

## 5. Convex Functions

In the following theorem, the book uses  $[0, 1]$  with Lebesgue measure for the probability space.

**THEOREM 5.1 (Jensen's Inequality).** *Let  $(X, \mathcal{A}, \mu)$  be a probability space (i.e.,  $\mu(X) = 1$ ), and fix an integrable function  $f$  on  $X$  with range contained in an open interval  $I$ . Let  $\varphi$  be a convex function on  $I$ . Then*

$$\varphi\left(\int_X f \, d\mu\right) \leq \int_X \varphi \circ f \, d\mu.$$

Note : It may be that  $\varphi \circ f$  is not integrable; in this case, the integral on the right equals  $+\infty$ .

**PROOF.** Let  $y = \int_X f \, d\mu$ . Then  $y \in I$  since  $y$  is a "general average" of the values of  $f$ . Another way to see this is to note that if  $a$  is the left endpoint of  $I$  and  $a$  is finite, then  $a = a \cdot \mu(X) \leq \int_X f \, d\mu$ , and equality means that  $f$  takes the value  $a$  almost everywhere, which it does not. A similar proof holds for a finite right endpoint of  $I$ . Now, if

$$\beta = \sup_{\substack{x < y \\ x \in I}} \frac{\varphi(y) - \varphi(x)}{y - x}$$

then for any  $z > y$  in  $I$ ,

$$\beta \leq \frac{\varphi(z) - \varphi(y)}{z - y}$$

It follows that for  $z \geq y$  in  $I$ ,

$$*) \quad \varphi(z) \geq \varphi(y) + \beta \cdot (z - y).$$

If  $z < y$  in  $I$ , then by the definition of  $\beta$ ,

$$\begin{aligned} \beta &\geq \frac{\varphi(y) - \varphi(z)}{y - z} = \frac{\varphi(z) - \varphi(y)}{z - y}, \text{ so} \\ \varphi(z) - \varphi(y) &\geq (z - y)\beta. \end{aligned}$$

This means that (\*) holds for all  $z$  in  $I$ . For any  $t \in X$ , we may let  $z = f(t)$ , and we get

$$\varphi(f(t)) \geq \varphi(y) + \beta(f(t) - y).$$

Recall  $\varphi$  is continuous so  $\varphi \circ f$  is measurable. The inequality means that  $\varphi(f(t))$  is bounded below by an integrable function. Integrating both sides of the inequality, we get

$$\int_X \varphi \circ f \, d\mu \geq \varphi\left(\int_X f \, d\mu\right) + \beta \cdot 0. \quad \blacksquare$$

EXAMPLE 5.2. A primary example is given by setting  $\varphi(x) = e^x$  on  $I = \mathbb{R}$ . If  $X$  is a set consisting of  $n$  points with each point having probability  $1/n$ , then letting  $x_i$  be the image under  $f$  of the  $n^{\text{th}}$  point, we have

$$\exp\left(\frac{x_1 + \cdots + x_n}{n}\right) \leq \frac{e^{x_1} + \cdots + e^{x_n}}{n}.$$

Putting  $y_i = e^{x_i}$ , this gives the classical inequality between the arithmetic and geometric mean for positive numbers:

$$(y_1 y_2 \cdots y_n)^{\frac{1}{n}} \leq \frac{y_1 + y_2 + \cdots + y_n}{n}.$$

A generalization works with positive weights  $\alpha_i$  such that  $\sum \alpha_i = 1$ . This gives

$$y_1^{\alpha_1} \cdots y_n^{\alpha_n} \leq \alpha_1 y_1 + \cdots + \alpha_n y_n.$$

For a countable number of points, we have

$$\prod_{n=1}^{\infty} y_i^{\alpha_i} \leq \sum_{n=1}^{\infty} \alpha_i y_i.$$

Infinite products are covered in the graduate complex variables course.