

→ fill out the post MT2 survey! (Due tomorrow night)

→ Today: a lot on LLSE.

1. MMSE for Jointly Gaussian RVs
2. Orthogonal LLSE
3. Linear Regression

orthogonality: $X \perp Y \Leftrightarrow E[XY] = 0$.

$L[X|Y]$ is LLSE $\Leftrightarrow X - L[X|Y] \perp Y$
 $g(Y) = L[X|Y]$

MMSE is orthogonal to all functions of Y .

$$\text{MMSE}[X|Y] \perp E[X|Y]$$

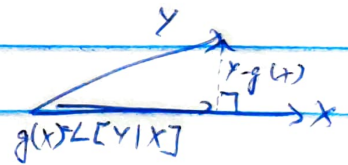
$$\text{and } E[E[X|Y]g(Y)] = 0 \quad \forall g(\cdot).$$

$$1. \quad g(X) = L[Y|X] = E[Y] + \frac{\text{cov}(X,Y)}{\text{var}(X)} (X - E[X])$$

(1)

$$\text{WTS: } E[(Y - g(X))X] = 0 \Leftrightarrow Y - g(X) \perp X$$

True because LLSE: residual $\perp X$.



$$(2) \text{ WTS } \text{cov}(Y - g(X), X) = 0.$$

$$\text{cov}(X, Y) = E[XY] - E[X]E[Y]$$

$\neq 0$ in general

$$= E[(Y - g(X))X] - E[Y - g(X)]E[X]$$

$$= \underset{\text{by eq 1.}}{0} - 0 = 0$$

LLSE is unbiased

$$E[g(X)] = E[Y]$$

$$E[E[Y] + (\dots)(E[X] - E[X])] = E[Y]$$

\perp orthogonal, \parallel indep.

$$(3) \text{cov}(Y-g(X), X) = 0 \rightarrow Y-g(X) \perp X$$

true for any RVs? No.

true for Gaussians? Yes.

X, Y are JG

Are $X, \underline{Y-g(X)}$ JG? Yes

Linear combination of X, Y .

$$X = A_1 Z_1 + \mu_1$$

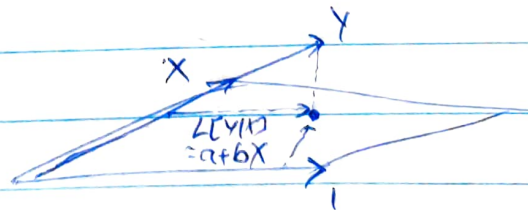
$$Y = A_2 Z_2 + \mu_2$$

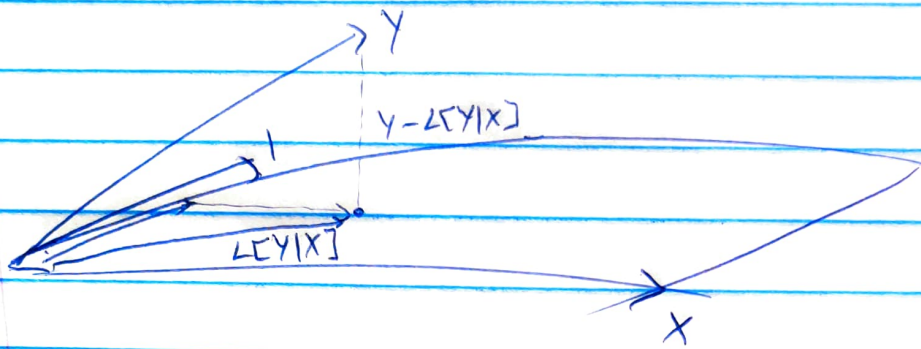
$$(4) Y-g(X) \perp f(X) \text{ for all } f(\cdot)$$

$$E[(Y-g(X))f(X)] = E[(Y-g(X))] E[f(X)] = 0.$$

(5) MMSE is the unique function $h(X)$ such that

$$Y-h(X) \perp f(X) \text{ for all } f(\cdot)$$





$$Y - L[Y|X] \perp 1$$

$$Y - L[Y|X] \perp X$$

2. Orthogonal LLSE.

$X - L[X|Y, Z]$ has to be orthogonal to all of $\{1, Y, Z\}$

idea: Y and Z give you "independent information" about X .

