w_i}{2w} \sum_j \frac{w_{ij}}{w_i} \ln \frac{w_{ij}}{w_i} = \frac{-1}{2w} \sum_{i,j} w_{ij} \ln \frac{w_{ij}}{w_i}$$

$$= - \sum_{i,j} \frac{w_{ij}}{(2w)} \ln \frac{w_{ij}}{(2w)} + \sum_{i,j} \frac{w_{ij}}{2w} \ln \frac{w_{ij}}{2w} =$$

$$= H(0, \dots, \frac{w_1}{2w}, \dots) - H(0, \dots, \frac{w_1}{2w}, \dots)$$

• IF ALL EDGES HAVE EQUAL WEIGHT, STATIONARY  
PROB DISTRIBUTION GIVES WEIGHT  $1/2E$  ON NODE  $i^{\circ}$ ,  
WHERE  $E$  IS NUMBER OF EDGES EXTRADUCING FROM  
NODE  $i^{\circ}$  AND  $E$  IS TOTAL NUMBER OF EDGES  
IN THE GRAPH.

$$- \sum_i \sum_j \frac{w_{ij}}{2w} \ln \frac{w_{ij}}{2w} = \sum_i \sum_j \frac{w_{ij}}{2E} \ln \frac{w_{ij}}{2E} =$$

$$(w_{ij} = w = \text{const}) = - \frac{w}{2E} \sum_i \sum_j \ln \frac{w}{2E} =$$

$$= \left( w = \frac{1}{E} \right) = - \frac{1}{2E} \cdot E \cdot \ln \frac{1}{2E} = \frac{1}{E} \ln 2E^2 = \frac{1}{E} \ln E^2$$

BY DEFINITION  
ZENITH  $w=1$

ORIGIN TO EXTRACT 4.2 :  $(2A w=1)$

$$\checkmark \quad \sum_{i=1}^2 \sum_{j=1}^7 \frac{w}{2E} \ln \frac{w}{2E} = \frac{w_{12}}{2E} \ln \frac{w_{12}}{2E} + \frac{w_{15}}{2E} \ln \frac{w_{15}}{2E} +$$

$$+ \frac{w_{21}}{2E} \ln \frac{w_{21}}{2E} + \frac{w_{25}}{2E} \ln \frac{w_{25}}{2E} + \frac{w_{27}}{2E} \ln \frac{w_{27}}{2E} + \frac{w_{32}}{2E} \ln \frac{w_{32}}{2E} +$$

$$+ \frac{w_{35}}{2E} \ln \frac{w_{35}}{2E} + \frac{w_{34}}{2E} \ln \frac{w_{34}}{2E} + \frac{w_{42}}{2E} \ln \frac{w_{42}}{2E} + \frac{w_{45}}{2E} \ln \frac{w_{45}}{2E} +$$

$$+ \frac{w_{51}}{2E} \ln \frac{w_{51}}{2E} + \frac{w_{52}}{2E} \ln \frac{w_{52}}{2E} + \frac{w_{53}}{2E} \ln \frac{w_{53}}{2E} + \frac{w_{54}}{2E} \ln \frac{w_{54}}{2E} = 2E \cdot \frac{1}{2E} = 1$$

$$\text{FOL } w=1 \Rightarrow - \sum_i \sum_j \frac{w_{ij}}{2w} \ln \frac{w_{ij}}{2w} = \ln 2E$$

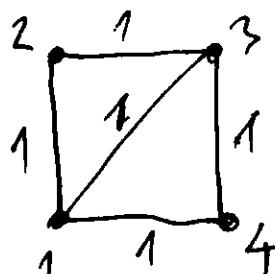
→ FOR EQUAL EDGES:

$$H(X) = \text{ld}(2E) - H\left(\frac{e_1}{2E}, \frac{e_2}{2E}, \dots, \frac{e_n}{2E}\right)$$

" " 9.00F 9.83 " " 9.00F 9.84 "

$$- H\left(\frac{e_1}{2E}, \frac{e_2}{2E}, \dots, \frac{e_n}{2E}\right) = \frac{e_1}{2E} \text{ld} \frac{2E}{e_1} + \frac{e_2}{2E} \text{ld} \frac{2E}{e_2} + \dots + \frac{e_n}{2E} \text{ld} \frac{2E}{e_n}$$

$$\sum_i \sum_j \frac{w_{ij}}{2W} \text{ld} \frac{w_i}{2W} = \sum_{i=1}^4 \sum_{j>1} \frac{1}{2E} \text{ld} \frac{w_i}{2E} =$$



$$= \frac{w_{12}}{2E} \text{ld} \frac{w_1}{2E} + \frac{w_{14}}{2E} \text{ld} \frac{w_1}{2E} + \\ + \frac{w_{13}}{2E} \text{ld} \frac{w_1}{2E} + \frac{w_{21}}{2E} \text{ld} \frac{w_2}{2E} +$$

$$+ \frac{w_{23}}{2E} \text{ld} \frac{w_2}{2E} + \frac{w_{31}}{2E} \cdot \text{ld} \frac{w_1}{2E} + \frac{w_{32}}{2E} \text{ld} \frac{w_3}{2E} + \frac{w_{34}}{2E} \text{ld} \frac{w_3}{2E}$$

$$+ \frac{w_{41}}{2E} \text{ld} \frac{w_4}{2E} + \frac{w_{42}}{2E} \text{ld} \frac{w_4}{2E} = \textcircled{3} \frac{1}{2E} \text{ld} \frac{w_1}{2E} + \textcircled{2} \frac{1}{2E} \text{ld} \frac{w_2}{2E}$$

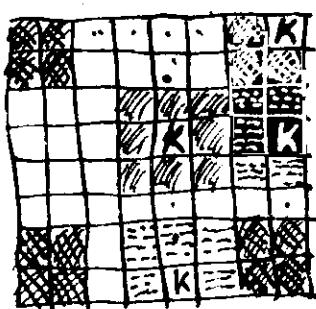
$$+ \textcircled{3} \frac{1}{2E} \text{ld} \frac{w_3}{2E} + \textcircled{2} \frac{1}{2E} \text{ld} \frac{w_4}{2E} = \frac{w_1}{2E} \text{ld} \frac{w_1}{2E} + \frac{w_2}{2E} \text{ld} \frac{w_2}{2E} +$$

$$+ \frac{w_3}{2E} \text{ld} \frac{w_3}{2E} + \frac{w_4}{2E} \text{ld} \frac{w_4}{2E} = \sum_{i=1}^4 \frac{w_i}{2E} \text{ld} \frac{w_i}{2E} =$$

$$= -H\left(\frac{w_1}{2E}, \frac{w_2}{2E}, \frac{w_3}{2E}, \frac{w_4}{2E}\right)$$

DONE!!!

EXAMPLE 4.3.1 (Random Walk on a Chessboard) Let a KING move at random on an 8x8 chessboard.



$$H(X) = \text{ld} 2E - H\left(\frac{e_1}{2E}, \frac{e_2}{2E}, \dots, \frac{e_n}{2E}\right)$$

i = ■ — INTERIOR 4 komorakca vörös  
e = ■ — EDGE 6 2 komorakca fekete  
c = ■ — CORNER 4x1 komorakca = 4

INTERIOR ≈ EDGE  $w_{ij}$

VÉVŐNÖK MINDEN KERÉK EDGES ( $w_{ij}$ ) = ① 4x8 = 32 edges  
②  $8 \times 5 = 40$  edges ②  $4 \times 7 = 12$  edges \$ 72  
④  $W = E = 32 + 40 + 12 = 72 + 12 = 84$  edges \$ 18.5.

$$H(x) = \text{ld} 168 + 4 \left( \frac{8}{168} \text{ld} \frac{8}{168} \right) + 8 \left( \frac{5}{168} \text{ld} \frac{5}{168} \right) + 4 \left( \frac{3}{168} \text{ld} \frac{3}{168} \right)$$

$$\mu_i = \frac{W_i}{2W}$$

$$\mu_i = \frac{8}{168}$$

$$\mu_e = \frac{5}{168} \quad \mu_e = \frac{3}{168}$$

$$\textcircled{1} \quad 36 \text{ pozicii} \times 8 \text{ linii} = 288 \text{ linii} \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \text{OVALE}$$

$$\textcircled{2} \quad 24 \text{ pozicii} \times 5 \text{ linii} = 120 \text{ linii} \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \text{TRIUNGH.}$$

$$\textcircled{3} \quad 4 \text{ pozicii} \times 3 \text{ linii} = 12 \text{ linii} \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \text{KNEE}$$

$$2E = \underline{\underline{\text{TOTAL}}} = 420 \text{ linii} \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} 076105$$

ZNAČÍ:

$$\mu_i = \frac{8}{420} \quad \mu_e = \frac{5}{420} \quad \mu_c = \frac{3}{420}$$

STACIONARNA VELOCITATE

$$H(x) = \text{ld} 420 + 36 \frac{8}{420} \text{ld} \frac{8}{420} + 24 \frac{5}{420} \text{ld} \frac{5}{420} + 4 \frac{3}{420} \text{ld} \frac{3}{420}$$

$$= \frac{1}{35} (\text{ld} 420 + 163 + 101d5) = \frac{1}{35} \text{ld} (2^{72} \cdot 3 \cdot 5^{10}) =$$

$$= \cancel{\text{ld} 420} + \frac{288}{420} \text{ld} 8 - \cancel{\frac{288}{420} \text{ld} 420} + \frac{120}{420} \text{ld} 5 - \cancel{\frac{120}{420} \text{ld} 420} +$$

$$\frac{12}{420} \text{ld} 3 - \cancel{\frac{12}{420} \text{ld} 3} = \frac{288}{420} \text{ld} 8 + \frac{120}{420} \text{ld} 5 + \frac{12}{420} \text{ld} 3$$

$$= \frac{288}{420} \cdot 3 + \frac{2}{7} \text{ld} 5 + \frac{2}{35} \text{ld} 3 = \frac{144}{210} \cdot 3 + \frac{2}{7} \text{ld} 5 + \frac{1}{35} \text{ld} 3$$

$$= \frac{72}{35} + \frac{2}{7} \text{ld} 5 + \frac{1}{35} \text{ld} 3 = \frac{72}{35} + \text{ld} \left( 5^{\frac{2}{7}} \cdot 3^{\frac{1}{35}} \right) = \underline{\underline{2.76584}}$$

$$= \frac{70}{35} + \frac{2}{35} \text{ld} 2 + \frac{40}{35} \text{ld} 5 + \frac{1}{35} \text{ld} 3 = 2 + \frac{1}{35} \text{ld} (4 \cdot 5^{10} \cdot 3) =$$

$$= 2 + \frac{1}{35} \text{ld} 12 + \frac{40}{35} \text{ld} 5 = 2 + \frac{1}{35} \text{ld} 12 + \frac{2}{7} \text{ld} 5 =$$

$$= 2 + \frac{1}{35} \text{ld} (12 \cdot 5^{10}) = \boxed{0.32 \cdot \text{ld} 8} \rightarrow \begin{array}{l} \text{VO KNIGHT} \\ \text{GO DOVEDURANT} \\ \text{DO DATA FORMA} \end{array}$$

ENRICOZ PATE OF THE ROOKS

$$\mu = \frac{14}{2W}$$

$$2W = 14 \cdot 8^2 = 14 \cdot 64 = 896$$

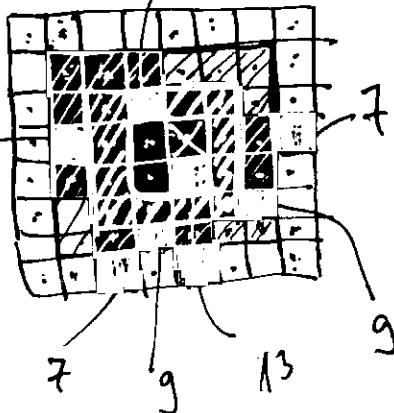
$$\mu = \frac{14}{896}$$

$$H(x) = \text{ld} 896 + 64 \cdot \frac{14}{896} \cdot \text{ld} \frac{14}{896}$$

$$= 16896 \cdot \frac{14}{896} = 18 \frac{896}{14} = 3.8$$

DATI XA 21821

## BISHOP



- I: 7 - MOVES      ( $\frac{7 \cdot 28}{28}$  NODES) KONCENTRATEN  
II: 9 - MOVES      ( $\frac{9 \cdot 20}{12+8}$  NODES)  
III: 11 - MOVES      ( $\frac{11 \cdot 72}{8+4}$  NODES)  
IV: 13 - MOVES      ( $\frac{13 \cdot 4}{4}$  NODES)

$$\frac{11 \cdot 12}{22} \\ \frac{11}{13} \\ \underline{\underline{132}}$$

$$2E = 7 \cdot 28 + 9 \cdot 10 + 11 \cdot 12 + 13 \cdot 4 = 196 + 180 + 132 + 52 \\ = 376 + 134 = \underline{\underline{510}} \text{ LINES}$$

$$M_I = \frac{7}{560} \quad M_{II} = \frac{9}{560} \quad M_{III} = \frac{11}{560} \quad M_{IV} = \frac{13}{560}$$

$$H(x) = \text{ld } 2E + 28 \cdot \frac{7}{560} \text{ld } \frac{7}{560} + 20 \cdot \frac{9}{560} \text{ld } \frac{9}{560} + \\ + 12 \cdot \frac{11}{560} \text{ld } \frac{11}{560} + 4 \cdot \frac{13}{560} \text{ld } \frac{13}{560}$$

$$= \frac{196}{560} \text{ld } 7 + \frac{180}{560} \text{ld } 9 + \frac{132}{560} \text{ld } 11 + \frac{52}{560} \text{ld } 13$$

$$= \frac{1}{560} (196 \text{ld } 7 + 180 \text{ld } 9 + 132 \text{ld } 11 + 52 \text{ld } 13)$$

$$= \underline{\underline{3.16053}}$$

## QUEEN

$$I: 7+14=\underline{\underline{21}} \text{ MOVES} \quad \frac{21 \cdot 28}{28} \text{ NODES}$$

$$\frac{21 \cdot 28}{28}$$

$$\frac{168}{42}$$

$$\frac{588}{42}$$

$$II: 9+14=\underline{\underline{23}} \text{ MOVES} ; 20 \text{ NODES}$$

$$\frac{25 \cdot 12}{12}$$

$$\frac{300}{12}$$

$$III: 11+14=\underline{\underline{25}} \text{ MOVES} ; 12 \text{ NODES}$$

$$\frac{25 \cdot 12}{12}$$

$$IV: 13+14=\underline{\underline{27}} \text{ MOVES} ; 4 \text{ NODES}$$

$$\frac{300}{4}$$

$$2E = 21 \cdot 28 + 23 \cdot 20 + 25 \cdot 12 + 27 \cdot 4 = 588 + 460 + 300 + 108$$

$$= 1048 + 408 = 1456$$

$$M_I = \frac{21}{1456} \quad M_{II} = \frac{23}{1456} \quad M_{III} = \frac{25}{1456}$$

$$M_{IV} = \frac{27}{1456}$$

$$H(x) = \text{ld } 1456 + \frac{28 \cdot 21}{1456} \text{ld } \frac{21}{1456} + 20 \frac{23}{1456} +$$

$$\cdot \text{ld } \frac{25}{1456} + 12 \cdot \frac{25}{1456} \text{ld } \frac{25}{1456} + 4 \cdot \frac{27}{1456} \text{ld } \frac{27}{1456} =$$

$$= \frac{1}{1456} (588 \text{ld } 21 + 460 \text{ld } 23 + 300 \text{ld } 25 + 108 \text{ld } 27) = \underline{\underline{4.5125}}$$

KERZICATHIMA MIGAZETTA ENTROPY RATE. Mat zoku-  
VATJE & DECA VARTETELE E LATE IC SE NAPPE A SO  
TOA. EMA KOT SADMAT SCORODA NA DU ZEKE.

## 4.4. SECOND LAW OF THERMODYNAMICS

ENTROPY OF AN ISOLATED SYSTEM IS NON DECREASING.

