

# Appendix A

## Families of Random Variables

### A.1 Discrete Random Variables

#### — Bernoulli $(p)$ —

For  $0 \leq p \leq 1$ ,

$$P_X(x) = \begin{cases} 1-p & x=0 \\ p & x=1 \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = 1-p+pe^s$$

$$E[X] = p$$

$$\text{Var}[X] = p(1-p)$$


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#### — Binomial $(n, p)$ —

For a positive integer  $n$  and  $0 \leq p \leq 1$ ,

$$P_X(x) = \binom{n}{x} p^x (1-p)^{n-x} \quad \phi_X(s) = (1-p+pe^s)^n$$

$$E[X] = np$$

$$\text{Var}[X] = np(1-p)$$


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#### — Discrete Uniform $(k, l)$ —

For integers  $k$  and  $l$  such that  $k < l$ ,

$$P_X(x) = \begin{cases} 1/(l-k+1) & x=k, k+1, \dots, l \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = \frac{e^{sk} - e^{s(l+1)}}{(l-k+1)(1-e^s)}$$

$$E[X] = \frac{k+l}{2}$$

$$\text{Var}[X] = \frac{(l-k)(l-k+1)}{12}$$


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**Geometric** ( $p$ )For  $0 < p \leq 1$ ,

$$P_X(x) = \begin{cases} p(1-p)^{x-1} & x = 1, 2, \dots \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = \frac{pe^s}{1 - (1-p)e^s}$$

$$E[X] = 1/p$$

$$\text{Var}[X] = (1-p)/p^2$$

**Multinomial**For integer  $n > 0$ ,  $p_i \geq 0$  for  $i = 1, \dots, n$ , and  $p_1 + \dots + p_n = 1$ ,

$$P_{X_1, \dots, X_r}(x_1, \dots, x_r) = \binom{n}{x_1, \dots, x_r} p_1^{x_1} \dots p_r^{x_r}$$

$$E[X_i] = np_i$$

$$\text{Var}[X_i] = np_i(1 - p_i)$$

**Pascal** ( $k, p$ )For positive integer  $k$ , and  $0 < p < 1$ ,

$$P_X(x) = \binom{x-1}{k-1} p^k (1-p)^{x-k} \quad \phi_X(s) = \left( \frac{pe^s}{1 - (1-p)e^s} \right)^k$$

$$E[X] = k/p$$

$$\text{Var}[X] = k(1-p)/p^2$$

**Poisson** ( $\alpha$ )For  $\alpha > 0$ ,

$$P_X(x) = \begin{cases} \frac{\alpha^x e^{-\alpha}}{x!} & x = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = e^{\alpha(e^s - 1)}$$

$$E[X] = \alpha$$

$$\text{Var}[X] = \alpha$$

**Zipf**  $(n, \alpha)$ 

For positive integer  $n > 0$  and constant  $\alpha \geq 1$ ,

$$P_X(x) = \begin{cases} \frac{c(n, \alpha)}{x^\alpha} & x = 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

where

$$c(n, \alpha) = \left( \sum_{k=1}^n \frac{1}{k^\alpha} \right)^{-1}$$

**A.2 Continuous Random Variables****Beta**  $(i, j)$ 

For positive integers  $i$  and  $j$ , the beta function is defined as

$$\beta(i, j) = \frac{(i+j-1)!}{(i-1)!(j-1)!}$$

For a  $\beta(i, j)$  random variable  $X$ ,

$$f_X(x) = \begin{cases} \beta(i, j) x^{i-1} (1-x)^{j-1} & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$E[X] = \frac{i}{i+j}$$

$$\text{Var}[X] = \frac{ij}{(i+j)^2(i+j+1)}$$

**Cauchy**  $(a, b)$ 

For constants  $a > 0$  and  $-\infty < b < \infty$ ,

$$f_X(x) = \frac{1}{\pi} \frac{a}{a^2 + (x-b)^2} \quad \phi_X(s) = e^{bs - a|s|}$$

Note that  $E[X]$  is undefined since  $\int_{-\infty}^{\infty} x f_X(x) dx$  is undefined. Since the PDF is symmetric about  $x = b$ , the mean can be defined, in the sense of a principal value, to be  $b$ .