$L[X|Y]$ should be independent of / orthogonal to Z
and $L[X|Z]$ should be independent of / orthogonal to Y .

Is their sum, orthogonal to both?

$$E[Z \cdot \underset{\text{residual}}{L[X|Y]}] = 0, \quad E[Y \cdot \overset{(a+bZ)}{L[X|Z]}] = 0$$

$$g(Y, Z) = L[X|Y] + L[X|Z]$$

$$\text{WTS } g(Y, Z) = L[X|Y, Z] \leftarrow *g(Y, Z) \perp \{1, Y, Z\}$$

$$X - g(Y, Z) \perp 1 \iff E[(X - g(Y, Z)) \cdot 1] = 0$$

$$E[X - L[X|Y] - L[X|Z]] = 0$$

$$E[X] - E[L[X|Y]] - E[L[X|Z]] = 0$$

$$0 - 0 - 0 = 0$$

$$A \perp B$$

$$E[AB] = 0$$

$$g(Y, Z) = L[X|Y] + L[X|Z]$$

$$X - g(Y, Z) \perp Y \Leftrightarrow E[(X - L[X|Y] - L[X|Z])Y] = 0$$

$$E[(X - L[X|Y]) \cdot Y] - E[L[X|Z]Y] = 0$$

↳ 0 because

it's an LLSE

↳ 0 because

$Y \perp Z$

Similarly, $X - g(Y, Z) \perp Z$

$$\therefore X - g(Y, Z) \perp \{1, Y, Z\}$$

$$g(Y, Z) = L[X|Y, Z]$$

$$X \perp Y \Leftrightarrow E[XY] = 0$$

$$X - g(Y) \perp 1 \Leftrightarrow E[(X - g(Y)) \cdot 1] = 0$$

$$E[X] = E[g(Y)]$$

$$b) L[X|Y, Z] = L[X|Y] + L[X|Z - L[Z|Y]]$$

$$\text{if } Y, Z \text{ indep } L[Z|Y] = E[Z]$$

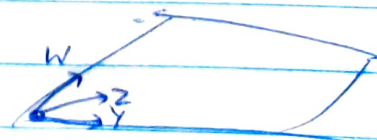
$$\therefore L[X|Y, Z] = L[X|Y] + L[X|Z - E[Z]]$$

$$= L[X|Y] + L[X|Z]$$

$$\text{Say } W = Z - L[Z|Y]$$

$$Y \perp W$$

(LLSE property)

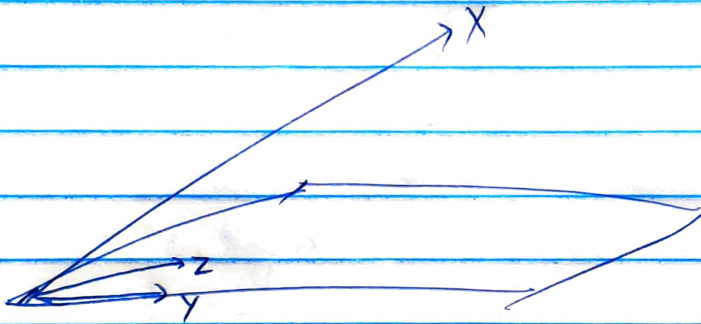
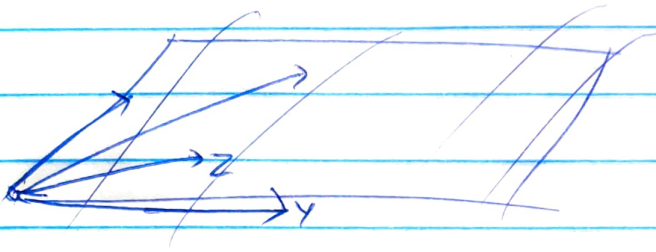


$$L[X|Y, W] = L[X|Y] + L[X|W]$$

$$= L[X|Y] + L[X|Z - L[Z|Y]]$$

why is this $L[X|Y, Z]$? Yes: $(Y, W) \leftrightarrow (Y, Z)$ reversible.

$Z \rightarrow$ something w/ only the info unique to Z .



$E[YZ] \neq 0$ - correlated as RVs
not orthogonal as vectors

$$W = Z - L[Z|Y]$$

$$L[X|Y, W] = L[X|Y] + L[X|W]$$

$$L[X|Y, Z] = L[X|Y, W]$$