- IN STATISTICAL DYNAMICS ENTROPY IS  $\log^2$  OF NUMBER OF MICROSTATES IN THE SYSTEM

- ① RELATIVE ENTROPY  $D(\mu_1 || \mu_1')$  DECREASES WITH  $t$ .  
LET  $\mu_1$  &  $\mu_1'$  BE TWO PROBABILITY DISTRIBUTIONS ON THE STATE SPACE OF MARKOV CHAIN AT TIME  $t_1$ , AND  $\mu_{t+1}$  AND  $\mu'_{t+1}$  BE THE CORRESPONDING DISTRIBUTIONS AT TIME  $t+1$ . LET CORRESPONDING MEAS FUNCTIONS BE DENOTED AS  $r$  &  $g$ . THUS:

$$g(t_1, t_{t+1}) = g(t_1) r(t_{t+1} | t_1) \quad g(t_1, t_{t+1}) = g(t_1) r(t_{t+1} | t_1)$$

WHERE  $r(\cdot | \cdot)$  IS PROBABILITTY TRANSITION FUNCTION FOR MARKOV CHAIN.

$$\begin{aligned} D(g(t_1, t_{t+1}) || g(t_1) r(t_{t+1} | t_1)) &= D(g(t_1) || g(t_1)) + \\ &+ D(g(t_{t+1} | t_1) || r(t_{t+1} | t_1)) = D(g(t_{t+1}) || g(t_{t+1})) + \\ &+ D(g(t_1 | t_{t+1}) || g(t_1 | t_{t+1})) \end{aligned}$$

$$g(t_{t+1} | t_1) = g(t_{t+1} | t_1) = r(t_{t+1} | t_1) \Rightarrow$$

$$D(g(t_{t+1} | t_1) || g(t_{t+1} | t_1)) = 0$$

$$D(g(t_1) || g(t_1)) \geq D(g(t_{t+1}) || g(t_{t+1})) \quad \text{i.e.}$$

$$D(\mu_1 || \mu_1') \geq D(\mu_{t+1} || \mu'_{t+1})$$

Consequently the distance between the probability mass functions is decreasing with time  $t_1 = t$  for any markov chain. PRIMER: AND SE PREDOMINATE PROBABILITIES IN SYSTEM ARE REDUCING IN A SLOW MANNER

THE LST VO UK, CANADA, SO TEK MA VELME DISTRIBUTION CIZADA IT POKRADOVO VO THE ONE ZEPIDI  
i.e. STANE SLICNA.

- ② RELATIVE ENTROPY  $D(\mu_1 || \mu)$  BETWEEN A DISTRIBUTION  $\mu_1$  ON-THE STATES AT TIME  $t_1$  AND A STATIONARY DISTRIBUTION  $\mu$  DECREASES WITH  $t_1$   
e.g.  $\mu_1 = \mu$   $D(\mu_1 || \mu) \geq D(\mu_{t+1} || \mu)$

WHICH MEANS THAT ANY ~~STATIONARY~~ DISTRIBUTION CONVERGES  
TOWARDS STATIONARY DISTRIBUTOR AS TIME PASSES.

### 3. ENTROPY INCRESES IF STATIONARY DISTRIBUTION IS UNIFORM

IN GENERAL, THE FACT THAT REATIVE ENTROPY DECREASES DOESN'T IMPLY THAT ENTROPY INCREASES.

- IF WE START THE MARKOV CHAIN FROM UNIFORM DISTRIBUTION, THE DISTRIBUTION WILL TEND TO THE STATIONARY DISTRIBUTION, HENCE ENTROPY DECREASES WITH TIME.

- IF STATIONARY DISTRIBUTION IS UNIFORM

$$D(\gamma || \theta) = \sum_{i=1}^n \gamma \ln \frac{\gamma}{\theta_i} = \sum_{i=1}^n \gamma \ln \frac{\gamma}{\frac{1}{n}} = \ln(n) + \sum_{i=1}^n \gamma \ln \gamma$$

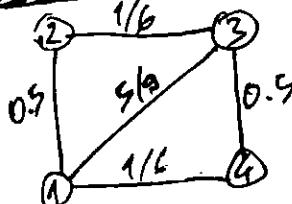
$$\boxed{D(\gamma || \theta) = \ln(n) - H(\gamma)}$$

$$D(\mu_{\text{null}} || \mu) = \ln(n) - H(\mu_{\text{null}}) = \ln(n) - H(\pi)$$



SINCE FOR MARKOV CHAIN WE HAVE MONOTONIC DECREASE OF REACTIVE ENTROPY IT IMPLIES MONOTONIC INCREASE OF ENTROPY. THIS EXPLANATION TIES MOST CLOSELY TO STATISTICAL THERMODYNAMIC WHERE ALL MICROSTATES ARE EQUALLY LIKELY.

### • PORTFOLIO AND OF A MARKOV PROCESS (Ex. 2)



$$\epsilon = 5 \quad 2\epsilon = 10$$

$$\mu_1 = ? \quad \mu_1 = \frac{3}{10} \quad \mu_2 = \frac{2}{10} \quad \mu_3 = \frac{3}{10} \quad \mu_4 = ?$$

$$P_{ij} = 0.2$$

$$\mu_j = \sum_i \mu_i \cdot P_{ij}$$

$$\begin{aligned} j = 2 \quad \mu_2 &= \mu_1 \cdot P_{12} + \mu_3 \cdot P_{32} = \frac{3}{10} \cdot 0.2 + \frac{3}{10} \cdot 0.2 = \\ &= 0.2 \cdot 2 \cdot \frac{3}{10} = 0.4 \cdot \frac{3}{10} = \frac{1.2}{10} \end{aligned}$$

$$\mu_2 = \mu_1 \cdot p + \mu_3 \cdot r = \frac{3}{10} \cdot p \quad p = \frac{10}{3} \cdot \mu_2 = \frac{10}{3} \cdot \frac{2}{10} = \frac{2}{3}$$

$$\mu_3 = \mu_1 \cdot p + \mu_2 \cdot r + \mu_4 \cdot p = \left( \frac{3}{10} + \frac{4}{10} \right) \cdot p = \frac{7}{10} \cdot \frac{2}{3} = \frac{14}{30}$$

$$[\mu_1 \mu_2 \mu_3 \mu_4] = [\mu_1 \mu_2 \mu_3 \mu_4]$$

$$\begin{bmatrix} 0 & 0.2 & 0.2 & 0.2 \\ 0.2 & 0 & 0.2 & 0 \\ 0.2 & 0.2 & 0 & 0.2 \\ 0.2 & 0 & 0.2 & 0 \end{bmatrix}$$

$$\begin{aligned} \mu_1 &= 0.2(\mu_2 + \mu_3 + \mu_4) & \mu_3 &= 0.2(\mu_1 + \mu_2 + \mu_4) \\ \mu_2 &= 0.2(\mu_1 + \mu_3) & \mu_4 &= 0.2(\mu_1 + \mu_2) \end{aligned}$$

$$\begin{aligned} 5\cdot\mu_1 &= 0.2(\mu_1 + \mu_2) + \mu_3 + 0.2(\mu_1 + 0.2(\mu_1 + \mu_2)) = \\ &= \underline{0.2\mu_1} + \underline{0.2\mu_2} + \underline{\mu_3} + \underline{0.2\mu_1} + \underline{0.04\mu_1} + \underline{0.04\mu_3} = \\ &= (2\cdot 0.2 + 0.04)\mu_1 + (1.2\mu_2 + 0.04\mu_2) = 0.44\mu_1 + 1.24\mu_2 \\ \boxed{1.56\mu_1 = 1.24\mu_2} \end{aligned}$$

$$[0.3, 0.2, 0.3, 0.2] = [0.3, 0.2, 0.3, 0.2] \begin{bmatrix} P_{11}, P_{12}, P_{13}, P_{14} \\ P_{21}, P_{22}, P_{23}, P_{24} \\ P_{31}, P_{32}, P_{33}, P_{34} \\ P_{41}, P_{42}, P_{43}, P_{44} \end{bmatrix}$$

$$\begin{aligned} &= [0.3, 0.2, 0.3, 0.2] \begin{bmatrix} 0 & P_{12} & P_{13} & P_{14} \\ P_{12} & 0 & P_{23} & 0 \\ P_{13} & P_{23} & 0 & P_{34} \\ P_{14} & 0 & P_{34} & 0 \end{bmatrix} = \quad \text{...} \\ &= [0.3, 0.2, 0.3, 0.2] \begin{bmatrix} 0 & P_1 & P_2 & P_3 \\ P_1 & 0 & P_3 & 0 \\ P_2 & P_3 & 0 & P_4 \\ P_3 & 0 & P_4 & 0 \end{bmatrix} \quad \frac{18}{10} = \frac{9}{5} \end{aligned}$$

$$0.3 = 0.2P_1 + 0.3P_2 + 0.2P_3 \quad 0.3 = 0.3P_2 + 0.2P_3 + 0.2P_4$$

$$0.2 = \underline{0.3P_1 + 0.3P_2} \quad 0.2 = 0.2P_3 + 0.3P_4$$

$$P_1 = P_4, \quad P_2 = 0.556, \quad P_3 = 0.67 - P_4, \quad \left. \right\} \text{match}$$

$$\boxed{P_4 = 0.5 \quad P_1 = 0.5; P_2 = 0.556 \quad P_3 = 0.17}$$

$$\begin{array}{c} 0.15 \\ 0.051 \\ 0.02 \end{array}$$

$$0.3 \cdot P_1 + 0.3P_2 = 0.3 \cdot 0.2 + 0.3 \cdot 0.17 = 0.015 + 0.051$$

$$P_1 = 0.5; P_2 = \frac{5}{9} \quad ; \quad P_3 = \frac{1}{6}; P_4 = 0.5$$

$$x' = \{x_1, x_2, x_3, x_4\}$$

$$\begin{aligned} P(x_2|x_1) &= 0.5; P(x_3|x_2) = 1/6; P(x_3|x_1) = 5/9; \\ P(x_4|x_1) &= 1.6; P(x_3|x_4) = 0.5 \end{aligned}$$

$$P_{ij} = \begin{bmatrix} 1 & 0.5 & 3/13 & 4/13 \\ 2 & 0.5 & 0 & 1/13 \\ 3 & 5/13 & 1/13 & 0 & 0.5 \\ 4 & 1/13 & 0 & 0.5 & 0 \end{bmatrix}$$

$$\mu = [0.3, 0.2, 0.3, 0.2]$$

$$H(x) = -\sum_i \mu_i \sum_j P_{ij} \ln P_{ij}$$

$$H(x) = 1.2135 \quad \boxed{\text{maximieren}}$$

$$x_1: X_1 = \{1, 0, 0, 0\}$$

$$x_2 = \{0, 1, 0, 0\}$$

$$x_3 = \{0, 0, 1, 0\}$$

$$x_4 = \{0, 0, 0, 1\}$$

$$\gamma(x_1) = \mu_1 (1-\mu_2) (1-\mu_3) (1-\mu_4) = 0.3 \cdot 0.8 \cdot 0.7 \cdot 0.8 = 0.1344$$

$$\gamma(x_2) = \mu_2 (1-\mu_1) (1-\mu_3) (1-\mu_4) = 0.2 \cdot 0.7 \cdot 0.7 \cdot 0.8 = 0.0784$$

$$\gamma(x_3) = \mu_3 (1-\mu_1) (1-\mu_2) \cdot (1-\mu_4) = 0.3 \cdot 0.7 \cdot 0.8 \cdot 0.8 = 0.1344$$

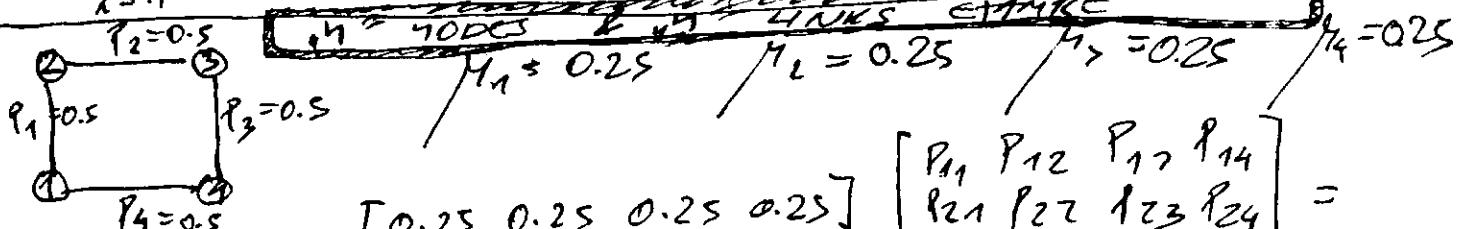
$$\gamma(x_4) = \mu_4 (1-\mu_1) (1-\mu_2) (1-\mu_3) = 0.2 \cdot 0.7 \cdot 0.8 \cdot 0.7 = 0.0784$$

$$H(x_1, x_2, x_3, x_4) = -\sum_{i=1}^4 \mu_i \ln \gamma(x_i) = \underline{0.97867}$$

$$H(x) = \frac{1}{4} H(x_1, x_2, x_3, x_4) = \frac{1.35421}{4} = \underline{0.3386}$$

two sets ~~events~~  $\varphi(x_i) = \mu_i$  totals

$$-\sum_{i=1}^4 \mu_i \ln \mu_i = \underline{1.97095} \Rightarrow \frac{H(x_1, x_2, x_3, x_4)}{4} = \frac{1.97095}{4} = \underline{0.3}$$



~~HIER VON P1 P2 P3 P4 E  
ISOLGIERT A KER  
KREI 100% 20%  
E POUDET STOMSTE!!~~

$$= [0.25 \ 0.25 \ 0.25 \ 0.25] \cdot \begin{bmatrix} 0 & P_1 & 0 & P_4 \\ P_1 & 0 & P_2 & 0 \\ 0 & P_2 & 0 & P_3 \\ P_4 & 0 & P_3 & 0 \end{bmatrix} = [0.25, 0.25, 0.25, 0.25]$$

$$0.25 P_1 + 0.25 P_4 = 0.25$$

$$0.25 P_1 + 0.25 P_2 = 0.25$$

$$P_1 + P_4 = 1$$

$$P_1 + P_2 = 1$$

$$0.25 P_2 + 0.25 P_3 = 0.25$$

$$0.25 P_3 + 0.25 P_4 = 0.25$$

$$P_2 + P_3 = 1$$

$$P_3 + P_4 = 1$$

- KVADRAT:  $P_1 = p_3$ ,  $P_2 = 1 - P_3$ ,  $\underline{P_3 = 1 - P_3}$
- PERTOGEMIK:  $P_1 = 1 - P_3$ ;  $P_2 = P_3$ ;  $P_3 = 1 - P_2$ ;  $\underline{P_4 = 1 - P_3}$ ;  $\underline{P_5 = P_3}$

(FOR  $\underline{P_3 = 0.5}$ )

$\boxed{P_1 = 0.5 \quad P_2 = 0.5 \quad P_3 = 0.5 \quad P_4 = 0.5}$

$$\rightarrow P_3 = 0.5 \Rightarrow P_1 = 0.5 \quad P_2 = 0.5 \quad P_3 = 0.5, \quad P_4 = 0.5, \quad P_5 = 0.5$$

• İESİDÖKÜLKÜK:

$$P_1 = P_3; \quad P_2 = 1 - P_3; \quad P_3 = P_5; \quad P_4 = 1 - P_3; \quad P_5 = P_3; \quad P_6 = 1 - P_3$$

• PERTOGEMIK

$$\begin{bmatrix} P_1 & P_2 & P_3 & P_4 & P_5 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0.5 & 0 & 0 \\ 0 & 0.5 & 0 & 0.5 & 0 \\ 0 & 0 & 0.5 & 0 & 0 \\ 0 & 0 & 0 & 0.5 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} = \underline{\underline{[0.2, 0.2, 0.2, 0.2, 0.2]}}$$

