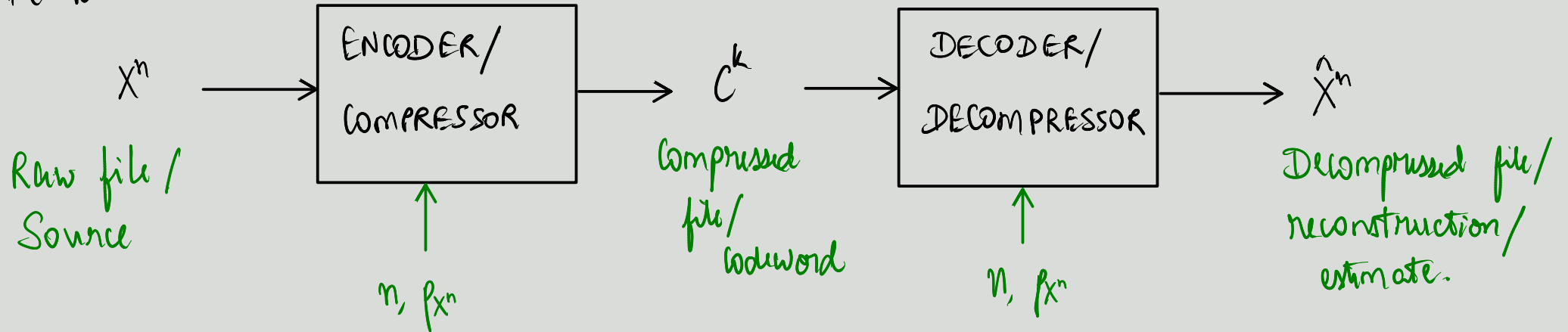


Source Coding/Data Compression

Claude Shannon, "A Mathematical Theory of Communication"

X_1, \dots, X_n
 $X_i \in \mathcal{X}$



p_{X^n} : Source distribution

Universal

Memoryless: $X^n \sim \text{iid}(p_X)$

Rate: $R = \frac{k}{n}$

Assumptions:

① X^n is random ✓

② $X^n \sim \text{iid}(p_x)$ ✗

③ p_x is known to ENC DEC ✗

f_{X^n}

1. Lossless Compression:

$$\hat{X}^n \cong X^n$$

Fixed-length compression:

k depends only on n, P_{X^n}

P_e = Probability of error

Harry Potter I	1 MB	<u>100 kB</u>
II	1 MB	100 kB

$$R = \frac{k}{n}$$

Variable-length compression:

$$\underline{\hat{X}^n \cong X^n}$$

→ Expected length = $E k(X^n)$

$$R_{avg} = \frac{E k(X^n)}{n}$$

k can vary even if n, P_{X^n} fixed

1 MB	100 kB
1 MB	102 kB

2. Lossy Compression:

$$\hat{X}^n \neq X^n$$

Distortion measure

$$E d(X^n, \hat{X}^n) < \delta$$

$$MSD = E \| \hat{X}^n - X^n \|^2 = E \sum_{i=1}^n (X_i - \hat{X}_i)^2$$

$n=2$

0 0	0.5	→	0	→	0 0
0 1	0.25	→	1	→	0 1
1 0	0.125	→	0 0	→	1 0
1 1	0.125	→	0 1	→	1 1
	↑		c^k		
	$P(x^n)$				
2 bits					

Expected length: $E k(x^n)$

$$= 0.5 \times 1 + 0.25 \times 1 + 0.125 \times 2 + 0.125 \times 2$$

$$= 0.5 + 0.25 + 0.5$$

$$= 1.25 \text{ bits} < 2 \text{ bits.}$$

$$R_{avg} = \frac{1.25}{2} < 1$$

0 0	0.5	ENC →	0
0 1	0.25	→	1
<u>1 0</u>	0.125	→	0
<u>1 1</u>	0.125	→	1

DEC

$$0 \rightarrow 00$$

$$1 \rightarrow 01$$

$k=1$ bit

P_e

Probability of error

$$\geq \Pr[\hat{X}^n \neq X^n]$$

$$\geq P(10) + P(11) \geq 0.25$$

$$R_{avg} = \frac{E_k(X^n)}{n}$$

Optimal variable-length compression

Assumptions: (n, p_{X^n}) is known

Computational power - free.

