

**Definition.** A random variable  $X$  has a **uniform continuous distribution** with parameters  $\alpha$  and  $\beta$  (with  $\alpha < \beta$ ) if and only if the following function is a probability density for  $X$ :

$$f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta - \alpha} & \text{if } \alpha < x < \beta \\ 0 & \text{elsewhere} \end{cases}.$$

**Definition.** The **gamma function** is defined as  $\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy$  for  $\alpha > 0$ .

**Proposition.** For any positive integer  $n$ ,  $\Gamma(n) = (n-1)!$

**Definition.** A random variable  $X$  has a **gamma distribution** with parameters  $\alpha > 0$  and  $\beta > 0$  if and only if the following function is a probability density for  $X$ :

$$g(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-\frac{x}{\beta}} & \text{for } x > 0 \\ 0 & \text{elsewhere} \end{cases}.$$

**Proposition.** A gamma distribution with parameters  $\alpha$  and  $\beta$  has moment-generating function  $M_X(t) = (1 - \beta t)^{-\alpha}$ , mean  $\mu = \alpha\beta$ , and variance  $\sigma^2 = \alpha\beta^2$ .

**Definition.** A random variable  $X$  has an **exponential distribution** with parameter  $\theta > 0$  if and only if the following function is a probability density for  $X$ :

$$g(x; \theta) = \begin{cases} \frac{1}{\theta} e^{-\frac{x}{\theta}} & \text{if } x > 0 \\ 0 & \text{elsewhere} \end{cases} \quad (\text{a gamma distribution with } \alpha = 1 \text{ and } \beta = \theta).$$

**Proposition.** An exponential distribution with parameter  $\theta$  has mean  $\mu = \theta$  and variance  $\sigma^2 = \theta^2$ .

**Definition.** A random variable  $X$  has a **chi-square distribution** with parameter  $\nu > 0$  if and only if the following function is a probability density for  $X$ :

$$f(x) = \begin{cases} \frac{1}{2^{\nu/2} \Gamma(\nu/2)} x^{\frac{\nu-2}{2}} e^{-\frac{x}{2}} & \text{if } x > 0 \\ 0 & \text{elsewhere} \end{cases} \quad (\text{a gamma distribution with } \alpha = \frac{\nu}{2} \text{ and } \beta = 2).$$

**Proposition.** A chi-square distribution with parameter  $\nu$  has mean  $\mu = \nu$  and variance  $\sigma^2 = 2\nu$ .

**Definition.** A random variable  $X$  has a **beta distribution** with parameters  $\alpha > 0$  and  $\beta > 0$  if and only if the following function is a probability density for  $X$ :

$$f(x) = \begin{cases} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} & \text{if } 0 < x < 1 \\ 0 & \text{elsewhere} \end{cases}.$$

**Proposition.** A beta distribution with parameters  $\alpha$  and  $\beta$  has mean  $\mu = \frac{\alpha}{\alpha + \beta}$  and variance  $\sigma^2 = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$ .

**Definition.** A random variable  $X$  has a **normal distribution** with parameters  $\mu$  and  $\sigma > 0$  if and only if the following function is a probability density for the  $X$ :

$$n(x; \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \text{ for all } x \in \mathbb{R}.$$

**Proposition.** A normal distribution with parameters  $\mu$  and  $\sigma$  has a moment-generating function  $M_X(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2}$ , mean  $\mu = \mu$  and variance  $\sigma^2 = \sigma^2$ .

Of these distributions, only the uniform continuous distribution and the exponential distribution allow one to compute probabilities by hand. Calculations of probabilities for the rest of the distributions generally rely on a table or computational device. For a normally distributed random variable  $X$  this usually means **standardizing** the random

variable:  $Z = \frac{X - \mu}{\sigma}$  has a **standard normal distribution** with mean 0 and variance 1 (cf. homework problem 4.23).