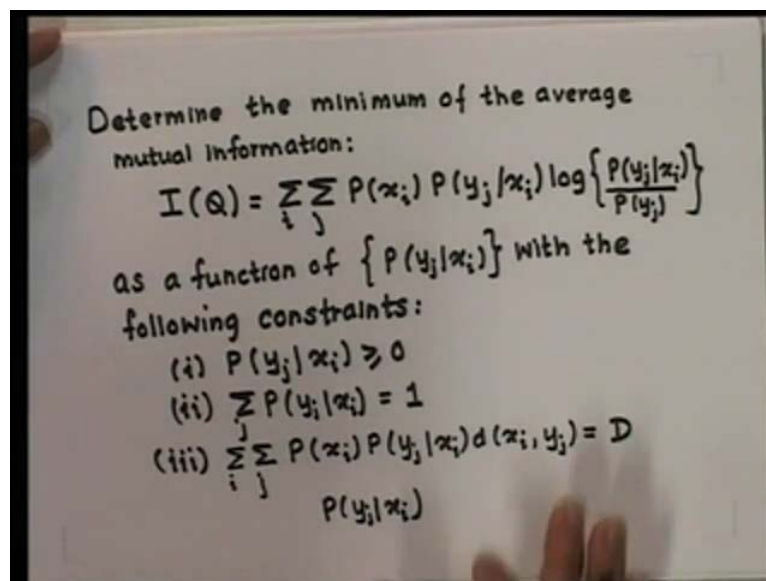


**Information Theory and Coding**  
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**Lecture - 34**  
**Computational Approach for Calculation of Rate-Distortion Functions**

In the earlier classes, we have studied an approach to evaluate the rate distortion function for both discrete and continuous sources. In this approach, essentially we calculated the lower bound for the average mutual information, and then showed that this bound is achievable. In today's class, we will look at the computational approach to evaluate the rate distortion function. We will restrict our discussion to discrete sources, without going into the details of the derivation, we will examine the salient features of this derivation. The determination of the rate distortion function can be regarded as a problem from the calculus of variations. We can formulate the problem as follows.

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Determine the minimum of the average mutual information that is  $I$  of  $Q$ , where based on the earlier notation,  $Q$  is the conditional or transitional matrix is equal to double summation over  $i, j$  of probability of the source multiplied by the conditional probabilities  $\log$  of. So we have to minimize  $I$  of  $Q$  as a function of conditional probabilities with the following constraints.

First conditional probability should be all greater than equal to 0, summation of the conditional probabilities, that is probability of  $y_j$  given  $x_i$  is equal to 1 and the average distortion, which is double summation over  $i, j$  of probability of the source multiplied by conditional probabilities distortion between the source and reconstructing symbols, this is equal to specified distortion level  $D$ . So, if we ignore the constraint one, a solution is possible by differentiating the average mutual information with respect to conditional probabilities that is, probability of  $y_j$  given  $x_i$  and then equating this result to 0. Now, we will not go into the derivation of this, but using Lagrangian's method, we will obtain the following expression for the conditional probabilities.

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The image shows a whiteboard with the following handwritten equations:

$$P(y_j | x_i) = \frac{P(y_j) e^{\lambda d(x_i, y_j)}}{\sum_j P(y_j) e^{\lambda d(x_i, y_j)}}$$

$$\beta(i) = \frac{1}{\sum_j P(y_j) e^{\lambda d(x_i, y_j)}}$$

$$P(y_j | x_i) = \beta(i) P(y_j) e^{\lambda d(x_i, y_j)}$$

$$P(y_j) = \sum_i P(x_i) P(y_j | x_i)$$

Probability of  $y_j$  given  $x_i$  is equal to probability  $y_j$  multiply  $e$  raise to exponential. Now, if we introduce another parameter called  $\beta(i)$ , which is equal to 1 by summation of  $P(y_j) e^{\lambda d(x_i, y_j)}$ , then this expression that is conditional probability can be written as probability of  $y_j$  given  $x_i$  is equal to  $\beta(i)$  times probability  $y_j$ ,  $e$  raise to  $\lambda d(x_i, y_j)$ . So, this procedure yields a set of equations for each  $i$  and  $j$  where probability of  $y_j$  given  $x_i$  is expressed in terms of probabilities of  $y_j$  it therefore, now remains to determine the probability  $p(y_j)$  and in general probability  $p(y_j)$  is equal to summation over  $i$  of  $P(x_i)$  multiplied by conditional probabilities  $P(y_j | x_i)$ .

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$$\sum_i \beta(i) P(x_i) e^{\lambda d(x_i, y_j)} = 1$$

or

$$\sum_i \frac{P(x_i) e^{\lambda d(x_i, y_j)}}{\sum_j P(y_j) e^{\lambda d(x_i, y_j)}} = 1$$

N equations  $\rightarrow$  P(y<sub>j</sub>)  
 $\mathcal{Q}$  R(D) and D(Q)

So, if we divide this equation by probability of y<sub>j</sub> and use this relationship, we get the following relationship is equal to 1 or same thing can be rewritten as summation over i of P x i, e raise to lambda d x i, y j. Now, it must still be verified, if constraint one is satisfied. So, in this manner we get n equations assuming that the size of the reconstruction alphabet is N for the N probabilities P y j, which now may be solved and after this the conditional or transitional probability matrix q can be determined. So, with the equation so obtained it is now possible to obtain an expression for R D and D Q as follows.

