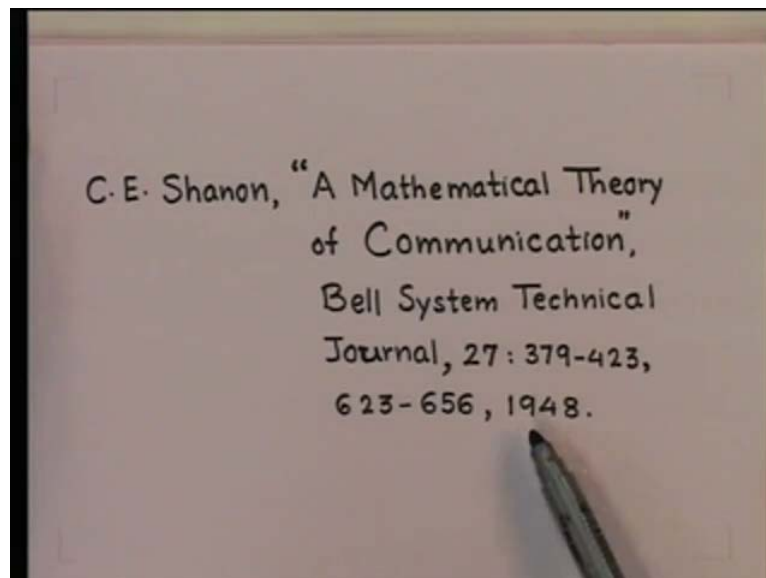


Information Theory and Coding
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Lecture - 33
Calculation of Rate - Distortion Functions

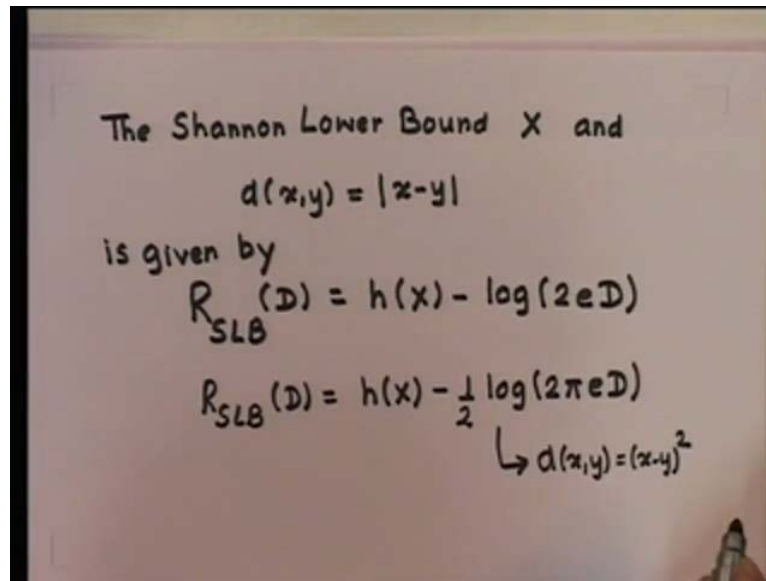
Based on the mathematical definition of the rate distortion function, we have shown that like the differential entropy for a Gaussian source. The rate distortion function for a Gaussian source also has a distinction of being larger than the rate distortion function, for any other source with continuous distribution and the same variance. Now, this is a very important result, because for many sources it can be very difficult to calculate the rate distortion functions. In such situations it is helpful to have an upper bound for the rate distortion function. Now, it will be very useful if you could also find the lower bound for the rate distortion function. Shannon, in his 1948 paper has described this lower bound and appropriately called Shannon's lower bound.

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The references, C. E. Shannon, a Mathematical Theory of Communication, Bell system technical journal, volume 27. There were 2 parts and published in 1948. We will state the lower bound here without derivation, if 1 is interested in detailed derivation then it could be referred to this reference. T. Berger, rate distortion theory a mathematical basis for data compression, published by Prentice Hall in 1971.

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The Shannon Lower Bound X and
 $d(x,y) = |x-y|$
is given by
 $R_{SLB}(D) = h(x) - \log(2eD)$
 $R_{SLB}(D) = h(x) - \frac{1}{2} \log(2\pi eD)$
 $\hookrightarrow d(x,y) = (x-y)^2$

So, the Shannon's lower bound for a random variable X and the magnitude error criterion, that is $D \times y$ equal to the mod of the difference between x and y is given by Shannon lower bound is equal to differential entropy minus \log of $2eD$, where D denotes the specified distortion level. Now, if we use this squared error criteria then Shannon's lower bound is given by differential entropy minus half $\log 2\pi eD$.

So, this result is for squared error criterion that is $D \times y$ is equal to x minus y squared. So, to summarize we have provided a mathematical definition for the rate distortion function and also calculated the same for 2 important sources, 1 binary source and the other was the Gaussian source. We have also obtained the upper bounds and lower bounds for any arbitrary independent identically distributed source. Now, these functions and results are specially useful when we want to know if it is possible to design compression schemes to provide the specified rate, and distortion for a given source.

