$$\begin{split} F_X(x) &= \int_{-\infty}^x f(t) dt \\ P(a < X \le b) &= F(b) - F(a), \quad a < b \\ \mu_n &= \int_{-\infty}^\infty (x - c)^n f(x) d_x \\ \mu_1 &= E(x) = \int_{-\infty}^\infty x f(x) dx \\ V(X) &= \sigma^2 = \frac{\sum (x_i - mean)^2}{n} = E\left[\left(x - E(X)\right)^2\right] \\ V(X) &= \frac{\sum x_i^2}{n} - \left(\frac{\sum x_i}{n}\right)^2 = E(X^2) - E(X)^2 \\ P_{ber}(x) &= \begin{cases} p, & x = 1 \\ 1 - p, & x = 0 \end{cases}, \quad E[X] = p, \quad V(X) = pq \\ P_{bin}(x) &= \binom{n}{x} p^x q^{n-x}, \quad E[X] = np, \quad V(X) = npq \\ P_{geo}(x) &= q^{x-1}p, \quad E[X] = \frac{1}{p}, \quad V(X) = \frac{q}{p^2} \\ P_{-bin}(x) &= \left(\frac{x - 1}{k - 1}\right) q^{x-k} p^k, \quad E[X] = \frac{k}{p}, \quad V(X) = \frac{kq}{p^2} \\ P_{poi}(x) &= \frac{e^{-\alpha} \alpha^x}{x!}, \quad \alpha = \lambda t, \quad E[X] = V(X) = \alpha \\ f_{uni}(x) &= \frac{1}{n} = \frac{1}{b-a}, \quad a \le x \le b \\ f_{exp}(x) &= \lambda e^{-\lambda x}, (x \ge 0), \quad \theta = \frac{1}{\lambda}; E(X) = \theta; V(X) = \theta^2 \\ F_{x}(x) &= 1 - e^{-\lambda x} \\ f_{gam}(x) &= \frac{\beta \theta}{\Gamma(\beta)} (\beta \theta x)^{\beta - 1} e^{-\beta \theta x}, \quad \Gamma(\beta) &= \int_0^\infty x^{\beta - 1} e^{-x} dx \\ \Gamma(\beta) &= (\beta - 1)!, \quad E(X) &= \frac{1}{\theta}, \quad V(X) &= \frac{1}{\beta \theta^2} \\ F_{erl}(x) &= 1 - \sum_{l=0}^{k-1} \frac{e^{-k\theta x} (k\theta x)^l}{i!}; E(X) &= \frac{1}{\theta}; V(X) &= \frac{1}{k\theta^2} \\ f_{nor}(x) &= \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x - \mu}{\sigma}\right)^2}, \quad Z &= \frac{X - \mu}{\sigma} \end{split}$$

$$f_{gam}(x) = \frac{\beta \theta}{\Gamma(\beta)} (\beta \theta x)^{\beta - 1} e^{-\beta \theta x}, \qquad \Gamma(\beta) = \int_{0}^{\infty} x^{\beta - 1} e^{-x} dx$$

$$, \Gamma(\beta) = (\beta - 1)!, \qquad E(X) = \frac{1}{\theta}, \qquad V(X) = \frac{1}{\beta \theta^{2}}$$

$$F_{erl}(x) = 1 - \sum_{i=0}^{k-1} \frac{e^{-k\theta x} (k\theta x)^{i}}{i!}; \quad E(X) = \frac{1}{\theta}; \quad V(X) = \frac{1}{k\theta^{2}}$$

$$f_{nor}(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x - \mu}{\sigma}\right)^{2}}, \qquad Z = \frac{X - \mu}{\sigma}$$

$$f_{wei}(x) = \frac{\beta}{\alpha} \left(\frac{x - \nu}{\alpha}\right)^{\beta - 1} \exp\left[-\left(\frac{x - \nu}{\alpha}\right)^{\beta}\right]$$

$$, E(X) = \nu + \alpha \Gamma\left(\frac{1}{\beta} + 1\right), V(X) = \alpha^{2} \left[\Gamma\left(\frac{2}{\beta} + 1\right) - \Gamma\left(\frac{1}{\beta} + 1\right)^{2}\right]$$

$$\begin{split} f_{Inor}(x) &= \frac{1}{\sqrt{2\pi}\sigma x} \exp\left[-\frac{(\ln x - \mu)^2}{2\sigma^2}\right] \\ f_{Inor}(x) &= \frac{1}{\sqrt{2\pi}\sigma x} \exp\left[-\frac{(\ln x - \mu)^2}{2\sigma^2}\right] \\ f_{EX}(x) &= e^{\mu + \frac{\sigma^2}{2}}, \quad V(X) = e^{2\mu + \frac{\sigma^2}{2}}(e^{\sigma^2} - 1) \\ f_{tri}(x) &= \begin{cases} \frac{2(x - a)}{(b - a)(c - a)}, a \le x \le b \\ 2(c - x), b < x \le c \end{cases} \\ f_{EX}(x) &= \begin{cases} \frac{(x - a)^2}{(b - a)(c - a)}, a < x \le b \\ 1 - \frac{(c - x)^2}{(c - b)(c - a)}, b < x \le c \end{cases} \\ f_{bet}(x) &= \frac{x^{\beta_1 - 1}(1 - x)^{\beta_2 - 1}}{8(\beta_1, \beta_2)}, \quad B(\beta_1, \beta_2) = \frac{\Gamma(\beta_1)\Gamma(\beta_2)}{\Gamma(\beta_1 + \beta_2)} \\ f_{EX}(x) &= \frac{\beta_1}{\beta_1 + \beta_2}, \quad V(X) = \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1}{\beta_1 + \beta_2}, \quad V(X) = \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1}{\beta_1 + \beta_2}, \quad V(X) = \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1}{\beta_1 + \beta_2}, \quad V(X) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &= \frac{\beta_1\beta_2}{(\beta_1 + \beta_2)^2(\beta_1 + \beta_2 + 1)} \\ f_{EX}(x) &=$$