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$$D(Q) = \sum_i \sum_j P(x_i) P(y_j | x_i) d(x_i, y_j)$$

$$= \sum_i \sum_j \beta(i) P(x_i) P(y_j) e^{\lambda d(x_i, y_j)} d(x_i, y_j)$$

Since I(Q) has been minimised

$$R(D) = I(Q)$$

$$= \sum_i \sum_j P(x_i) P(y_j | x_i) \log \left\{ \frac{P(y_j | x_i)}{P(y_j)} \right\}$$

From  $\frac{P(y_j | x_i)}{P(y_j)} = \beta(i) e^{\lambda d(x_i, y_j)}$

We know that the distortion as a function of conditional probability matrix  $Q$  is equal to double summation  $i, j$  probability  $x_i$  multiplied by conditional probability, multiplied by distortion between  $x_i, y_j$ . This can be rewritten based on the previous results as summation over  $i, j$   $\beta_i, P(x_i), P(y_j), e^{\lambda d(x_i, y_j)}, d(x_i, y_j)$ . Now, since  $I(Q)$  has been minimized, we can write the rate distortion function  $R(D)$  equal to  $I(Q)$ , which is equal to double summation over  $i, j$  of the quantity  $P(x_i), P(y_j)$  given  $x_i$  log of  $P(y_j)$  given  $x_i$  divided by probability  $y_j$ . From the relationship, which we have derived from  $P(y_j)$  given  $x_i$ , over  $P(y_j)$  equal to  $\beta_i, e^{\lambda d(x_i, y_j)}$ . We can write the rate distortion function  $R(D)$  as follows.

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$$R(D) = \sum_i \sum_j P(x_i) P(y_j|x_i) \log \{ \beta_i e^{\lambda d(x_i, y_j)} \}$$

$$= \lambda D \log e + \sum_i P(x_i) \log \beta_i$$

$\lambda \rightarrow R(D)$  w.r.t.  $D$

$$\lambda = \frac{dR(D)}{dD} \times \frac{1}{\log e}$$

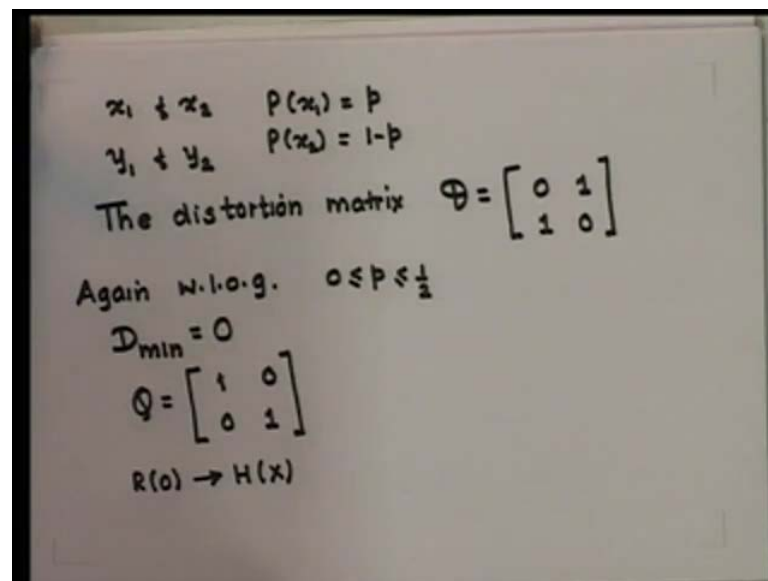
$0 \leq D \leq D_{\max} \cdot \lambda$  is not positive

$R(D)$  is equal to double summation of  $i, j, P(x_i), P(y_j)$  given  $x_i$  log of  $\beta_i, e^{\lambda d(x_i, y_j)}$  and this can be further simplified to get the following result. So, we have now found an expression for the rate of the source as a function of the permissible distortion  $D$ . Now, the solution is found in implicit form, why the parameter  $\lambda$ ? And explicit relation is not possible excepting in a few simple cases. Now, a value of  $\lambda$  yields a value for  $D$  and for  $R(D)$  and thus gives a point on the  $R(D)$  curve.

Now, it can be shown that the parameter  $\lambda$  is proportional to the first derivative of  $R(D)$  with respect to  $D$  and is therefore, connected to the slope of the  $R(D)$  curve at a certain point. So, we have the relationship as  $\lambda$  is equal to derivative of  $R(D)$  with respect to  $D$  with a constant multiplying factor  $1/\log e$ . Therefore, the  $R(D)$  is a

continuous monotone decreasing function for  $D$  greater than equal to 0 and less than equal to  $D_{\max}$ . The parameter  $\lambda$  is continuous for  $D$  greater than 0, less than  $D_{\max}$  and is not positive. In order to get a better understanding of the concepts of the computational approach, let us apply this approach to evaluate the rate distortion function for a binary source. Earlier we have already derived the rate distortion function for a binary case; we will redo the problem using the computational approach.

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$x_1 \text{ \& } x_2 \quad P(x_1) = p$   
 $y_1 \text{ \& } y_2 \quad P(x_2) = 1-p$   
 The distortion matrix  $D = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$   
 Again w.l.o.g.  $0 \leq p \leq \frac{1}{2}$   
 $D_{\min} = 0$   
 $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$   
 $R(0) \rightarrow H(x)$

So, let us assume that the source generates symbols  $x_1$  and  $x_2$  with the probabilities  $P(x_1)$  equal to small  $p$  and  $P(x_2)$  equal to 1 minus  $p$ . Let us assume that the reproducing symbols are  $y_1$  and  $y_2$ , again using the earlier notation. Let us assume that the distortion matrix is given by 0 1, 1 0, which means that if  $x_1$  is reproduced as  $y_1$ , there is no distortion, but if  $x_1$  is reproduced as  $y_2$  there is a distortion 1 and similarly, for  $x_2$ .

