### 4.1 4.2 Joint, Marginal, Conditional

$$\begin{split} &f(x,y)=f_X(x)f_Y(y)\ 4.2.5-\operatorname{Indep}\ f(x,y)=g(x)h(y)\ 4.2.7\\ &F(x,y)=P(X\leq x,Y\leq y)=\int_{-\infty}^y\int_{-\infty}^x f(s,t)dsdt-\operatorname{Indep}\ F_{X,Y}(x,y)=F_X(x)F_Y(y)\\ &f(x|y)=P(X=x|Y=y)=\frac{f(x,y)}{f_Y(y)}\ 4.2.1\\ &f_X(x)=\sum_{y\in\mathbf{R}}f_{X,Y}(x,y)=\int_{-\infty}^\infty f(x,y)dy\\ &M_Z(t)=M_X(t)M_Y(t)\ 4.2.12-M_Z(t)=(M_X(t))^n\ 4.6.7-M_{\bar{X}}(t)=[M_X(\frac{t}{n})]^n\ 5.2.7\\ &M_W(t)=M_X(t)M_Y(t)=e^{\mu_1(e^t-1)}e^{\mu_2(e^t-1)}=e^{(\mu_1+\mu_2)(e^t-1)}\ \operatorname{Use}\ 4.2.12\ \operatorname{proof}\ 4.3.2\\ &X,Y\ \operatorname{indep}\ \operatorname{r.v.}\ g(X),h(Y)\ \operatorname{indep}\ 4.3.5\\ &U\sim Geom(p=\frac{1}{2}),u=1,2...\ V\sim NBin(p=\frac{1}{2},r=2),v=2,3..\ (\mathrm{U,V})\\ &\{(u,v):u=1,2,...;v=u+1,u+2,...\}\ \operatorname{not\ indep} \end{split}$$

### Expectations

$$\begin{split} E[X] &= \int_{-\infty}^{\infty} x f(x) dx = E[E(X|Y)] \ 4.4.3 \ E[g(x)] = \sum_{x \in D} h(x) p(x) = \int_{-\infty}^{\infty} g(x) f(x) dx \\ E[(X - \mu)^n] &= \mu_n = \sum (x - \mu)^n p(x) = \int (x - \mu)^n f(x) dx - E(aX + b) = aE(X) + b \\ E[g_1(X_1) \cdots g_n(X_n)) &= (E(g_1(X_1)] \cdots (E(g_n(X_n)) \ X_1, ..., X_n \ \text{indep } 4.6.6 \\ E[g(\vec{X})] &= \sum \cdots \sum_{all \vec{x}} g(\vec{x}) p(\vec{x}) = \int \cdots \int_{\mathbf{R}^n} g(\vec{x}) f(\vec{x}) d\vec{x} \ E[\sum_{i=1}^n g(X_i)] = nE(g(X_1)) \ 5.2.5 \\ E[g(X,Y)] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f(x,y) dx dy \ 4.1.10 \\ E[XY] &= E[X] E[Y] \ E[g(X)h(Y)] = E[g(x)] E[h(y)] \ P(X \le x, Y \le y) = P(X \le x) P(Y \le y) \ 4.2.10 \\ E[g(X)|y] &= \sum_x g(x) f(x|y) = \int_{-\infty}^{\infty} g(x) f(x|y) dx \ \text{r.v.}(y) \ 4.2.3 \end{split}$$

#### Variances

$$\begin{split} V[X] &= \sigma_x^2 = E(X^2) - [E(X)]^2 = E[V(X|Y)] + V[E(X|Y)] \ 4.4.7 \\ V[X|Y] &= E[(X - E[X|Y])^2|Y] = E[X^2|y] - (E[X|y])^2 \ 4.2.4 \\ V[aX \pm bY] &= a^2VX + b^2VY \pm 2abCov(X,Y) \ 4.5.6 \ V[aX + b] = a^2\sigma^2 \ \sigma_{ax+b} = |a|\sigma_x \\ V[X \pm Y] &= VX + VY \ \text{Indep } 4.5.6 \ V[\sum_{i=1}^n g(X_i)] = nVar(g(X_1)) \ 5.25 \end{split}$$

