Und Extension coits probabilities p, & P2 where P1+P2=1. Second extension of a binary source will have 2=4 symbols given by. 8,8, with probabilities; pipi = Pi 3 Mario Sasionas production of and 200020 P2 P2 = P2 , 1.2. - Norman J. L. L. Merida. Sum of all probabilities of the dad extended some. PitoPilest Pall + Pad = Pit +2pipa + pt = (Pi+12) = 1. Entropy of basicubinary douce;

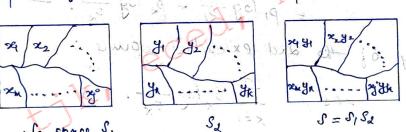
H(s) = > pk log pk P, 608 p + P2 609 P2 Entropy of the and extended source; HCsx) = I Px log PR. = p, 169 px + P1P2 69 pip2 + P1P2 69pp2 + P2 69 px = 2 p, d log pi + 2 pipe log pipe + 2ped log pe. = 2p/ (0g/p+ 2p/2 (0g/p+2p/2 (0g/p+2p) (0g/p) = dp (p+p2) log p + & p2 (p+p2) log p2 = 2[P, 68 P, + P2 68 P2] (H(sd) = 2 H(s). from leg (). Similarly the entropy of 3rd extended source can be $H(s^3) = 3.H(s)$

Source will have an symbols and the entropy of the noth extended source in given by

TOINT ENTROPY AND CONDITIONAL ENTROPY

So far we considered only a single probability scheme is probability belong either to the transmitter on to the receiver. That means we are able to study the behaviour of either transmitter or receiver.

But to study the Behaveour of a communication system, we must simultaneously study the behavior of the transmitter and receiver. This will give the concept of two dimensional peobability scheme.



Sample space, S, sur l'accert sample spaces s, and S, There one two finite discute sample spaces s, and S, and S, and stheir product space will be J=5, S,.

Let $(X) = [x_1, x_2, \dots, x_m]$

Each event ze of S, occur in connection (selation)

with any exterent ye of Sz.

The complete set of events in $S = S_1 S_2$ is (XY).

The three sets of complète probability ochemes:

$$p(xy) = \begin{bmatrix} p(xy, y_k) \\ y_1 \\ y_2 \\ p(x_1, y_k) \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ p(x_1, y_k) \\ p(x_2, y_2) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p(x_2, y_k) \\ p(x_2, y_k) \end{bmatrix} = \begin{bmatrix} p(x_1, y_k) \\ p$$

P(xy) is called "Tout Probability Matrix". Properties of

1) By adding the elements of JPM coloumnavise, we can obtain the peobabilitées peobability of output $\sum_{k=1}^{\infty} b(x^{k}, A^{k}) = b(A^{k}).$

@ By adding the etements of JPM rowwise, we can obtain the probability of input symbols.

$$\sum_{k=1}^{n} p(x_j, y_k) = p(x_j).$$

3 The sum of all the elements of JPM is equal to unity $\sum_{i=1}^{m}\sum_{k=1}^{n}p(x_i,y_k)=1$