Now, again without the loss of generality, we assume that  $p$  is greater than 0 and less than equal to half. So, in this case it is easy to see that,  $D_{\min}$  is equal to 0, that is the minimum distortion is equal to 0 and this occurs, if we choose for the conditional probability matrix  $Q$  equal to 1 0 0 1. The  $R(D)$  function  $R(0)$  is equal to  $H(x)$  in this case.

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$$\begin{aligned} R(0) &= \min_{Q \in Q_0} I(Q) \\ &= \min_{Q \in Q_0} [H(X) - H(X|Y)] \\ &= H(X) \\ &= -p \log p - (1-p) \log(1-p) \end{aligned}$$

Now, this is not difficult to show  $R(0)$  is equal to minimum  $I(Q)$ ,  $Q$  belonging to  $Q_D$  and if  $D$  is equal to 0 this is equal to minimum  $Q$  belonging to  $Q_D$  of  $H(X)$  minus  $H(X|Y)$ . Since, the equivocation is 0 this is equal to  $H(X)$ , which is equal to  $-p \log p - (1-p) \log(1-p)$ .

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$$\begin{aligned} D &= p & I(Q) &= 0 \\ D &= 1-p \\ Q &= \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \\ R(D) &= P(y_j) \end{aligned}$$

Now, the maximum distortion is  $D$  equal to  $p$  now, this is the best choice that can be made if  $I(Q)$  is equal to 0, because the alternative  $D$  equal to  $1-p$ , always yields a larger distortions, since  $p$  is less than equal to half. So, the matrix  $Q$  of the conditional

probabilities, in this case is equal to 0 1 0 1. Now, in order to determine the other points of the R D curve, the probabilities  $P_{y_j}$  must first be found.

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The image shows a whiteboard with the following handwritten text:

$$\sum_i \beta(i) P(x_i) e^{\lambda d(x_i; y_j)} = 1$$

Let  $e^\lambda = a$ . It follows that

$$\left. \begin{aligned} \beta(1) \cdot p + \beta(2) \cdot (1-p) \cdot a &= 1 \\ \beta(1) \cdot p \cdot a + \beta(2) \cdot (1-p) &= 1 \end{aligned} \right\}$$

$$\beta(1) = \frac{1}{p(1+a)} \quad \beta(2) = \frac{1}{(1-p)(1+a)}$$

So, we have seen that summation over  $i$  of  $\beta(i) P(x_i) e^{\lambda d(x_i; y_j)}$  is equal to 1. Now, let us assume  $e^\lambda = a$ , then it follows that  $\beta(1) \cdot p + \beta(2) \cdot (1-p) \cdot a = 1$  and  $\beta(1) \cdot p \cdot a + \beta(2) \cdot (1-p) = 1$ . If these two equations are solved simultaneously, we get  $\beta(1) = \frac{1}{p(1+a)}$  and  $\beta(2) = \frac{1}{(1-p)(1+a)}$ . Next, we determine probability  $y_j$  from  $\beta(i)$  now, this can be done on the basis of the relation  $\beta(i) = \frac{1}{\sum_j P(y_j) e^{\lambda d(x_i; y_j)}}$ . So, using this relationship we find,  $P(y_1) \cdot p(1+a) + P(y_2) \cdot (1-p)(1+a) = 1$ , which is equal to  $p(1+a) + (1-p)(1+a) = 1$ .

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We determine  $P(y_j)$  from  $\beta(i)$ .

$$\beta(i) = \frac{1}{\sum_j P(y_j) e^{\lambda d(x_i, y_j)}}$$

We find,

$$P(y_1) + a P(y_2) = \frac{1}{\beta(2)} = p(1+a)$$

$$a P(y_1) + P(y_2) = \frac{1}{\beta(2)} = (1-p)(1+a)$$

Another relationship which we get as follows, a times probability y 1 plus probability y 2 is equal to 1 by beta 2 is equal to 1 minus p times 1 plus a. If this two equations are solved we get probability of y 1 is equal to p minus a times 1 minus p divided by 1 minus a, and probability of y 2 is equal to 1 minus p minus a times p divided by 1 minus a.

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$$P(y_1) = \frac{p - a(1-p)}{(1-a)}$$

$$P(y_2) = \frac{1-p-ap}{(1-a)}$$

$$D = \sum_i \sum_j \beta(i) P(x_i) P(y_j) e^{\lambda d(x_i, y_j)}$$

$$= \beta(1) P(x_1) P(y_2) e^{\lambda} + \beta(2) P(x_2) P(y_1) e^{\lambda}$$

$$= \frac{a}{(1+a)}$$

Now, if we use this result and substitute in the expression for the permissible distortion, we get the result as follow D is equal to double summation over i j of beta i, P x i, P y j, e raise to lambda d x i y j times d x i y j and this can be simplified as beta 1 time P x 1



times  $P y_2$ ,  $e$  raise to  $\lambda + \beta_2$ ,  $p \times 2$ ,  $p y_1$   $e$  raise to  $\lambda$ . If we substitute the values for  $p y_1$ ,  $p y_2$   $\beta_1 \beta_2$ ,  $p \times 1$ ,  $p \times 2$ , we can show that this simplifies to a by 1 plus a.