### 4.5 Covariance and Correlation

$$\begin{aligned} &Cov(X,Y) = E[(X-\mu_X)(Y-\mu_Y)] = E(XY) - E(X)E(Y) = E[XY] - \mu_X\mu_Y = \sigma_{XY} \ 4.5.1/3 \\ &Cov(aX,bY) = abCov(X,Y) \ Cov(X,Y+Z) = Cov(X,Y) + Cov(X,Z) \ Cov(X,c) = 0 \\ &\rho_{XY} = \frac{\sigma_{XY}}{\sigma_X\sigma_Y} \ Corr(X,Y) = \frac{Cov(X,Y)}{\sqrt{Var(X)Var(Y)}} \ 4.5.2 \end{aligned}$$

#### 4.3 Transform

$$f_{U,V}(u,v) = f_{X,Y}(h_1(u,v), h_2(u,v))|J| \quad 4.3.2 = \sum_{i=1}^k f_{X,Y}(h_{1i}(u,v), h_{2i}(u,v))|J_i| \quad 4.3.5$$

$$\begin{vmatrix} u = g_1(x,y) & x = h_1(u,v) \\ v = g_2(x,y) & y = h_2(u,v) \end{vmatrix} J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial y}{\partial v} \\ \frac{\partial x}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \frac{\partial x}{\partial u} \frac{\partial y}{\partial v} - \frac{\partial y}{\partial u} \frac{\partial x}{\partial v}$$

$$f_Z(z) = \int_{-\infty}^{\infty} f_X(w) f_Y(z-w) dw \quad Z = X + Y \quad 5.2.9 \text{ Convolution}$$

#### Distribution

$$\begin{split} X &\sim n(\mu,\sigma^2), \ Y \sim n(\gamma,\tau^2) - X + Y \sim n(\mu + \gamma,\sigma^2 + \tau^2) \text{indep } 4.2.14 \\ X,Y &\sim n(0,1) - X + Y, X - Y \sim n(0,2), \ X/Y \sim cauchy(0,1) \ \text{indep } 4.3.4/6 \\ X &\sim Poisson(\theta), \ Y \sim Poisson(\lambda) - X + Y \sim Poisson(\theta + \lambda) \ \text{indep } 4.3.2 \\ X &\sim Beta(\alpha,\beta), \ Y \sim Beta(\alpha+\beta,\gamma) - XY \sim n(\alpha,\beta+\gamma) \ \text{indep } 4.3.3 \\ X|Y &\sim Bin(Y,p), Y|\Lambda \sim Pois(\Lambda), \Lambda \sim Expo(\beta) \ \text{or} \ X|Y \sim Bin(Y,p), Y \sim NBin(1,\frac{1}{1+\beta}) \ 4.4.5 \\ Y|\Lambda &\sim Pois(\Lambda), \Lambda \sim Gamma(\alpha,\beta) \ \text{then} \ Y \sim NBin(\alpha,\frac{1}{1+p\beta}) \ \text{Pois-Gamma} \\ X|P &\sim Bin(n,P), P \sim Gamma(\alpha,\beta) \ EX = E[E(X|P)] = E[nP] = n\frac{\alpha}{\alpha+\beta} \ \text{Beta-Bin } 4.4.6 \\ V[X] &= V[E(X|P)] + E[V(X|P)] = \frac{n^2\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)} + \frac{n\alpha\beta}{(\alpha+\beta)(\alpha+\beta+1)} = \frac{n\alpha\beta(\alpha+\beta+n)}{(\alpha+\beta)^2(\alpha+\beta+1)} \ \text{Beta-Bin } 4.4.8 \\ f_X(x) &\sim n(\mu_X,\sigma_X^2) \ f_Y(y) \sim n(\mu_Y,\sigma_Y^2), \ \rho_{XY} = \rho \ \text{bivarialte normal } 4.5.10 \end{split}$$

$$\begin{split} f_{Y|X}(y|x) &\sim n \left( \mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X), \sigma_Y^2 (1 - \rho^2) \right) \\ aX + bY &\sim n (a\mu_X + b\mu_Y, a^2\mu_X^2 + b^2\mu_Y^2 + 2ab\rho\sigma_X\sigma_Y \\ X_1, ...X_n &\sim Gamma(\alpha, \beta), \ X_1 + ...X_n \sim Gamma(\alpha_1 + ...\alpha_n, \beta) \ \bar{X} \sim Gamma(n\alpha, \beta/n) \ \text{indep 4.6.8} \\ X &\sim Cauchy(0, \sigma), Y \sim Cauchy(0, \tau) - X + Y \sim Cauchy(0, \sigma + \tau) \ 5.2.10 \end{split}$$

