

Discussion 12

Fall 2017

1. Orthogonal LLSE

- (a) Consider zero-mean random variables X, Y, Z such that Y, Z are orthogonal. Show that $L[X | Y, Z] = L[X | Y] + L[X | Z]$.
- (b) Show that for any zero-mean random variables X, Y, Z it holds that:

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

Solution:

- (a) Let $U(Y) = L[X | Y]$, $V(Z) = L[X | Z]$. X , $U(Y)$, and $V(Z)$ are all zero-mean. Observe that $V(Z)$ and Y are orthogonal. To see this, observe that Y is orthogonal to 1 (this is the statement that Y is zero-mean) and to Z , and hence to any affine function of Z (in particular, Y is orthogonal to $V(Z)$). A similar argument establishes that $U(Y)$ and Z are orthogonal as well. Now,

$$\begin{aligned}\mathbb{E}[X - U(Y) - V(Z)] &= 0, \\ \mathbb{E}[(X - U(Y) - V(Z))Y] &= \mathbb{E}[V(Z)Y] = 0, \\ \mathbb{E}[(X - U(Y) - V(Z))Z] &= \mathbb{E}[U(Y)Z] = 0,\end{aligned}$$

since $X - U(Y)$ is orthogonal to Y and $X - V(Z)$ is orthogonal to Z . Therefore, $X - U(Y) - V(Z)$ is orthogonal to any linear function of 1, Y , and Z , and hence it is the LLSE of X given Y, Z .

- (b) Let $W = Z - L[Z | Y]$, so W and Y are orthogonal. From the previous part we know $L[X | Y] + L[X | W] = L[X | W, Y]$, so it remains to argue that $L[X | W, Y] = L[X | Y, Z]$. This is intuitively clear since (W, Y) and (Y, Z) are linear functions of each other.

2. Noisy Guessing

Let X, Y , and Z be i.i.d. with the standard Gaussian distribution. Find $\mathbb{E}[X | X + Y, X + Z, Y - Z]$.

Hint: Argue that the observation $Y - Z$ is redundant.

Solution:

Since $Y - Z = X + Y - (X + Z)$, we have

$$\mathbb{E}(X | X + Y, X + Z, Y - Z) = \mathbb{E}(X | X + Y, X + Z).$$

First, we calculate $\mathbb{E}(X \mid X + Y) = (X + Y)/2$ by symmetry. Also,

$$\mathbb{E}(X + Z \mid X + Y) = \mathbb{E}(X \mid X + Y) = \frac{X + Y}{2},$$

so the innovation is $X + Z - (X + Y)/2 = (X - Y + 2Z)/2$. Thus,

$$\begin{aligned} \text{cov}\left(X, \frac{X - Y + 2Z}{2}\right) &= \frac{1}{2}, \\ \text{var} \frac{X - Y + 2Z}{2} &= \frac{3}{2}, \end{aligned}$$

and so $\mathbb{E}(X \mid (X - Y + 2Z)/2) = (X - Y + 2Z)/6$. Hence,

$$\begin{aligned} \mathbb{E}(X \mid X + Y, X + Z) &= \frac{X + Y}{2} + \frac{X - Y + 2Z}{6} = \frac{2}{3}X + \frac{1}{3}Y + \frac{1}{3}Z \\ &= \frac{1}{3}(X + Y + X + Z). \end{aligned}$$

3. Joint Gaussian Probability

Let $X \sim \mathcal{N}(1, 1)$ and $Y \sim \mathcal{N}(0, 1)$ be jointly Gaussian with covariance ρ . What is $\mathbb{P}(X > Y)$?

Solution:

Let $\bar{X} = X - 1$.

We can write $Y = \rho\bar{X} + \sqrt{1 - \rho^2}Z$, where $Z \sim \mathcal{N}(0, 1)$ is independent of \bar{X} . (To check that this is correct, observe that $\text{cov}(\bar{X}, \rho\bar{X} + \sqrt{1 - \rho^2}Z) = \rho$ and also $\text{var}(\rho\bar{X} + \sqrt{1 - \rho^2}Z) = \rho^2 + (1 - \rho^2) = 1$ as required.)

So, $\mathbb{P}(X > Y) = \mathbb{P}(\bar{X} > Y - 1) = \mathbb{P}((1 - \rho)\bar{X} - \sqrt{1 - \rho^2}Z > -1)$. But

$$(1 - \rho)\bar{X} - \sqrt{1 - \rho^2}Z \sim \mathcal{N}(0, (1 - \rho)^2 + 1 - \rho^2) = \mathcal{N}(0, 2(1 - \rho))$$

by independence so

$$\mathbb{P}(X > Y) = \mathbb{P}\left(\mathcal{N}(0, 1) > -\frac{1}{\sqrt{2(1 - \rho)}}\right) = \Phi\left(\frac{1}{\sqrt{2(1 - \rho)}}\right).$$