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The image shows a whiteboard with the following handwritten mathematical steps:

$$R(D) = \lambda D \log e + \sum_i P(x_i) \log \beta(i)$$

$$\because a = \frac{D}{1-D}$$

$$\therefore \lambda = \frac{\log_2 a}{\log_2 e}$$

$$R(D) = D \log \left( \frac{D}{1-D} \right) + P(x_1) \log \beta(1) + P(x_2) \log \beta(2)$$

$$= D \log \left( \frac{D}{1-D} \right) + p \log \left[ \frac{1}{p(1+a)} \right] + (1-p) \log \left[ \frac{1}{(1-p)(1+ta)} \right]$$

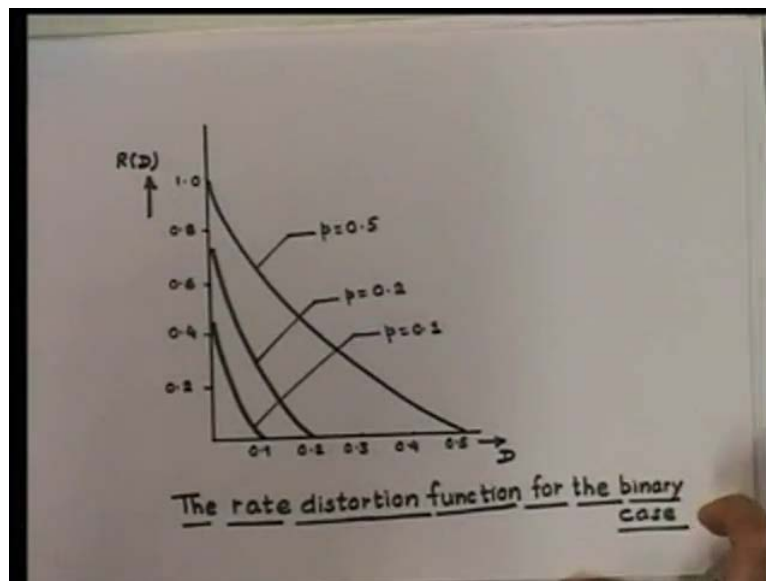
Now, we know that  $R D$  is equal to  $\lambda$  times  $D$  multiplied by  $\log$  to the base 2 of  $e$  plus summation of  $P \times i$  times  $\log \beta(i)$ . Now, because  $a$  is equal to  $D$  by  $1$  minus  $D$ , from this relationship and  $\lambda$  is equal to  $\log_2 a$  divided by  $\log_2 e$   $R D$  can be rewritten as follows, is equal to  $D$  times  $\log$  of  $D$   $1$  minus  $D$ , plus  $P \times 1$ ,  $\log \beta(1)$  plus,  $p \times 2 \log \beta(2)$ . This can be further simplified as  $D \log$  of  $D$   $1$  minus  $D$  plus  $p$  times  $\log$  of  $1$  by  $p$  times  $1$  plus  $a$  plus,  $1$  minus  $p$  times  $\log$  of  $1$  by  $1$  minus  $p$  times  $1$  plus  $a$ .

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A hand-drawn equation on a whiteboard. The equation is written in black marker and shows the derivation of the rate distortion function. It starts with  $R(D) = p \log \frac{1}{p} + (1-p) \log \frac{1}{(1-p)} + D \log D + (1-D) \log (1-D)$  and then simplifies to  $= H(X) - H(D)$ . A hand holding a green marker is visible at the bottom of the frame.

Now, it is not very difficult to show that this can be reduced to the following expression,  $R(D)$  is equal to  $p \log \frac{1}{p}$ , plus  $(1-p) \log \frac{1}{(1-p)}$ , plus  $D \log D$ , plus  $(1-D) \log (1-D)$  and this is equal to  $H(X) - H(D)$ . So, we have thus found a relation from which we can directly determine the mutual information, for a given value of  $D$  which must be conveyed in order to achieve an average distortion  $D$ .

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So, this figure provides the rate distortion curve for a few values of  $p$ , from the figure it can be seen that a smaller average distortion can only be achieved by increasing the rate.

The rate distortion function  $R(D)$  for  $p = 0.5$  is larger than for  $p < 0.5$  for every value of  $D$ , which is again intuitively true. So, each point on the curve is reached by a matrix of conditional or transitional probabilities, which give rise to both an average distortion  $D_Q = D$  and average mutual information  $R(D)$ .

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Handwritten mathematical derivation on a whiteboard:

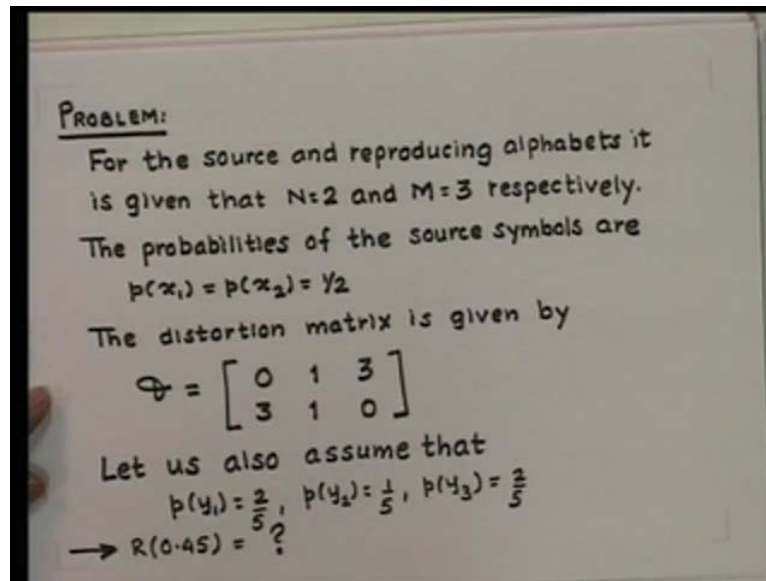
$$P(y_i | x_i) = \beta(x_i) P(y_j) e^{\lambda d(x_i, y_j)}$$

