

Assignment 9 Papoulis ex 6.63

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Question

For any two random variables x and y , let

$$\sigma_x^2 = \text{Var}\{x\}, \sigma_y^2 = \text{Var}\{y\} \text{ and } \sigma_{x+y}^2 = \text{Var}\{x+y\}$$

(a) show that

$$\frac{\sigma_{x+y}}{\sigma_x + \sigma_y} \leq 1$$

(b) More generally, show that for $p \geq 1$

$$\frac{\{E(|x+y|^p)\}^{1/p}}{\{E(|x|^p)\}^{1/p} + \{E(|y|^p)\}^{1/p}} \leq 1$$

Solution (a)

For any two random variables X and Y we have

$$\begin{aligned}\sigma_{X+Y} &= \text{Var}(X + Y) = E \left[\{(X - \mu_X) + (Y - \mu_Y)\}^2 \right] \\ &= \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y) = \sigma_X^2 + \sigma_Y^2 + 2\sigma_X\sigma_Y\rho_{XY} \\ &\leq (\sigma_X + \sigma_Y)^2\end{aligned}$$

Since $|\rho_{XY}| \leq 1$. Thus

$$\sigma_{X+Y} \leq \sigma_X + \sigma_Y,$$

and hence it easily follows that

$$\frac{\sigma_{X+Y}}{\sigma_X + \sigma_Y} \leq 1.$$

Solution (b)

We shall prove this result in three parts by making use of Holder's inequality.

(i) **Holder's inequality:** The function $\log x$ is concave, for $0 < \alpha < 1$, and hence we have

$$\log [\alpha x_1 + (1 - \alpha)x_2] \geq \alpha \log x_1 + (1 - \alpha) \log x_2$$

or

$$x_1^\alpha x_2^{1-\alpha} \leq \alpha x_1 + (1 - \alpha)x_2, \quad 0 < \alpha < 1 \quad (3.1)$$

Let

$$x_1 = |x|^p, \quad \alpha = \frac{1}{p}, \text{ so that } 1 - \alpha = 1 - \frac{1}{p} \triangleq \frac{1}{q}, \quad x_2 = |y|^q \quad (3.2)$$

so that (3.1) becomes

$$|xy| \leq \frac{|x|^p}{p} + \frac{|y|^q}{q}, p > 1, \quad (3.3)$$

the Holder's inequality. From (3.2), note that

$$\frac{1}{p} + \frac{1}{q} = 1, \quad p > 1, \quad q > 1 \quad (3.4)$$

(ii) Define

$$x = X (E \{|X|^p\})^{-1/p}, \quad y = Y (E \{|Y|^q\})^{-1/q}$$

where p and q are as in (3.4). Substituting these into the Holder's inequality in (3.3), we get

$$\begin{aligned} |XY| &\leq p^{-1} |X|^p (E \{|X|^p\})^{1/p-1} (E \{|Y|^q\})^{1/q} \\ &\quad + q^{-1} |Y|^q (E \{|Y|^q\})^{1/q-1} (E \{|X|^p\})^{1/p} \end{aligned} \quad (3.5)$$

Taking expected values on both sides of (3.5), we get

$$E \{|XY|\} \leq (E \{|X|^p\})^{1/p} (E \{|Y|^q\})^{1/q} \quad (3.6)$$

which represents the generalization of the Cauchy-Schwarz inequality.

(Note $p=q=2$ corresponds to Cauchy-Schwarz inequality) (iii) To prove the desired inequality, notice that

$$\begin{aligned} |X + Y|^p &= |X + Y| |X + Y|^{p-1} \\ &\leq |X| |X + Y|^{p-1} + |Y| |X + Y|^{p-1}, \quad p > 1 \end{aligned}$$

and taking the expected values on both sides we get

$$E \{|X + Y|^p\} \leq E \{|X| |X + Y|^{p-1}\} + E \{|Y| |X + Y|^{p-1}\} \quad (3.7)$$

Applying (3.6) to each term on the right side of (3.7) we get

$$E \{|X| |X + Y|^{p-1}\} \leq (E \{|X|^p\})^{1/p} \left(E \{|X + Y|^{(p-1)q}\} \right)^{1/q} \quad (3.8)$$

and

$$E \{ |Y| |X + Y|^{p-1} \} \leq (E \{ |Y|^p \})^{1/p} \left(E \{ |X + Y|^{(p-1)q} \} \right)^{1/q} \quad (3.9)$$

Using (3.8) and (3.9) together with $(p-1)q = p$ in (3.7) we get

$$E \{ |X + Y|^p \} \leq \left[(E \{ |X|^p \})^{1/p} + (E \{ |Y|^p \})^{1/p} \right] \cdot \left(E \{ |X + Y|^{(p-1)q} \} \right)^{1/q}$$

or for $p > 1$

$$(E \{ |X + Y|^p \})^{1/p} \leq (E \{ |X|^p \})^{1/p} + (E \{ |Y|^p \})^{1/p}$$

the desired inequality. Since $p = 1$ follows trivially, we get

$$\frac{(E \{ |X + Y|^p \})^{1/p}}{(E \{ |X|^p \})^{1/p} + (E \{ |Y|^p \})^{1/p}} \leq 1, \quad p \geq 1$$