 $X_1,...,X_n \sim Cauchy(0,\sigma) \longrightarrow \bar{X} \sim Cauchy(0,\sigma), \sum_{i=1}^n X \sim Cauchy(0,n\sigma)$ 

## 5.3 Sampling from N

$$\begin{split} \bar{X} &= \frac{X_1 + \ldots + X_n}{n} = \frac{1}{n} \sum_{i=1}^n X_i \ 5.2.2 \ \bar{X}_{n+1} = \frac{\sum_{i=1}^{n+1} X_i}{n+1} = \frac{\sum_{i=1}^n X_i + X_{n+1}}{n+1} = \frac{n\bar{X}_n + X_{n+1}}{n+1} \ 5.3.1 \\ S^2 &= \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i^2 - n\bar{X}^2) \ 5.2.3 \ nS_{n+1}^2 = (n-1)S_n^2 + (\frac{n}{n+1})(X_{n+1} - \bar{X}_n)^2 \\ \frac{5.2.6 \quad X_1, \ldots, X_n \ \text{iid} \quad E[\bar{X}] = \mu \quad E[S^2] = \sigma^2 \quad V[\bar{X}] = \frac{\sigma^2}{n} \quad W/o \ \text{Normal} \\ \hline 5.3.1 \quad \sim n(\mu, \sigma^2) \quad \bar{X}, S^2 \ \text{indep} \quad \bar{X} \sim N(\mu, \frac{\sigma^2}{n}) \quad \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2 \quad W \ \text{Normal} \end{split}$$

$$X \sim n(\mu, \sigma^2), \ \frac{x - \mu}{\sigma}, \frac{\bar{x} - \mu}{\sigma / \sqrt{n}} \sim n(0, 1), \ \frac{\bar{X} - \mu}{S / \sqrt{n}} = \frac{\frac{\bar{x} - \mu}{\sigma / \sqrt{n}}}{\sqrt{S^2 / \sigma^2}} = \frac{U}{\sqrt{\frac{\chi_{n-1}^2}{n-1}}} \sim t_{n-1} \ 5.3.4 - t_1 = Cauchy(0, 1)$$

$$x_i \sim n(0,1), \sum_{i=1}^n x_i^2 \sim \chi_n^2 - x_i \sim n(0,\sigma^2), \sum_{i=1}^n x_i^2 \sim \sigma^2 \chi_n^2, \sum_{i=1}^n (x_i - \bar{x})^2 \sim \sigma^2 \chi_{n-1}^2$$
  
 $\chi_2^2 \Leftrightarrow Expo(2) \chi_n^2 \sim Gamma(\frac{p}{2}, 2) \ 4.6.8$ 

$$X_1, ... X_n \sim \chi^2_{p_i}, X_1 + ... X_n \sim \chi^2_{p_1 + ... p_n}$$
 5.3.2  $U \sim \chi^2_m, V \sim \chi^2_n, U + V \sim \chi^2_{m+n}$ 

$$X_i \sim n(\mu_X, \sigma_X^2) Y_j \sim n(\mu_Y, \sigma_Y^2), \ X_1..X_n, Y_1..Y_m \ \text{indep}, \frac{S_X^2}{\sigma_X^2} \sim \chi^2 - \frac{S_X^2/\sigma_X^2}{S_Y^2/\sigma_Y^2} \sim F \ 5.3.5$$

$$X \sim F_{p,q}, \ \frac{1}{X} \sim F_{q,p} - X \sim T_q, \ X^2 \sim F_{1,q} - X \sim F_{p,q}, \ \frac{\frac{p}{q}X}{\frac{1+\frac{p}{2}X}{q}} \sim Beta(\frac{p}{2},\frac{q}{2})$$

$$Var\chi_{n-1}^2 = Var[\frac{(n-1)S^2}{\sigma^2}] = \frac{(n-1)^2}{\sigma^4} Var[S^2] = 2(n-1) \implies Var[S^2] = \frac{2\sigma^4}{n-1}$$