$$Q = \begin{bmatrix} \frac{a(p-1)+p}{p(1-a^2)} & -\frac{a(p-1)-a^2p}{p(1-a^2)} \\ \frac{ap-a^2(1-p)}{(1-p)(1-a^2)} & \frac{(1-p)-ap}{(1-p)(1-a^2)} \end{bmatrix}$$

$$P(y_i | x_i) = \beta(x_i) P(y_i) = \frac{1}{p(1+a)} \cdot \frac{p-a(1-p)}{(1-a)} = \frac{a(p-1)+p}{p(1-a^2)}$$

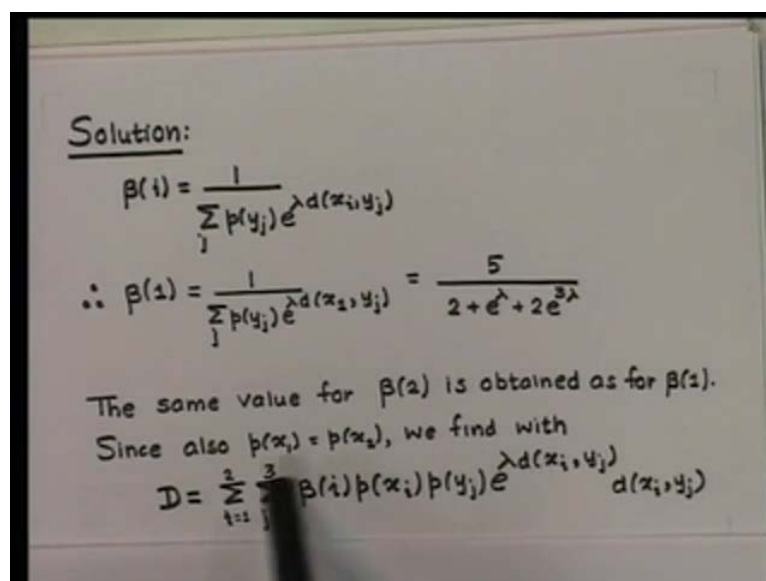
Now, for the conditional or transition probabilities, we have probability of  $y_j$  given  $x_i$  is equal to  $\beta(x_i)$  times probability  $P(y_j)$ ,  $e^{\lambda d(x_i, y_j)}$  and this yields the following matrix  $Q$  and finally,  $\frac{1-p-ap}{(1-p)(1-a^2)}$ . Now, the procedure to evaluate any of this term is very simple for example, probability of  $y_1$  given  $x_1$ , that is this quantity is equal to  $\beta(x_1)$  times  $P(y_1)$ . This is equal to  $\frac{1}{p(1+a)}$  multiplied by  $p - a(1-p)$  over  $(1-a)$ , and this simplifies to  $\frac{a(p-1)+p}{p(1-a^2)}$ . So similarly, other terms can be evaluated. Now, let us extend the application of the computational approach to a discrete source which is not a binary.

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So, let us take another problem, where we have a source and reproducing alphabet. The source alphabet size is 2 and reproducing alphabet size is equal to 3. The probabilities of the source symbols are given as follows, probability of  $x_1$  is equal to probability of  $x_2$  is equal to half and the distortion matrix is given as shown here. Let us also assume that, the output or the reproducing symbols probabilities are specified, this is to simplify the problem. So,  $p$  of  $y_1$  is equal to  $\frac{2}{5}$ ,  $p$  of  $y_2$  is equal to  $\frac{1}{5}$  and  $p$  of  $y_3$  is equal to  $\frac{2}{5}$  and the problem is to calculate the rate distortion function for  $D$  equal to 0.45.

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So, the solution is as follows, we know that beta i is given by this expression therefore, beta 1 can be rewritten as based on the fact that probability of P y j has been specified and distortion between x 1 and y j has been also specified. Now, the same value for beta 2 is obtained as for beta 1 since also p x 1 is equal to p x 2, we find that the distortion D which is given by this expression can be rewritten as follows.

