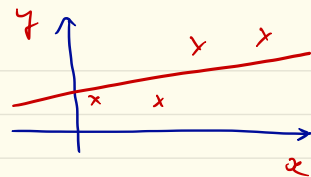


ECE 313: Lecture 34

ML estimation of linear regression

Joint pdfs of functions of random variables (Ch 4.7)

Recognizing independence



Setup: Given training data $(x_n, y_n)_{n=1}^N$
 Model: $y_n =$

"Noise" $y_n = \underbrace{f_{(a,b)}(x_n)}_{\theta} = ax_n + b + w_n$

$\Rightarrow w_n = \underbrace{y_n - (ax_n + b)}_{w_n}$ w_n has dist'n $N(0, \sigma^2)$
 \uparrow independent

$\Rightarrow f_{w_n}(w_n) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{w_n^2}{2\sigma^2}\right)$

ML $f(w_1, w_2, \dots, w_N) = \prod_{n=1}^N f_{w_n}(w_n) = \left(\frac{1}{\sqrt{2\pi}\sigma}\right)^N \exp\left(-\frac{1}{2\sigma^2} \sum_{n=1}^N w_n^2\right)$

ML estimator

$$\max_{w_1, w_2, \dots, w_N} f_{w_1, w_2, \dots, w_N} (w_1, w_2, \dots, w_N)$$

$$\Rightarrow \min \sum_{n=1}^N w_n^2 \quad \Rightarrow \min \underbrace{\sum_{n=1}^N (y_n - (ax_n + b))^2}_{J(a, b)}$$

Hence, under Gaussian "noise" model

ML estimator \equiv min sum of square errors

$$\frac{\partial J}{\partial a} = \sum_{n=1}^N 2(y_n - (ax_n + b)) x_n = 0$$

$$\frac{\partial J}{\partial b} = 0$$

$$\Rightarrow a_{ML} = \frac{\sum_n y_n x_n}{\sum_n x_n^2}$$

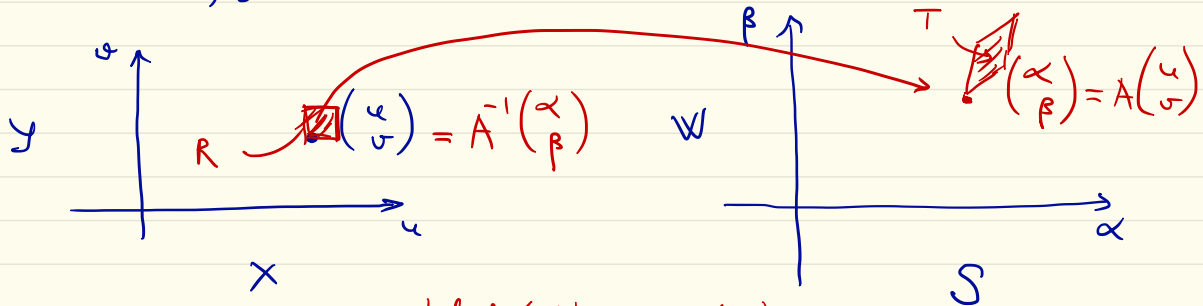
⊛ Prob setting

$$\begin{bmatrix} S \\ W \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} X \\ Y \end{bmatrix} \Rightarrow \begin{bmatrix} X \\ Y \end{bmatrix} = A^{-1} \begin{bmatrix} S \\ W \end{bmatrix}$$

linear transformation A

$$\begin{matrix} X, Y \\ f_{X,Y}(u, v) \end{matrix}$$

$$\begin{matrix} S, W \\ f_{S,W}(\alpha, \beta) = ? \end{matrix}$$



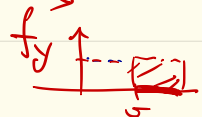
$$\text{area}(T) = |\det(A)| \text{area}(R)$$

$$f_{S,W}(\alpha, \beta) = \frac{1}{|\det(A)|} f_{X,Y}\left(A^{-1}\begin{pmatrix} \alpha \\ \beta \end{pmatrix}\right)$$

Recall:



$$\begin{matrix} y = ax \\ v = au \end{matrix} \Rightarrow \begin{matrix} y = \frac{a}{a} x \\ v = \frac{a}{a} u \end{matrix} \Rightarrow \begin{matrix} f_y\left(\frac{y}{a}\right) = \frac{1}{a} \cdot f_x\left(\frac{y}{a}\right) \end{matrix}$$



[8+4 points] Suppose X and Y are ~~independent~~ random variables with joint pdf:

$$f_{X,Y}(u,v) = \begin{cases} 2e^{-u}e^{-2v} & \text{if } u \geq 0, v \geq 0 \\ 0 & \text{else.} \end{cases}$$

(a) Find the joint pdf of $S = X + Y$ and $W = Y - X$.

$$\begin{pmatrix} S \\ W \end{pmatrix} = \underbrace{\begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}}_A \begin{pmatrix} X \\ Y \end{pmatrix}$$

$\begin{pmatrix} \alpha \\ \beta \end{pmatrix}$
 $\begin{pmatrix} u \\ v \end{pmatrix}$

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$$

$$A^{-1} = \frac{1}{\det(A)} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

$$\left. \begin{aligned} \det(A) &= 2 \\ A^{-1} &= \frac{1}{2} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \\ \begin{pmatrix} u \\ v \end{pmatrix} &= \frac{1}{2} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \end{aligned} \right\}$$

$$f_{S,W}(\alpha, \beta) = \frac{1}{\det(A)} f_{X,Y}(A^{-1} \begin{pmatrix} \alpha \\ \beta \end{pmatrix})$$

$$= \frac{1}{2} f_{X,Y} \left(\underbrace{\frac{1}{2}(\alpha - \beta)}_u, \underbrace{\frac{1}{2}(\alpha + \beta)}_v \right)$$

$$= \begin{cases} e^{-\frac{1}{2}(\alpha - \beta)} e^{-(\alpha + \beta)} & \text{for } \frac{1}{2}(\alpha - \beta) \geq 0; \frac{1}{2}(\alpha + \beta) \geq 0 \\ 0 & \text{else} \end{cases}$$

(b) Are S and W independent? Explain.

No, (not a product set)

4.7. [Recognizing independence]

Decide whether X and Y are independent for each of the following three joint pdfs. If they are independent, identify the marginal pdfs f_X and f_Y . If they are not, give a reason why.

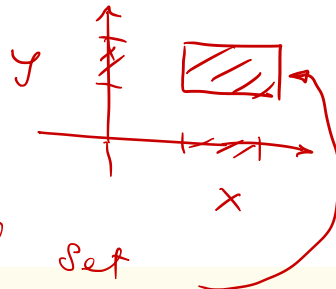
Y (a) $f_{X,Y}(u,v) = \begin{cases} \frac{4}{\pi} e^{-(u^2+v^2)} & u, v \geq 0 \\ 0 & \text{else.} \end{cases}$

Y (b) $f_{X,Y}(u,v) = \begin{cases} -\frac{\ln(u)v^2}{21} & 0 \leq u \leq 1, 1 \leq v \leq 4 \\ 0 & \text{else.} \end{cases}$

N (c) $f_{X,Y}(u,v) = \begin{cases} \frac{(96)u^2v^2}{\pi} & u^2 + v^2 \leq 1 \\ 0 & \text{else.} \end{cases}$

\iff ① $f_{X,Y}(u,v) = g(u)h(v)$

② Support = product region



not product set