

Named Probability Distributions

Discrete Probability Distributions

pmf: $p(y)$ cdf: $F(y) = \sum_{z=-\infty}^y p(z)$
 $0 \leq p(y) \leq 1; \sum_{y=-\infty}^{\infty} p(y) = 1$
 $P(Y = y) = p(y); P(a \leq Y \leq b) = \sum_a^b p(y)$

Binomial – $Y \sim \text{Binom}(n, p)$

$p(y) = \frac{n!}{y!(n-y)!} p^y (1-p)^{n-y}, y \in [0, n], p \in [0, 1]$
 $\mathbb{E}[Y] = np$
 $\mathbb{V}[Y] = np(1-p)$
 $m(t) = [pe^t + (1-p)]^n$

Geometric – $Y \sim \text{Geom}(p)$

$p(y) = (1-p)^{y-1} p, y \in [1, \infty), p \in [0, 1]$
 $\mathbb{E}[Y] = 1/p$
 $\mathbb{V}[Y] = (1-p)/p^2$
 $m(t) = \frac{p}{1-qe^t}$

Hypergeometric – $Y \sim \text{HG}(N, K, n)$

$p(y = k) = \frac{\binom{K}{k} \binom{N-K}{n-k}}{\binom{N}{n}}, k \in \{\max(0, n+K-N), \dots, \min(n, K)\}, K \leq N; n \leq N$
 $\mathbb{E}[Y] = \frac{nK}{N}$
 $\mathbb{V}[Y] = \frac{nK(N-K)(N-n)}{N^2(N-1)}$

Negative Binomial – $Y \sim \text{NBinom}(r, p)$

$P(Y = k) = \binom{r+k-1}{r-1} p^r (1-p)^k, k \in [r, \infty), r \in \mathbb{Z}^+, p \in [0, 1]$
 $\mathbb{E}[Y] = rq/p$
 $\mathbb{V}[Y] = rq/p^2$
 $m(t) = (\frac{p}{1-qe^t})^r \text{ for } qe^t < 1$

Poisson – $Y \sim \text{Poi}(\lambda)$

$p(y) = \frac{\lambda^y}{y!} e^{-\lambda}, y \in [0, \infty);$
 $\mathbb{E}[Y] = \mathbb{V}[Y] = \lambda$
 $m(t) = e^{\lambda(e^t-1)}$

Continuous Probability Distributions

pdf: $f(y) = \frac{d}{dy}(y)$ cdf: $F(y) = \int_{-\infty}^y f(z) dz$
 $f(y) \geq 0; \int_{-\infty}^{\infty} f(y) dy = 1; P(Y = y) = 0$
 $P(a \leq Y \leq b) = \int_a^b f(y) dy = F(b) - F(a)$

Uniform – $Y \sim \text{Uniform}(a, b)$

$f(y) = (b-a)^{-1}, y \in [a, b]$
 $\mathbb{E}[Y] = (a+b)/2$
 $\mathbb{V}[Y] = (b-a)^2/12$
 $m(t) = (e^{bt} - e^{at})/[t(b-a)]$

Normal – $Y \sim N(\mu, \sigma^2)$

$f(y) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(y-\mu)^2/2\sigma^2} y \in (-\infty, \infty), \mu \in \mathbb{R}, \sigma \in \mathbb{R}^+$
 $\mathbb{E}[Y] = \mu;$
 $\mathbb{V}[Y] = \sigma^2$
 $m(t) = \exp(\mu t + t^2 \sigma^2/2)$
If $Y \sim N(\mu, \sigma)$, then $Z = (Y - \mu)/\sigma; Z \sim N(0,1).$
 $P(Y \leq y) = \Phi\left(\frac{y-\mu}{\sigma}\right) = \Phi(z)$ (non-analytic function)

Exponential – $Y \sim \text{Exponential}(\lambda)$

$f(y) = \lambda e^{-\lambda y}, y \in [0, \infty), \lambda \in \mathbb{R}^+$
 $\mathbb{E}[Y] = 1/\lambda$
 $\mathbb{V}[Y] = 1/\lambda^2$
 $m(t) = \frac{\lambda}{\lambda-t} \text{ for } t < \lambda$

Gamma – $Y \sim \text{Gamma}(\alpha, \beta)$

$f(y) = y^{\alpha-1} e^{-y/\beta} / [\beta^\alpha \Gamma(\alpha)], y \in [0, \infty), \alpha \in \mathbb{R}^+, \beta \in \mathbb{R}^+$
 $\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy = (\alpha-1)\Gamma(\alpha-1)$
If n is a positive integer, $\Gamma(n) = (n-1)!$
 $\mathbb{E}[Y] = \alpha\beta$
 $\mathbb{V}[Y] = \alpha\beta^2$
 $m(t) = (1-\beta t)^{-\alpha}$
 $\alpha = 1 \Rightarrow \text{exponential distribution}$
 $\beta = 2, \alpha = \nu/2, \nu \in \mathbb{Z}^+ \Rightarrow \text{chi-square distribution}$

Beta – $Y \sim \text{Beta}(\alpha, \beta)$

$f(y) = y^{\alpha-1} (1-y)^{\beta-1} / B(\alpha, \beta), y \in [0, 1], \alpha \in \mathbb{R}^+, \beta \in \mathbb{R}^+$
 $B(\alpha, \beta) = \Gamma(\alpha)\Gamma(\beta)/\Gamma(\alpha+\beta)$
 $\mathbb{E}[Y] = \alpha/(\alpha+\beta)$
 $\mathbb{V}[Y] = \alpha\beta/[(\alpha+\beta)^2(\alpha+\beta+1)]$