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The image shows a handwritten derivation of the rate distortion function. The equations are as follows:

$$D = \beta(2) p(x_1) \left\{ \sum_{j=1}^3 p(y_j) e^{\lambda d(x_1, y_j)} d(x_1, y_j) + \sum_{j=2}^3 p(y_j) e^{\lambda d(x_2, y_j)} d(x_2, y_j) \right\}$$

$$= \frac{\lambda/2}{2 + e^\lambda + 2e^{3\lambda}} \left\{ 2 \left( \frac{1}{5} e^\lambda + \frac{6}{5} e^{3\lambda} \right) \right\}$$

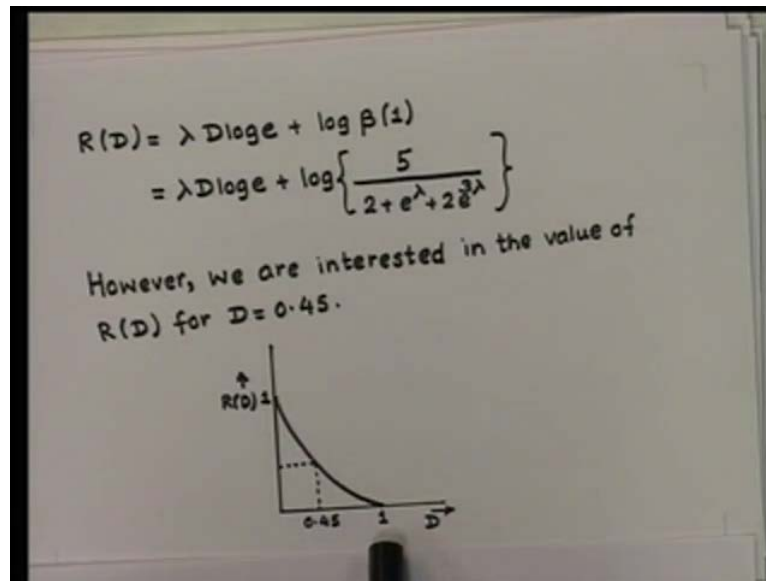
$$= \frac{\lambda}{2 + e^\lambda + 2e^{3\lambda}} \left( e^\lambda + 3e^{3\lambda} \right)$$

The rate distortion function as function of D and  $\lambda$  becomes with the help of equation

$$R(D) = \lambda D \log e + \sum p(x_i) \log \beta(i)$$

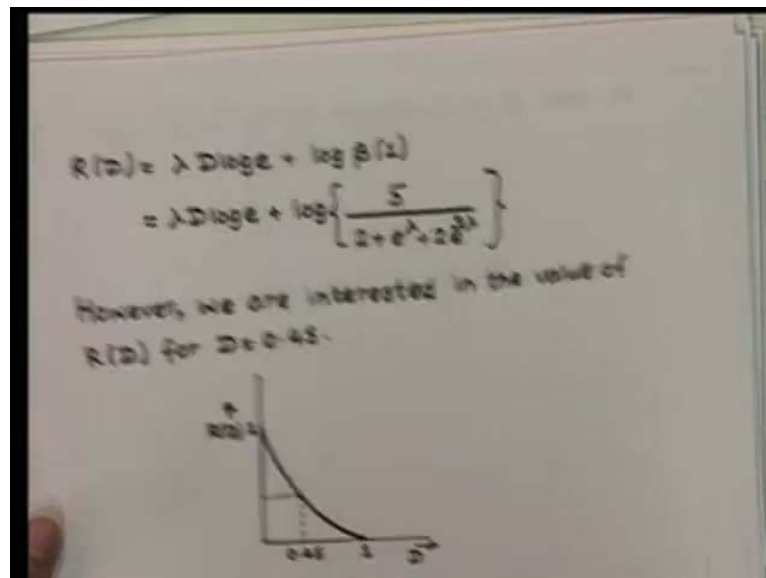
D is equal to beta 1 times probability of x 1 multiplied by this two terms and if we substitute the values for this terms, we get the following expression. So, the rate distortion function, as function of D and lambda becomes with the help of the equation R D equal to lambda D log e plus summation of p x i log b i as follows.

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$R(D)$  is equal to  $\lambda D \log e$  plus  $\log \beta(1)$ ; because  $\beta(1)$  and  $\beta(2)$  are same and this can be simplified to this expression. Now, however we are interested in the value of  $R(D)$  for  $D$  equal to 0.45 the rate distortion curve would be as follows, the for  $D$  equal to 0,  $R(D)$  is equal to 1 and for  $R(D)$  equal to 0 maximum distortion is equal to 1.

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So, if we use  $x$  equal to  $e^\lambda$ , in the expression for  $D$ , this leads to  $D$  is equal to this quantity and we can solve this quantity as follows.

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Using  $x = e^\lambda$  in the expression for  $D$  leads to

$$D = \frac{x + 3x^3}{2 + x + 2x^3}$$

and thus

$$x + 3x^3 = 0.45(2 + x + 2x^3),$$

or

$$2.1x^3 + 0.55x - 0.9 = 0$$

The solution of this equation is  $x \cong 0.64$ .

$\therefore \lambda = \ln 0.64 \cong -0.45$

Now the value of  $R(0.45)$  can be found.

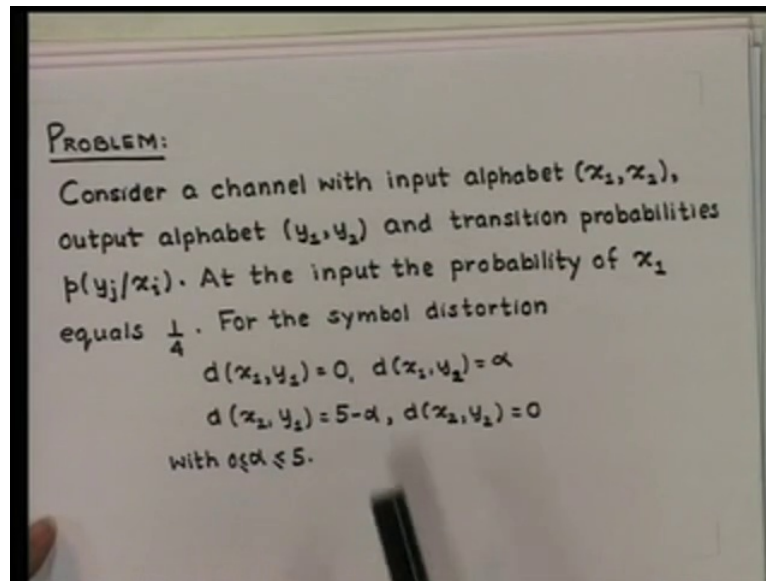
The solution of this equation is,  $x$  is approximately equal to 0.64, therefore  $\log$  is logarithmic of this term, which is equal to minus 0.45. Now, the value of rate distortion function for  $D$  equal to 0.45 can be found as follows.

