

Mean, Variance, and Moment Function of Discrete Distributions

Bernoulli $f(x) = p^x(1-p)^{1-x}, \quad x = 0, 1$

$$M(t) = 1 - p + pe^t; \quad \mu = p, \quad \sigma^2 = p(1-p)$$

Binomial $f(x) = \frac{n!}{x!(n-x)!}p^x(1-p)^{n-x}, \quad x = 0, 1, 2, \dots, n$

$$b(n, p) \quad M(t) = (1-p+pe^t)^n; \quad \mu = np, \quad \sigma^2 = np(1-p)$$

Geometric $f(x) = (1-p)^{x-1}p, \quad x = 1, 2, \dots$

$$M(t) = \frac{pe^t}{1-(1-p)e^t}, \quad t < -\ln(1-p)$$

$$\mu = \frac{1}{p}, \quad \sigma^2 = \frac{1-p}{p^2}$$

Hypergeometric $f(x) = \frac{C(N_1, x)C(N_2, n-x)}{C(N, n)}, \quad x \leq n, x \leq N_1, n-x \leq N_2$

$$M(t) = \times$$

$$\mu = n \left(\frac{N_1}{N} \right), \quad \sigma^2 = n \left(\frac{N_1}{N} \right) \left(\frac{N_2}{N} \right) \left(\frac{N-n}{N-1} \right)$$

Negative Binomial $f(x) = C(x-1, r-1)p^r(1-p)^{x-r}, \quad x = r, r+1, r+2, \dots$

$$M(t) = \frac{(pe^t)^r}{[1-(1-p)e^t]^r} \quad t < -\ln(1-p)$$

$$\mu = \frac{r}{p}, \quad \sigma^2 = \frac{r(1-p)}{p^2}$$

Poisson $f(x) = \frac{\lambda^x e^{-\lambda}}{x!}, \quad x = 0, 1, 2, \dots$

$$M(t) = e^{\lambda(e^t-1)}; \quad \mu = \lambda, \quad \sigma^2 = \lambda$$

Uniform $f(x) = \frac{1}{m}, \quad x = 1, 2, \dots$

$$M(t) = \frac{1}{m} \cdot \frac{e^t(1-e^{mt})}{1-e^t}; \quad \mu = \frac{m+1}{2}, \quad \sigma^2 = \frac{m^2-1}{12}$$

Uniform, Exponential, Gamma, Chi-Square, Normal, Beta, Student-t, and F Distributions

Uniform $U(a, b)$ $f(x) = \frac{1}{b-a}$, $a \leq x \leq b$

$$E(X) = \frac{a+b}{2}, \text{Var}(X) = \frac{b^2-a^2}{12}, M(t) = \frac{e^{tb}-e^{ta}}{t(b-a)}, t \neq 0.$$

Exponential $f(x) = \frac{1}{\theta}e^{-x/\theta}$, $0 < x < \infty$

$$M(t) = \frac{1}{1-\theta t}, t < \frac{1}{\theta}, E(X) = \theta, \text{Var}(X) = \theta^2.$$

Gamma $f(x) = \frac{1}{\Gamma(\alpha)\theta^\alpha}x^{\alpha-1}e^{-x/\theta}$, $0 < x < \infty$, $\alpha > 0$.

$$M(t) = \frac{1}{(1-\theta t)^\alpha}, t < \frac{1}{\theta}, E(X) = \alpha\theta, \text{Var}(X) = \alpha\theta^2.$$

$\chi^2(r)$ **Chi-Square** $f(x) = \frac{1}{\Gamma(r/2)2^{r/2}}x^{(r/2)-1}e^{-x/2}$, $0 < x < \infty$

$$M(t) = \frac{1}{(1-2t)^{r/2}}, t < \frac{1}{2}, E(X) = r, \text{Var}(X) = 2r.$$

$N(\mu, \sigma^2)$ **Normal** $f(x) = \frac{1}{\sqrt{2\pi}\sigma}e^{-(x-\mu)^2/(2\sigma^2)}$, $-\infty < x < \infty$

$$M(t) = \phi(t) = e^{\mu t + \frac{\sigma^2 t^2}{2}}, E(X) = \mu, \text{Var}(X) = \sigma^2.$$

Beta Distribution $f(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}x^{\alpha-1}(1-x)^{\beta-1}$, $0 < x < 1$

$$E(X) = \frac{\alpha}{\alpha+\beta}, \text{Var}(X) = \frac{\alpha\beta}{(\alpha+\beta+1)(\alpha+\beta)^2}.$$

Let $Z \sim N(0, 1)$, $X \sim \chi^2(n)$, $Y \sim \chi^2(m)$, define $T = \frac{Z}{\sqrt{\chi^2(n)/n}}$ and $F = \frac{\chi^2(n)/n}{\chi^2(m)/m}$, then

Student-t Distribution $f_T(t) = \frac{\Gamma((n+1)/2)}{\sqrt{\pi n} \Gamma(n/2) [1+(t^2/n)]^{(n+1)/2}}$, $-\infty < t < \infty$

F Distribution $f_F(w) = \frac{\Gamma((n+m)/2)(n/m)^{n/2}w^{(n/2)-1}}{\Gamma(n/2)\Gamma(m/2)[1+nw/m]^{(n+m)/2}}$, $0 < w < \infty$

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2} \quad (1)$$

$$\square \quad (X - \mu)/\sigma \sim N(0, 1)$$

$$\square \quad Z \sim N(0, 1) \Rightarrow Y = Z^2 \sim \chi^2(1)$$

When $\mu = 0$, $\sigma = 1$, $X \sim N(0, 1)$ is said to have the *standard normal distribution*. The cumulative distribution is denoted as

$$\diamond \quad \Phi(z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx, \quad -\infty < z < \infty$$