These results are also useful in determining the amount of performance improvement that we could obtain by designing better compression schemes. Therefore, the rate distortion function plays the same role for lossy compression scheme that entropy plays for lossless compression. Now, in the last class we had discussed 2 approaches to calculate the rate distortion function for a source.

One approach which we utilize to calculate the rate distortion function for the binary source and the Gaussian source was to calculate the lower bound, for the average mutual

information and then show that this bound is achievable. The other approach is based on a computational approach. Now, we will not go into the detail derivation of this approach, but we will present the salient features of this approach. The algorithm by itself is very simple. Now, in order to do that we will use simplified notations as follows.

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The image shows handwritten mathematical notation on a slide. It defines a set of conditional probabilities $\{P(y_j|x_i)\}$ as a matrix Q . The source alphabet X is defined as $\{x_1, x_2, \dots, x_N\}$ and the reconstruction alphabet Y is defined as $\{y_1, y_2, \dots, y_M\}$. The matrix Q is then defined as a matrix where the element in the i -th row and j -th column is $P(y_j|x_i)$. The matrix is written as:

$$Q \triangleq \begin{bmatrix} P(y_1|x_1) & P(y_2|x_1) & \dots & P(y_M|x_1) \\ P(y_1|x_2) & P(y_2|x_2) & \dots & P(y_M|x_2) \\ \vdots & \vdots & \ddots & \vdots \\ P(y_1|x_N) & P(y_2|x_N) & \dots & P(y_M|x_N) \end{bmatrix}$$

Conditional probabilities that is probability of y_j given x_i by a matrix Q , where we assume that the source alphabet is of size N , that is there are N elements and reconstruction alphabet is of size M . So, in this case Q matrix by definition is given by probability of y_1 given x_1 , probability of y_2 given x_1 to probability of y_M given x_1 , probability of y_1 given x_2 , probability of y_2 given x_2 , probability of y_M given x_2 . And finally, the last row is probability of y_1 given x_N , probability of y_2 given x_N and the last term is probability of y_M given x_N . Now, using this notation, we can write the average distortion for a given source and specified distortion measure as the function of the matrix Q as follows.

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$$\begin{aligned} \text{source} &\rightarrow \{p(x_i)\} \text{ \& } d(x_i, y_j) \\ D &\rightarrow D(\{p(y_j|x_i)\}) = D(Q) \\ I(X; Y) &= I(Q) \\ Q_D &\rightarrow \text{matrices } Q \text{ } \{p(y_j|x_i)\} \text{ for which} \\ &D(Q) \leq D \\ \text{i.e. } Q_D &= \{Q: D(Q) \leq D\} \end{aligned}$$

Distortion D , which we have seen is basically function of probabilities, can be written as D function of Q and similarly, the average mutual information which is again function of conditional probabilities can be written as the function of the matrix Q . Now, we introduce the set Q_D which consists of the matrices Q of the conditional probabilities, for which distortion as the function of Q is less than or equal to some specified distortion level D , that is Q_D is equal to set of matrices Q will satisfy the distortion criterion.

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$$\begin{aligned} R(D) &= \min_{\{p(y_j|x_i)\} \in \Gamma} I(X; Y) \\ \Gamma &= \{ \{p(y_j|x_i)\}, \text{ s.t. } D(\{p(y_j|x_i)\}) \leq D \} \\ R(D) &= \min_{Q \in Q_D} I(Q) \end{aligned}$$

Now, to recollect the mathematical definition for the rate distortion function $R(D)$ is equal to minimum of $I(X; Y)$. This minimum is over conditional probabilities belonging to the set where the set is equal to set of conditional probabilities is such that distortion criterion is satisfied. So, based on our new notation, we can write the rate distortion function as $R(D)$ is equal to minimum of $I(X; Q)$, where Q belongs to the set $\mathcal{Q}(D)$. Now, based on this new notation, let us solve some few examples in order to understand these concepts in much better manner.

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EXAMPLE: SOURCE $\rightarrow N=2$
 RECONSTRUCTION $\rightarrow M=3$

$P(x_1) = P(x_2) = 1/2$

The distortion matrix, \mathcal{D} is given by

$$\mathcal{D} = \begin{bmatrix} d(x_1, y_1) & d(x_1, y_2) & d(x_1, y_3) \\ d(x_2, y_1) & d(x_2, y_2) & d(x_2, y_3) \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 & 3 \\ 3 & 1 & 0 \end{bmatrix} \quad D = 0.45$$

So, let us assume that we have a source alphabet of size N equal to 2 and reconstruction alphabet of size 3 that is M is equal to 3. Let us assume that the probabilities of the source symbols that is P of x_1 , equal to P of x_2 is equal to half. Let us specify the distortions between a source symbol and a reconstruction symbol in the form of a distortion matrix given by. This is by definition, $D_{x_1 y_1}$, $D_{x_1 y_2}$, $D_{x_1 y_3}$, $D_{x_2 y_1}$, $D_{x_2 y_2}$.

And finally, $D_{x_2 y_3}$ and this let us assume to be given as 013310. What does distortion matrix means that there is no distortion if source symbol x_1 is encoded into symbol y_1 and x_2 is encoded into y_3 . Otherwise, there is distortion and let us assume that the average permissible symbol distortion be less than equal to D equal to 0.5. Now, let the two conditional probability matrices Q 1.