### 5.4 Order statistics

$$5.4.4 f_K(x) = \frac{n!}{(j-1)!(n-j)!} K[F_X(x)]^{k-1} [1 - F_X(x)]^{n-k} f(x)$$
 1-29p9

$$5.4.6 \ f_{X_{(1)},...,X_{(n)}}(x_1,...,x_n) = \begin{cases} n! f_X(x_1) \cdot ... \cdot f_X(x_n) & -\infty < x_1 < ... < x_n < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$f_{X_{(i)},X_{(j)}}(u,v) = \frac{n!}{(i-1)!(j-1-i)!(n-j)!} f_X(u) f_X(v) [F_X(u)]^{i-1} [F_X(v) - F_X(u)]^{j-1-i} [1 - F_X(v)]^{n-j}$$

$$f_{X_{(i)},X_{(i)}}(u,v) = \frac{n!}{(i-1)!(i-1-i)!(n-i)!} f_X(u) f_X(v) [F_X(u)]^{i-1} [F_X(v) - F_X(u)]^{j-1-i} [1 - F_X(v)]^{n-1-i} [1$$

### 1 from N to T to Chi to F

Given some function of these, find the distribution.

# 2 Transformation of pairs of r.v.s

Given 2 r.v.s, f(x,y), and a function of them, find its distribution. 1-10p1

$$f(x,y) = \frac{1}{4}e^{-\frac{x+y}{2}}, 0 < x < \infty, 0 < y < \infty, u = \frac{X-Y}{2}$$

$$1. \begin{vmatrix} U = \frac{x-y}{2} & X = 2u + v \\ V = y & Y = v \end{vmatrix} 2. J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 2 & 1 \\ 0 & 1 \end{vmatrix} = 2$$

3. 
$$g(u,v) = f(x,y)|J| = \frac{1}{4}e^{-\frac{x+y}{2}}2 = \frac{1}{2}e^{-\frac{2u+v+v}{2}} = \frac{1}{2}e^{-(u+v)}$$

$$4. \ 0 < x < \infty, 0 < y < \infty \implies 0 < 2u + v < \infty, 0 < v < \infty \implies v > -2u$$

5. Double Exponential(Laplace)

$$\begin{array}{l|l} G_{0}(u) = \left| \int_{-2u}^{\infty} \frac{1}{2} e^{-(u+v)} dv = \frac{1}{2} e^{-u} \int_{-2u}^{\infty} e^{-v} dv = \frac{1}{2} e^{-u} \left[ -e^{-v} \right]_{-2u}^{\infty} = \frac{1}{2} e^{-u} \left[ 0 + e^{2u} \right] \right| \quad u < 0$$

$$= \frac{1}{2} e^{|u|} \left| \int_{0}^{\infty} \frac{1}{2} e^{-(u+v)} dv = \frac{1}{2} e^{-u} \int_{0}^{\infty} e^{-v} dv = \frac{1}{2} e^{-u} \left[ -e^{-v} \right]_{0}^{\infty} = \frac{1}{2} e^{-u} \left[ 0 + 1 \right] \right| \quad u \ge 0$$

Given f(x,y), find the

$$f_X(x) = --E[X] = --V[X] =$$

$$f_{Y}(y) = -E[Y] = -V[Y] =$$

$$f(x,y) = -E[XY] =$$

$$Cov(x, y) = E[XY] - EXEY =$$

$$f(X|Y) = --E[X|Y] = --V[X|Y] =$$

$$\rho = Cov(x, y) / \sqrt{V[X]V[Y]} =$$

$$4.5.7 \ E[Y|X] = a + bx, \ E[Y] = E[E[Y|X]] = E[a + bX] = a + bE[X] \text{ by } 4.4.3,$$

$$\begin{array}{l} 4.5.7 \; E[Y|X] = a + bx, \; E[Y] = E[E[Y|X]] = E[a + bX] = a + bE[X] \; \text{by } 4.4.3, \\ E[XE[Y|X]] = E[X(a + bX)] = aE[X] + bE[X^2], \; E[XE[Y|X]] = \int_{-\infty}^{\infty} x E[Y|x] f_X(x) dx \; \text{by } 2.2.1, \end{array}$$