• TRİGONOMETRİK

$$\begin{bmatrix} \mu_1 & \mu_2 & \mu_3 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0.5 & 0.5 \\ 0.5 & 0 & 0.5 \\ 0.5 & 0.5 & 0 \end{bmatrix}$$

$$\boxed{H(X) = 1}$$

$$- \begin{bmatrix} \mu_1 & \mu_2 & \mu_3 \end{bmatrix} * \left( P_{ij} \cdot \log P_{ij} \right) = \textcircled{1} \quad \begin{array}{l} \text{seviye} \\ \text{en} \quad \text{yüksek} \\ \text{normal} \quad \text{matriks} \end{array}$$

$$- \sum_{j=1}^N \frac{1}{N} \cdot \log \frac{1}{N} = + \frac{N \cdot \log N}{N} = \underline{\underline{\log N}} = H(x_1, x_2, \dots, x_N)$$

$$\underline{\underline{H(x_1, x_2, \dots, x_N)}} = \frac{\log N}{N} \rightarrow 0$$

$$p(x_i) = \mu_1 (1 - \mu_2) (1 - \mu_3) \dots (1 - \mu_N) = \underline{\underline{\mu (1 - \mu)^{N-1}}} = p(x_1) = \dots = p(x_N)$$

$$- \sum_{i=1}^N p(x_i) \log p(x_i) = -N \cdot p(x_1) \log p(x_1) = -N \cdot \mu (1 - \mu)^{N-1} \log \mu (1 - \mu)^{N-1}$$

$$\lim_{N \rightarrow \infty} -N \mu (1 - \mu)^{N-1} \log \mu (1 - \mu)^{N-1} = -\lim_{N \rightarrow \infty} \mu (1 - \mu)^{N-1} \log \mu (1 - \mu)^{N-1}$$

$$-\lim_{\mu \rightarrow 0} \mu (1 - \mu)^{N-1} \log \mu - \lim_{\mu \rightarrow 1} \mu (1 - \mu)^{N-1} \log (1 - \mu)^{N-1} = \left[ \lim_{\mu \rightarrow 0} (N-1) \mu (1 - \mu)^{N-2} \right] \cdot \mu \cdot \log (1 - \mu)$$

$$\lim_{n \rightarrow \infty} \mu(1-\mu)^{n-1} \ln((1-\mu)^{n-1}) = \lim_{n \rightarrow \infty} \frac{1}{n} \left(1 - \frac{1}{n}\right)^{n-1} \ln\left(\frac{n-1}{n}\right)^{n-1}$$

$$= \lim_{n \rightarrow \infty} \frac{1}{n} \left(\frac{n-1}{n}\right)^{n-1} \ln\left(\frac{n-1}{n}\right)^{n-1} = \lim_{n \rightarrow \infty} \left(\frac{n-1}{n}\right)^n \cdot \cancel{\ln\left(\frac{n-1}{n}\right)^{n-1}}$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \left(\frac{n-1}{n}\right)^{n-1} \ln\frac{1}{n} = \lim_{n \rightarrow \infty} \frac{1}{n} \left(\frac{n-1}{n}\right)^{n-1} \cancel{\ln\frac{1}{n}}$$

• Ako se odi do fiksacije

$$H(X) = \underbrace{\ln 2e}_{0.5} - H\left(\frac{E_1}{2e}, \frac{E_2}{2e}, \dots, \frac{E_7}{2e}\right) \quad 4.667.501.082$$

$$= 0.5 \ln \frac{2e}{0.5} = 0.5 \ln 4e$$

- TRAPEZNIK  $\Rightarrow$   $E_1 = 2 \cdot 0.5 = 1$

$$H(X) = 0.5 \ln 4.3 + \cancel{3} \cdot \left( \frac{1}{6} \ln \frac{1}{6} \right) =$$

$$= 0.5 \ln 12 + 0.5 \ln \frac{1}{6} = \frac{0.5 \ln 1 + 0.5 \ln 6 - 0.5 \ln 6}{1 + 0.5 \ln 3}$$

- CETRIPATAGONIK

$$H(X) = 0.5 \ln(4 \cdot 4) + \cancel{4} \left( \frac{1}{8} \ln \frac{1}{8} \right) = 0.5 \ln 2 + 0.5 \ln 8 -$$

$$= \underline{\underline{0.5}} \quad \text{NUMBER OF DOGS}$$

$\Rightarrow$  PREMENI ZEMENO VO PREVIO DOKA  $2e = 2 \cdot N_e \cdot 0.5 = N_e$

$$H(X) = -2 \cdot N_e \frac{0.5}{N_e} \ln \frac{0.5}{N_e} - H\left(\frac{E_1}{N_e}, \frac{E_2}{N_e}, \dots, \frac{E_7}{N_e}\right) \quad \text{SUM OF WEIGHTS}$$

$$= \ln 2(N_e) - H\left(\frac{E_1}{N_e}, \frac{E_2}{N_e}, \dots, \frac{E_7}{N_e}\right)$$

- TRAPEZNIK

$$H(X) = \ln 2 \cdot 3 + \cancel{3} \cdot \frac{1}{3} \ln \frac{1}{3} = \ln 2 + \ln 3 - \ln 3 = 1$$

- KVADRAT

$$H(X) = \ln 2 \cdot 4 + \cancel{4} \cdot \frac{1}{4} \ln \frac{1}{4} = \ln 2 + \ln 4 - \ln 4 = 1$$

- PENTAGONIK

$$H(X) = \ln 2 \cdot 5 + \cancel{5} \cdot \frac{1}{5} \ln \frac{1}{5} = \ln 2 + \ln 5 - \ln 5 = 1$$

PR. 30  $\star \quad X \in \{1, 2, 3, 4\} \quad m=4 \rightarrow$  DAŽNA NA KODATA  
IZVOROT GENERATA 10 even znak so dolzina  $\boxed{4=1}$

$$P(X) = \{P_1, P_2, P_3, P_4\} = \{0.1344, 0.0784, 0.1344, 0.0784\}$$

ENTROPIJA NA SEKVENCATA E:

$$H(x) = \frac{1}{n} H(x_1) = \frac{1}{1} H(x) = -\sum_{i=1}^n p_i \log p_i = \underbrace{1.35921}_{D.P. 7.1}$$

1.35921 ~ 1:2135 ||

ZNAPI ZA PRIMEROT 5 LIMA 4 221 DOPIVANJE TONICA NA TEOLGYATA 4.24 1.0 FETULTADOT DOBIEN TONICA DOGOVARTA NA REZULTATOT DOBIEN SOGLASNO DEFI-NICIJATA ZA ENTROR RAVE. SEPTA ZA PRIMEROTE LINIJA & "N" NODES NE USLJEM PA SO DOBLJAVI TONICA.

DEFINITION A PROBABILITET TRANSITION MATRIX  $[P_{ij}]$ , IF

$$P_{ij} = P \{ X_{t+1} = j | X_t = i \}, \text{ IS CALLED DOUALT STO-} \\ \sum P_{ij} = 1 \quad j = 1, 2, 3, \dots \text{ AND } \sum_i P_{ij} = 1 \quad i = 1, 2, 3.$$

REMARK: THE UNIFORM DISTRIBUTION IS <sup>i</sup> STATIONARY DISTRIBUTION OF P IF AND ONLY IF THE PROBABILITY TRANSITION MATRIX IS DOUALT STO- (PROBLEM 4.1)

④ THE CONDITIONAL ENTROPY  $H(x_t | x_1)$  INCREASES WITH  $t = 1, 2, \dots$  FOR A STATIONARY MARKOV PROCESS. IF THE MARKOV PROCESS IS STATIONARY  $H(x_t)$  IS CONSTANT, SO THE ENTROPY IS NONINCREASING, HOWEVER  $H(x_t | x_1)$  IS INCREASING WITH  $t = 1, 2, \dots$ . THUS THE CONDITIONAL UNCERTAINTY OF THE FUTURE INCREASES.

$$H(x_t | x_1) \geq H(x_t | x_1, x_2) \quad (\text{CONDITIONING REDUCES ENTROPY}) = \\ = H(x_t | x_2) = \underbrace{H(x_{t-1} | x_1)}_{\text{BY MARKOVITY}} \quad \text{BY STATIONARITY}$$



$$H(x_t | x_1) \geq H(x_{t-1} | x_1) \quad \begin{array}{l} \text{LOGIČNO } M_1 \in II \\ \text{SO TEK NA VIKORE} \\ \text{NEVJEZNOŠTA PASTE.} \end{array}$$

- ALTERNATIVCE JE DATA - REČO SING INEQUALITY

$$I(X; Y | Z) = H(X|Z) - \underbrace{H(X|ZY)}_{H(X|Y)} = H(Y|Z) - H(Y|X, Z)$$

$$I(X; Y | Z) = -I(X; Y) + I(Z; Y | X)$$

$$I(X, Z; Y) = I(X; Y) + I(Z; Y | X) = \underbrace{I(Z; Y)}_{I(Z; Y)} + \underbrace{I(X; Y | Z)}_{I(X; Y | Z)}$$

$$I(X, Y; Z) = I(X; Z) + I(Y; Z | X) = \underbrace{I(X; Z)}_{I(X; Z)} + \underbrace{I(Y; Z | X)}_{I(Y; Z | X)}$$

$$\boxed{I(Y; Z) \geq I(X; Z)}$$

$$I(X; YZ) = I(X; Y) + I(X; Z | Y) = I(X; Z) + I(X; Y | Z)$$

$$\boxed{I(X; Y) \geq I(X; Z)}$$

DATA LOGOLOGIČNO INEQUALITY  $\rightarrow$

$$x_1 \rightarrow x_{n-1} \rightarrow x_n$$

$$I(x_1; x_{n-1}) \geq I(x_1; x_n)$$

$$\boxed{H(x_1 | x_{n-1}) \leq H(x_1 | x_n)}$$

$$\begin{aligned} H(x_1) - H(x_1 | x_{n-1}) &\geq \\ H(x_1) - H(x_1 | x_n) &\quad \leftarrow \\ H(x_n | x_{n-1}) &\geq H(x_n | x_n) \end{aligned}$$

- 1. OPERATOR:

$$H(x_{n-1}) - H(x_{n-1} | x_n) \geq H(x_n) - H(x_n | x_n).$$

STATIONARITY

$$\boxed{H(x_n) = H(x_{n-1})} \Rightarrow$$

0	1	1
1	0	1
1	1	0

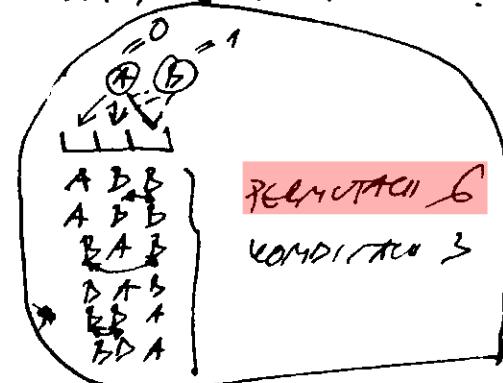
$$\binom{3}{2} = \frac{6}{2} = 3$$

$$H(x_{n-1} | x_n) \leq H(x_n | x_n) \quad \boxed{H(x_n | x_n) \geq H(x_{n-1} | x_n)}$$

⑤ SHUFFLES INCREASE ENTROPY. IF  $T$  IS A SHUFFLE (PERMUTATIONS) OF A DECK OF CARDS AND  $X$  IS THE INITIAL (RANDOM) POSITION OF THE CARDS IN THE DECK, AND IF THE CHOICE OF THE SHUFFLE  $T$  IS INDEPENDENT OF  $X$ , THEN

$$H(TX) \geq H(X)$$

WHERE  $TX$  IS PERMUTATION OF THE DECK INDUCED BY THE SHUFFLE  $T$  ON THE INITIAL PERMUTATION  $X$ . PROOF BY 4.3 OUTLINES THE PROOF.



## 4.5 FUNCTIONS OF MARKOV CHAINS

LET  $x_1, x_2, \dots, x_n$  BE STATIONARY MARKOV CHAIN AND LET  $\xi_i = \phi(x_i)$  BE A PROCESS EACH TERM OF WHICH IS A FUNCTION OF THE CORRESPONDING STATE IN THE MARKOV CHAIN. WHAT IS  $H(\xi)$ ? IT WOULD SIMPLY MARKOV CHAIN IF  $\xi_1, \xi_2, \dots, \xi_n$  ALSO FORMED MARKOV CHAIN (MC), BUT IN MOST CASES IT IS NOT TRUE. SINCE MC IS STATIONARY SO IS  $\xi_1, \xi_2, \dots, \xi_n$ .

$$H(\xi_n | \xi_{n-1}, \xi_{n-2}, \dots, \xi_1) \rightarrow H(\xi)$$

FOR STATIONARY PROCESSES  
 $H(\xi) = H(\xi_n)$  TH. 4.2.1

- FOR COOLER ROUND WE WILL USE  
 $H(\xi_n | \xi_{n-1}, \dots, \xi_1)$  BASED ON THE IDEA THAT  $x_1$  CONTAINS AS MUCH INFORMATION FOR  $\xi_n$  AS  $\xi_1, \xi_2, \xi_3, \dots$

[LEMMA 4.5.1]

$$\boxed{H(\xi_n | \xi_{n-1}, \dots, \xi_1, x_1) \leq H(\xi)}$$

PROOF: for  $k=1, 2, \dots$   $H(Y_k | Y_{k-1}, \dots, Y_1, X_1) = \frac{H(Y_k | T_{k-1}, \dots, T_1, f(X_1))}{\text{since } Y_i = f(T_i)}$

$$H(Y_k | Y_{k-1}, \dots, Y_1, X_1) = \star$$

$T = f(X)$ 
 $H(Y | T) = 0$ 
 $H(Y, Z | X) = H(Y | X) + H(Z | X, Y)$

$$H(Z | X, Y) = H(Y, Z | X) - H(Y | X)$$

$$H(Z, Y | X) = H(Z | X) + H(Y | \star, Z) \quad \begin{matrix} \text{two so same } X \\ \text{so same } Z \end{matrix} = H(Z | X) + H(Y | Z)$$

$$\Rightarrow H(Z | X, Y) = \cancel{H(Z | X)} + \cancel{H(Y | X, Z)} = H(Y, Z | X)$$

$$H(X_k | X_1, Y_1) = H(Y_1 | \star) + H(Y_1 | Y_k).$$

$$H(Y_1 | X_1) = H(Y_1, Y_2 | X_1) - H(Y_1 | \star, Y_2) \quad \begin{matrix} \text{two so same } X_1 \\ \text{so same } Y_1 \end{matrix}$$

by definition  $= H(Y_1 | X_1, Y_1)$

$$H(Y_1, Y_2 | X_1) = H(Y_1 | X_1) + H(Y_2 | X_1, Y_1) =$$

$$\Rightarrow H(Y_2 | X_1, Y_1) = H(Y_2 | Y_1) + H(Y_2 | X_1, Y_1, Y_1) = H(Y_2 | Y_1)$$

$$H(Y_2 | X_1, Y_1) = -H(Y_2 | Y_1) + H(X_1, Y_1 | Y_1)$$

use defn:  $H(Y_2 | X_1, Y_1) = H(Y_2 | X_1) - H(Y_2 | \star)$   $\leftarrow$   
 $\boxed{Y_1 = f(X_1)}$

so formula becomes:

$$H(Y_2 | X_1, Y_1) = H(Y_2 | X_1) + H(Y_2 | X_1, Y_1) \quad \begin{matrix} \text{two so same } X_1 \\ \text{so same } Y_1 \end{matrix}$$

$$\Rightarrow \boxed{H(Y_2 | X_1, Y_1) = H(Y_2 | Y_1)} \quad \begin{matrix} \text{IF } Y_1 = f(X_1) \\ \text{MMV} \end{matrix}$$

$$\star = H(Y_k | Y_{k-1}, \dots, Y_1, X_1, X_0, X_{-1}, \dots, X_{-k}) =$$

due to MARKOVITY OF  $X$

$$= H(Y_k | Y_{k-1}, \dots, Y_1, X_1, X_0, X_{-1}, \dots, X_{-k}, \underbrace{X_0, X_{-1}, \dots, X_{-k}}_{\text{constant values because stationary}}) \quad \begin{matrix} \text{since } X_i = f(T_i) \\ \text{constant values because stationary} \end{matrix}$$

$$\leq H(Y_k | Y_{k-1}, \dots, Y_1, X_0, X_{-1}, \dots, X_{-k}) = H(Y_{k+k+1} | Y_{k+k}, \dots, Y_1)$$

$$H(Y_k | Y_{k-1}, \dots, Y_1, X_1) \leq \lim_{k \rightarrow \infty} H(Y_{k+k}, Y_{k+k-1}, \dots, Y_1) = H(Y) \quad \begin{matrix} \text{for all } k \\ \rightarrow \end{matrix}$$

**LEMMA 4.5.2**

$$H(Y_4 | Y_{4-1}, \dots, Y_1) - H(Y_4 | Y_{4-1}, \dots, Y_1, X_1) \rightarrow 0$$

**PROOF:**

$$H(X_1 | Y_{4-1}, \dots, Y_1) - H(X_1 | Y_{4-1}, \dots, Y_1, X_1) = I(X_1; Y_1 | Y_{4-1}, \dots, Y_1)$$

$$\underbrace{I(X_1; Y_1, Y_2, \dots, Y_4)}_{\text{r}} = I(Y_1; Y_1) + I(X_1; Y_2 | Y_1) + \dots + I(X_1; Y_4 | Y_{4-1}, \dots, Y_1) = \sum_{i=1}^4 I(X_1; Y_i | Y_{i-1}, \dots, Y_1)$$

$$I(X_1; Y) = H(X_1) - H(X_1 | Y) \quad I(X_1; Y) \leq H(X_1)$$

$$I(X_1; Y_1, Y_2, \dots, Y_4) \leq H(X_1)$$

$I(X_1; Y_1, Y_2, \dots, Y_4) \rightarrow$  increases with  $y^n$  thus  
 $\lim_{n \rightarrow \infty} I(X_1; Y_1, Y_2, \dots, Y_4)$  exists and

$$\lim_{n \rightarrow \infty} I(X_1; Y_1, Y_2, \dots, Y_4) \leq H(X_1)$$

$$H(X_1) \geq \lim_{n \rightarrow \infty} I(X_1; Y_1, Y_2, \dots, Y_4) = \lim_{n \rightarrow \infty} \sum_{i=1}^n I(X_1; Y_i | Y_{i-1}, \dots, Y_1)$$

$$H(X_1) \geq \sum_{i=1}^{\infty} I(X_1; Y_i | Y_{i-1}, \dots, Y_1)$$

SINCE THIS INFINITE SUM IS FINITE AND THE TERMS ARE NON-NEGATIVE, TERMS MUST TEND TO  $0^+$ ; P.E.O.:

$$\lim_{n \rightarrow \infty} I(X_1; Y_4 | Y_{4-1}, \dots, Y_1) = 0 \quad \text{and i.e.}$$

$$H(Y_4 | Y_{4-1}, \dots, Y_1) - H(Y_4 | Y_{4-1}, \dots, Y_1, X_1) \rightarrow 0 \quad \underline{\text{because!!!}}$$

COMBINING LEMMA 4.5.1 AND 4.5.2 WE HAVE:

**THEOREM 4.5.1** If  $X_1, X_2, \dots, X_n$  form a stationary MARKOV CHAIN, AND  $Y_i = \phi(X_i)$ , THEN:

$$H(Y_4 | Y_{4-1}, \dots, Y_1, X_1) \leq H(Y) \leq H(Y_4 | Y_{4-1}, \dots, Y_1)$$

AND:  $\lim_{n \rightarrow \infty} H(Y_4 | Y_{4-1}, \dots, Y_1, X_1) = H(Y) = \lim_{n \rightarrow \infty} H(Y_4 | Y_{4-1}, \dots, Y_1)$

In general we could also consider case where  $\tau_i$  is STOCHASTIC FUNCTION (as opposed to a DETERMINISTIC FUNCTION OF  $X_i$ ).