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Q_1 and Q_2 be given by

$$Q_1 = \begin{bmatrix} 0.7 & 0.2 & 0.1 \\ 0.1 & 0.2 & 0.7 \end{bmatrix}, \quad Q_2 = \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.1 & 0.1 & 0.8 \end{bmatrix}$$

$D(Q) \leq D$?

And Q_2 be given as follows 0.7 0.2 0.1 0.1 0.2 0.7 and Q_2 is equal to 0.8 0.1 0.1 0.1 0.1 0.8. So, the sum of the row elements is always equal to 1. Now, for both these matrices, we may confirm whether or not they give an average symbol distortion satisfying the given constrain, that is $D(Q)$ is less than or equal to D this is what we want to verify.

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We find

$$D(Q_1) = \sum_{i=1}^2 \sum_{j=1}^3 P(x_i) P(y_j | x_i) d(x_i, y_j)$$
$$= \frac{1}{2} \times 0.2 \times 1 + \frac{1}{2} \times 0.1 \times 3 + \frac{1}{2} \times 0.1 \times 3 + \frac{1}{2} \times 0.2 \times 1$$
$$= 0.1 + 0.15 + 0.15 + 0.1$$
$$= 0.5$$

$D(Q_1) \not\leq D = 0.45$

So, we find $D(Q_1)$ as follows. Double summation i is equal to 1 to 2, j is equal to 1 to 3, probability of x_i , conditional probability of y_j given x_i and the distortion measure between x_i and y_j . Now, let us plug in the values which have been provided. This is

equal to half multiplied by 0.2 multiplied by 1 plus half multiplied by 0.1 into 3 plus half multiplied by 0.1 into 3 plus half into 0.2 multiplied by 1.

Other two terms are 0 because $D \times 1 \times 1$ is equal to 0 and $D \times 2 \times y$ (Refer Time 19:27) and this can be simplified as 0.1 plus 0.15, 0.15 plus 0.1 is equal to 0.5. This clearly shows that D of Q_1 is not less than the specified distortion level that is D equal to 0.5. Now, let us similarly, calculate the distortion for the given matrix Q_2 .

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The image shows a handwritten calculation on a piece of paper. The first line is the formula for the distortion of matrix Q2: $D(Q_2) = \frac{1}{2} \times 0.1 \times 1 + \frac{1}{2} \times 0.1 \times 3 + \frac{1}{2} \times 0.1 \times 3 + \frac{1}{2} \times 0.1 \times 1$. The second line shows the result of the calculation: $= 0.4$. The third line shows the comparison: $D(Q_2) \leq D = 0.45$. The fourth line shows the mapping: $Q_2 \rightarrow Q_D$. The fifth line shows the non-mapping: $Q_1 \not\rightarrow Q_D$.

If you plug in the values similarly to what we have done here then we can show that D of Q_2 is equal to half multiplied by 0.1 multiplied by 1 plus half multiplied by 0.1 multiplied by 3 plus half multiplied by 0.1 into 3 plus half into 0.1 multiplied by 1, and this is equal to 0.4. Clearly, D of Q_2 satisfies the constrain. Therefore, the matrix Q_2 belongs to Q_D as opposed to Q_1 which does not belong to Q_D . Now, we can calculate the average mutual information for all those matrices which belong to Q_D and select the lowest 1 of this. Now, for the given example, we have only 1 matrix that is Q_2 that belongs to the set Q_D . So, let us calculate I of Q_2 .

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$$\text{For } I(Q) = \sum_{i=1}^2 \sum_{j=1}^3 P(x_i) P(y_j | x_i) \log \frac{P(y_j | x_i)}{P(y_j)}$$

$$P(y_j) = \sum_{i=1}^2 P(y_j | x_i) P(x_i)$$

$$P(Y=1) = \frac{0.9}{2} = 0.45, \quad P(Y=2) = 0.1, \quad P(Y=3) = \frac{0.9}{2} = 0.45$$

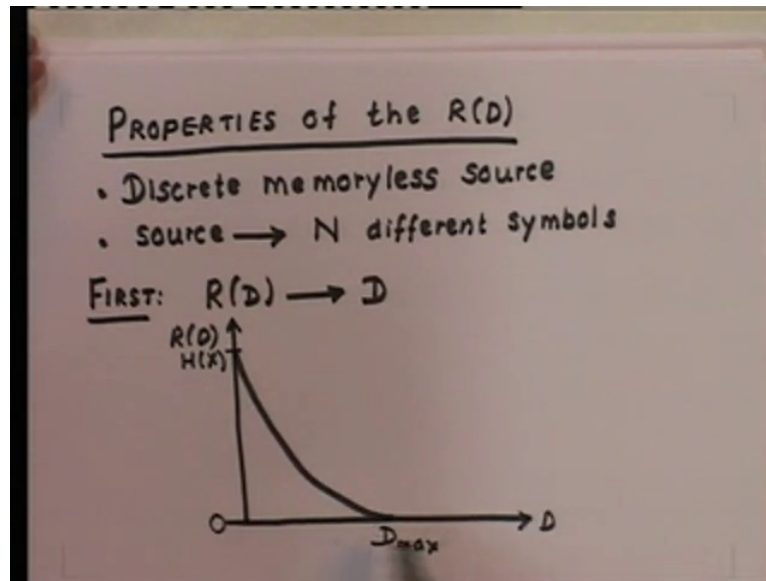
$$I(Q_2) = 2 \left\{ \frac{1}{2} \cdot \frac{8}{10} \cdot \log \frac{8/10}{0.45} + \frac{1}{2} \cdot \frac{1}{10} \cdot \log \frac{1/10}{0.1} + \frac{1}{2} \cdot \frac{1}{10} \cdot \log \frac{1/10}{0.45} \right\}$$