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$$R(D) = \lambda D \log e + \log \left\{ \frac{5}{2 + e^\lambda + 2e^{3\lambda}} \right\}$$
$$= -0.45 \times 0.45 \times \log e + \log \left\{ \frac{5}{2 + 0.64 + 2(0.64)^3} \right\}$$
$$= \frac{1}{\ln 2} (-0.20 + 0.46)$$
$$= \frac{0.26}{\ln 2}$$
$$\cong 0.37$$
$$\Rightarrow R(0.45) \cong 0.37$$

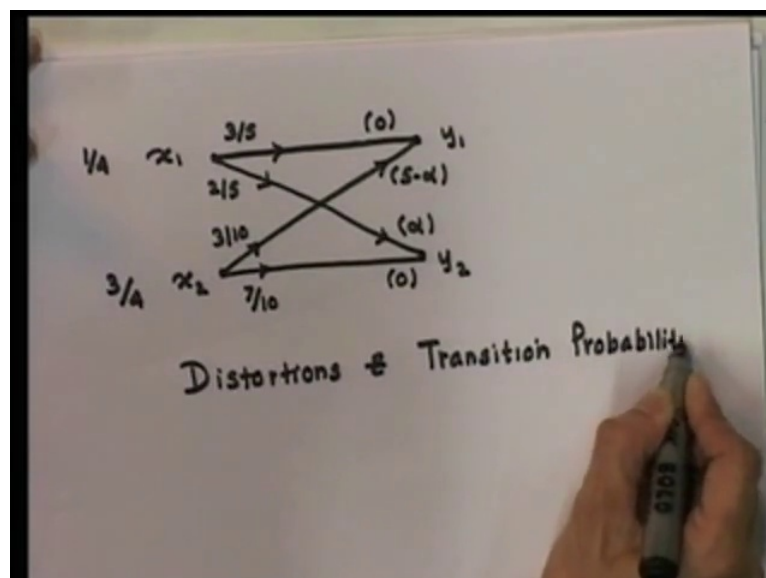
RD is equal to this expression, substitute the values for  $\lambda$  and  $D$ , we can show that this reduces to 0.37. So, this shows how to apply the computational approach to evaluate the rate distortion function in the example discussed. Let us take one more example to understand the concepts of evaluation of the rate distortion function.

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So, let us consider a channel with input alphabet given by  $x_1, x_2$ , output alphabet  $y_1, y_2$  and transition probabilities. Probability  $y_j$  given  $x_i$  at the input the probability of  $x_1$  equals one-fourth. For the symbol distortion we assume that distortion between  $x_1, y_1$  is equal to 0 that between  $x_1$  and  $y_2$  is given by a variable factor  $\alpha$  and between  $x_2, y_1$  is equal to 5 minus  $\alpha$  and distortion between  $x_2, y_2$  is equal to 0, where we assume that  $\alpha$  is greater than equal to 0 and less than equal to 5.

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So, distortions and transition probabilities between the inputs and the reproducing symbols can be denoted by this diagram,  $x_1 \times 2 \ y_1 \ y_2$ . Probability of this is one-fourth probability of  $x_2$  is three-fourth probability of  $y_1$  given  $x_1$  is 3 by 5. This would be 2 by 5 this is 3 by 10 and this is 7 by 10 and distortion between  $x_1$  and  $y_1$  is denoted in brackets this 0. This is 5 minus alpha between  $x_1$  and  $y_2$  is alpha and that between  $x_2$  and  $y_2$  is 0.

(Refer Slide Time: 43:15)

(a) Calculate the amount of information at the output of the channel.  
 Assume  $p(y_1/x_1) = \frac{3}{5}$  and  $p(y_1/x_2) = \frac{3}{10}$

Solution: For the marginal probabilities  $p(y_1)$  and  $p(y_2)$  we find

$$p(y_1) = \sum_{i=1}^2 p(x_i) p(y_i/x_i) = \frac{1}{4} \times \frac{3}{5} + \frac{3}{4} \times \frac{3}{10} = \frac{3}{8}$$

$$p(y_2) = 1 - \frac{3}{8} = \frac{5}{8}$$

Now, the amount of information  $H(Y)$  becomes

$$H(Y) = -\frac{3}{8} \log \frac{3}{8} - \frac{5}{8} \log \frac{5}{8} = 0.954 \text{ bit.}$$

So, this is distortions and transition probabilities, problem is calculate the amount of information at the output of the channel assuming, that probability of  $y_1$  given  $x_1$  is 3 by 5 and probability of  $y_1$  given  $x_2$  is equal to 3 by 10 and this is very simple. So, for the marginal probabilities  $p y_1$ ,  $p y_2$ , we find using this expression is equal to 3 by 8, 5 by 8. So, the amount of information  $H Y$  becomes 0.954 bit.

