

① let $Y_1 = X + N_1$ where $X = \pm 1$ w.p $\frac{1}{2}$
 $Y_2 = 2X + N_2$ N_1, N_2 indep. of X
 $\& N_1, N_2$ i.i.d $\sim N(0, 1)$

$Y_1 + 2Y_2 = 5X + N_1 + 2N_2$
 $Y_1 + 4Y_2 = 9X + N_1 + 4N_2$
 $C = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

\hat{X} = MMSE estimate of X from Y_1 & Y_2 .

$\hat{X} = E[X|Y] = E[X|Y_1, Y_2] = P_{\Omega}[X=1|Y_1, Y_2] - P_{\Omega}[X=-1|Y_1, Y_2]$

$P_{\Omega}[X=1|Y_1=y_1, Y_2=y_2] = \frac{f(y_1, y_2/x=1) P[X=1]}{f(y_1, y_2)}$
 $= \frac{1}{(2\pi)(1)} e^{-\frac{1}{2}[(y_1-1)^2 + (y_2-2)^2]} \cdot \frac{1}{2}$
 $= \frac{\frac{1}{2} \left(\frac{1}{2\pi} e^{-\frac{1}{2}[(y_1-1)^2 + (y_2-2)^2]} \right) + \frac{1}{2} \left(\frac{1}{2\pi} e^{-\frac{1}{2}[(y_1+1)^2 + (y_2+2)^2]} \right)}{e^{-\frac{1}{2}[(y_1-1)^2 + (y_2-2)^2]} + e^{-\frac{1}{2}[(y_1+1)^2 + (y_2+2)^2]}}$

$P_{\Omega}[X=-1|Y_1=y_1, Y_2=y_2] = \frac{e^{-\frac{1}{2}[(y_1+1)^2 + (y_2+2)^2]}}{e^{-\frac{1}{2}[(y_1-1)^2 + (y_2-2)^2]} + e^{-\frac{1}{2}[(y_1+1)^2 + (y_2+2)^2]}}$

$\hat{X} = \frac{e^{-\frac{1}{2}[(y_1-1)^2 + (y_2-2)^2]} - e^{-\frac{1}{2}[(y_1+1)^2 + (y_2+2)^2]}}{e^{-\frac{1}{2}[(y_1-1)^2 + (y_2-2)^2]} + e^{-\frac{1}{2}[(y_1+1)^2 + (y_2+2)^2]}}$

$e^{-\frac{1}{2}(y_1^2 - 2y_1 + 1 + y_2^2 - 4y_2 + 4)} - e^{-\frac{1}{2}(y_1^2 + 2y_1 + 1 + y_2^2 + 4y_2 + 4)}$

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$\frac{e^{+2y_1 + 2y_2} - e^{-2y_1 - 2y_2}}{e^{y_1 + 2y_2} + e^{-y_1 - 2y_2}} = \tanh(y_1 + 2y_2)$

LMMSE estimate of X from Y_1 and Y_2 .

(a) $E[X] = 0$

$E[Y_1] = 0$

$E[Y_2] = 0$

$\hat{X} = K_0 Y$ where $K_0 R_Y = R_{XY}$.

$Y_1 + 2Y_2 = 5X + N_1 + 2N_2$
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$$R_Y = \begin{bmatrix} 1+1 & 2 \\ 2 & 4+1 \end{bmatrix} = \begin{bmatrix} 2 & 2 \\ 2 & 5 \end{bmatrix}$$

$$R_{XY} = \begin{bmatrix} 1 & 2 \end{bmatrix}$$

$$K_0 \begin{bmatrix} 2 & 2 \\ 2 & 5 \end{bmatrix} = \begin{bmatrix} 1 & 2 \end{bmatrix} \Rightarrow K_0 = \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} 5 & -2 \\ -2 & 2 \end{bmatrix} \frac{1}{6}$$

$$= \begin{bmatrix} 1 & 2 \end{bmatrix} \frac{1}{6}$$

$$\hat{X} = \frac{1}{6} \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \frac{Y_1 + 2Y_2}{6}$$

(b) MSE achieved by LMMSE estimate.

$$E[|\tilde{X}|^2] = E[(X - \hat{X})(X - \hat{X})] = E[X(X - \hat{X})] = E[X^2] - E[\hat{X}^2]$$

$$E[X^2] = 1$$

$$E[X\hat{X}] = E[X K_0 Y] = K_0 R_{XY} = K_0 R_Y K_0^H = \frac{1}{6} \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \frac{5}{6}$$

$$MSE = 1 - \frac{5}{6} = \frac{1}{6}$$

(3)

$$Y_1 = X_1 + N_1$$

$$Y_2 = X_2 + N_2$$

$$X_2 = \rho X_1 + \sqrt{1-\rho^2} Z$$

- X_1 zero mean, variance 1.

