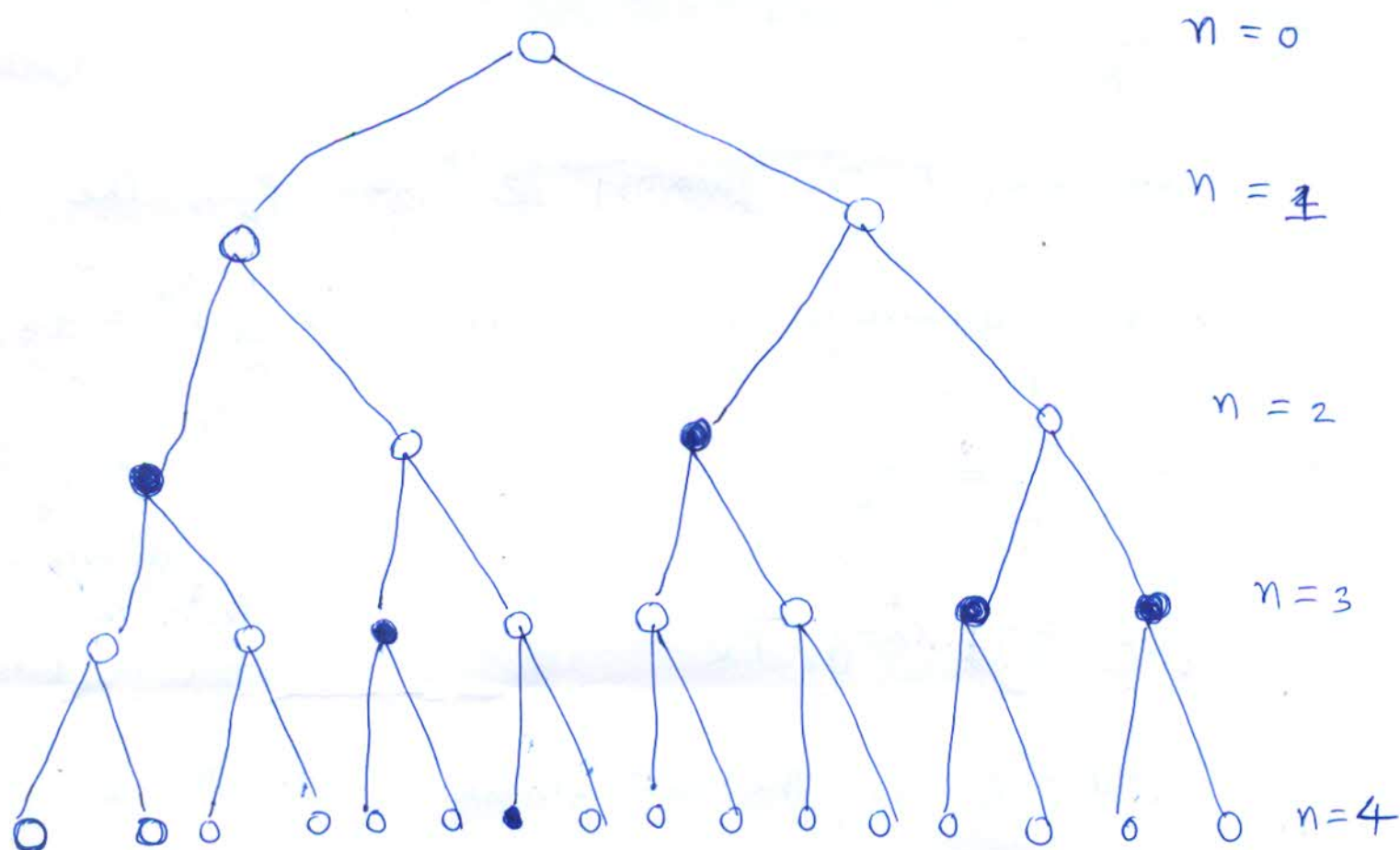


## A Binary Tree Representation of Prefix-free Codes:



### Kraft's Inequality:

→ Let the height of  $k^{\text{th}}$  prefix-free codeword be denoted as  $n_k$ .