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(b). Calculate the average distortion as a function of  $\alpha$ . What is the smallest average distortion obtainable?  
Assume  $p(y_1/x_1) = \frac{3}{5}$  and  $p(y_1/x_2) = \frac{3}{10}$

Solution:  
The average distortion can be found easily:  
$$D(Q) = \sum p(x_i) p(y_j/x_i) d(x_i, y_j)$$
$$= \frac{1}{4} \cdot \frac{3}{5} \cdot \alpha + \frac{3}{4} \cdot \frac{3}{10} \cdot (5-\alpha) = -\frac{1}{8}\alpha + \frac{9}{8}$$
The average distortion has a minimum for  $\alpha = 5$   
$$\min_{\alpha} D(Q) = \frac{1}{2}$$

The next problem is, calculate the average distortion as a function of alpha and what is the smallest average distortion obtainable? Again assume conditional probabilities as specified here. The average distortion can be found easily  $D(Q)$  is equal to this expression and when we plug in the values, we get  $D$  is equal to function of alpha and the average distortion has a minimum value for alpha equal to 5, so minimum  $D(Q)$  is equal to half.

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(c) Calculate  $R(0)$  and give the corresponding channel matrix.

Solution If  $D=0$   
$$H(x) = -\frac{1}{4} \log \frac{1}{4} - \frac{3}{4} \log \frac{3}{4}$$
$$= 0.811$$
$$Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Next is calculate  $R(0)$  and give the corresponding channel matrix. Now, the solution to this is, if  $D$  is equal to 0, the rate distortion function achieves a maximum value equal to

H X, which is the information of the source. In this case, this is equal to minus one-fourth, log one-fourth minus three-fourth, log three-fourth is equal to 0.811 and the corresponding channel matrix Q is equal to 1 0 0 1. That means there is a 1 to 1 correspondence between input and output symbol.

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[d] Calculate  $D_{\max}$ , the smallest distortion possible when no information from the source is received. What value of  $\alpha$  will obtain the largest  $D_{\max}$ ?

Solution:

$$D_{\max} = \min_j \sum_{i=1}^2 p(x_i) d(x_i, y_j)$$

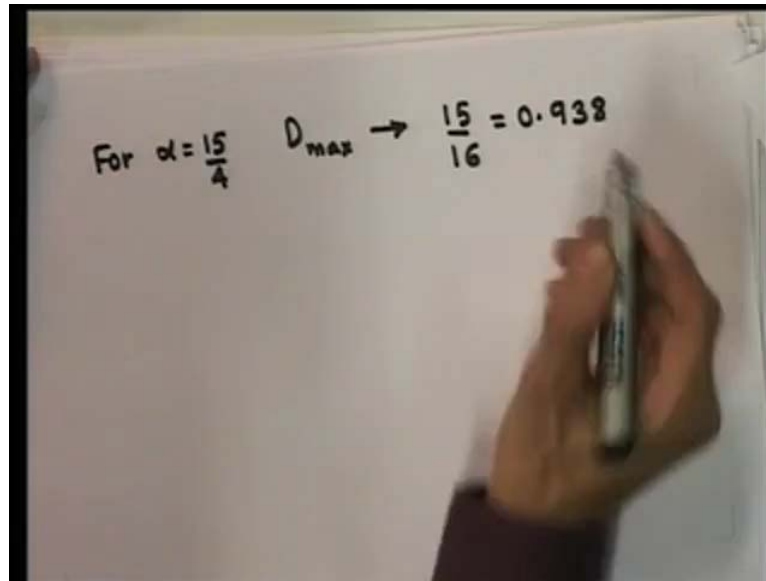
$$= \min \left( \frac{1}{4} \alpha, \frac{3}{4} (5 - \alpha) \right)$$

$$D_{\max} = \frac{1}{4} \alpha \quad \text{for } 0 \leq \alpha \leq \frac{15}{4}$$

$$= \frac{3}{4} (5 - \alpha) \quad \text{for } \frac{15}{4} \leq \alpha \leq 5$$

Finally, let us calculate  $D_{\max}$ , which is the smallest distortion possible, when no information from the source is received and what value of alpha will obtain the largest  $D_{\max}$ . Now, the solution again is as follows, on the basis of the theory we have discussed  $D_{\max}$  is equal to minimum over j of summation i equal to 1 to two,  $P x_i, d x_i y_j$  and in the present case we get this, minimum of  $\frac{1}{4} \alpha$   $\frac{3}{4} (5 - \alpha)$  and is equal to  $\frac{1}{4} \alpha$ , for alpha between 0 and  $\frac{15}{4}$  and is equal to  $\frac{3}{4} (5 - \alpha)$  for alpha between  $\frac{15}{4}$  and 5.

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For  $\alpha = \frac{15}{4}$   $D_{\max} \rightarrow \frac{15}{16} = 0.938$

For alpha equal to 15 by 4  $D_{\max}$  will achieve its absolute maximum, which is 15 by 16, is equal to 0.938. So, after having examined some of these problems, we can come to the following conclusion, that determining the rate distortion function is generally not easy. And because of this a lower limit is often used for calculating the rate distortion function, numerical techniques also do exist for calculation of rate distortion function. Over next few classes, we will examine how the concepts of rate distortion function can be utilised to design efficient lossy compression schemes. In the next class, we will begin this study with contestation of a continuous source in finite discrete levels.