Consider a Markov process  $x_1, x_2, \dots, x_n$  and define a new process  $y_1, y_2, \dots, y_n$  where each  $y_i$  is drawn according to  $p(y_i | x_i)$ , conditionally independent of all the others  $x_j, j \neq i$ ; that is:

$$P(x^*, y^*) = P(x_1) \prod_{i=1}^{n-1} P(x_{i+1} | x_i) \prod_{i=1}^n P(y_i | x_i)$$

Such process is called Hidden Markov Model (HMM) and is used in speech and handwriting recognition.

### SUMMARY

**ENTROPY RATE: TWO DEFINITIONS**

$$H(X) = \lim_{n \rightarrow \infty} \frac{1}{n} H(x_1, x_2, \dots, x_n)$$

$$H'(X) = \lim_{n \rightarrow \infty} H(x_n | x_{n-1}, \dots, x_1)$$

- FOR STATIONARY STOCHASTIC PROCESS

$$H(X) = H'(X)$$

• ENTROPY RATE OF STATIONARY MARKOV CHAIN

$$H(X) = - \sum_{ij} p_{ij} \log p_{ij}$$

**SECOND LAW OF THERMODYNAMICS FOR MARKOV CHAIN:**

1.) REACTIVE ENTROPY  $D(\pi_0 || \pi_t)$  DECREASES WITH TIME

2.) REACTIVE ENTROPY BETWEEN DISTRIBUTION AND THE STATIONARY DISTRIBUTION DECREASES WITH TIME.

3.) ENTROPY INCREASES IF STATIONARY DISTRIBUTION

IS UNIFOLY

4.)  $H(x_1 | x_1)$  INCREASES WITH TIME

5.)  $H(x_0 | x_n)$  INCREASES FOR THE MARKOV CHAIN.

FUNCTIONS OF MARKOV CHAIN. If  $x_1, x_2, \dots, x_n$  forms a stationary Markov chain and  $\tilde{x}_i = p(x_i)$

$$H(x_1 | x_{n-1}, \dots, x_1, x_1) \leq H(\tilde{x}) \leq H(x_1 | x_{n-1}, \dots, x_1)$$

$$\lim H(x_1 | x_{n-1}, \dots, x_1, x_1) = H(\tilde{x}) = \lim_{n \rightarrow \infty} H(x_1 | x_{n-1}, \dots, x_1)$$

### PROBLEMS

**4.1** DOUBLY STOCHASTIC MATRICES. An  $4 \times 4$  matrix  $P = [P_{ij}]$  is said to be DOUBLY STOCHASTIC if  $P_{ij} \geq 0$  AND  $\sum_j P_{ij} = 1$  FOR ALL  $i$  AND  $\sum_i P_{ij} = 1$  FOR ALL  $j$ . An  $4 \times 4$  MATRIX  $P$  is said to be PERMUTATION MATRIX if it is DOUBLY STOCHASTIC AND THERE IS EXACTLY ONE  $P_{ij} = 1$  IN EACH ROW AND EACH COLUMN. IT CAN BE SHOWN THAT EVERY DOUBLY STOCHASTIC MATRIX CAN BE WRITTEN AS THE CONVEX COMBINATION OF PERMUTATION MATRICES.

(a) LET  $a^t = (a_1, a_2, \dots, a_n)$ ,  $a_i \geq 0$   $\sum a_i = 1$  BE A PROBABILITY VECTOR. LET  $b = a.P$  WHERE  $P$  IS DOUBLY STOCHASTIC. SHOW THAT  $b$  IS PROBABILITY VECTOR AND THAT:

$$H(b_1, b_2, \dots, b_n) \geq H(a_1, a_2, \dots, a_n)$$

THUS STOCHASTIC MIXING INCREASES THE ENTROPY.

(b) SHOW THAT STATIONARY DISTRIBUTION  $\pi^u$  FOR DOUBLY STOCHASTIC MATRIX  $P$  IS THE UNIFORM DISTRIBUTION

(c) CONVERSELY, PROVE THAT IF THE UNIFORM IS STATIONARY DISTRIBUTION FOR MARKOV TRANSITION MATRIX  $P$ , THEN  $P$  IS DOUBLY STOCHASTIC.

(a)

$$[a_1, a_2, \dots, a_n] \begin{bmatrix} P_{11} & P_{12} & \dots & P_{1n} \\ P_{21} & P_{22} & \dots & P_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ P_{n1} & P_{n2} & \dots & P_{nn} \end{bmatrix} = \left[ \underbrace{\sum_{i=1}^n a_i P_{i1}}_{b_1}, \underbrace{\sum_{i=1}^n a_i P_{i2}}, \dots, \underbrace{\sum_{i=1}^n a_i P_{in}}_{b_n} \right]$$

Given  $\Sigma a_i = 1$

$$b_1 = \sum_{i=1}^n a_i P_{i1} = a_1 P_{11} + a_2 P_{21} + \dots + a_n P_{n1}$$

$$P_{11} = 1 - \sum_{i=2}^n P_{ii} \quad P_{21} = 1 - \sum_{i=2}^{n-1} P_{ii} \quad \dots \quad P_{n1} = 1 - \sum_{i=1}^{n-1} P_{ii}$$

$b_1$  IS PROBABILITY VECTOR hence:

$$b_1 = [b_1, b_2, \dots, b_n] \quad b_1 + b_2 + \dots + b_n = 1$$

$$\sum_{i=1}^n a_i P_{i1} + \sum_{i=1}^n a_i P_{i2} + \dots + \sum_{i=1}^n a_i P_{in} = \sum_{j=1}^n \sum_{i=1}^n a_i P_{ij} =$$

$$= \sum_{i=1}^n a_i \underbrace{\sum_{j=1}^n P_{ij}}_{1 (\text{DUE TO DOUBLY STOCHASTIC})} = \sum_{i=1}^n a_i = 1 \quad \text{PROVED!}$$

1 (DUE TO DOUBLY STOCHASTIC)

1 (DUE TO PROBABILITY VECTOR)

(b)

$$\sum_{i=1}^n \mu_i P_{ij} = \mu_j$$

$$\mu_1 = \sum_{i=1}^n \mu_i P_{i1}$$

$$[\mu_1 \mu_2] \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} = [\mu_1 \mu_2]$$

$$\begin{aligned}\mu_1 &= \mu_1 P_{11} + \mu_2 P_{21} \\ \mu_2 &= \mu_1 P_{12} + \mu_2 P_{22}\end{aligned}$$

$$\begin{aligned}\mu_1(1-P_{11}) &= \mu_2 P_{12} \\ \mu_2(1-P_{22}) &= \mu_1 P_{12}\end{aligned}$$

$$\mu_1 = \frac{\mu_2 P_{12}}{1-P_{11}}$$

$$\mu_2(1-P_{22}) = \frac{\mu_2 P_{21} P_{12}}{(1-P_{11})}$$

$$\begin{aligned}P_{11} &= P_{21} + P_{22} = 1 \Rightarrow P_{21} = 1 - P_{22} \\ P_{12} + P_{22} &= 1 \Rightarrow P_{12} = 1 - P_{22} \\ \mu_1 P_{12} &= \mu_2 P_{21} \\ \mu_2 P_{21} &= \mu_1 P_{12} \\ \mu_1(1-P_{11}) &= \mu_2 P_{21} \\ \mu_1(1-P_{11}) &= \mu_2(1-P_{22}) \\ P_{11} + P_{21} &= 1 \\ P_{21} &= 1 - P_{11} \\ \mu_1(1-P_{11}) &= \mu_2(1-P_{22})\end{aligned}$$

OVA & OVAZ 2x2  $\begin{matrix} \mu_1 = \mu_2 \\ \text{AND} \\ \mu_1(1-P_{11}) = \mu_2(1-P_{22}) \end{matrix}$   
- V0 greater success

$$\mu_1 = \sum_{i=1}^n \mu_i P_{i1} = \mu_1 P_{11} + \mu_2 P_{21} + \mu_3 P_{31} + \dots + \mu_n P_{n1}$$

$$\mu_1(1-P_{11}) = \mu_2 P_{21} + \mu_3 P_{31} + \dots + \mu_n P_{n1}$$

$$1-P_{11} = P_{21} + P_{31} + \dots + P_{n1}$$

$$\begin{aligned}\mu_1 P_{21} + \mu_1 P_{31} + \dots + \mu_1 P_{n1} &= \mu_2 P_{21} + \mu_3 P_{31} + \dots + \mu_n P_{n1} \\ \Rightarrow \mu_1 &= \mu_2 \quad \mu_1 = \mu_3 \quad \mu_1 = \mu_n\end{aligned}$$

converse on p. 100

$$(c) [\mu_1 \mu_2 \dots \mu_n] \begin{bmatrix} P_{11} & P_{12} & \dots & P_{1n} \\ P_{21} & P_{22} & \dots & P_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ P_{n1} & P_{n2} & \dots & P_{nn} \end{bmatrix} = [\mu_1 \mu_2 \dots \mu_n]$$

$$\mu \cdot \sum_{i=1}^n p_{ii} = \mu$$

$$\sum_{i=1}^n p_{ii} = 1$$

$$\mu \cdot \sum_{i=1}^n p_{i2} = \mu$$

$$\sum_{i=1}^n p_{i2} = 1$$

$$\sum_{i=1}^n p_{i3} = 1$$

$$\mu \cdot \sum_{i=1}^n p_{i4} = \mu$$

$$\sum_{i=1}^n p_{i4} = 1$$

$$\sum_{i=1}^n p_{ij} = 1 \quad \text{for } j = 1, \dots, 4$$

(b) even greater generalization:

~~proof~~

$$\mu_1 p_{21} + \mu_1 p_{31} + \dots + \mu_1 p_{n1} = \mu_2 p_{21} + \mu_3 p_{31} + \dots + \mu_n p_{n1}$$

$$\mu_1 \sum_{i=2}^n p_{ii} = \sum_{i=2}^n \mu_i p_{ii} \quad \sum_{i=2}^n \mu_i p_{ii} - \sum_{i=2}^n \mu_1 p_{ii} = 0$$

$$\sum_{i=2}^n (\mu_i - \mu_1) p_{ii} = 0 \quad \Rightarrow \quad \underline{\mu_i = \mu_1 \quad i = 2, \dots, n}$$

$$(a) H(b_1, b_2, \dots, b_n) = - \left( \sum_{i=1}^n a_i p_{ii} \ln \sum_{j=1}^n q_j p_{ji} + \sum_{i=1}^n a_i p_{ii} \ln \sum_{j=1}^n q_j p_{ji} \right)$$

$$+ \dots + \sum_{i=1}^n q_i p_{ii} \ln \sum_{j=1}^n q_j p_{ji}$$

$$\sum_{i=1}^n a_i p_{ii} = E[a_i] \quad \ln E[a_i] \geq \ln E[\ln a_i]$$

$$\sum_{i=1}^n a_i p_{ii} \ln \sum_{j=1}^n q_j p_{ji} \geq \sum_{i=1}^n a_i p_{ii} \underbrace{\sum_{j=1}^n \ln(q_j p_{ji})}_{\text{concave function}}$$

$$(\ln a_1 p_{11} + \ln a_2 p_{21} + \dots + \ln a_n p_{n1})$$

$$H(a_1, a_2, \dots, a_n) = - \sum_{i=1}^n q_i \ln a_i$$

$$\begin{aligned} & \text{log concave} \\ & \sum_{i=1}^n a_i \ln a_i \geq (\sum_{i=1}^n a_i) \ln (\sum_{i=1}^n a_i) \end{aligned}$$

$$\geq - \sum_{i=1}^n (p_{ii} a_i) \ln (p_{ii} a_i) - \sum_{i=1}^n (p_{i2} a_i) \ln (p_{i2} a_i) - \dots - \sum_{i=1}^n (p_{in} a_i) \ln (p_{in} a_i)$$

$$= - \sum_{j=1}^n \sum_{i=1}^n p_{ij} a_i \ln (p_{ij} a_i) = - \sum_{j=1}^n \sum_{i=1}^n p_{ij} a_i \ln p_{ij} - \sum_{j=1}^n \sum_{i=1}^n p_{ij} a_i \ln a_i$$

$$\begin{aligned}
 &= - \sum_{i=1}^n a_i \sum_{j=1}^n p_{ij} \ln p_{ij} - \sum_{i=1}^n a_i \ln a_i \underbrace{\sum_{j=1}^n p_{ij}}_1 = \\
 &= \underbrace{\sum_{i=1}^n p_{ij} \ln \frac{1}{p_{ij}}}_{\geq 0} + \sum_{i=1}^n a_i \ln \frac{1}{a_i} \geq \sum_{i=1}^n a_i \ln \frac{1}{a_i} = \\
 &\quad = H(a_1, a_2, \dots, a_n) \quad \text{PROVED !!!}
 \end{aligned}$$

**EQUATION 1 SORROWS**

SKOVO IDENTITET OG DOMMELTE

(a)  $H(b) - H(a) = - \sum_{i=1}^n b_i \ln b_i + \sum_{i=1}^n a_i \ln a_i =$

$$= - \sum_{j=1}^n \sum_{i=1}^n a_i p_{ij} \ln \underbrace{\sum_{k=1}^n a_k p_{kj}}_{\text{OVERVURAS}} + \sum_{i=1}^n a_i \ln a_i$$

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SO "1" (VIO U 100)

$$\sum_{i=1}^n a_i \ln a_i = \sum_{i=1}^n a_i \ln a_i \sum_{j=1}^n p_{ij} = \sum_{i=1}^n \sum_{j=1}^n a_i p_{ij} \ln a_i$$

$$= \sum_{i=1}^n \sum_{j=1}^n a_i p_{ij} \ln \frac{a_i}{\sum_{k=1}^n a_k p_{kj}} = \left( \sum_{i=1}^n a_i p_{ij} \right) \ln \frac{\sum_{i=1}^n a_i}{\sum_{i=1}^n b_i}$$

LOGARITHM INEQUALITY:

$$\sum_{i=1}^n a_i \ln \frac{a_i}{b_i} \geq \left( \sum_{i=1}^n a_i \right) \ln \frac{\sum_{i=1}^n a_i}{\sum_{i=1}^n b_i}$$

$$= \underbrace{\sum_{i=1}^n a_i \sum_{j=1}^n p_{ij}}_1 \cdot \ln \frac{1}{m} = 1 \cdot \ln 1 = 0$$