$$= \int_{-\infty}^{\infty} x \left[ \int_{-\infty}^{\infty} y f_Y(y|x) dy \right] f_X(x) dx \text{ by } 4.2.3, = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x y f(x,y) dy dx = E[XY] \text{ by } 4.1.10,$$

$$\sigma_{XY} = E[XY] - \mu_X \mu_Y = a\mu_X + bE[X^2] - \mu_X \mu_Y = a\mu_X + b(\sigma_X^2 + \mu_X^2) - \mu_X (a + b\mu_X) = b\sigma_X^2$$

$$\sigma_{XY} = E[XY] - \mu_X \mu_Y = a\mu_X + bE[X^2] - \mu_X \mu_Y = a\mu_X + b(\sigma_X^2 + \mu_X^2)$$

$$\rho_{XY} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y} = \frac{b\sigma_X^2}{\sigma_X \sigma_Y} = b\frac{\sigma_X}{\sigma_Y}$$

# 4 Order statistics

Find the distribution of X(k) or  $X_{(i)}, X_{(k)}$ joint pmf 1-31p2-3

5.4.5 uniform order pdf 1-31p4-8

### pdf/pmf

CDF: 
$$F(x) = P(X \le x) = \sum_{y:y \le x} p(y) = \int_{-\infty}^{x} f(y) dy$$
 probabilities:  $a \le b$ :  
PMF:  $p(x) = P(X = x) = P(\forall w \in W : X(w) = x)$   
 $P(a < X < b) = F(b) - F(a^{-}); P(a < X < b) = F(b) - F(a); P(a < X < a) = p(a);$ 

 $P(a < X < b) = F(b^{-}) - F(a)$ ; (where  $a^{-}$  is the largest possible X value strictly less than a); Taking a = b yields P(X = a) = F(a) - F(a - 1) as desired.

PDF:  $P(\forall w \in \mathcal{W} : a \leq X(w) \leq b) = \int_a^b f(x) dx$ 

 $P(a \le X \le b) = P(a < X \le b) = P(a < X \le b) = P(a < X \le b); P(X > a) = 1 - F(a); P(a \le X \le b) = F(b) - F(a)$ 

### CDF

Condition:  $f(x) \ge 0 \forall x \text{ pmf } \sum_{x} f_X(x) = 1, pdf \int_{-\infty}^{\infty} f(x) dx = 1$ 

### mgf

$$\begin{split} E(e^{tx}) &= \int e^{tx} f(x) dx = \sum e^{tx} f(x); \ M_{aX+b}(t) = e^{tb} M_{aX}(t) \\ M_X(t) &= E(e^{tx}) \\ M_X'(t) &= E(xe^{tx}) \\ M_X''(t) &= E(x^2 e^{tx}) \\ M_X^n(t) &= E(x^n e^{tx}) \\ M_X^n(t) &= E(x^n e^{tx}) \end{split} \quad \begin{aligned} M_X(0) &= E(X) \\ M_X''(0) &= E(X^2) \\ M_X''(0) &= E(X^n) \end{aligned}$$

#### transform

$$\begin{array}{l} g(x) \uparrow & F_Y(y) = F_X(g^{-1}(y)) \\ g(x) \downarrow & F_Y(y) = 1 - F_X(g^{-1}(y)) \\ \text{not monotone} \because X \leq 0 \text{ is } \emptyset \therefore P(X \leq -\sqrt{y}) = 0 \\ F_Y(y) = P(Y \leq y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq \sqrt{y}) = P(X \leq \sqrt{y}) - P(X \leq -\sqrt{y}) = F_X(\sqrt{y}) = \sqrt{y}, \ 0 < \sqrt{y} < 1 \end{array}$$

monotone:  $f_Y(y) = f_X(g^{-1}(y)) |\frac{d(g^{-1}(y))}{dy}|$ 

# Series

Finite 
$$\begin{array}{ll} \sum_{k=1}^{n} k = \frac{n(n+1)}{2} & \text{Binomial} \\ \sum_{k=1}^{n} (2k-1) = n^2 & \sum_{k=0}^{n} \binom{n}{k} = 2^n \\ \sum_{k=1}^{n} k^2 = \frac{n(n+1)(2n+1)}{6} & \sum_{k=0}^{n} \binom{r+k}{k} = \binom{r+n+1}{n} \\ \sum_{k=1}^{n} k^3 = (\frac{n(n+1)}{2})^2 & \sum_{k=0}^{r} \binom{m}{k} \binom{n}{r-k} = \binom{m+n}{r} \\ \sum_{k=0}^{n} c^k = \frac{c^{n+1}-1}{c-1} & \sum_{k=0}^{n} \binom{n}{k} a^{n-k} b^k = (a+b)^n \end{array}$$