Want $\mathbb{E} k(X^n)$ to be as small as possible.

$$p_{X^n}(X^n) \uparrow \Rightarrow k(X^n) \downarrow$$

$$X^n \in \mathcal{X}^n \quad \mathcal{X} = \{0, 1, 2, \dots, a\}$$

$$n_1 \quad \text{---} \quad \text{---} \quad \text{---} \quad n_n$$

$$X^n(1)$$

$$X^n(2)$$

⋮

$$X^n(a+1)^n$$

$$p_{X^n}(X^n(1))$$

∨

$$p_{X^n}(X^n(2))$$

⋮

⋮

⋮

c^k

$(a+1)^n$

$\mathcal{N}^n(1)$ 1 0

$\mathcal{N}^n(2)$ 0 1

$\mathcal{N}^n(3)$ 0 0

$\mathcal{N}^n(4)$ 1 1

⋮

⋮

⋮

0 \emptyset

1 0

00 1

01 00

10 01

11 10

000 11

001 000

010 !

⋮

111

0000

⋮

⋮

$E k(X^n)$

$$= \sum_{i=1}^{(2^n)-1} k(\mathcal{N}^n(i))$$

$P_{X^n}(\mathcal{N}^n(i))$

$n(i)$

$$k(n(1)) = 0$$

$$k(n(2)) = 1$$

$$k(n(3)) = 1$$

$$4 = 2$$

$$5 = 2$$

$$6 = 2$$

$$7 = 2$$

$$8 = 3$$

⋮

$$15 = 3$$

$$16 = 4$$

⋮

$$31 = 4$$

⋮

$$\underline{\underline{k(n(i)) = \lfloor \log_2 i \rfloor}}$$

$$k(w(i)) = \lfloor \log_2 i \rfloor$$

$$p_{X^n}(w(1)) \geq p_{X^n}(w(2)) \geq \dots$$

Claim: $p_{X^n}(w(i)) \leq \frac{1}{i}$

$$p_{X^n}(1) \leq 1 \quad p_{X^n}(2) \leq \frac{1}{2}$$

$$p_{X^n}(3) \leq \frac{1}{3}$$

Assume: $p_{X^n}(2) > \frac{1}{2} \Rightarrow p_{X^n}(1) \geq p_{X^n}(2) > \frac{1}{2}$
 $\Rightarrow p_{X^n}(1) + p_{X^n}(2) > \frac{1}{2} + \frac{1}{2} = 1$

EXPECTED

LENGTH

$$E k(X^n) = \sum_{i=1}^{(a+1)^n} k(w(i)) p_{X^n}(w(i))$$

$$\leq \sum_{i=1}^{(a+1)^n} p_{X^n}(w(i)) \lfloor \log_2 i \rfloor$$

$$\leq \sum_{i=1}^{(a+1)^n} p_{X^n}(w(i)) \left\lfloor \log_2 \frac{1}{p_{X^n}(w(i))} \right\rfloor$$

$$\leq \sum_{i=1}^{(a+1)^n} p_{X^n}(w(i)) \log_2 \frac{1}{p_{X^n}(w(i))}$$

$$= \sum_{w} p_{X^n}(w) \log \frac{1}{p_{X^n}(w)}$$

$H(X^n)$

ENTROPY OF p_{X^n}

In general, if $P_{X^n}(i) > \frac{1}{i}$ for some i

$$\Rightarrow P_{X^n}(1) \geq P_{X^n}(2) \geq \dots \geq P_{X^n}(i) > \frac{1}{i}$$

$$\sum_{j=1}^i P_{X^n}(j) = P_{X^n}(1) + P_{X^n}(2) + \dots + P_{X^n}(i)$$
$$> \frac{1}{i} + \frac{1}{i} + \frac{1}{i} + \dots + \frac{1}{i}$$

$$\geq i \times P_{X^n}(i) > i \times \frac{1}{i} = 1 \Rightarrow \text{NOT POSSIBLE}$$

Contradiction!