$$= 0.44706$$

For I Q calculation, we use the following relationship. Probability of x i, probability of y j given x i, log of probability y j give x i divided by probability of y j and we also know that probability of y j is equal to summation of over i equal to 1 to 2 of probability y j given x i multiplied by probability of x i. So, we calculate I Q 2 for Q 2 since it belongs to Q D and probability of y equal to 1 from the given values can be evaluated as follows.

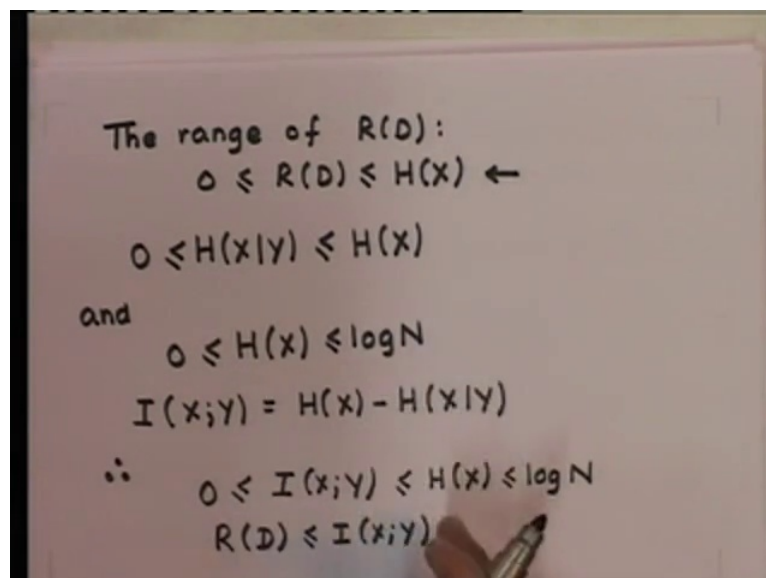
Probability of y equal to 2 is equal to 0.1 and probability of y equal to 3 is equal to 0.9 by 2 is equal to 0.45. And if we use this result and plug in for the relationship of I Q 2 we get the following result plus half into 110 th into log of 110 th by 110 th plus half to 110 th log of 110 th divided by 0.45 and this on simplification is equal to 0.44706. So, having solved this example, now let us proceed and try to look in general what are the properties of the rate distortion function.

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So, we will investigate the properties of the rate distortion function. We will assume that we have discrete memory less information source and let us assume that the source has N different symbols. Now, first let us see how the rate distortion curve proceeds as a function of distortion level D . Now, we can show based on what we derive from the binary source and the Gaussian source that in general that the rate distortion function has the following shape. So, it is a monotonically decreasing function with increasing D .

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The range of rate distortion is given by $R(D)$ is greater than equal to 0 and is less than equal to entropy of discrete memory less source. Now, this result can be shown as follows. We know that equivocation that is $H(X|Y)$ is always greater than equal to 0 and is always less than equal to entropy of the variable X . And we also know based on our previous study that $H(X)$ is greater than equal to 0, and less than equal to $\log N$ based on the assumptions that we have, a discrete memory less information source with the source alphabet of size N .

Now, average mutual information is equal to $H(X) - H(X|Y)$. Therefore, using the previous results it implies that $I(X;Y)$ is greater than equal to 0 is less than equal to $H(X)$ is less than equal to $\log N$. Now, since the rate distortion function, that is $R(D)$, is the minimum of the average mutual information, that is $I(X;Y)$, we have rate distortion function less than equal to average mutual information, and based on the previous results we can say.

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Handwritten mathematical derivation on a whiteboard:

$$0 \leq R(D) \leq H(X)$$

SECOND:

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

D_{\min} $x_i \rightarrow y_j, d(x_i, y_j) \rightarrow \text{minimum}$
 $P(y_j|x_i) = 1$
 x_i

So, $R(D)$ is greater than equal to 0 and less than equal to $h(x)$. Next, let us determine the permissible distortion D may take on. So, the next property of the rate distortion function is to find the permissible distortion D . So, we know that average distortion $D(Q)$ is defined as double summation over i, j of the quantity $P(x_i)$, probability of y_j given x_i , distortion measure between x_i, y_j , this quantity is less than equal to the specified distortion level D .