**4.2** Times Anew. Let  $\{x_i\}_{i=-\infty}^{\infty}$  be a stationary stochastic process move that:

$$H(x_0 | x_{-1}, x_{-2}, \dots, x_n) = H(x_0 | x_1, x_2, \dots, x_n)$$

$$H(x_{0+n+1} | x_1, x_2, \dots, x_n) = H(x_0 | x_1, x_2, \dots, x_n)$$

$$H(x_{n+1} | x_1, x_2, \dots, x_n) = H(x_0 | x_1, x_2, \dots, x_n)$$

$$H(\underline{x_{n+1}} | x_1, \dots, x_n) = H(x_0 | x_1, \dots, x_n)$$

$$\boxed{H(x_1 | t_0) = H(x_0 | x_{-1})}$$

$$\underline{H(x_1, x_2, \dots, x_{n+1})} = \sum_{i=1}^{n+1} H(x_i | x_{i-1}, \dots, t_1) = \\ = H(x_1) + H(x_2 | x_1) + H(x_3 | x_1, x_2) \dots + \underline{H(x_{n+1} | t_0, x_1, \dots, x_n)}$$

$$H(x_0, x_1, \dots, x_n) = \sum_{i=0}^n H(x_i | x_{i-1}, \dots, x_0) =$$

$$= H(x_0) + H(x_1 | x_0) + \dots + H(x_n | x_{n-1}, \dots, x_0) =$$

$$= H(x_1, \dots, x_n, x_0) \rightarrow H(x_1) + H(x_2 | x_1) + H(x_3 | x_1, x_2) + \dots \\ + \dots + \underline{H(t_0 | x_1, x_2, \dots, x_n)}$$

$$\underline{H(x_0, x_1, \dots, x_n)} = H(x_0 + t, x_0 + t, \dots, x_n + t) = \left\{ \begin{array}{l} t = 1 \\ \dots \end{array} \right\} = \\ = H(x_1 + x_2, \dots, x_{n+1})$$

Zurück:  $\underline{H(x_1, x_2, \dots, x_{n+1})} = H(x_1, x_2, \dots, x_n, t_0)$

$$\rightarrow H(x_1) + H(x_2 | x_1) + H(x_3 | x_1, x_2) \dots + \underline{H(x_n | x_1, x_2, \dots, x_{n-1})} + \\ + H(x_{n+1} | x_1, x_2, \dots, x_n) = H(x_1) + H(x_2 | x_1) + H(x_3 | x_1, x_2) + \dots \\ + \underline{H(x_n | x_{n-1}, x_{n-2}, \dots, x_1)} + H(t_0 | x_1, x_2, \dots, x_n)$$

Zurück:  $\underline{H(x_{n+1} | x_1, x_2, \dots, x_n)} = H(t_0 | x_1, x_2, \dots, x_n)$   
PROVED !!!

T.E. BEIDESI  $H(x_{n+1} | x_1, x_2, \dots, x_n) = H(x_0 | x_{-1}, x_{-2}, \dots, x_n)$ .

$$\Rightarrow \boxed{H(x_0 | x_{-1}, x_{-2}, \dots, x_n) = H(x_0 | x_1, x_2, \dots, x_n)}.$$

~~QUESTION 1 SOLUTION:~~  $\underline{H(x_0 | x_{-1}, x_{-2}, \dots, x_{-n})} = H(x_0, x_1, \dots, x_n)$

$$\therefore H(x_{-1}, x_{-2}, \dots, x_{-n}) = H(x_n, \dots, x_1, t_0) - H(x_1, x_2, \dots, x_n) \\ = \underline{H(x_0 | x_1, x_2, \dots, x_n)}$$

4.3 Shuffles increase entropy. Argue that for any distribution on shuffles  $T$  and the distribution on card positions  $X$  that:

$$H(TX) \geq H(TX|T) = H(T^{-1}TX|T) = H(X|T) = H(X)$$

If  $X$  and  $T$  are independent

$$X = \{x_1, x_2, \dots, x_i, \dots, x_{51}, x_2\} \quad x_i \in \{1, 2, \dots, 52\}$$

$$\text{e.g. } X = \{3, 4, 1, 2, \dots, 52, 51\}$$

INITIAL RANDOM POSITION OF THE CARDS IN THE DECK

$$x_1 \rightarrow x_2 \rightarrow x_3 \rightarrow \dots \rightarrow x_n$$

$$I(x_1; x_2) \geq I(x_1; x_n)$$

$$[X_2 = TX_1]$$

$$x_1 \rightarrow Tx_1 \rightarrow Tx_2 \rightarrow \dots \rightarrow Tx_n$$

$$I(x_1; Tx_1) \geq I(x_1; Tx_n)$$

$$H(x_1) - H(x_1|Tx_n) \geq H(x_1) - H(x_1|Tx_1) \Rightarrow$$

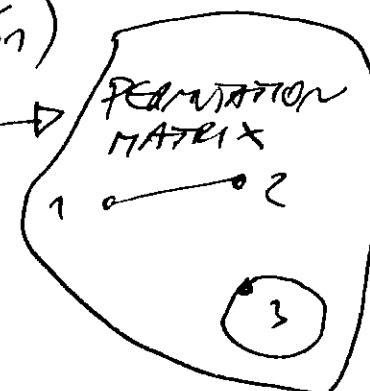
$$H(x_1|Tx_1) \leq H(x_1|Tx_n)$$

THE CONDITIONAL ENTROPY DECREASES.

$$H(Tx_1) - H(Tx_1|x_1) \geq H(Tx_n) - H(Tx_n|x_n)$$

$$[H(Tx_1|x_1) \leq H(Tx_n|x_n)]$$

$$\begin{bmatrix} 1 & 2 & 3 \\ x & & \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 3 \\ T & & \end{bmatrix}$$



$$[x_1, x_2, \dots, x_{52}] \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1,52} \\ p_{21} & p_{22} & \dots & p_{2,52} \\ \vdots & \vdots & \ddots & \vdots \\ p_{52,1} & p_{52,2} & \dots & p_{52,52} \end{bmatrix} = [x_{11}, x_{12}, \dots, x_{52}]$$

$$\boxed{x_{ik} \in \{1, 2, \dots, 52\}}$$

$$\Rightarrow f(x) = ? \quad P(x_i) = \frac{1}{n!}$$

$$- \text{and thus } 3 \text{ ways}$$

$$x_1 = 1, 2, 3; \quad x_2 = 2, 1, 3; \quad x_3 = 2, 3, 1; \quad x_4 = 3, 2, 1;$$

$$x_5 = 3, 1, 2; \quad x_6 = 1, 3, 2$$

$$P_k^4 = \frac{n!}{(n-k)!}$$

$$P_4^4 = \frac{4!}{0!} = 4!$$

$$[H(x) = \sum_{i=1}^n \frac{1}{n!} \log(n!) = \log(n!)]$$

$$[n=52]$$

$$P_3^3 = \frac{n!}{(n-k)!} = \frac{3!}{0!} = 6$$

$$T_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$T_2 = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$T_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$T_4 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$T_5 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$T_6 = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & 0 & 0 \end{bmatrix}$$

$$T_7 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$T_8 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$T_9 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$T_{10} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$T_{11} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$T_{12} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$T_{13} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix}$$

$$T_{14} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}$$

$$T_{15} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$T_{16} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

• Berechne Punktanzahl

$$t_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad t_2 = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad t_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad t_4 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$t_5 = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}, \quad t_6 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$P_2^1 = P_3^2 = \frac{3!}{0!} = 6$$

$$H(x) = \ln 4! = \ln 6 = 2.585.$$

• Unter Punktanzahl Parameter ( $n=2$  müssen also nur 4 Werte ausrechnen)

$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} T_{11} & T_{12} \\ T_{21} & T_{22} \end{bmatrix}$$

$$T_1 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$T_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\begin{array}{c} x_1^n = \begin{bmatrix} 1 & 2 \end{bmatrix} \\ \hline H(x^n) = \ln 2 = 1 \end{array} \quad \begin{array}{c} x_2^n = \begin{bmatrix} 2 & 1 \end{bmatrix} \\ \hline H(x^n) = \ln 1 = 0 \end{array}$$

$$H(x^n) = \sum_{i=1}^n \frac{1}{x_i^n} \ln x_i^n = \ln 1 = 0$$

$$H(TX|T) = \underbrace{P(T=t_1)}_{1/2} \cdot H(t_1 X | t_1) + \underbrace{P(T=t_2)}_{1/2} \cdot H(t_2 X | t_2)$$

PERMUTAȚORII MATRICEI SE GENERAȚIAZĂ CU UNIFORMITATEA.

- $H(t_1 X | t_1) = ?$

$$X_1^n = \{1, 2\}$$

$$t_1 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$x_1^n = [2, 1]$$

$$t_1 \cdot x_1^n = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$t_1 \cdot x_1^n = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$$

$$P(t_1 x_1^n) = \frac{1}{2!}$$

$$P(t_1 x_1^n) = \frac{1}{2!}$$

$$H(t_1 X | t_1) = \frac{1}{2!} \cdot \log 2! + \frac{1}{2!} \cdot \log 2! = \frac{1}{2} + \frac{1}{2} = 1$$

$$H(t_2 X | t_2) = \frac{1}{2} \log 2 + \frac{1}{2} \log 2 = 1$$

$$H(TX|T) = \frac{1}{2} + \frac{1}{2} = 1$$

- SE NOUA XEAMĂ NU PERMUTATOR SOU  $4=3$  VOM VEDA ÎMPOTRIVĂ

$$H(TX|T) = \sum_{i=1}^6 P(T=t_i) \underline{H(t_i X | t_i)}$$

NOU TERMENATĂRIE MODULUI SE GENERAȚIA UNIFORME  
DEZBĂ:

$$H(TX|T) = \frac{1}{6} \sum_{i=1}^6 H(t_i X | t_i) \quad X = [x_1, x_2, x_3]$$

$$H(t_1 X | t_1) = ?$$

$$x_1^n = [1, 2, 3] \quad x_2^n = [1, 3, 2] \quad x_3^n = [2, 1, 3] \quad x_4^n = [2, 3, 1]$$

$$\underline{x_5^n = [3, 1, 2]} \quad x_6^n = [3, 2, 1]$$

$$t_1 x_1^n = [2, 1, 3]; \quad t_1 x_2^n = [3, 1, 2]; \quad t_1 x_3^n = [1, 2, 3]; \quad t_1 x_4^n = [3, 2, 1]$$

$$t_1 x_5^n = [1, 3, 2]; \quad t_1 x_6^n = [2, 3, 1]$$

$$H(t_1 X | t_1) = \left( \frac{1}{6} \log 6 \right) \cdot 6 = \log 6 = \cancel{\log(3!)}$$

- DURĂ NE MODUL UNIFORME CĂ SE PASCOAZĂ  
PERMUTAȚORII MATRICEI, VOR FI

$$H(TX|T) = \sum_{i=1}^6 P(T=t_i) \cdot \underbrace{H(t_i X | t_i)}_{= \log 6} = 6 \cdot \sum_{i=1}^6 P(T=t_i) = \underline{6} = H(X)$$

OVAZ DOARE MODUL A MODULUI VD GENERAȚIA  
SCURTAJ !!

$$H(TX|T) = \sum_{i=1}^{n!} P(T=t_i) H(TX|T=t_i) = \frac{\text{ld}(n!)^2}{\text{ld}(n!)} \sum_{i=1}^{n!} P(T=t_i)$$

④  $H(TX|T) = \text{ld}(n!) = H(X)$  !!!  $H(fX) \geq H(TX|T)$

JE PRAZDNU ČÍTAVOSTA !!!

- SO OGLED NA DOŠE ŽE VLOŽIT ZA  
NAKON LAGE (MORE IN COMMON WORK AREA -  
KUTA DA SE NAKONI UVEDE, T = C  
REVIEWER TAKO ZA GORE  
MATRICA)



$$H(f_1 X | X) \leq H(f_n X | X)$$



$$I(X;Y) \geq I(X;Z) \quad H(X) - H(X|Y) \geq H(X) - H(X|Z)$$

$$H(Z) - H(Z|X) \geq H(Z) - H(Z|X) \quad H(X|Y) \leq H(X|Z)$$

$H(Y) = H(Z) \rightarrow$  STATIONARNA MNOŽINA

$$H(Y|X) \leq H(Z|X)$$

$$H(TX) \geq \underbrace{H(TX|T)}_{\textcircled{B}} = H(T^{-1}TX|T) = H(X|T) = H(X) \quad \textcircled{A}$$

②  $\Rightarrow$  CONSIDERING PROBABILITIES ON T?

$$H(X,T) = H(X) + H(T|X)$$

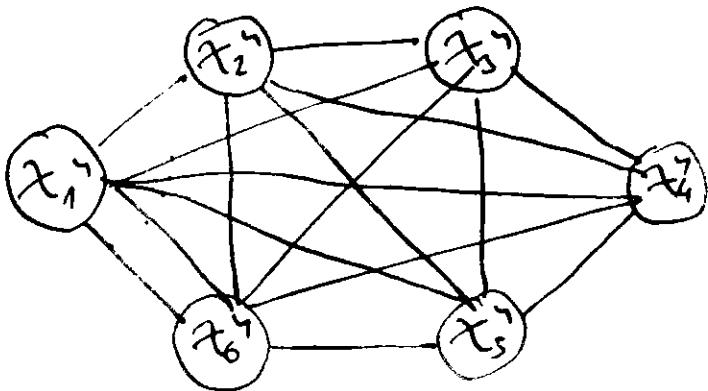
①  $\Rightarrow$  X & T ARE STATIONARY INDEPENDENT

$$H(TX) = \sum_{t,x} p(T,t) \text{ld} p(T,x) \quad ?$$

$$p(T,x) = p(T) \cdot p(x) = \frac{1}{n!} \cdot \frac{1}{n!}$$

$$H(TX) = \sum_{t,x} \frac{1}{(n!)^2} \text{ld} (n!)^2 = (n!)^2 \cdot \frac{1}{(n!)^2} \text{ld} (n!)^2$$

$$H(TX) = 2 \underbrace{(\text{ld} n!)}_{H(X)} = 2H(X)$$



$$x_1^{\gamma} = \{x_1, x_2, x_3\}$$

$$x_2^{\gamma} = \{x_1, x_3, x_2\}$$

$$x_3^{\gamma} = \{x_2, x_1, x_3\}$$

$$x_4^{\gamma} = \{x_2, x_3, x_1\}$$

$$x_5^{\gamma} = \{x_3, x_1, x_2\}$$

$$x_6^{\gamma} = \{x_1, x_2, x_3\}$$

$$\mu_j = \sum_{i=1}^6 p_i \underline{p_{ij}} \quad j = 1, 2, \dots, 4$$

$$p_{ij} = P(t_i)$$

$$H(X) = \sum_{i=1}^6 \mu_i \sum_{j=1}^6 p_{ij} \ln p_{ij} = \sum_{j=1}^4 \sum_{i=1}^6 \mu_i p_{ij} \ln p_{ij}$$

$$= \sum_{i=1}^6 \sum_{j=1}^6 \frac{1}{6!} \frac{1}{6!} \cdot 1/6! = \frac{(6!)^2}{6!^2} \ln 6! = \ln(6!)$$

• Ako oviči se formiraju u istom redu:

$$H(X) = \ln 26 - H\left(\frac{e_1}{26}, \frac{e_2}{26}, \dots, \frac{e_6}{26}\right).$$

$$\boxed{26 = 6 \cdot 6 = 36} \quad \frac{e_1}{26} = \frac{e_2}{26} = \dots = \frac{e_6}{26} = \frac{6}{36} = \frac{1}{6}$$

$$H(X) = \ln 36 - 6 \cdot \frac{1}{6} \cdot \ln 6 = \ln 6 - \ln 6 =$$

$$= \ln 6 + \ln 6 - \frac{6 \ln 6}{6} = \ln 6$$

SWITZERLAND KONSTANTNA VODA SLOVII OD MAJIC ZA VODU NA PR. 105/106 JE DESETA:

$$\boxed{H(TX) = H(X)}$$

ZADOVA STO SAMO SE PREDSTAVI, SOVODATE, A NE JE FAKTOM VEROVATNOSTI TE NA SOVODATE.

$$H(TX) \geq H(TX|T) = \sum_t p(t) H(TX|T=t) =$$

$$= \sum_t p(t) H(TX) = H(X) \sum_t p(t) = H(X)$$

ista vrednost ako učit, t.e. = H(X)

→ ONE TO CONDITIONS REDUCES ENTROPY.