- N_1, N_2 i.i.d. zero mean, variance σ^2 indep. of X_1

- Z zero-mean, variance 1 indep. of X_1, X_2, N_1, N_2

$$E[X_2^2] = \rho^2 + 1 - \rho^2 = 1$$

(a) Estimate X_1 from Y_1

$$\hat{X}_1 = k_0 Y_1 \quad k_0(1 + \sigma^2) = 1$$

$$\Rightarrow \hat{X}_1 = \frac{Y_1}{1 + \sigma^2}$$

(b) Estimate Y_2 from Y_1

$$\hat{Y}_2 = k Y_1 \quad k(1 + \sigma^2) = R_{Y_1 Y_2} = E[X_1 X_2] = \rho$$

$$\hat{Y}_2 = \hat{X}_2 + \hat{N}_2 = \rho \hat{X}_1 = \frac{\rho Y_1}{1 + \sigma^2}$$

(c) let $E_2 = Y_2 - \hat{Y}_2 = Y_2 - \frac{pY_1}{1+\sigma^2}$

Estimate X_2 from E_2 .

$$\hat{X}_{2e} = k_e E_2$$

$$k_e R_{E_2} = R_{E_2 X_2}$$

$$R_{E_2 X_2} = E \left[\left(Y_2 - \frac{pY_1}{1+\sigma^2} \right) X_2 \right] = 1 - \frac{p \cdot p}{1+\sigma^2}$$

$$= 1 - \frac{p^2}{1+\sigma^2}$$

$$R_{E_2} = E \left[(Y_2 - \hat{Y}_2)(Y_2 - \hat{Y}_2) \right] = E \left[Y_2(Y_2 - \hat{Y}_2) \right]$$

$$= E[Y_2^2] - E[Y_2 \hat{Y}_2]$$

$$= 1 + \sigma^2 - \frac{p}{1+\sigma^2} E[Y_2 Y_1] = 1 + \sigma^2 - \frac{p}{1+\sigma^2} E[X_1 X_2]$$

$$= 1 + \sigma^2 - \frac{p^2}{1+\sigma^2}$$

$$k_e = \frac{1 - \frac{p^2}{1+\sigma^2}}{1 + \sigma^2 - \frac{p^2}{1+\sigma^2}}$$

$$\hat{X}_{2e} = \frac{1 - \frac{p^2}{1+\sigma^2}}{1 + \sigma^2 - \frac{p^2}{1+\sigma^2}} \left(Y_2 - \frac{pY_1}{1+\sigma^2} \right)$$

(d) Estimate X_2 from Y_1

$$\hat{X}_{21} = p \hat{X}_1 = \frac{pY_1}{1+\sigma^2}$$

(e) $\hat{X}_2 = \hat{X}_{21} + \hat{X}_{2e} = \frac{pY_1}{1+\sigma^2} + \frac{1 - \frac{p^2}{1+\sigma^2}}{1 + \sigma^2 - \frac{p^2}{1+\sigma^2}} \left(Y_2 - \frac{pY_1}{1+\sigma^2} \right)$

$$= \frac{\sigma^2}{1 + \sigma^2 - \frac{p^2}{1+\sigma^2}} \left(\frac{pY_1}{1+\sigma^2} \right) + \frac{1 - \frac{p^2}{1+\sigma^2}}{1 + \sigma^2 - \frac{p^2}{1+\sigma^2}} Y_2$$

$$= \frac{p\sigma^2}{(1+\sigma^2)^2 - p^2} Y_1 + \frac{1 + \sigma^2 - p^2}{(1+\sigma^2)^2 - p^2} Y_2$$

\hat{X}_2 from Y_1, Y_2 .

$$\hat{X}_2 = [K_0] \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix}$$

$$[K_0] = R_{X_2Y} R_Y^{-1}$$

$$= [p \quad 1] \begin{bmatrix} 1+\sigma^2 & p \\ p & 1+\sigma^2 \end{bmatrix}^{-1}$$

$$= [p \quad 1] \begin{bmatrix} 1+\sigma^2 & -p \\ -p & 1+\sigma^2 \end{bmatrix} \frac{1}{(1+\sigma^2)^2 - p^2}$$

$$\hat{X}_2 = \frac{p\sigma^2}{(1+\sigma^2)^2 - p^2} Y_1 + \frac{1+\sigma^2 - p^2}{(1+\sigma^2)^2 - p^2} Y_2 = [p\sigma^2 \quad -p^2 + 1 + \sigma^2] \frac{1}{(1+\sigma^2)^2 - p^2}$$

Direct answer
for (e)