So, the smallest possible average distortion D minimum is obtained if for each source symbol, that is x_i , we seek the reconstructed symbol y_j where $d(x_i, y_j)$ is minimum and subsequently set this conditional probability that is y_j given x_i is equal to 1. Thus set the remaining conditional probabilities of the reconstructed symbol given that particular source symbol x_i equal to 0.

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The image shows a whiteboard with handwritten mathematical derivations. At the top, it states $d(i) = \min_j d(x_i, y_j)$. Below this, it says "hence $D_{min} = \sum_{i=1}^N P(x_i) d(i)$ ". To the right of this equation, there is a mapping $x_i \rightarrow y_j$ and the expression $d(x_i, y_j)$. Below the main equation, it says "w.l.o.g. $\rightarrow \underline{D_{min} = 0}$ ". At the bottom right, there is a small diagram showing $x_i \rightarrow y_j$ and $d(x_i, y_j)$ with a hand holding a marker pointing to it.

So, let us define d_i to be minimum of $d(x_i, y_j)$ over j . Hence, D minimum is equal to summation of $p(x_i) d_i$ i is equal to 1 to N . Now, without loss of generality, we will assume that D minimum is equal to 0. If this is not the case, we modify the distortion $d(x_i, y_j)$ in such a manner that D minimum becomes equal to 0 again. This means that that for every source symbol x_i there must be a symbol y_j such that $d(x_i, y_j)$ is equal to 0.

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Handwritten notes on a whiteboard:

$$P(y_j|x_i) = \begin{cases} 1 & \text{if } x_i \text{ and } y_j \text{ correspond to each other (i.e. } d(x_i, y_j) = 0) \\ 0 & \text{otherwise} \end{cases}$$

Below this, there is a circled 0 with an equals sign pointing to $H(x|y)$.

$$I(Q) = H(X)$$
$$R(0) = I(Q) = H(X)$$

Now, since for the conditional probabilities we have y_j given x_i equal to 1 if x_i and y_j correspond to each other that is distortion measure between x_i and y_j is equal to 0 is equal to 0 otherwise. The equivocation will be equal to 0. So, that $I(Q)$ is equal to $H(X)$. Now, Q no longer appears here. So, that it directly follows that $R(0)$ is equal to $I(Q)$ is equal to $H(X)$. So, in order to achieve 0 distortion all of the information of the source must therefore, be constructed at the destination. Now, let us determine the maximum possible average distortion.

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Handwritten notes on a whiteboard:

D_{max} ?

$R(D) \rightarrow$ monotone decreasing for increasing D

$D_{max} \rightarrow$ minimal $R(D)$, $R(D) = 0$

D_{max} $y_j \rightarrow$ smallest

$$I(Q) = \sum_i \sum_j P(x_i) P(y_j|x_i) \log \frac{P(y_j|x_i)}{P(y_j)}$$

$I(Q) = 0$ if $P(y_j|x_i) = P(y_j)$

So, we are interested in finding out D_{\max} and this can be determined as follows. It can be shown that the rate distortion function is monotone decreasing for increasing distortion D . So, the maximum D then occurs for minimal $R(D)$ that is $R(D)$ equals to 0. This means that the destination receives no information from the source. Let us denote smallest possible distortion that is then still possible by D_{\max} . This value is achieved by constantly choosing that value y_j for which, the average distortion between an arbitrary source symbol and the symbol y_j in question is smallest and every other choice would lead to a larger average distortion.

Now, the mutual information $I(X;Y)$ is equal to $H(Y) - H(Y|X)$, probability of y_j given x_i , log of probability y_j given x_i over probability of y_j . So, if $I(X;Y)$ is equal to 0 then probability of y_j given x_i is equal to probability of y_j . So, if we choose conditional probability is equal to probability of y_j which means, the occurrence of x_i does not have any effect on the reconstruction of y_j . So, that no information is reproduced in that case the average distortion $D(Q)$ becomes.

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

$$D_{\max} = \min_j \sum_i P(x_i) d(x_i, y_j)$$

So, the minimum value of $D(Q)$ is obtained by setting probability y_j equal to 1 for that value of j for which this quantity is smallest. So, this gives D_{\max} equal to minimum over j of the quantity $\sum_i P(x_i) d(x_i, y_j)$ summed over i . Let us try to appreciate this result with the help of an example. Let us consider again the earlier example, but now we want

to find D minimum and D maximum, and the corresponding values of the rate distortion function.

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Handwritten mathematical derivation on a whiteboard:

(i) $D_{\min} = 0$ appear if $\Phi = \begin{bmatrix} 0 & 1 & 3 \\ 3 & 1 & 0 \end{bmatrix}$
 $Q = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ $P(x_1) = P(x_2) = \frac{1}{2}$

$R(0) = I(Q) = H(X) = 1$

(ii) $D_{\max} = \min_j \sum_{i=1}^2 P(x_i) d(x_i, y_j)$
 $= \frac{1}{2} \min_j \{ d(x_1, y_j) + d(x_2, y_j) \}$
 $= \frac{1}{2} \min [0+3, \underline{1+1}, 3+0] = \frac{1}{2} \times 2 = 1$

So, going back to the earlier example which we had discussed in the starting of the class, we can show that D minimum is equal to 0 and it appears, if the matrix Q is of the form 100001 where the distortion matrix was given as follows 013310 and probability of x_1 equal to probability of x_2 was specified to be half. Now, this corresponds to $R(0)$ equal to $I(Q)$ is equal to $H(X)$ is equal to based on this one. So, the minimum value is equal to 0 and the rate for the minimum distortion is equal to the entropy of the source that is given by 1.