- OVA JE VJUJEST PREDVIRAN DOVOLJE NA OVA STO LUKA 90  
DOVOLJTE NA PR. 105 - 106

4.4 SECOND LAW OF THERMODYNAMICS. Let  $x_1, x_2, x_3, \dots$  be a stationary first-order Markov chain. In section 4.4 it was shown that:

$$H(x_n | x_1) \geq H(x_{n-1} | x_1) \text{ for } n = 2, 3, \dots$$

Thus conditional uncertainty about the future grows with time. This is true although the unconditional uncertainty  $H(x_n)$  remains constant. However, show by example that  $H(x_n | x_1 = x_1)$  does not necessarily grow with  $n$  for fixed  $x_1$ .

$$H(x_n | x_1) = p(x_1 = x_1) H(x_n | x_1 = x_1) + p(x_1 = x_2) H(x_n | x_1 = x_2) \\ + \dots + p(x_1 = x_n) H(x_n | x_1 = x_n)$$

EXAMPLE IF  $n=3$

$$x_1 \rightarrow x_2 \rightarrow x_3 \quad I(x_1; x_2) \geq I(x_1; x_3)$$

$$\cancel{H(x_2)} - H(x_2 | x_1) \geq \cancel{H(x_3)} - H(x_3 | x_1)$$

$$\boxed{H(x_3 | x_1) \geq H(x_2 | x_1)}$$

$$\text{if } p(x_1) = \frac{1}{|X_1|} \Rightarrow \text{UNIFORM DISTRIBUTION}$$

$$H(x_2 | x_1) = \frac{1}{|X_1|} H(x_2 | x_1 = \frac{1}{|X_1|}) + \frac{1}{|X_1|} \cdot H(x_2 | x_1 = \frac{1}{|X_1|}) + \dots$$

$$H(x_2 | x_1) = \cancel{|X_1|} \cdot \frac{1}{\cancel{|X_1|}} H(x_2 | x_1 = \frac{1}{|X_1|}) = H(x_2 | x_1 = \frac{1}{|X_1|})$$

$$H(x_3 | x_1) = H(x_3 | x_1 = \frac{1}{|X_1|}) \Rightarrow H(x_3 | \frac{1}{|X_1|}) = H(x_2 | \frac{1}{|X_1|})$$

$$D(\varphi || \varphi) = \sum_x p(x) \log \frac{p(x)}{\cancel{p(x)}} = - \sum_x p(x) \log \frac{1}{p(x)} + \sum_x p(x) \log \cancel{p(x)}$$

$$= \log |\varphi| - H(\varphi) = \log |\varphi| - \underline{H(\varphi)} \quad \textcircled{2}$$

POTENTIALS:

$$\mu_{x_1}, \mu_{x_2}; \mu_{x_{n+1}}, \mu_{x_n}$$

TRANSITION PROBABILITY

$$P(x_n, x_{n+1}) = P(x_n) \cdot \cancel{r(x_{n+1} | x_n)}$$

$$Q(x_n, x_{n+1}) = q(x_n) \cdot r(x_{n+1} | x_n)$$

$$D(\varphi(x_1, x_{n+1}) || Q(x_n, x_{n+1})) = D(\varphi(x_n) || Q(x_n)) +$$

$$D(\underbrace{p(x_{n+1}|x_n)}_{\theta} \parallel q(x_{n+1}|x_n)) = D(p(x_{n+1}) \parallel q(x_{n+1})) +$$

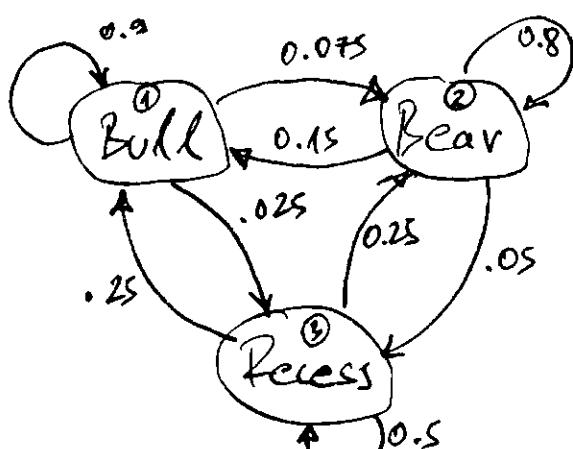
$$+ D(p(x_n|x_{n+1}) \parallel q(x_n|x_{n+1}))$$

$$p(x_{n+1}|x_n) = q(x_{n+1}|x_n) = p(x_{n+1}|x_n) \Rightarrow D(p(x_{n+1}|x_n) \parallel q(x_{n+1}|x_n)) = 0$$

$$\boxed{D(p(x_n) \parallel q(x_n)) \geq D(p(x_{n+1}) \parallel q(x_{n+1}))} \quad \text{i.e.}$$

$$D(M_n \parallel M_n) \geq D(M_{n+1} \parallel M_{n+1})$$

- Períodos za riscov cratim os NIKIRENDA



• BULL WEEK IS FOLLOWED BY ANOTHER BULL WEEK 90% OF TIME

$$\mu_j = \sum_{i=1}^3 \mu_i \underline{p_{ij}} \quad j = 1$$

$$\mu_1 = \mu_1 \cdot p_{11} + \mu_2 \cdot p_{21} + \mu_3 \cdot p_{31}$$

$$\mu_1(1-p_{11}) = \mu_2 p_{21} + \mu_3 p_{31}$$

$$\boxed{0.1 \mu_1 = 0.15 \mu_2 + 0.25 \mu_3}$$

$$\mu_2 = \sum_{i=1}^3 \mu_i \underline{p_{i2}} = \mu_1 \underline{p_{12}} + \mu_2 \underline{p_{22}} + \mu_3 \underline{p_{32}}$$

$$\mu_2(1-p_{22}) = 0.075 \mu_1 + 0.25 \mu_3 \quad \boxed{0.2 \mu_2 = 0.075 \mu_1 + 0.25 \mu_3}$$

$$\mu_3 = \mu_1 \cdot p_{13} + \mu_2 \cdot p_{23} + \mu_3 \cdot p_{33} \quad \boxed{0.5 \mu_3 = 0.025 \mu_1 + 0.05 \mu_2}$$

$$\mu_1 = 0.625 \quad \mu_2 = 0.3125 \quad \mu_3 = 0.0625$$

$$D(M_n \parallel M_n) \geq D(M_{n+1} \parallel M_{n+1}) \quad \text{IF stationary distribution}$$

$$\textcircled{2} \Rightarrow |H(X) - H(Y)| \geq |H(X) - H(Y_{n+1})|$$

IF stationary distribution is  $p(x) = y = \frac{1}{2^n}$



$$D(M_n \parallel M) \geq D(M_{n+1} \parallel M)$$

$$D(M_{\text{full}} || \mu) = \sum_{x \in X} p(x) \log \left( \frac{p(x)}{q(x)} \right) \quad I(x) = \frac{1}{2^x}$$

$$= -H(x^n) + \sum_{x \in X} p(x) \log 2^n = n - H(x^n)$$

const

$$D(M_{n+1} || \mu) = (n+1) - H(x^n) \quad D(M_{n+1} || \mu) \geq D(M_n || \mu)$$

not possible!!!

$$\underline{H(x_1 | t_1)} \geq H(x_1 | x_1, x_2) \stackrel{\text{Markovity}}{=} H(x_1 | t_2) = H(x_m | t_1)$$

$$H(x_1 | x_{1-1}) = H(x_1 | x_{n-1}, x_{n-2}) =$$

↓ due to markovity i.e.  
knowing the present future  
doesn't depend from past

$$\vdash H(\underline{x_{n-1} | x_{n-2}}, x_{1-1}) = H(x_{n-1} | x_{n-2})$$

IF:  $p(x_1) = p(x_{n-1}) \Rightarrow$

$$H(x_1 | x_1) = H(x_{n-1} | x_{n-2}) \quad \text{NE \in for dots!!!}$$

• IF:  $x_{n-1} = f(x_n)$

MADE TO PREDICT FUTURE  
ACTUALLY ASSISTED BY READING PAST

$$H(x_1 | x_1) = H(x_1 | x_1, \underline{x_{n-1}}) = H(x_1 | x_m)$$

↑ MARKOVITY

∴ SEE MARKOV PROPERTY!!!

PROPERTY IS USEFUL TO DETERMINE DIVERSITY!!!

DEFINITION 1:  $x_1 \rightarrow x_{n-1} \rightarrow t_n$

$$I(x_1; x_{n-1}) \geq I(x_1; t_n) \quad H(x_m) - H(x_m | t_1) \geq$$

$$\geq H(x_1) - H(x_1 | x_1); \quad H(x_1) = H(x_{n-1}) \Rightarrow$$

$$H(x_{n-1} | x_1) \leq H(x_1 | x_1) \quad \text{i.e. } H(x_1 | x_1) \geq H(x_m | t_1)$$

**4.5 GENERATION OF RANDOM TREE** Consider the following method of generating a random tree with  $n = 10$  nodes.

110 FIRST CHOOSE THE ROOT NODE:

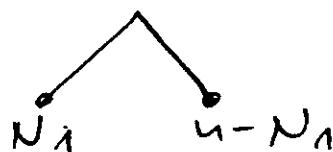
Then expanding one of the terminal nodes at random:



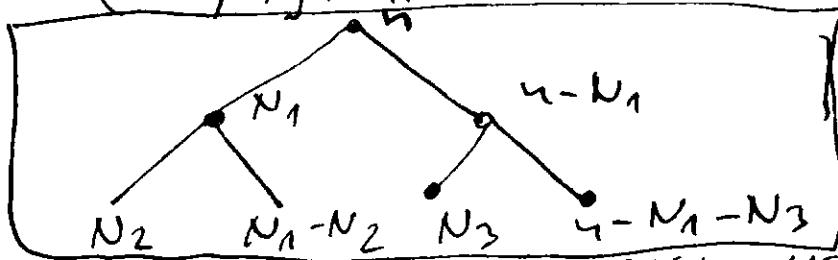
At time  $k = 1$ , chose one of the  $k-1$  terminal nodes according to the uniform distribution and expand it. Continue until  $u =$  terminal nodes have been generated. Thus sequence leading to a tree node tree might look like this:



Surprisingly the following method of generating random trees yields the same probability distribution on trees with  $u =$  terminal nodes. First chose an integer  $N_1$  uniformly distributed on  $\{1, 2, \dots, u-1\}$ . We then have picture:

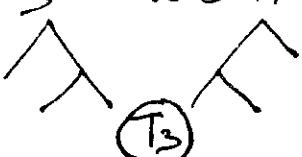


Then chose an integer  $N_2$  uniformly distributed over  $\{1, 2, \dots, N_1-1\}$  and independently chose another integer  $N_3$  uniformly over  $\{1, 2, \dots, (u-N_1)-1\}$ . The picture is now



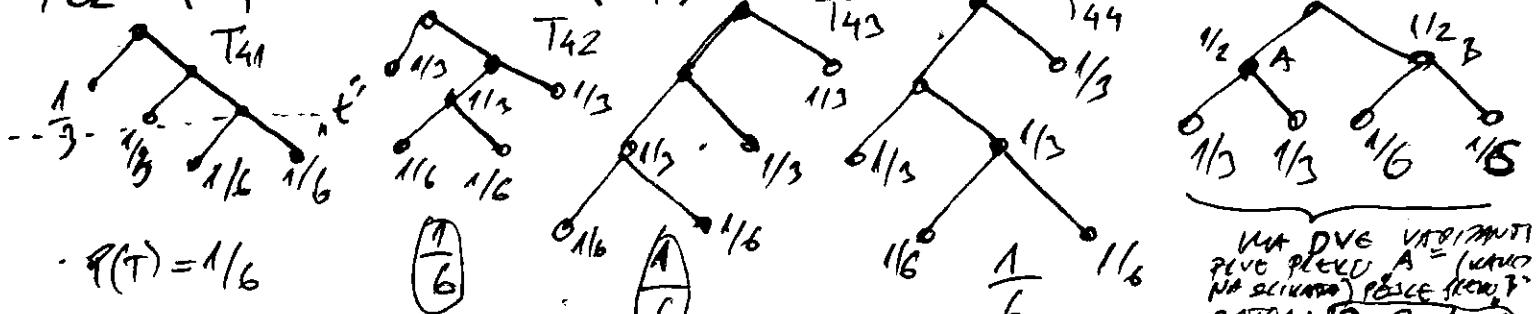
Continue the process until no further subdivision can be made. (The equivalence of these two tree generation schemes follows, for example from Polya's urn model.)

Now let  $T_u$  denote a random  $u$ -node tree generated as described. The probability of such trees seems difficult to describe, but we can find the entropy of this distribution in recursive form. First some examples. For  $u=2$  we have only one tree thus  $H(T_2) = 1 \cdot \log 2$ . For  $u=3$  we have two equal probable trees:



$$\text{Thus } H(T_3) = \left(\frac{1}{2} \log 2\right) \cdot 2 + \frac{1}{2} \log 2$$

For  $q=4$  we have ( $T_4$ ):



$$P(T) = \frac{1}{6}$$

- Võ momentot "t" määriks 00 kui TO UNIFORMNA  
DISTRIKUTA SITKA EGA, GO ESTIMATI

$$8 = 4 \cdot \frac{1}{3} + \frac{1}{3} = \frac{2}{3} + \frac{1}{3} = 1$$

MIT DVE VÄGI MÜTT  
PEDE PLECU A = (VÄGI  
PA SÜVADU PESKE KERVA)  
SATOA!  $P = 2 \cdot \frac{1}{6} = \frac{1}{3}$

$P(T_4)$  NEMA VÄGI SO  
GRANNEEDU NA PROTO  
TULL SO VÄIKET PEG NA  
PEVA, BODU NA PÖÖDUVAD NA PA

NOW FOR THE RECURRANCE RELATION. LET  $N_1(T_n)$  DENOTE  
THE NUMBER OF TERMINAL NODES OF  $T_n$  IN THE RIGHT HALF  
OF THE TREE.

$$\begin{aligned} H(T_4) &\stackrel{(a)}{=} H(N_1, T_n) \stackrel{(e)}{=} H(N_1) + H(T_n | N_1) \stackrel{(c)}{=} \text{ld}(n-1) + H(T_n | N_1) = \\ &\stackrel{(d)}{=} \text{ld}(n-1) + \frac{1}{n-1} \sum_{k=1}^{n-1} (H(T_k) + H(T_{n-k})) \stackrel{(e)}{=} \text{ld}(n-1) + \frac{2}{n-1} \sum_{k=1}^{n-1} H(T_k) \\ &= \text{ld}(n-1) + \frac{2}{n-1} \sum_{k=1}^{n-1} H_k. \end{aligned} \quad (b)$$

(f) USE THIS TO SHOW THAT:

$$(n-1)H_n = nH_{n-1} + (n-1)\text{ld}(n-1) - (n-2)\text{ld}(n-2) \quad \text{OR}$$

$$\frac{H_n}{n} = \frac{H_{n-1}}{n-1} + c_n \quad \text{FOR APPROXIMATELY DEFINED } c_n^2.$$

SINCE  $\sum c_n = C < \infty$ , YOU HAVE SHOWN THAT  $\frac{1}{n} H(T_n)$   
CONVERGES TO A CONSTANT. THUS, THE EXPECTED  
NUMBER OF BITS NECESSARY TO DESCRIBE RANDOM  
TREE  $T_n$  GROWS LINEARLY WITH  $n =$

(g) AT USAGE OF OTHER FORMULA  
 $H(N_1, T_n) = H(N_1) + H(T_n | N_1)$

(h)  $N_1$  IS UNIFORMLY DISTRIBUTED OVER  $\{1, 2, \dots, n-1\}$

$$\text{i.e. } P(N_1) = \frac{1}{n-1} \Rightarrow H(N_1) = \sum_{i=1}^{n-1} \frac{1}{n-1} \text{ld} \frac{1}{n-1}$$

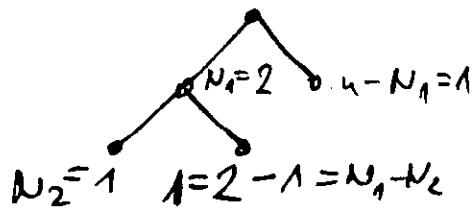
$$N_1 \in \{1, 2, \dots, n-1\}$$

- PROOF NA RECURRANCE PRISTANOT ET  $n=3$   
(i)  $N_1 \in \{1, 2\}$   $P(N_1) = \frac{1}{2}$  e.g.  $[N_1=2]$

$$N_1=2$$

$$1=n-N_1$$

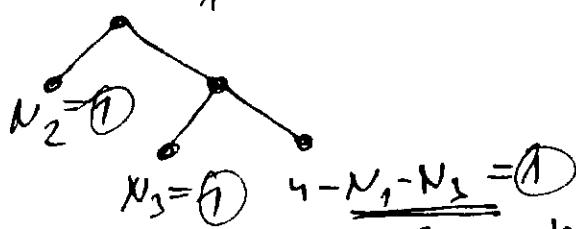
$$2^{\circ} \quad N_2 \in \{1\} \quad N_3 \in \{1..3-2-1\} \in \{0\}$$



