

Math/Stat 251 Fall 2015  
Summary of Continuous Random Variables

**Example 1.** The random variable  $X$  has an *exponential distribution with parameter*  $\lambda > 0$  if the density of  $X$  is

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0, \\ 0, & x < 0. \end{cases}$$

The distribution function of  $X$  is

$$F(x) = \begin{cases} 1 - e^{-\lambda x}, & x \geq 0, \\ 0, & x < 0. \end{cases}$$

We often write  $X \sim \text{Exp}(\lambda)$  for such a random variable.

**Example 2.** The random variable  $X$  has a *uniform distribution on*  $[a, b]$ ,  $-\infty < a < b < \infty$ , if the density of  $X$  is

$$f(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b, \\ 0, & \text{otherwise.} \end{cases}$$

The distribution function of  $X$  is

$$F(x) = \begin{cases} 0, & x < a, \\ \frac{x-a}{b-a}, & a \leq x \leq b, \\ 1, & x > b. \end{cases}$$

We often write  $X \sim \text{Unif}(a, b)$  for such a random variable.

**Example 3.** The random variable  $X$  has a *Cauchy distribution with parameter*  $\theta \in \mathbb{R}$  if the density of  $X$  is

$$f(x) = \frac{1}{\pi} \cdot \frac{1}{1 + (x - \theta)^2}$$

for  $-\infty < x < \infty$ . The distribution function of  $X$  is

$$F(x) = \frac{1}{2} + \frac{1}{\pi} \arctan(x - \theta)$$

for  $-\infty < x < \infty$ . We often write  $X \sim \text{Cauchy}(\theta)$  for such a random variable.

**Example 4.** The random variable  $X$  has a *Normal distribution with parameters*  $\mu \in \mathbb{R}$  and  $\sigma > 0$  if the density of  $X$  is

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp \left\{ -\frac{(x - \mu)^2}{2\sigma^2} \right\}$$

for  $-\infty < x < \infty$ . The distribution function of  $X$  is

$$F(x) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^x \exp \left\{ -\frac{(t - \mu)^2}{2\sigma^2} \right\} dt$$

for  $-\infty < x < \infty$ .

Note that there is no closed-form expression for the distribution function of a normal random variable. In order to evaluate  $F(x)$  for a particular  $x$  it is necessary to resort to a numerical approximation. This is why tables of normal probabilities have been compiled. Also note that sometimes  $\Phi$  is used for the normal distribution function so that  $\Phi(x) = F(x)$ . We often write  $X \sim \mathcal{N}(\mu, \sigma^2)$  for such a random variable.

**Example 5.** The random variable  $X$  has a *Gamma distribution with parameters*  $\alpha > 0$  and  $\lambda > 0$  if the density of  $X$  is

$$f(x) = \begin{cases} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x}, & x \geq 0, \\ 0, & x < 0. \end{cases}$$

Here,  $\Gamma$  is the gamma function defined by

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx.$$

We often write  $X \sim \text{Gamma}(\alpha, \lambda)$  for such a random variable. The distribution function of  $X$  cannot be evaluated in closed-form and so it is rarely used.

Let  $n = 1, 2, \dots$ . We sometimes say that the random variable  $X$  has a *Chi-square distribution with  $n$  degrees of freedom* and write  $X \sim \chi^2(n)$  if  $X \sim \text{Gamma}(n/2, 1/2)$ .

Also note that  $X \sim \text{Exp}(\lambda)$  if and only if  $X \sim \text{Gamma}(1, \lambda)$ .

**Example 6.** The random variable  $X$  has a *Beta distribution with parameters*  $a > 0$  and  $b > 0$  if the density of  $X$  is

$$f(x) = \begin{cases} \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} x^{a-1} (1-x)^{b-1}, & 0 \leq x \leq 1, \\ 0, & \text{otherwise.} \end{cases}$$

We often write  $X \sim \text{Beta}(a, b)$  for such a random variable. The distribution function of  $X$  cannot be evaluated in closed-form and so it is rarely used.

Note that  $X \sim \text{Unif}(0, 1)$  if and only if  $X \sim \text{Beta}(1, 1)$ .