Now, let us evaluate the maximum value of the distortion. So, D_{\max} is equal to minimum over j from the quantity $\sum_{i=1}^2 P(x_i) d(x_i, y_j)$ summed over 1 to 2 is equal to, $P(x_1)$ is equal to half therefore, is equal to half times minimum over j of $d(x_1, y_j) + d(x_2, y_j)$ and this is equal to half based on this matrix times minimum of 0 plus 3 1 plus 13 plus 0. So, the minimum is 2 is equal to half times 2 is equal to 1.

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Handwritten notes on a whiteboard:

$$R(D_{\max}) = R(1) = 0$$

corresponding Q

$$Q = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

So, for this D_{\max} is equal to 1, the corresponding value of $R D$ satisfies the relationship given here and the corresponding conditional probability matrix Q is given as shown here 010010. In order to get a better and a more clearer understanding of rate distortion function, let us solve another problem. The problem is given as follows.

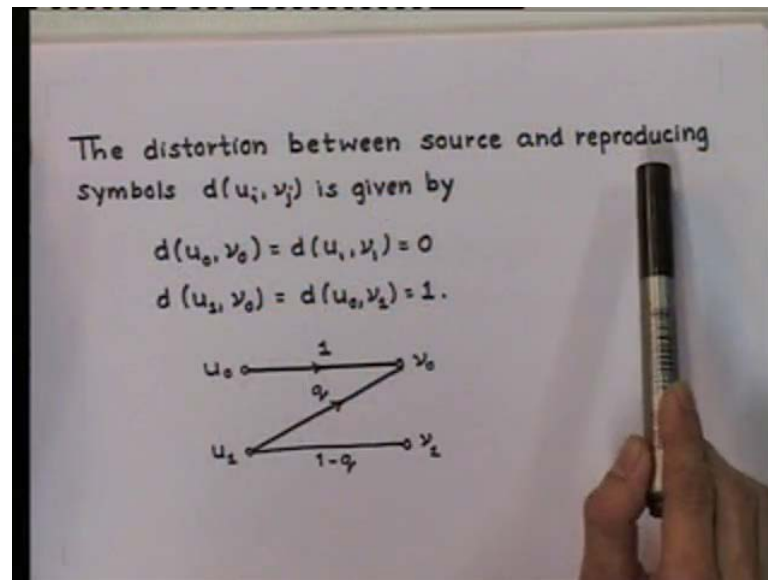
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PROBLEM:
Given a binary source which generates symbols u_0 and u_1 from the alphabet \mathcal{U} . The source symbols occur with the probabilities $p(u_0) = p$ and $p(u_1) = 1 - p$ ($0 \leq p \leq \frac{1}{2}$). The Z-channel is given as well. $\{v_0, v_1\}$ are the destination symbols from the alphabet \mathcal{V} . The transition probabilities $p(v_j/u_i)$ are given by $p(v_0/u_0) = 1$, $p(v_1/u_1) = 1 - q$.

Given a binary source which generates symbols u_0 , and u_1 from the alphabet \mathcal{U} . The source symbols occurs with the probabilities, probability of u_0 equal to p and probability of u_1 equal to $1 - p$ where p is greater than 0 and less than equal to half. The Z

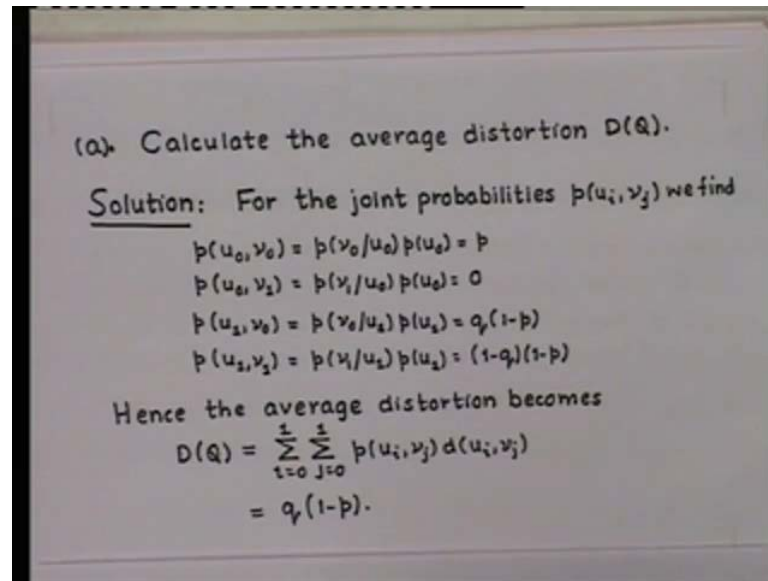
channel for this is given as well v_0 and v_1 are the destination symbols from the alphabet V . The transition probabilities, that is the probability of v_j given u_i are also given as follows. Probability of v_0 given u_0 is equal to 1 and probability of v_1 given u_1 is equal to $1 - q$.