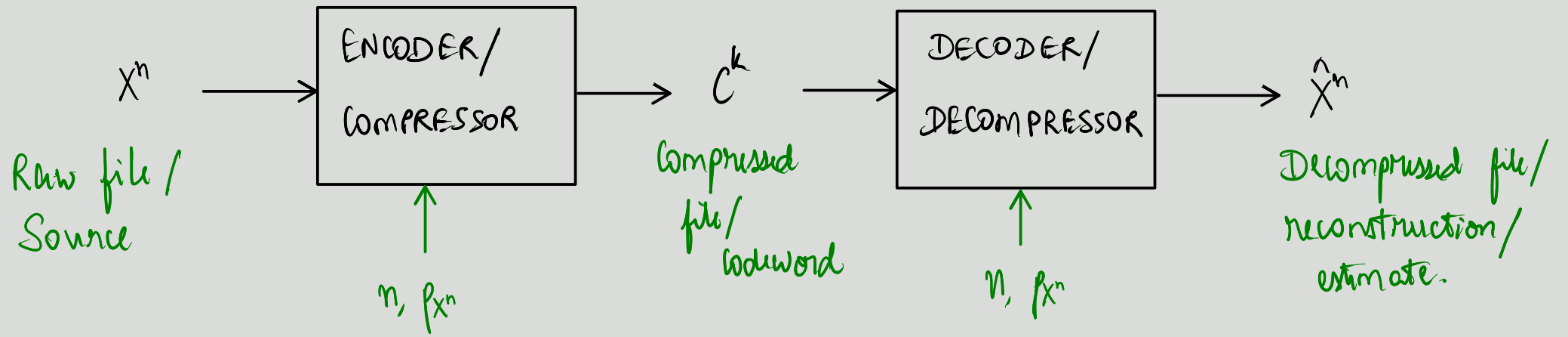
$$P_{X^n}(i) \leq \frac{1}{i} \Rightarrow i \leq \frac{1}{P_{X^n}(i)} = \frac{1}{P_{X^n}(X^n(i))}$$

$$R_{\text{avg}} = \lim_{n \rightarrow \infty} \frac{E k(X^n)}{n} = \lim_{n \rightarrow \infty} \frac{H(X^n)}{n} \rightarrow \underline{\text{ENTROPY RATE}}$$

Can construct compressor for which $R_{\text{avg}} = \text{ENTROPY RATE}$.

For iid X^n ,

$$\lim_{n \rightarrow \infty} \frac{H(X^n)}{n} = H(X) = \sum_{x \in \mathcal{X}} p_X(x) \log_2 \frac{1}{p_X(x)}$$



Lossless source coding theorem:

$$\textcircled{1} X^n \sim \text{iid}(p_X)$$

for every $\epsilon > 0$, (fixed length compressor)

Achievability: We can construct (ENC, DEC) s.t.

$$R \geq H(X) + \epsilon$$

$$P_e \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

Converse: for every (ENC, DEC) with $R < H(X)$,

$$P_e \rightarrow 1 \quad \text{as } n \rightarrow \infty$$

fixed
length
compression

$$X^n \sim \text{iid}(p_X) \Rightarrow H(X^n) = n H(X) \Rightarrow \frac{H(X^n)}{n} = H(X)$$

$$H(X^n) = \sum_{x^n} p_{x^n} \log_2 \frac{1}{p_{x^n}}$$

$$= \sum_{x^n} \left(\prod_{i=1}^n p_X(x_i) \right) \log_2 \frac{1}{\left(\prod_{i=1}^n p_X(x_i) \right)} \quad (\text{iid})$$

$$= \sum_{x^n} \prod_{i=1}^n p_X(x_i) \sum_{j=1}^n \log \frac{1}{p_X(x_j)}$$

$$= \sum_{j=1}^n \sum_{x^n} \left(\prod_{i=1}^n p_X(x_i) \right) \left(\log \frac{1}{p_X(x_j)} \right)$$

$$= \sum_{j=1}^n \sum_{x_j} p_X(x_j) \log_2 \frac{1}{p_X(x_j)} = n \sum_{x_j} p_X(x_j) \log_2 \frac{1}{p_X(x_j)}$$

$$H(X^n) = H(X_1) + H(X_2) + \dots + H(X_n) \quad \text{if } X^n \text{ is iid}$$