$$\begin{aligned} E[X] &\equiv b \\ \text{Var}[X] &= \infty \end{aligned}$$

**Erlang**  $(n, \lambda)$ 

For  $\lambda > 0$ , and a positive integer  $n$ ,

$$f_X(x) = \begin{cases} \frac{\lambda^n x^{n-1} e^{-\lambda x}}{(n-1)!} & x \geq 0 \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = \left( \frac{\lambda}{\lambda - s} \right)^n$$

$$\begin{aligned} E[X] &= n/\lambda \\ \text{Var}[X] &= n/\lambda^2 \end{aligned}$$

**Exponential**  $(\lambda)$ 

For  $\lambda > 0$ ,

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = \frac{\lambda}{\lambda - s}$$

$$\begin{aligned} E[X] &= 1/\lambda \\ \text{Var}[X] &= 1/\lambda^2 \end{aligned}$$

**Gamma**  $(a, b)$ 

For  $a > -1$  and  $b > 0$ ,

$$f_X(x) = \begin{cases} \frac{x^a e^{-x/b}}{a! b^{a+1}} & x > 0 \\ 0 & \text{otherwise} \end{cases} \quad \phi_X(s) = \frac{1}{(1 - bs)^{a+1}}$$

$$\begin{aligned} E[X] &= (a+1)b \\ \text{Var}[X] &= (a+1)b^2 \end{aligned}$$

**Gaussian**  $(\mu, \sigma)$ 

For constants  $\sigma > 0$ ,  $-\infty < \mu < \infty$ ,

$$f_X(x) = \frac{e^{-(x-\mu)^2/2\sigma^2}}{\sigma\sqrt{2\pi}} \quad \phi_X(s) = e^{s\mu + s^2\sigma^2/2}$$

$$\begin{aligned} E[X] &= \mu \\ \text{Var}[X] &= \sigma^2 \end{aligned}$$

**Laplace**  $(a, b)$ 

For constants  $a > 0$  and  $-\infty < b < \infty$ ,

$$\begin{aligned} f_X(x) &= \frac{a}{2} e^{-a|x-b|} & \phi_X(s) &= \frac{a^2 e^{bs}}{a^2 - s^2} \\ E[X] &= b \\ \text{Var}[X] &= 2/a^2 \end{aligned}$$


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**Log-normal**  $(a, b, \sigma)$ 

For constants  $-\infty < a < \infty$ ,  $-\infty < b < \infty$ , and  $\sigma > 0$ ,

$$\begin{aligned} f_X(x) &= \begin{cases} \frac{e^{-(\ln(x-a)-b)^2/2\sigma^2}}{\sqrt{2\pi}\sigma(x-a)} & x > a \\ 0 & \text{otherwise} \end{cases} \\ E[X] &= a + e^{b+\sigma^2/2} \\ \text{Var}[X] &= e^{2b+\sigma^2} (e^{\sigma^2} - 1) \end{aligned}$$


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**Maxwell**  $(a)$ 

For  $a > 0$ ,

$$\begin{aligned} f_X(x) &= \begin{cases} \sqrt{2/\pi} a^3 x^2 e^{-a^2 x^2/2} & x > 0 \\ 0 & \text{otherwise} \end{cases} \\ E[X] &= \sqrt{\frac{8}{a^2 \pi}} \\ \text{Var}[X] &= \frac{3\pi - 8}{\pi a^2} \end{aligned}$$


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**Pareto**  $(\alpha, \mu)$ 

For  $\alpha > 0$  and  $\mu > 0$ ,

$$\begin{aligned} f_X(x) &= \begin{cases} (\alpha/\mu) (x/\mu)^{-(\alpha+1)} & x \geq \mu \\ 0 & \text{otherwise} \end{cases} \\ E[X] &= \frac{\alpha\mu}{\alpha-1} & (\alpha > 1) \\ \text{Var}[X] &= \frac{\alpha\mu^2}{(\alpha-2)(\alpha-1)^2} & (\alpha > 2) \end{aligned}$$


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**Rayleigh** ( $a$ )For  $a > 0$ ,

$$f_X(x) = \begin{cases} a^2 x e^{-a^2 x^2/2} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$E[X] = \sqrt{\frac{\pi}{2a^2}}$$

$$\text{Var}[X] = \frac{2 - \pi/2}{a^2}$$

**Uniform** ( $a, b$ )For constants  $a < b$ ,

$$f_X(x) = \begin{cases} \frac{1}{b-a} & a < x < b \\ 0 & \text{otherwise} \end{cases}$$

$$\phi_X(s) = \frac{e^{bs} - e^{as}}{s(b-a)}$$

$$E[X] = \frac{a+b}{2}$$

$$\text{Var}[X] = \frac{(b-a)^2}{12}$$