DUGIOT PTT

$$1^{\circ}) \quad N_1 \in \{1, 2\} \quad P(N_1) = \frac{1}{2} \quad N_1 = 1$$

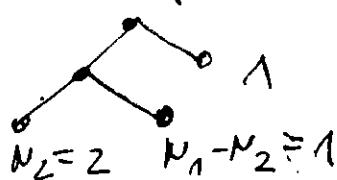
$$2^{\circ}) \quad N_2 \in \{1, \dots, \underline{N_1-1}\} \in \{\} \quad \underline{N_3 = \{1, \dots, \underline{n-N_1-1}\}} \in \{1\}$$



• Recurrence P(11111) =  $\frac{1}{3}$   $n=4$

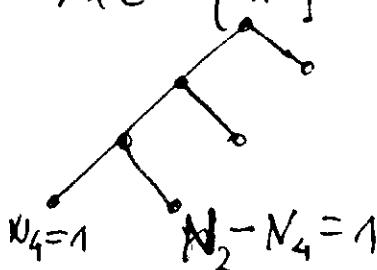
$$1^{\circ}) \quad N_1 \in \{1, 2, 3\} \quad \boxed{N_1=3} \quad \boxed{P(N_1) = \frac{1}{3}}$$

$$2^{\circ}) \quad N_1=3 \quad n-N=1 \quad N_2 \in \{1, 2\}; \quad P(N_2) = \frac{1}{2}; \quad N_2 \in \{1..4-3-1\} = \{\}$$

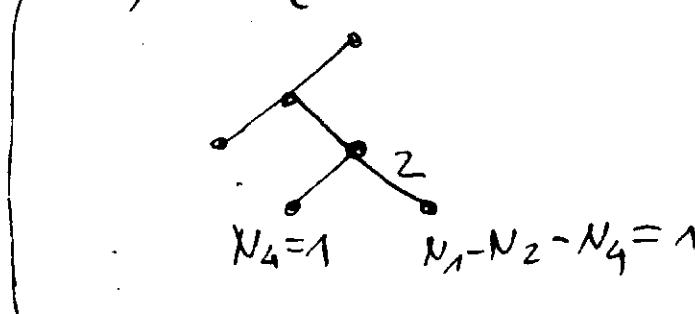


4.)  $\boxed{N_2=2}$  6.)  $\boxed{N_2=1}$

$$3^{\circ}) \quad N_4 \in \{1, 2, \dots, N_2-1\} \quad N_4 \in \{1\}$$



$$3^{\circ}) \quad N_4 \in \{1, 2, \dots, N_1-N_2-1\} \in \{1\}$$



$$1^{\circ}) \quad \boxed{N_1=2}$$

$$N_1=2 \quad n-N_1=2$$

$$2^{\circ}) \quad N_2 \in \{1\} \quad N_3 \in \{1\}$$

$$N_2=1 \quad N_1-N_2=1 \quad n-N_1-N_2=1$$

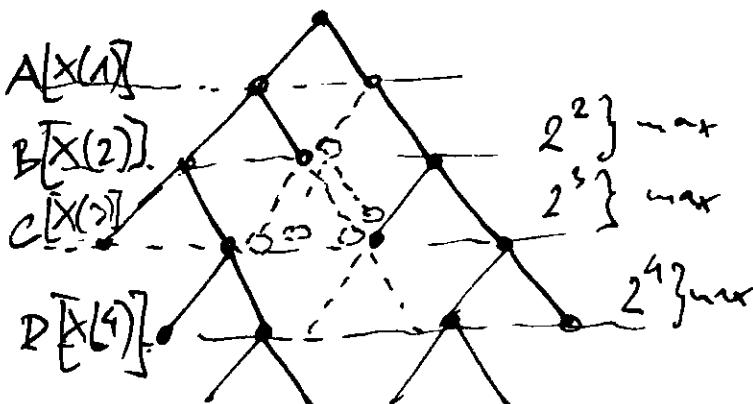
$$N_1=1 \quad 1^{\circ}) \quad n=4 \quad n-N_1=3$$

$$2^{\circ}) \quad N_2 \in \{\} \quad N_3 \in \{1..4-1-1\} \in \{1, 2\}$$

$$N_4 \in \{\} \quad N_5 \in \{1..2-1\} = \{1\}$$

$$N_1=1 \quad N_3=1 \quad n-N_1=2$$

$$n-N_1-N_3-N_5=4-3=1$$



$(n=4)$

	1	2	3	4	5	6	7	8
$T_{41}$				1	3			
					1	2		
						1	1	3

$T_{45} =$

	1	2	3	4	5	6	7	8
		2	3					
			1	1	1	1		

$$\overbrace{\begin{array}{|c|c|c|c|c|} \hline y & = & 31 \\ \hline 1 & 2 & 3 & 4 & 5 \\ \hline 0 & 1 & 2 & 1 & 0 \\ \hline 1 & 1 & 0 & 0 & 2 \\ \hline \end{array}}^{2^{4-1}} \quad \left\{ \begin{array}{l} n-1 \\ 1 \end{array} \right\}$$

$$X'' = \overbrace{\begin{array}{|c|c|c|c|c|} \hline 1 & 2 & 3 & 4 & 5 \\ \hline 0 & 1 & 2 & 1 & 0 \\ \hline 0 & 0 & 1 & 1 & 1 \\ \hline \end{array}}^{2^4} \quad \left\{ \begin{array}{l} 1 \\ 2 \end{array} \right\}$$

$N_1$

	1	2	3	4	5	6	7	8
$T_{42}$			1	3				
				2	1			
					1	1		

$$\begin{aligned}
 H(N_1) &= P(N_1=1) \log \frac{1}{P(N_1=1)} + \\
 &P(N_1=2) \log \frac{1}{P(N_1=2)} + P(N_1=3) \log \frac{1}{P(N_1=3)} \\
 &= \left(\frac{1}{3} \log 3\right) \cdot 3 = 1.098
 \end{aligned}$$

ELIMINACION DE DERIVADAS

1122343      121212 ; 21212 ; 212 ; 213

$N_1$

	1	2	3	4	5	6	7	8
$T_{43}$		2	3	1				
		1	1					

$N_1$

	1	2	3	4	5	6	7	8
$T_{44}$			3	1	2			
				1	2			
					1	1		

• Se incrementan los  $T_3$  y  $T_4$  para obtener la medida de generación de la RAM.

$$(a) H(N_1, T_4) = H(T_4) + H(N_1 | T_4) = \left| \begin{array}{l} N_1 = f(T_4) \\ \Rightarrow H(N_1 | T_4) = 0 \end{array} \right| = H(T_4)$$

$$H(T_4) = H(N_1, T_4) = H(N_1) + H(T_4 | N_1)$$

$$(d) H(T_4 | N_1) = ? \quad H(T_4 | N_1) = \frac{1}{n-1} \sum_{k=1}^{n-1} (H(T_k) + H(T_{n-k}))$$

e.g.  $N_1 = 3$   $\boxed{n=4}$  (FIG. 11.112)

$$\begin{aligned}
 H(T_4 | N_1) &= \frac{1}{3} \sum_{k=1}^3 H(T_k) + H(T_{4-k}) = \frac{1}{3} [H(T_1) + H(T_3)] + \\
 &+ H(T_2) + H(T_2) = \underline{H(T_2)} + \underline{H(T_3)} + \underline{H(T_1)} = \frac{2}{3} [H(T_1) + H(T_2) + H(T_3)]
 \end{aligned}$$

• Direct + via ~~reverentia~~ NT:

$$H(T_4 | N_1=3) = H(T_{41}, T_{42}) = \frac{1}{2} \text{ld}_2 + \frac{1}{2} \text{ld}_2 = 1$$

$$T_{43} \cup H(T_4 | N_1=2) = 1 \text{ld}_1 = 0$$

$$H(T_4 | N_1=1) = \frac{1}{2} \text{ld}_2 + \frac{1}{2} \text{ld}_2 = 1$$

$$H(T_4 | N_1) = P(N_1=3) \cdot H(T_4 | N_1=3) + P(N_1=2) \cdot H(T_4 | N_1=2) + P(N_1=1) \cdot H(T_4 | N_1=1)$$

$$= \frac{1}{3} \cdot 1 + \frac{1}{3} \cdot 1 + \frac{1}{3} \cdot 0 = \frac{2}{3}$$

$$H(T_4) = 4 \cdot \frac{1}{6} \text{ld}_6 + \frac{1}{3} \text{ld}_3 = \frac{2}{3} \text{ld}_2 + \frac{2}{3} \text{ld}_3 + \frac{1}{3} \text{ld}_2$$

$$= \frac{2}{3} + \text{ld}_2$$

$$H(T_4) = H(N_1) + H(T_4 | N_1) = \left(\frac{1}{3} \text{ld}_3\right) \cdot 3 + \frac{2}{3} = \text{ld}_3 + \frac{2}{3}$$

PROBABILITÄT NT P. 117

- DREAM ABOUT CONDITIONAL PROBABILITY

$P(X, Y)$					
X	◊	♥	♠	♣	$P(X)$
RED	$\frac{1}{2} \cdot \frac{1}{2}$	$\frac{1}{4}$	0	0	$\frac{1}{2}$
BLACK	0	0	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$
$P(Y)$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{4}$	

X - COLOUR

Y - CARD FIGURE

$$\text{e.g. } P(X, Y) = (\text{RED}, \diamond) = \\ = P(\text{RED}) \cdot P(\diamond) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$$

$P(Y|X)$

$P(Y X)$					
X	◊	♥	♠	♣	
RED	$\frac{1}{2} \cdot \frac{1}{2}$	$\frac{1}{2}$	0	0	
BLACK			$\frac{1}{2}$	$\frac{1}{2}$	

• ANDERER ZEICHEN HABT  $N_1$  NEU E SANTO ZA VERSUCH

SONDERT - OP PROBABILITY IN POSSIBEL:

$$H(T_4 | N_1=3) = H(T_{41}, T_{42}, T_{43}, T_{44}) = 4 \cdot \frac{1}{4} \text{ld}_4 = 2$$

$$• \text{ANDERER ZEICHEN SO VERSUCHSWEISE: } \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}$$

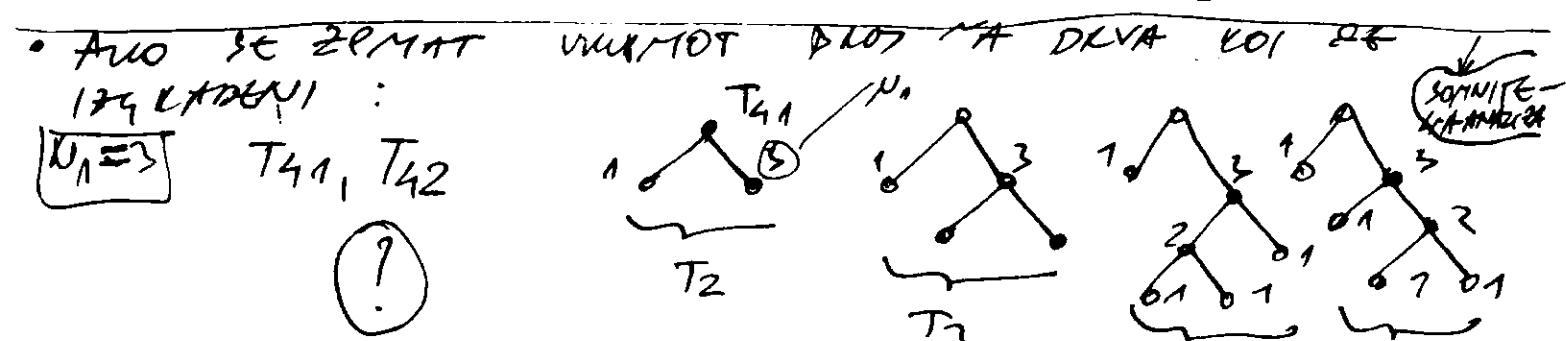
$$H(T_4) = \left(\frac{1}{5} \text{ld}_5\right) 5 = \text{ld}_5$$

$$H(T_4 | N_1=3) = \left(\frac{1}{2} \text{ld}_2\right) \cdot 2 = 1 \quad H(T_4 | N_1=2) = 1 \text{ld}_1 = 0$$

$$H(T_4 | N_1=1) = \left(\frac{1}{2} \text{ld}_2\right) \cdot 2 = 1$$

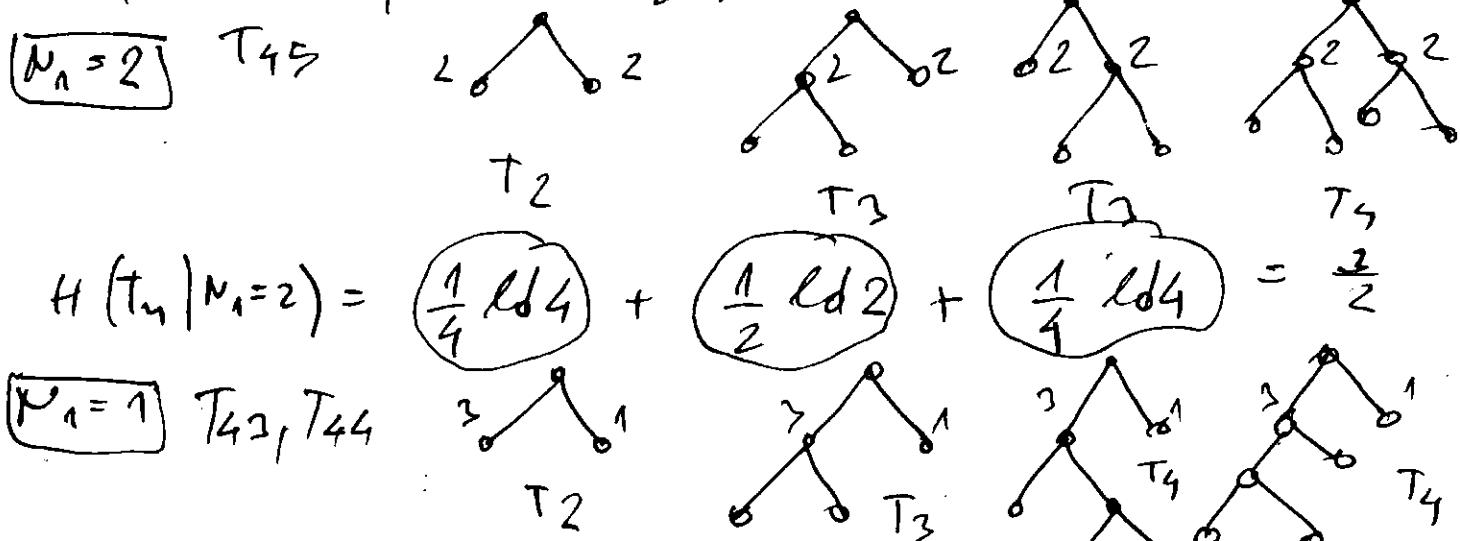
$$H(T_4, N_1) = P(N_1=3) \cdot H(T_4 | N_1=3) + P(N_1=2) \cdot H(T_4 | N_1=2) +$$

$$+ P(N_1=1) \cdot H(T_4 | N_1=1) = \frac{1}{3} + \frac{1}{3} = \frac{2}{3}$$



$$H(T_4 | N_1 = 3) = P(T_2) \text{ld} \frac{1}{P(T_2)} + P(T_3) \text{ld} \frac{1}{P(T_3)} + P(T_4) \text{ld} \frac{1}{P(T_4)}$$

$$= \frac{1}{4} \text{ld} 4 + \frac{1}{4} \text{ld} 4 + \frac{1}{2} \text{ld} 2 = \frac{3}{2}$$



$$H(T_4 | N_1 = 1) = \frac{1}{4} \text{ld} 4 + \frac{1}{4} \text{ld} 4 + \frac{1}{2} \text{ld} 2 = \frac{3}{2}$$

$$H(T_4 | N_1) = P(N_1 = 3) H(T_4 | N_1 = 3) + P(N_1 = 2) H(T_4 | N_1 = 2) + P(N_1 = 1) H(T_4 | N_1 = 1)$$

$$= \frac{1}{3} \cdot \frac{2}{4} \text{ld} 4 + \frac{1}{3} \cdot \frac{1}{4} \text{ld} 4 + \frac{1}{3} \cdot \frac{2}{4} \text{ld} 2 = \frac{1}{3} + \frac{1}{6} + \frac{1}{3} = \frac{5}{6}$$

$$H(T_3 | N_1) = \frac{1}{3} \cdot (H(T_2 | N_1 = 3) + H(T_2 | N_1 = 2) + H(T_3 | N_1 = 1)) = \frac{2}{3}$$

$$T(T_2 | N_1) = \frac{1}{3} (H(T_2 | N_1 = 3) + H(T_2 | N_1 = 2) + H(T_3 | N_1 = 1)) = \frac{1}{2}$$

$$H(T_2) = \frac{1}{2} \text{ld} 3 = \frac{1}{2} \text{ld} 3 \quad T(T_2) = \frac{1}{4} \text{ld} 4 = \frac{1}{2}$$

$$H(T_4) = \frac{5}{12} \text{ld} 12$$

$$= \frac{1}{2} \text{ld} 3 + \frac{5}{6}$$

$$H(T_4) = \frac{5}{12} \text{ld} 12 \cdot 3 = \frac{5}{12} \text{ld} 4 + \frac{5}{12} \text{ld} 3 \quad ?$$