Entropy is non-negative

$$X \sim P_X$$

$$H(X) = \sum_{\omega \in \Omega} P_X(\omega) \log_2 \frac{1}{P_X(\omega)}$$

By convention,
 $P_X(\omega) = 0$

$$P_X(\omega) \log_2 \frac{1}{P_X(\omega)} = 0$$

$$0 < P_X(\omega) \leq 1$$

$$\Rightarrow 1 \leq \frac{1}{P_X(\omega)}$$

$$H(X) = \sum_{\omega} P_X(\omega) \log_2 \frac{1}{P_X(\omega)} \geq 0$$

$$\lim_{\substack{p \rightarrow 0 \\ p \downarrow}} p \log_2 \frac{1}{p} = 0$$

Entropy:

$H(X)$

$H(P_X)$

→ Better notation for entropy

↓

actually a function of P_X & not X .

Entropy is invariant to relabeling

$$\mathcal{X} = \{0, 1, 2\}, \quad p_X(0) = \frac{1}{2}$$

$$p_X(1) = \frac{1}{4}$$

$$p_X(2) = \frac{1}{4}$$

$$H(X) = 1.5 \text{ bits}$$

Entropy is a function only of the probability multiset $\left\{ \frac{1}{2}, \frac{1}{4}, \frac{1}{4} \right\}$

$$q_X(0) = \frac{1}{4}$$

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

$$q_X(2) = \frac{1}{4}$$

$$H(q_X) = 1.5 \text{ bits}$$

$$H(X) = \sum_{u \in \mathcal{X}} p_X(u) \log_2 \frac{1}{p_X(u)} \quad \underline{\underline{\text{bits}}}$$

$$(\) \log_e (\) \quad \underline{\underline{\text{nats}}}$$

$$\textcircled{1} \quad X \sim \text{Unif}(\mathcal{X}) \quad |\mathcal{X}| = m$$

$$\begin{aligned} H(X) &= \sum_{x} \frac{1}{m} \log_e \frac{1}{\frac{1}{m}} \\ &= \sum_{x} \frac{1}{m} \log_2 m \\ &= \log_2 m \end{aligned}$$

$$P_X(x) = \frac{1}{m} \quad \forall x \in \mathcal{X}$$

$$H(X) = \log |\mathcal{X}|$$

$$\textcircled{2} \quad \mathcal{X} = \{1, 2, 3, \dots\} \quad Y = \mathbb{Z}_{>0}$$

Geometric rv:

$$P_X(x) = \left(\frac{1}{2}\right)^x$$

$$\sum_{x=1}^{\infty} \left(\frac{1}{2}\right)^x = 1$$

$$H(X) = \sum_{x=1}^{\infty} \frac{1}{2^x} \log_2 \frac{1}{\frac{1}{2^x}} = \sum_{x=1}^{\infty} \frac{x}{2^x} = 2 \text{ bits.}$$

$$\sum_{k=1}^{\infty} \frac{1}{2^k}$$

$$\sum_{n=1}^{\infty} p^n = \frac{p}{1-p}$$

$$\sum_{n=1}^{\infty} n p^n$$

$$= p \sum_{n=1}^{\infty} n p^{n-1} = p \frac{d}{dp} \left(\sum_{n=1}^{\infty} p^n \right)$$

$$= p \frac{d}{dp} \left(\frac{p}{1-p} \right) = p \left(\frac{1}{1-p} + \frac{p}{(1-p)^2} \right)$$