→ Let number of leaf nodes that are precluded from consideration as codewords be denoted as  $A_k$ .

→  $A_k = 2^{n-n_k}$ , where  $n$  is the height of the binary tree.

→ Now  $\sum_k A_k \leq 2^n \Rightarrow \sum_k 2^{n-n_k} \leq 2^n$

→ This proves Kraft's Inequality:  $\sum_k 2^{-n_k} \leq 1$

## Kraft's Inequality : An Interpretation :

Let us consider <sup>⑥</sup> a case when inequality is replaced by equal sign.

$$\sum_K 2^{-n_K} = 1$$

One can interpret the quantity  $2^{-n_K}$  as  $P_K$ , the

probability of occurrence of  $K^{\text{th}}$  symbol.

Now, given that

$$2^{-n_K} = P_K$$

$$\begin{aligned} \Rightarrow \sum_K 2^{-n_K} &= 1 \\ \Rightarrow \sum P_K &= 1 \end{aligned}$$

↑ This is as expected.

Probabilities should sum to 1!

$$\Rightarrow n_K = -\log_2 P_K$$

However  $-\log_2 P_K$  is the self-information, of  $K^{\text{th}}$  symbol.

Thus, the length of the codeword ~~should~~ equals its self-information

## ENTROPY & Asymptotic Equipartition Property or AEP

→ Consider a sequence of length  $N$  emitted by a binary iid ergodic source, with "0" probability of  $p$  and "1" probability of  $(1-p)$ .

→ If  $N$  is sufficiently large, this sequence should have approximately  $Np$  "0"s and  $N(1-p)$  "1"s.

→ Since the source is iid, probability of occurrence of each such string is  $p^{NP} \times (1-p)^{N(1-p)} = P_{\text{typ}}$

## AEP and Entropy :

→ How many such strings are There?

→ Max # of strings :  $2^N$

→ Actual # of strings  $M < 2^N$

if  $P \neq 1/2$ .

→ With  $N$  sufficiently large, it is almost guaranteed that all strings will have  $NP$  zeros and  $N(1-P)$  ones.

→ Prob. of each such string is  $P_{\text{typ}}$ .

$$\rightarrow M \times P_{\text{typ}} \approx 1$$

$$\Rightarrow M = \frac{1}{P_{\text{typ}}} = P^{-NP} \times (1-P)^{-N(1-P)}$$

$$\begin{aligned}\Rightarrow \log_2 M &= N(-P \log_2 P - (1-P) \log_2 (1-P)) \\ &= N \times H(X)\end{aligned}$$



1	0	0	0	0
2	0	0	0	1
3	0	0	1	0
4	0	1	0	0
5	1	0	0	0

$m = 5$        $2^N = 16$   
 $\lceil \log_2 m \rceil \approx 3$  bits to encode  
 This source...

16    1    1    1    1

If one takes length  $N$  binary strings, one is likely to see ONLY those strings for which # of zeros is  $N \times p$  and # of ones is  $N \times (1-p)$ , where  $p$  is prob. of zero.

All These strings can be Thought of as belonging to The "Typical set", and

They all occur with equal prob. given

as

$$P_{\text{typ}} = p^{N \times p} \cdot (1-p)^{N(1-p)} = \frac{1}{m}$$

where  $m$  is the size of the typical set.

$$\log_2 m = -\log_2 P_{\text{typ}}$$

$$= N H(x)$$

$$\Rightarrow m = 2^{N H(x)}$$

i.e., Instead of  $2^N$  strings,

We are going to get  $2^{N \times H(x)}$

strings.

i.e., Instead of  $N$  bits, we need  $N \times H(x)$  bits to represent the outcome of the source.

Therefore, average rate  $R$  of the source is  $H(x)$  bits. It cannot become smaller.