Aus der Zeit von  $T_0$  bis  $T_1$  =

$$H(T_4|N_1) = \frac{1}{3} \left( H(T_4|N_1=2) + H(T_4|N_1=1) + H(T_4|N_1=0) \right) =$$

$$\frac{2/5 \cdot \text{ld} \frac{2}{2}}{2} + \frac{1/5 \cdot \text{ld} \frac{1}{5}}{1} + \frac{2/5 \cdot \text{ld} \frac{1}{2}}{0}$$

$$= \frac{1}{3} \left( \text{ld} \frac{2}{2} \left( 2/5 \cdot 2 + \frac{1}{5} \right) - \text{ld} \frac{1}{5} \left( \frac{2}{5} + \frac{1}{5} \right) \right) = \frac{1}{3} \left( \text{ld} \frac{2}{2} - \frac{4}{5} \text{ld} \frac{1}{2} \right)$$

?

$$H(T_4) = \frac{1}{3} \text{ld} 2 + \frac{1}{3} \text{ld} \frac{1}{5}$$

$$H(T_4) = \frac{1}{3} \cdot \text{ld} 3$$

↑ # (Sommerzeit!!!)

$$\begin{aligned} H(N_1) &= P(N_1=2) \text{ld} \frac{1}{P(N_1=2)} + P(N_1=1) \text{ld} \frac{1}{P(N_1=1)} \\ &+ P(N_1=0) \text{ld} \frac{1}{P(N_1=0)} = 3 \frac{1}{3} \text{ld} \frac{1}{3} = \text{ld} 3 \end{aligned}$$

$$H(T_2) = 0 \quad H(T_3) = \left( \frac{1}{2} \text{ld} 2 \right) \cdot 2 = \text{ld} 2$$

rechnerisch  
mit PP.15

$$H(T_4) = 4 \frac{1}{6} \text{ld} 6 + \frac{1}{3} \text{ld} 2 = \frac{2}{3} \text{ld} 2 + \frac{2}{3} \text{ld} 3 + \frac{1}{3} \text{ld} 3$$

$$H(T_4) = \frac{2}{3} + \text{ld} 2$$

$$H(T_4|N_1) = \frac{2}{3} [H(T_1) + H(T_2) + H(T_3)] = \frac{2}{3} \text{ld} 2 = \frac{2}{3} \text{ld} 2$$

$$H(T_4|N_1) = \frac{1}{N-1} \sum_{k=1}^{N-1} [H(T_k) + H(T_{N-k})]$$

$$H(T_4|N_1) = P(N_1=1) H(T_1|N_1=1) + P(N_1=2) H(T_2|N_1=2) + \dots + P(N_1=N-1) H(T_{N-1}|N_1=N-1)$$

$$P(N_1) = \frac{1}{N-1}$$

$$H(T_4|N_1) = \frac{1}{N-1} \sum_{i=1}^{N-1} H(T_i|N_1=i)$$

$$\text{e.g. } H(T_4|N_1) = P(N_1=2) H(T_2|N_1=2) + P(N_1=1) H(T_3|N_1=1)$$

$$= \frac{1}{3} \left[ \underbrace{2 \frac{1}{2} \text{ld} 2}_{N_1=2} + \underbrace{1 \cdot \text{ld} 1}_{N_1=1} + \underbrace{2 \frac{1}{2} \text{ld} 2}_{N_1=0} \right] = \frac{2}{3} \text{ld} 2$$



$$H(T_4|N_1) = \frac{1}{3} \sum_{i=1}^3 H(T_4|N_1=i) = \frac{1}{3} \sum_{i=1}^3 [H(T_{n-i}) + H(T_i)]$$

$$H(T_4|N_1=2) = P(N_2=2) \cdot H(T_2|N_2=2) + P(N_2=1) \cdot H(T_3|N_2=1) = \frac{1}{2} H(T_2) = \text{ld} 2$$

$$H(T_4|N_1=1) = H(T_3) = \text{ld} 2$$

- Sústavom zároveň s deca:

$$H(T_4 | N_1 = i) = H(T_{4-i}) + H(T_i)$$

(\*)  $H(T_4 | N_1 = 3) = H(T_3)$   
GO WITH THE FLOW, DECA

$$H(T_5 | N_1) = P(N_1=4) \cdot H(T_5 | N_1=4) + P(N_1=3) \cdot H(T_5 | N_1=3) +$$

$$+ P(N_1=2) \cdot H(T_5 | N_1=2) + P(N_1=1) \cdot H(T_5 | N_1=1) =$$

$$= \frac{1}{4} \left[ H(T_4) + H(T_3) + H(T_2) + H(T_1) \right] \cdot (2)$$

$\boxed{n=5}$

$$\sum_{i=1}^4 H(T_i) + H(T_{4-i}) = H(T_1) + H(T_4) + H(T_2) + H(T_3) +$$

$$+ H(T_3) + H(T_2) + H(T_4) + H(T_1) =$$

$$= \frac{1}{2} [H(T_1) + H(T_2) + H(T_3) + H(T_4)]$$

GO WITH THE FLOW, DECA

(\*) Odeča posledna na rok sto je random za markov chain

Znaci:  $H(T_4 | N_1) = \frac{1}{n-1} \sum_{i=1}^{n-1} H(T_i) + H(T_{n-i}) = \frac{2}{n-1} \sum_{i=1}^{n-1} H(T_i)$

$H(T_3) = \left( \frac{1}{2} \cdot \log 2 \right) \cdot 2 = \log 2$

$H(T_4) = H(\textcircled{1}) + H(T_4 | N_1) = \log 3 + \frac{2}{3}$

$H(T_5) = \log(n-1) + \frac{2}{n-1} \sum_{i=1}^{n-1} H(T_i)$

PROBABILITY OF NUMBER  
COMBINATION AND NUMBER  
OF POSSIBLE WAYS  
IN RANDOM WALK  
ON NUMBER SCALE  
 $\frac{2}{n-1} = \frac{1}{2}$  THM V

$H(T_5) = \log 4 + \frac{2}{4} \sum_{i=1}^4 H(T_i) = \log 4 + \frac{1}{2} \left( 0 + 0 + 1 + \log 3 + \frac{2}{3} \right)$

$H(T_5) = 2 + \frac{1}{2} + \frac{1}{3} + \frac{1}{2} \log 3 = \frac{12+3+2}{6} + \frac{1}{2} \log 3 = \frac{19}{6} + \frac{1}{2} \log 3$

(f)  $H_n = \log(n-1) + \frac{2}{n-1} \sum_{i=1}^{n-1} H_K$

$(n-1)H_n = (n-1)\log(n-1) + 2 \sum_{K=1}^{n-2} H_K = (n-1)\log(n-1) + 2H_{n-1} + 2 \sum_{K=1}^{n-2} H_K$

$H_{n-1} = \log(n-2) + \frac{2}{n-2} \sum_{K=1}^{n-3} H_K \Rightarrow 2 \sum_{K=1}^{n-2} H_K = (n-2)H_{n-1} - (n-2)\log(n-2)$

$(n-1)H_n = (n-1)\log(n-1) + 2H_{n-1} + (n-2)H_{n-1} - (n-2)\log(n-2) =$

$= (n-1)\log(n-1) - (n-2)\log(n-2) + nH_{n-1}$

$(n-1)H_n = (n-1)\log(n-1) - (n-2)\log(n-2) + nH_{n-1}$

PROVED!!!

$$H_n = \text{ld}(n-1) - \frac{n-2}{n-1} \text{ld}(n-2) + \frac{1}{n-1} H_{n-1} \Rightarrow$$

$$\frac{H_n}{n} = \frac{H_{n-1}}{n-1} + \underbrace{\frac{\text{ld}(n-1)}{n} - \frac{n-2}{(n-1) \cdot n} \text{ld}(n-2)}_{C_n}$$

$$C_n = \frac{(n-1)\text{ld}(n-1) - (n-2)\text{ld}(n-2)}{n(n-1)}$$

$$\frac{H_n}{n} = \frac{H_{n-1}}{n-1} + C_n$$

$$\lim_{n \rightarrow \infty} C_n = 0$$

$$\lim_{n \rightarrow \infty} \left( \frac{H_n}{n} - \frac{H_{n-1}}{n-1} \right) = 0$$

Edition 1 - selecteaza future (este răbdător să alegi ceea ce urmărești)

$$H(T_n | N_1 = k) = H(T_k, T_{n-k} | N_1 = k) = H(T_k | N_1 = k) + H(T_{n-k} | N_1 = k) = H(T_k) + H(T_{n-k})$$

OVA & VOMA ZBOOGĂZĂ ACROPO POD-PROV, PESOPO  
POD-PROV SE IZBUCNEȘTE NEZAVISNO.

sega și oarecă definitoria nu acordă entropia:

$$H(T_n | N_1) = \sum_{k=1}^{n-1} P(N_1 = k) \cdot H(T_n | N_1 = k) = \frac{1}{n-1} \sum_{k=1}^{n-1} [H(T_k | N_1 = k) + H(T_{n-k} | N_1 = k)] = \frac{1}{n-1} \sum_{k=1}^{n-1} [H(T_k) + H(T_{n-k})] = \frac{1}{n-1} \sum_{k=1}^{n-1} \begin{cases} \text{LEVOPO} \\ \text{PESOPO POD-} \\ \text{PROV SE} \\ \text{IZBUCNEȘTE} \end{cases} = \frac{2}{n-1} \sum_{k=1}^{n-1} H(T_k)$$

$$\frac{1}{n-1} \sum_{k=1}^{n-1} [H(T_k) + H(T_{n-k})] = \frac{1}{n-1} \sum_{k=1}^{n-1} H(T_k) + \frac{1}{n-1} \sum_{k=1}^{n-1} H(T_{n-k})$$

$$= \left| \begin{array}{l} i=n-k \Rightarrow k=1 \quad i=n-1 \\ k=n-i \quad i=1 \end{array} \right| = \frac{1}{n-1} \sum_{k=1}^{n-1} H(T_k) + \frac{1}{n-1} \sum_{i=1}^{n-1} H(T_i) =$$

$$= \frac{2}{n-1} \sum_{k=1}^{n-1} H(T_k)$$

(f) Așa în Edition 1:

$$C_n = \frac{\text{ld}(n-1)}{n} - \frac{\text{ld}(n-2)}{n-1} + \frac{2 \text{ld}(n-2)}{n(n-1)}$$

$$\frac{H_{n-1}}{n-1} = \frac{H_{n-2}}{n-2} + C_{n-1} \quad C_{n-1} = \frac{\text{ld}(n-2)}{n-1} - \frac{\text{ld}(n-3)}{n-2} + \frac{2 \text{ld}(n-3)}{(n-2)(n-1)}$$

$$\frac{H_n}{n} = \frac{H_{n-2}}{n-2} + C_n + C_{n-1} = \frac{H_{n-2}}{n-2} + \frac{\text{ld}(n-1)}{n} + \frac{2 \text{ld}(n-2)}{n(n-1)} - \frac{\text{ld}(n-3)}{n-2} + \frac{2 \text{ld}(n-3)}{(n-2)(n-1)}$$

$$\text{II} = \frac{-(n-1) \text{ld}(n-2) + 2 \text{ld}(n-2)}{(n-2)(n-1)} = \frac{n-2(n-3) \text{ld}(n-2)}{(n-2)(n-1)} \cdot \frac{1}{n(n-1)}$$

070342764

072334049

Macro

- MIGRACIÓN
- ECOLOGÍA
- ANEXOS 2008
- TRANSPORTACIÓN
- OCUPA 2008

CE SE SISTEMA  
NO MIGRACIÓN  
INTERACCIÓN (TELECO-  
MING THE SWIM)

078270741 Igde

$$\textcircled{1} \quad \frac{H_n}{n} = \frac{H_{n-1}}{n-1} + \frac{ld(n-1)}{n} - \frac{ld(n-2)}{n-1} + \frac{2ld(n-2)}{n(n-1)}$$

$$\textcircled{2} \quad \frac{H_n}{n} = \frac{H_{n-2}}{n-2} + \frac{ld(n-1)}{n} + \frac{2ld(n-2)}{n(n-1)} - \frac{ld(n-3)}{n-2} + \frac{2ld(n-3)}{(n-2)(n-1)}$$

$$c_n = \frac{ld(n-1)}{n} - \frac{ld(n-2)}{n-1} + \frac{2ld(n-2)}{n(n-1)}$$

$$\sum_{j=n-1}^n \frac{2ld(j-2)}{j(j-1)} = \frac{2ld(n-3)}{(n-1)(n-2)} + \frac{2ld(n-2)}{n(n-1)} \quad \left. \begin{array}{l} \text{OD} \\ \textcircled{2} \end{array} \right\}$$

$$\sum_{j=n}^n \frac{2ld(j-2)}{j(j-1)} = \frac{2ld(n-2)}{n(n-1)} \quad \left. \begin{array}{l} \text{OD} \\ \textcircled{1} \end{array} \right\}$$

$$\textcircled{3} \quad \frac{H_{n-2}}{n-2} = \frac{H_{n-3}}{n-3} + c_{n-2} = \frac{H_{n-3}}{n-3} + \frac{ld(n-3)}{n-2} - \frac{ld(n-4)}{n-3} + \frac{2ld(n-4)}{(n-2)(n-3)}$$

$$\sum_{j=n-2}^n \frac{2ld(j-2)}{j(j-1)} = \frac{2ld(n-4)}{(n-2)(n-3)} + \frac{2ld(n-3)}{(n-1)(n-2)} + \frac{2ld(n-2)}{n(n-1)} \quad \left. \begin{array}{l} \text{CE SE SISTEMA} \\ \text{NO MIGRACIÓN} \\ \text{INTERACCIÓN} \end{array} \right\}$$

$$\textcircled{3} \Rightarrow \textcircled{2}$$

$$\frac{H_n}{n} = \frac{H_{n-3}}{n-3} + \frac{ld(n-1)}{n} + \frac{2ld(n-2)}{n(n-1)} + \frac{2ld(n-3)}{(n-1)(n-2)} - \frac{ld(n-4)}{n-3} + \frac{2ld(n-4)}{(n-2)(n-3)}$$

$$\frac{H_n}{n} = \frac{ld(n-1)}{n} + \sum_{j=3}^n \frac{2ld(j-2)}{j(j-1)} \quad \left. \begin{array}{l} \text{OD} \\ \text{CE SE SISTEMA} \\ \text{NO MIGRACIÓN} \end{array} \right\}$$

$$\lim_{n \rightarrow \infty} \frac{H_n}{n} = \lim_{n \rightarrow \infty} \frac{ld(n-1)}{n} + \sum_{j=3}^{\infty} \frac{2ld(j-2)}{j(j-1)} \leq \sum_{j=3}^{\infty} \frac{2ld(j-1)}{(j-1)^2} \quad \left. \begin{array}{l} \text{zeta} \\ \text{zeta} \end{array} \right\} \text{converges!}$$

$$= \sum_{K=2}^{\infty} \frac{2ldK}{K^2} \leq \sum_{j=2}^{\infty} \frac{2}{j^2} = \sum_{j=2}^{\infty} \frac{2}{j^2} \quad \left. \begin{array}{l} \text{zeta} \\ \text{zeta} \end{array} \right\} \lim \frac{H_n}{n} = 1.736 \text{ Gts}$$