$$= p \times \frac{1-p+p}{(1-p)^2}$$

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

$x_i \in \mathcal{X}$

$$\mathbb{E} k(X^n) \approx H(X^n)$$

$$X^n \text{ compressed } \underline{\underline{C^k}} \quad \underline{\underline{\mathbb{E} k}} \approx H(X^n) \stackrel{\text{iid}}{=} nH(X) \leq n \log |\mathcal{X}|$$

$$|\mathcal{X}| = m$$

$$\underline{\underline{n \lceil \log_2 m \rceil}}$$

iid Geometric ($1/2$) seq \rightarrow 2n bits on avg

1	006
2	001
3	010
4	011
5	100

$\lceil \log_2 n \rceil$

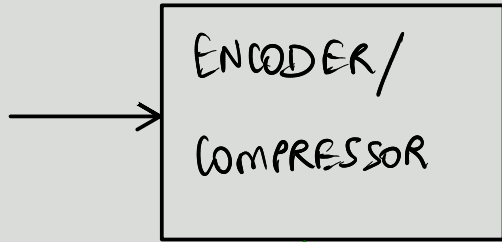
Fixed-length compression

$$k \rightarrow \text{fn of } n, p_X$$

$$X_1, \dots, X_n$$

$$X_i \in \mathcal{X}$$

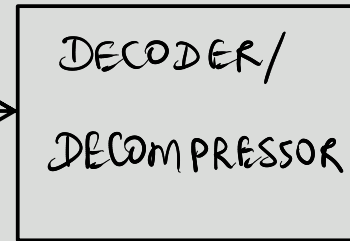
X^n
Raw file / Source



$$n, p_X$$

$$C^k \in \{0,1\}^k$$

Compressed file / codeword



$$n, p_X$$

\hat{X}^n
Decompressed file / reconstruction / estimate.

$$P_e = P_n[\hat{X}^n \neq X^n]$$

k - fixed depends only on n, p_X

$$X^n \sim \text{iid}(p_X)$$

$$X_i \sim p_X$$

$$R = \frac{k}{n}$$

P_e small.

$$n \gg 1$$

$$n \rightarrow \infty$$

$$\lim_{n \rightarrow \infty} \frac{k}{n} = R$$

$$= R$$

small,

$$P_e \rightarrow 0$$

as $n \rightarrow \infty$

Compressing Bernoulli(p) sequences

$$p_X(1) = p, \quad p_X(0) = 1-p.$$

$X^n = 0101101111000000$

↓

$C^h = 000010010$

$$R_{opt} = H(X)$$

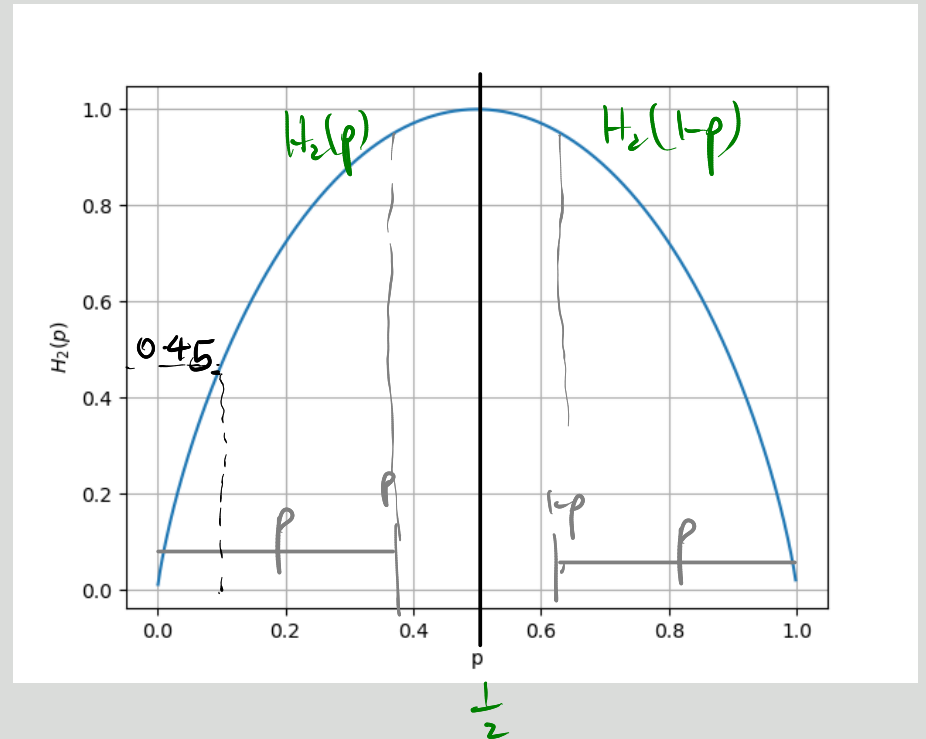
$$= p \log_2 \frac{1}{p} + (1-p) \log_2 \frac{1}{1-p}$$

