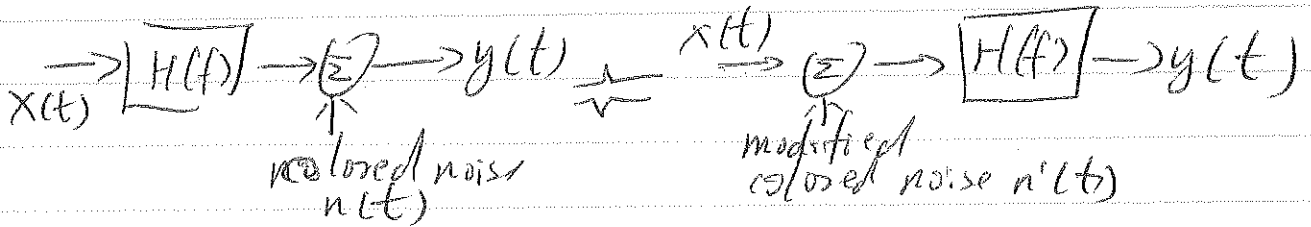


2/17/14

class 10

9.12 Information capacity of colored noisy channel



$$SN(f) = \frac{S_N(f)}{|H(f)|^2}$$

Suppose N channels

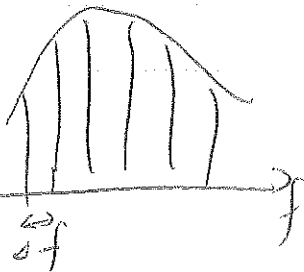
$$y_k(t) = x_k(t) + n_k(t), \quad k=1, 2, \dots, N$$

① $P_k = S_x(f_k) \Delta f, \quad k=1, 2, \dots, N$

② $\sigma_k^2 = \frac{S_N(f_k) \Delta f}{|H(f_k)|^2}, \quad k=1, 2, \dots, N$

$$C_k = \frac{1}{2} \Delta f \log_2 \left(1 + \frac{P_k}{\sigma_k^2} \right), \quad k=1, 2, \dots, N$$

$H(f)$



$$\max C \equiv \sum_{k=1}^N C_k = \frac{1}{2} \sum_{k=1}^N \Delta f \log_2 \left(1 + \frac{P_k}{\sigma_k^2} \right)$$

s.t. $\sum_{k=1}^N P_k = P = \text{constant}$

Lagrange multiplier λ

$$J = \frac{1}{2} \sum_{k=1}^N \Delta f \log_2 \left(1 + \frac{P_k}{\sigma_k^2} \right) + \lambda \left(P - \sum_{k=1}^N P_k \right)$$

$$\frac{\partial J}{\partial P_k} = \frac{\Delta f \log_2 e}{P_k + \sigma_k^2} - \lambda = 0$$

$$P_k + \sigma_k^2 = \lambda \Delta f \log_2 e = \text{constant}$$

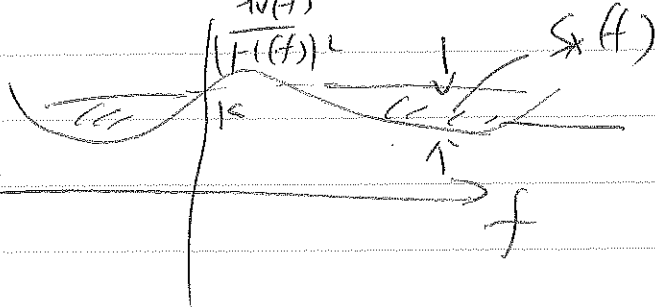
① $S_x = S_x(f_k) \Delta f + \left| \frac{S_N(f_k)}{H(f_k)} \right|^2 \Delta f$

$$\Rightarrow S_x(f_k) = \frac{P_k}{\Delta f} - \frac{S_N(f_k)}{|H(f_k)|^2} \quad \text{for } P_k = \frac{S_N(f_k)}{|H(f_k)|^2}$$

To calculate C : $P = \int \left(K - \frac{S_w(f)}{|H(f)|^2} \right) df$

$$C = \frac{1}{2} \sum_{k=1}^N \log \left(K \frac{|H(f_k)|^2}{S_w(f_k)} \right) = \frac{1}{2} \int_{-\infty}^{\infty} \log \left(K \frac{|H(f)|^2}{S_w(f)} \right) df$$

water filling Interpretation



shaded area is the power

Exam: give several channels, ^{wise level} overall power calculate the tx power for each channel to max the overall capacity

9.13 Rate Distortion Theory

rate distortion tx M -ary alphabet $X_i: 1 \dots M$ with probability P_i

R average code rate, H source entropy

rx N -ary alphabet $y_j = 1 \dots N$

$$P(X_i, y_j) = P(y_j | X_i) P(X_i)$$

single letter average distortion $\bar{d} = \sum_{i=1}^M \sum_{j=1}^N P(X_i) P(y_j | X_i) d(X_i, y_j)$

$$P_D = \{ P(y_j | X_i) : \bar{d} \leq D \}$$

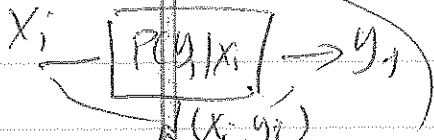
single letter distortion measure

$$I(X; Y) = \sum_{i=1}^M \sum_{j=1}^N P(X_i) P(y_j | X_i) \log \left(\frac{P(y_j | X_i)}{P(y_j)} \right)$$

rate distortion function

$$R(D) = \min_{P(y_j | X_i) \in P_D} I(X; Y)$$

$$s.t. \sum_{j=1}^N P(y_j | X_i) = 1 \quad i=1 \dots M$$



Ex 9.14 Gaussian Source

$$d(X, y) = (X - y)^2$$

$$R(D) = \begin{cases} \frac{1}{2} \log\left(\frac{\sigma^2}{D}\right) & 0 \leq D \leq \sigma^2 \\ 0 & D > \sigma^2 \end{cases}$$

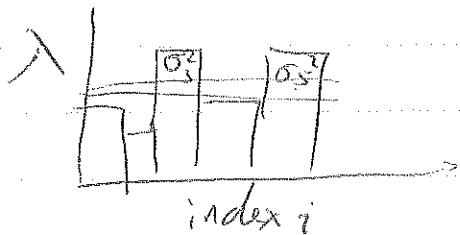
Example 9.15 set of Parallel Gaussian Source

$$d = \sum_{i=1}^N (x_i - \hat{x}_i)^2$$

$$R(D) = \sum_{i=1}^N \frac{1}{2} \log\left(\frac{\sigma_i^2}{D_i}\right)$$

$$D_i = \begin{cases} \lambda & \lambda < \sigma_i \\ \sigma_i & \lambda \geq \sigma_i \end{cases}$$

λ is selected s.t. $\sum D_i = D$



9.14 Data compression

Data is compressed and then transmitted. These two steps can be separated to be optimized without loss of the optimization, if there is no transmission deadline. If there is tx deadline like video conferencing