$$\begin{array}{ll} \text{Infinite } |p| < 1 & \sum_{k=0}^{\infty} kp^{k-1} = \frac{d}{dp} \left( \sum_{k=0}^{\infty} p^k \right) = \frac{1}{(1-p)^2} \\ \sum_{k=0}^{\infty} p^k = \frac{1}{1-p} & \sum_{k=0}^{\infty} \binom{r+k-1}{k} x^k = (1-x)^{-r} r \in \mathbb{N}^+ \\ \sum_{k=1}^{\infty} p^k = \frac{p}{1-p} & \sum_{k=0}^{\infty} \binom{\alpha}{k} p^k = (1+p)^{\alpha} & |p| < 1 \,, \, \alpha \in \mathbb{C} \\ \Gamma(n) = (n-1)! & \sum_{k=0}^{\infty} \binom{\alpha}{k} p^k = (1+p)^{\alpha} & |p| < 1 \,, \, \alpha \in \mathbb{C} \\ \Gamma(a+1) = a\Gamma(a) & \sum_{n=0}^{\infty} \frac{x^n}{n!} = e^x \\ \sum_{n=0}^{\infty} ar^n = \frac{a}{1-r}, |r| < 1 \\ \sum_{k=0}^{n-1} ar^n = \frac{a(1-r^n)}{1-r} & \sum_{k=0}^{n-1} ar^n = \frac{a(1-r^n)}{1-r} \\ \Gamma(0) = \Gamma(-1) = \infty & \int_0^1 x^{a-1} (1-x)^{b-1}, dx = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)} \\ \Gamma(-1/2) = -2\Gamma(1/2) & \int_0^{\infty} x^{t-1} e^{-x} dx = \Gamma(t) & \sum_{k=0}^{\infty} \frac{a}{n!} & \sum_{k=0}^{\infty}$$

### Integrals+c

Substitution 
$$u = g(x)$$
  $du = g'(x)dx$   $\int_a^b f(g(x))g'(x)dx = \int_{g(a)}^{g(b)} f(u)du$   $\int_a^b f(g(x))g'(x)dx = \int_a^b \frac{5}{3}\cos(u)du$ 

Integreation by parts 
$$\begin{pmatrix} u & du & dv & v \\ x & dx & e^{-x} & -e^{-x} \end{pmatrix} \int_{a}^{b} u dv \int_{x}^{a} x e^{-x} dx$$

$$\begin{aligned} &=uv|_a^b - \int_a^b v du \\ &= -xe^{-x} + \int xe^{-x} dx = -xe^{-x} - e^{-x} + c \\ &= xlnx|_3^5 - \int_3^5 dx = (xlnx - x)|_3^5 = 5ln5 - 3ln3 - 2 \\ &\int_a^b f(x) dx = F(b) - F(a) = -\int_b^a f(x) dx = \int_a^c f(x) dx + \int_c^b f(x) dx; \\ &\frac{d}{dx} \int_{v(x)}^{u(x)} f(t) dt = u'(x) f[u(x)] - v'(x) f[v(x)] \end{aligned}$$