$$R_{opt}(p) = H_2(p) \rightarrow \text{Binary entropy}(p)$$

$$\text{as } p \rightarrow 0 \quad p \log_2 \frac{1}{p} \rightarrow 0, \quad (1-p) \log_2 \frac{1}{1-p} \rightarrow 0 \Rightarrow H_2(p) \rightarrow 0$$

$$p \rightarrow 1$$

$$H_2(p) \rightarrow 0$$



GOALS Construct ENC, DEC st

$$\textcircled{1} \frac{k}{n} \rightarrow H(X) \quad \text{as } n \rightarrow \infty$$

$$\textcircled{2} p_e = P_n[\hat{X}^n \neq X^n] \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Structure of ANY fixed-length compressor

$$X^n \sim \text{iid Ber}(0.5)$$

$$H_2(0.5) = 0.45$$

$$n \rightarrow 0.45n$$

$$X^n \rightarrow C^{0.45n}$$

$$10^{-6}$$

$$10^7$$

$$\underline{X^{1000}} \rightarrow C^{450} \rightarrow \underline{\hat{X}^{1000}}$$

$$P_n[\hat{X}^{1000} \neq X^{1000}] \leq 10^{-2}$$

EXTREME:

$$p=0 \quad n$$

$$p=1$$

$$\begin{array}{c} 0000 \dots 0 \\ \leftarrow n \rightarrow \\ 11 \dots 1 \end{array}$$

$$k=0$$

$$k=0$$

X^n : ENC, DEC, know #1's = 1, n

$$n = 100,000$$

ENC : location of the 1.

$$k = \lceil \log_2 100,000 \rceil = \lceil \log_2 n \rceil \ll n$$

ASSUME: $P_n(\text{\#1's in } X^n > 1) \leq \frac{1}{n}$

$$\frac{k}{n} = \frac{\lceil \log_2 n \rceil}{n}$$

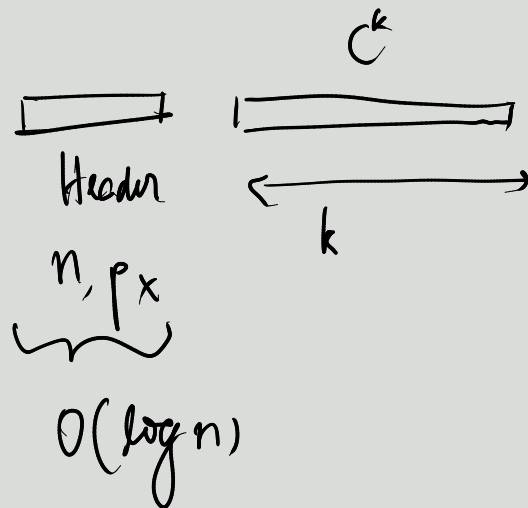
$$\xrightarrow{n \rightarrow \infty} 0$$

$P_e = \sum \frac{1}{n} \rightarrow 0$ as $n \rightarrow \infty$

$$p \neq 0 \quad k \neq 0 \quad p_c \neq 0$$

Assumptions: (n, p) known

Compressed file:



Can be ignored if $n \gg 1$.

$$a \rightarrow 0$$

$$b \rightarrow 10$$

$$c \rightarrow 110$$

$$d \rightarrow 1100$$

$$e \rightarrow 1110$$

$$\lceil \log_2 5 \rceil \approx 3 \text{ bits}$$

$$2^8$$

..

