

## A3 Characteristic function

The characteristic function  $\Phi(t)$  of a probability density function  $f(x)$ :

$$\Phi(t) \stackrel{\text{def}}{=} E[e^{itx}] = \int_{-\infty}^{\infty} e^{itx} f(x) dx \quad (\text{A3.1})$$

has some interesting properties. In fact,  $\Phi(t)$  is the Fourier transform of  $f(x)$ . This implies that the characteristic function of the *convolution*  $f_1 * f_2$  of two density functions  $f_1$  and  $f_2$  is the *product* of the two corresponding characteristic functions  $\Phi_1$  and  $\Phi_2$ . The convolution, defined by

$$f_1 * f_2(x) = \int_{-\infty}^{\infty} f_1(x - \xi) f_2(\xi) d\xi, \quad (\text{A3.2})$$

is the density distribution of the *sum* of two random variables  $x_1 + x_2$  when the density functions of  $x_1$  and  $x_2$  are resp.  $f_1$  and  $f_2$ . The *convolution theorem* of Fourier analysis states that the Fourier transform of a convolution equals the product of the Fourier transforms of the contributing terms. This product rule also applies to convolutions of  $n$  functions.

Another interesting property of the characteristic function is that its series expansion in powers of  $t$  generates the *moments* of the distribution. The characteristic function is therefore often called the *moment-generating function*. Since

$$e^{itx} = \sum_{n=0}^{\infty} \frac{(itx)^n}{n!}, \quad (\text{A3.3})$$

it follows that

$$\Phi(x) = E[e^{itx}] = \sum_{n=0}^{\infty} \frac{(it)^n}{n!} E[x^n] = \sum_{n=0}^{\infty} \frac{(it)^n}{n!} \mu_n. \quad (\text{A3.4})$$

The moments are also given by the *derivatives* of the characteristic function at  $t = 0$ :

$$\Phi^{(n)}(0) = \frac{d^n \Phi}{dt^n} \Big|_{t=0} = i^n \mu_n. \quad (\text{A3.5})$$

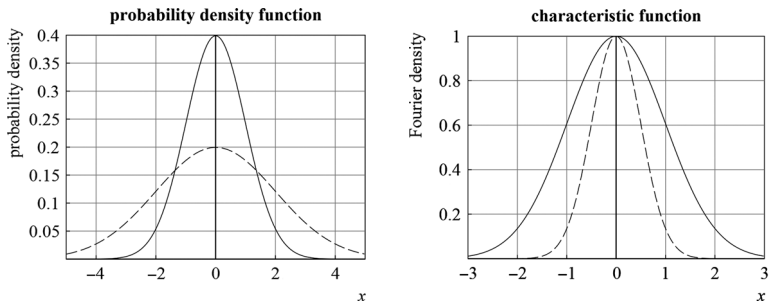


Figure A3.1 Left: a probability density function (in this case a normal distribution); right: its characteristic function. The dashed curve has a standard deviation twice that of the drawn curve.

The  $\mu_n$  are the moments, not the central moments. But you can always choose the origin of  $x$  at the position of the mean.

A special case is the variance  $\sigma^2$ :

$$\sigma^2 = -\frac{d^2\Phi}{dt^2}(0). \quad (\text{A3.6})$$

Figure A3.1 shows the relation between a density function and its characteristic function. The density function is normalized by its integral; the characteristic function is always equal to 1 for  $t = 0$ . The broader the density function, the narrower the characteristic function.