### Derivatives

Distribution	$\mathbf{CDF}$	P(X=x),f(x)	$\mu$	$EX^2$	Var	$\mathbf{MGF}$	M'(t)	M" $(t)$	$M^n(t)$
Bern(p)		$p^x q^{1-x}, x \in \{1, 0\}$	p	p	pq	$pe^t + q$			
Bin(n,p)	$I_{1-p}(n-x,x+1)$	$\binom{n}{x} p^x q^{n-x}; x \in \{0, 1n\}$	np	$\mu(\mu+q)$	$q\mu$	$(pe^t + q)^n$			
Geom(p)	$1 - q^x$	$pq^{x-1}, x \in 1, 2, \dots$	$\frac{1}{p}$	$\frac{p+2q}{p^2}$	$\frac{q}{p^2}$	$\frac{pe^t}{1-qe^t}, t < -\ln q$			
	$1 - q^{x+1}$	$pq^x, x \in 0, 1, \dots$	$\frac{q}{p}$	$\frac{p+2q}{p^2}$ $\frac{q^2+q}{p^2}$	$\frac{q}{p^2}$	$\frac{p}{1 - qe^t}, qe^t < 1$	$\frac{pqe^t}{(1-qe^t)^2}$	$\frac{2pqe^t}{(1-qe^t)^3} - M'(t)$	
$\operatorname{NBin}(r,p)$		$\binom{x-1}{r-1} p^r q^{x-r}, x \in r, r+1$	$\frac{r}{p}$		$\frac{rq}{p^2}$ $\frac{rq}{p^2}$ $\frac{rq}{p^2}$	$\left(\frac{pe^t}{1-qe^t}\right)^r$			
		$\binom{x+r-1}{r-1} p^r q^x, x \in 0, 1$	$\frac{rq}{p}$		$\frac{rq}{p^2}$	$(\frac{p}{1-qe^t})^r, qe^t < 1$			
$\overline{\text{HGeom}(N,m,k)}$		$\frac{\binom{x+r-1}{r-1}p^rq^x, x \in 0, 1}{\binom{m}{x}\binom{N-m}{k-x}}$	$\frac{km}{N}$		$\mu \frac{(N-m)(N-k)}{N(N-1)}$				
$\mathrm{HGeom}(w,b,k)$		$\frac{\binom{w}{x}\binom{b}{k-x}}{\binom{w+b}{k}}$	$\frac{kw}{w+b}$		$\mu_{\frac{b(w+b-k)}{(w+b)(w+b-1)}}$				
$Pois(\mu)$	$e^{-\mu} \sum_{i=0}^{x} \frac{\mu^{i}}{i!}$	$\frac{\mu^x}{x!}e^{-\mu}, x \in 0, 1$	$\mu$	$\mu^2 + \mu$	$\mu$	$e^{\mu(e^t-1)}$	$\mu e^t M(t)$	$\mu e^t (1 + \mu e^t) M(t)$	
$\mathrm{Unif}(n)$		$\frac{1}{n}, x \in 1, 2n$	$\frac{n+1}{2}$	$\frac{(n+1)(2n+1)}{6}$	$\frac{(n^2-1)}{12}$	$\frac{\sum_{i=1}^{n} e^{ti}}{n}$			
Unif $(a, b)$	$\frac{x-a}{b-a}$	$\frac{1}{b-a}, x \in (a,b)$	$\frac{a+b}{2}$		$\frac{(b-a)^2}{12}$	$\frac{e^{tb} - e^{ta}}{t(b-a)}$			
$\mathcal{N}(\mu, \sigma^2)$		$\frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$	μ	$\mu^2 + \sigma^2$	$\sigma^2$	$e^{\mu t + \frac{\sigma^2 t^2}{2}}$	$(\mu + \sigma^2 t) M(t)$	$[(\mu + \sigma^2 t)^2 + \sigma^2]M(t)$	
$\mathcal{N}(0,1)$		$\frac{1}{e^{-\frac{x^2}{2}}}$	0	1	1	$e^{\frac{t^2}{2}}$			
$\mathcal{LN}(\mu, \sigma^2)$		$\frac{1}{x\sigma\sqrt{2\pi}}e^{\frac{-(\ln x - \mu)^2}{2\sigma^2}}$	$e^{\mu + \frac{\sigma^2}{2}}$	$e^{2\mu+2\sigma^2}$	$\theta^2(e^{\sigma^2}-1)$	×			
$Cauchy(\theta, \sigma^2)$		$\frac{\pi\sigma}{1+(\frac{x-\theta}{2})^2}$	×	×	×				
$D$ Expo $(\mu, \sigma^2)$		$\frac{1}{2\pi\sigma}e^{-\left \frac{x-\mu}{\sigma}\right }$	$\mu$	$\mu^2 + 2\sigma^2$	$2\sigma^2$	$\frac{e^{\mu t}}{1 - \sigma^2 t^2}$			