$X^n \sim \text{iid Ber}(p)$

1's $\approx np$

$$\Pr[|\#1's \text{ in } X^n - np| > n\varepsilon] \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

$$\Pr[|(\text{fraction of 1's in } X^n) - p| > \varepsilon] \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

(WLLN)

$$2^n - 2^{nH_2(p)}$$

$$\mathcal{S}_{\varepsilon, p^n} = \left\{ x^n \in \{0,1\}^n : np(1-\varepsilon) \leq \mu_1(x^n) \leq np(1+\varepsilon) \right\} \quad (|\mathcal{S}| \approx 2^{nH_2(p)} \ll 2^n)$$

$$\Pr[X^n \in \mathcal{S}] \rightarrow 1 \quad \text{as } n \rightarrow \infty$$

$$\mathcal{N}(1) \quad \quad \quad 000 \dots 0$$

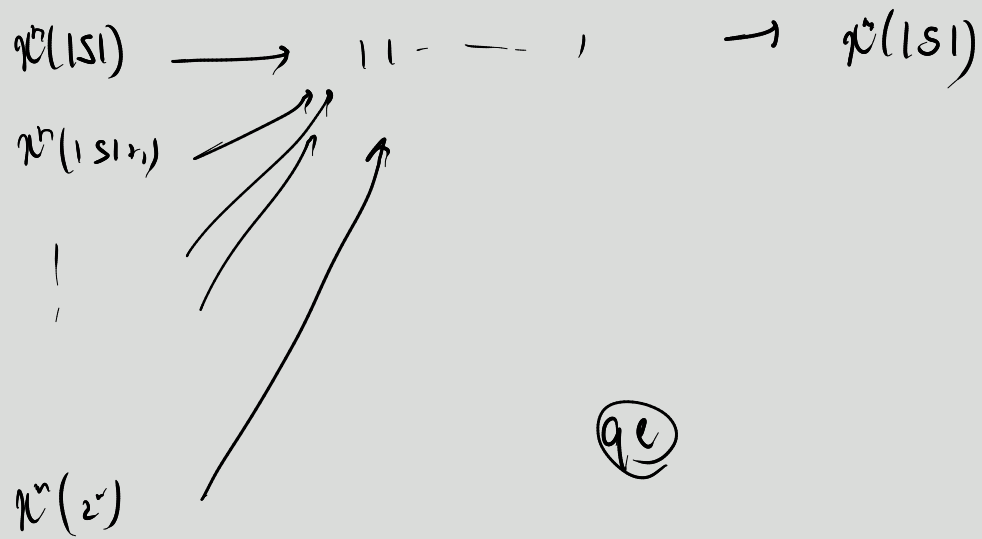
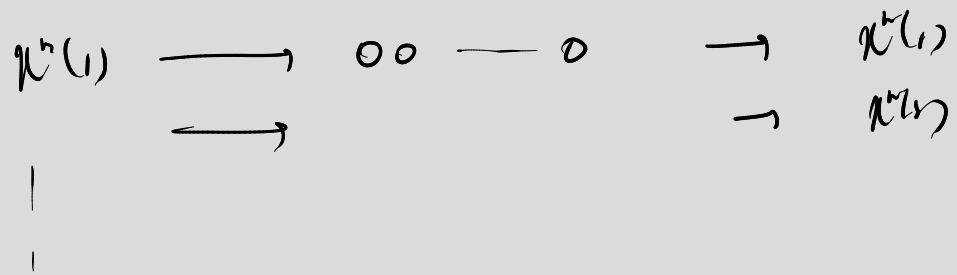
$$\vdots \quad \quad \quad 00 \dots 1$$

\vdots

$$\mathcal{N}(|\mathcal{S}|) \quad \quad \quad 11 \dots 1$$

$$k = \lceil \log_2 |\mathcal{S}| \rceil$$

$$\approx nH(X) = \underline{\underline{nH_2(p)}}$$



9e

an

as

9z