

VE414 Appendix 6

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Law of total expectation

Let $f_{X,Y}(x,y)$ be the joint probability mass/density function of X and Y , and

$$g(x,y)$$

be a smooth function of x and y , and the following expectations exist, then

$$\mathbb{E}[g(X,Y)] = \mathbb{E}[\mathbb{E}[g(X,Y) | Y]] = \mathbb{E}[\mathbb{E}[g(X,Y) | X]]$$

Proof

- Suppose X and Y are continuous, then by definition of expectation, we have

$$\begin{aligned}\mathbb{E}[g(X,Y)] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{X,Y}(x,y) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{X|Y}(x|y) \cdot f_Y(y) dx dy\end{aligned}$$

where the definition of condition PDF, and marginal PDF are used.

Proof

- Since f_Y does not depend on y ,

$$\begin{aligned}\mathbb{E}[g(X, Y)] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X|Y}(x | y) \cdot f_Y(y) dx dy \\ &= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} g(x, y) f_{X|Y}(x | y) dx \right) \cdot f_Y(y) dy\end{aligned}$$

- By the definition of conditional expectation, we have

$$\mathbb{E}[g(X, Y)] = \int_{-\infty}^{\infty} \mathbb{E}[g(X, Y) | Y] \cdot f_Y(y) dy = \mathbb{E}[\mathbb{E}[g(X, Y) | Y]]$$

- Of course, we could condition on X , and reach the following

$$\mathbb{E}[g(X, Y)] = \mathbb{E}[\mathbb{E}[g(X, Y) | X]]$$

- If X or Y is discrete, we would have summations instead of integrations in the above steps, but the idea is essentially the same.

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