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We have also been provided the distortion between source and reproducing symbols that is distortion between u_i and v_j is given as follows. Distortion between u_0 v_0 is equal to distortion between u_1 and v_1 is equal to 0. And distortion between u_1 v_0 is equal to distortion between u_0 v_1 equal to 1. The Z channel is given as shown here u_0 and u_1 are the input symbols v_0 and v_1 are output symbols.

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(a) Calculate the average distortion $D(Q)$.

Solution: For the joint probabilities $p(u_i, v_j)$ we find

$$p(u_0, v_0) = p(v_0/u_0) p(u_0) = p$$
$$p(u_0, v_1) = p(v_1/u_0) p(u_0) = 0$$
$$p(u_1, v_0) = p(v_0/u_1) p(u_1) = q_p(1-p)$$
$$p(u_1, v_1) = p(v_1/u_1) p(u_1) = (1-q_p)(1-p)$$

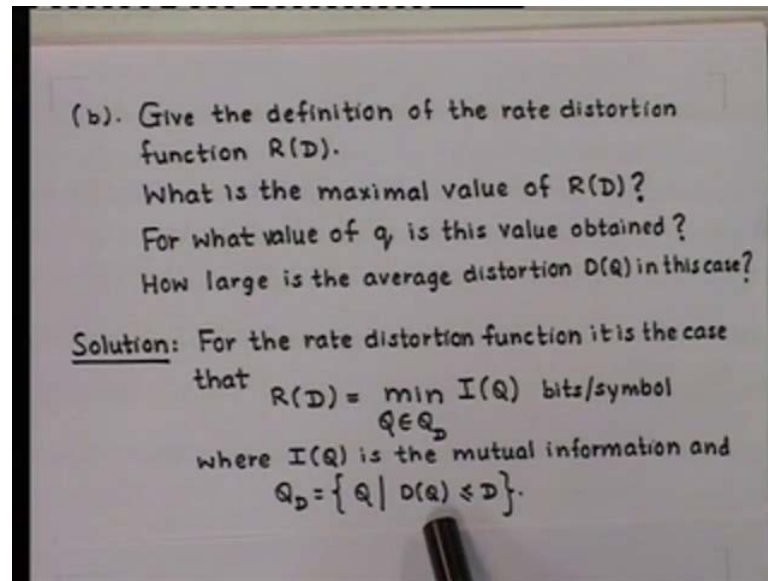
Hence the average distortion becomes

$$D(Q) = \sum_{i=0}^1 \sum_{j=0}^1 p(u_i, v_j) d(u_i, v_j)$$
$$= q_p(1-p).$$

Given this, we are required to calculate the average distortion $D(Q)$, where Q is the conditional probability matrix and the solution is as follows. First, let us calculate the joint probabilities, that is the probability of u_i and v_j . We calculate this as follows. Probability of $u_0 v_0$ is equal to conditional probability, that is probability of v_0 given u_0 multiplied by probability of u_0 and based on the information provided, this is equal to p .

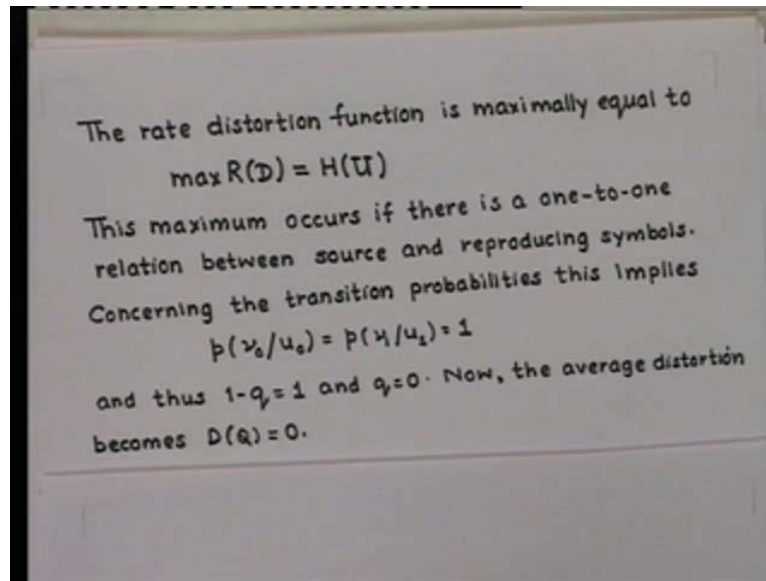
Similarly, we can calculate the joint probabilities $p(u_0 v_1)$ which comes out to be 0, probability $u_1 v_0$ is equal to Q into $1 - p$, probability of $u_1 v_1$ is equal to $1 - Q$ multiplied by $1 - p$. Given these joint probabilities we can plug this along with the distortions given between the source and reproducing symbols and the average distortion becomes as follows. $D(Q)$ given by this formula is equal to Q multiplied by $1 - p$.

