

# Markov's Inequality Proof

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## 1 Exercise

Let  $(\Omega, \mathbf{F}, \mathbf{P})$  be a probability space, and let  $\mathbf{X} : \Omega \rightarrow \mathbf{R}$  be a non-negative random variable. Then for any  $\alpha > 0$  the following inequality holds:

$$\mathbf{P}(\mathbf{X} > \alpha) \leq \frac{1}{\alpha} \mathbf{E}(\mathbf{X})$$

Proof: From Definition 1.2.1:

$$\mathbf{E}(\mathbf{X}) := \int_{\Lambda} \mathbf{x} \mu(\mathbf{x}) d\mathbf{x}$$

As  $\mathbf{X}$  is non-negative variable,  $\Lambda$  could be rewritten as  $[0, \infty)$

$$\mathbf{E}(\mathbf{X}) := \int_0^{\infty} \mathbf{x} \mu(\mathbf{x}) d\mathbf{x}$$

As  $\alpha > 0$ , we can write that  $\int_0^{\infty} \mathbf{x} \mu(\mathbf{x}) d\mathbf{x} \geq \int_{\alpha}^{\infty} \mathbf{x} \mu(\mathbf{x}) d\mathbf{x}$  (1)

Looking at the left side of the inequality, we are particularly interested in  $\mathbf{X} > \alpha$ ,

It follows from (1) that

$$\mathbf{E}(\mathbf{X}) \geq \int_{\alpha}^{\infty} \mathbf{x} \mu(\mathbf{x}) d\mathbf{x} \geq \int_{\alpha}^{\infty} \alpha \mu(\mathbf{x}) d\mathbf{x} = \alpha \int_{\alpha}^{\infty} \mu(\mathbf{x}) d\mathbf{x} \quad (2)$$

At the same time

$$\alpha \int_{\alpha}^{\infty} \mu(\mathbf{x}) d\mathbf{x} = \alpha \mathbf{P}(\mathbf{X} > \alpha)$$

Rewriting (2):

$$\begin{aligned} \Leftrightarrow \mathbf{E}(\mathbf{X}) &\geq \alpha \mathbf{P}(\mathbf{X} > \alpha) \\ \Leftrightarrow \frac{1}{\alpha} \mathbf{E}(\mathbf{X}) &\geq \mathbf{P}(\mathbf{X} > \alpha) \end{aligned}$$