$\text{Expo}(\lambda)$	$1 - e^{-\lambda x}$	$\lambda e^{-\lambda x}, x \in (0, \infty)$	$\frac{1}{\lambda}$		$\frac{1}{\lambda^2}$	$\frac{\frac{e^{r}}{1-\sigma^{2}t^{2}}}{\frac{\lambda}{\lambda-t}, t < \lambda}$			
$\text{Expo}(\beta)$		$\frac{1}{\beta}e^{-\frac{x}{\beta}}$	$\beta$		$\beta^2$	$\frac{1}{1-\beta t}$	$\beta(1-\beta t)^{-2}$	$2\beta^2(1-\beta t)^{-3}$	
$Gamma(a, \lambda)$		$\frac{\lambda^a}{\Gamma(a)} x^{a-1} e^{-\lambda x}, x \in (0, \infty)$	$\frac{a}{\lambda}$	$\frac{a}{\lambda^2}$	$\left(\frac{\lambda}{\lambda - t}\right)^a, t < \lambda$				
$\operatorname{Gamma}(\alpha,\beta)$		$\frac{1}{\Gamma(a)\beta^{\alpha}}x^{a-1}e^{-x/\beta}$	lphaeta	$\alpha \beta^2$	$\left(\frac{1}{1-\beta t}\right)^a, t < \frac{1}{\beta}$				
Beta(a,b)		$\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)}x^{a-1}(1-x)^{b-1}$	$\frac{a}{a+b}$		$\frac{\mu(1-\mu)}{(a+b+1)}$				$\frac{\Gamma(\alpha+n)\Gamma(\alpha+\beta)}{\Gamma(\alpha+\beta+n)\Gamma(\alpha)}$
$B(\alpha, \beta) =$		$\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} x^{a-1} (1-x)^{b-1}$ $\frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}, x \in (0,1)$		$\frac{a(a+1)}{(a+b)(a+b+1)}$	$\frac{ab}{(a+b)^2(a+b+1)}$				
$\chi_p^2$		$\frac{x^{\frac{p}{2}-1}}{\Gamma^{\frac{p}{2}}2^{\frac{p}{2}}}e^{-\frac{x}{2}}$	p	$2p + p^2$	2p	$(1-2t)^{-p/2}, t < \frac{1}{2}$			
$t_p$		$\frac{\Gamma(\frac{p+1}{2})}{\Gamma(\frac{p}{2})\sqrt{p\pi}}\left(1+\frac{x^2}{p}\right)^{-\frac{p+1}{2}}$	0, p > 1		$\frac{p}{p-2}, p > 2$	×			
$\overline{F}$	x > 0	$\frac{\Gamma(\frac{p+q}{2})}{\Gamma(\frac{p}{2})\Gamma(\frac{q}{2})} \left(\frac{p}{q}\right)^{\frac{p}{2}} \frac{x^{\frac{p}{2}-1}}{(1+\frac{p}{2}x)^{\frac{p+q}{2}}}$	$\frac{q}{q-2}$	q > 2	$2\left(\frac{q}{q-2}\right)^2 \frac{p+q-2}{p(q-4)}$	q > 4			
Arcsine	$\frac{1}{\pi \arcsin \sqrt{x}}$	$\frac{1}{\pi\sqrt{x(1-x)}}, x \in [0,1]$	$\frac{1}{2}$		$\frac{1}{8}$				$\operatorname{Beta}(\frac{1}{2},\frac{1}{2})$
Dirichlet	$B(a) = \frac{\prod_{i=1}^{k} \Gamma(a_i)}{\Gamma(\sum_{i=1}^{k} a_i)}$	$\frac{1}{B(a)} \prod_{i=1}^{k} x_i^{a_i - 1}, x \in (0, 1)$	$\frac{a_i}{\sum_k a_k}$	$\sum_{i=1}^{k} x_i = 1$	$\frac{a_i(a_0 - a_i)}{a_0^2(a_0 + 1)}$	$Cov(X_i, X_j) =$	$\frac{-a_i a_j}{a_0^2 (a_0 + 1)}$	$a_0 = \sum_{i=1}^k a_i$	

 $U \sim Geom(p=\frac{1}{2}), u=1,2..$  the number of trials needed to get the first head.  $V \sim NBin(p=\frac{1}{2},r=2), v=2,3..$  the number of trials needed to get two heads in repeated tosses of a fair coin.