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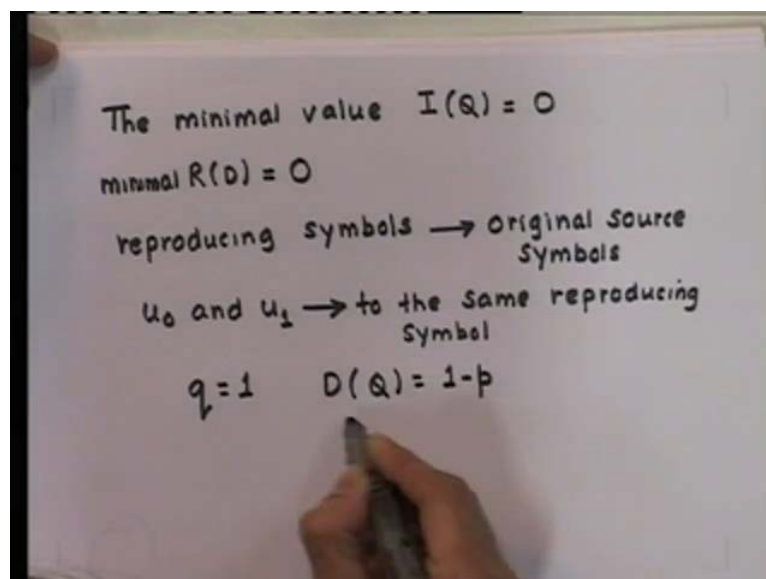
Next, we are required to provide the definition of the rate distortion function. Then for this problem what is the maximal value of $R(D)$ for what value of Q is this value obtained? And, in this case, how large is the average distortion $D(Q)$? The solution to this task, the solution for the same is as follows. For the rate distortion function, it is the case that $R(D)$ is defined as minimum of $I(Q)$. This minimum is over Q belonging to \mathcal{Q}_D where $I(Q)$ is the mutual information and \mathcal{Q}_D is the set of all conditional probability matrix, such that distortion for that conditional probability matrix is less than, or equal to the specified distortion level D . So, this is the definition of the rate distortion function which we had seen earlier.

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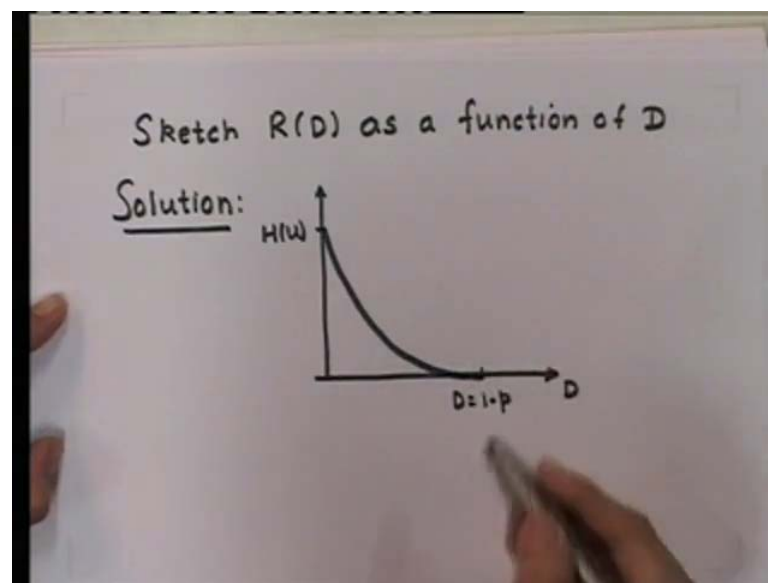
Now we have also seen that the rate distortion function is maximally equal to $H(U)$ and this maximum occurs if there is one-to-one relation between source and reproducing symbols. So, concerning the transition probabilities, this implies that we should have probability of v_0 given u_0 is equal to probability of v_1 given u_1 equal to 1 which implies that $1 - q$ equals to 1 which implies q is equal to 0 and. Now, in this case the average distortion becomes equal to 0. Next question is what is the minimal value of rate distortion function for what value of Q is this value obtained? And how large is the average distortion $D(Q)$ in this case?

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Now, the minimal value of the mutual information $I Q$ is equal to 0. This we have proved earlier and thus the minimal value of the rate distortion function also equals 0. This minimum appears, if the reproducing symbols give no information about the original source symbols and this is the case if both symbols u_0 and u_1 give rise to the same reproducing symbol. So, in this case, it leads to Q equal to 1 and now distortion is equal to $1 - p$ and the final question is...

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Let us sketch rate distortion function as a function of D and the solution to this is as follows. We have seen the maximum value is $H U$, minimum value is 0 and that value is achieved for D equal to $1 - p$ and in between this it will be monotonically decreasing function. So, for D equal to $1 - p$, $R D$ is equal to 0. There are two ways to calculate the rate distortion function.

We have looked at one approach, which is to calculate the mutual information, and find the lower bound for this mutual information and show that this lower bound is achievable. The other approach is a computational approach. In the next class, we will discuss the computational approach to calculate the rate distortion function and also, explore the application of the same to a binary case.