```
 \begin{array}{l} |V - \kappa(v, t), \theta \sim \kappa(v, t) \mid Y \sim N(0, 2) \mid \theta \mid Y \sim N(\frac{V}{2}, \frac{1}{2}) \\ \frac{Z}{2}; f(u) = \int_{-\infty}^{\infty} f(1) f(2) |v| dv = 2 \int_{0}^{\infty} = \frac{1}{\pi(u^{2} + 1)} \sim Cauch(0, 1) \\ \frac{Z}{2}; f(u) = \int_{-\infty}^{\infty} f(1) f(2) |v| dv = 2 \int_{0}^{\infty} = \frac{1}{\pi(u^{2} + 1)} \sim Cauch(0, 1) \\ \frac{Z}{2}; f(u) = \int_{-\infty}^{\infty} f(1) f(2) |v| dv = 2 \int_{0}^{\infty} = \frac{1}{\pi(u^{2} + 1)} \sim Cauch(0, 1) \\ \frac{Z}{2}; f(u) = \int_{0}^{\infty} f(1) f(2) |v| dv = 2 \int_{0}^{\infty} \frac{1}{\pi(u^{2} + 1)} \sim Cauch(0, 1) \\ \frac{Z}{2}; f(u) = \int_{0}^{\infty} f(1) f(2) |v| dv = 2 \int_{0}^{\infty} \frac{1}{\pi(u^{2} + 1)} \sim Cauch(0, 1) \\ \frac{Z}{2}; f(u) = \int_{0}^{\infty} f(1) f(2) |v| dv = 2 \int_{0}^{\infty} \frac{1}{\pi(u^{2} + 1)} \left( \frac{1}{\pi(u^{2} + 1)} + \frac{1}{\pi(u^{2
calculous (\frac{u}{v})' = \frac{u'v - uv'}{v^2}; (uv)' = u'v - uv', \int udv = uv - \int vdu; (a^x)' = a^x \ln(a), \int a^x \ln a = a^x; \int xe^x dx = xe^x - \int e^x dx = xe^x - e^x; (e^{ax})' = ae^{ax}, \int e^{au} = \frac{1}{a}e^{au}; (x \ln x - x)' = \ln x, \int \ln u = u \ln u - u; (x^n)' = nx^{n-1}, \int x^n = \frac{x^{n+1}}{x^{n+1}}; (\ln x)' = \frac{1}{x}, \int \frac{1}{ax+b} = \frac{\ln |ax+b|}{a}; \int \frac{1}{ax^n} = \frac{\ln |ax+b|}{a};
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             \theta x^{\theta-1} \sim Beta(\theta, 1), -\ln x \sim Expo(\theta), -2\theta \sum \ln x \sim \Gamma(n, 2), \chi^2_{2n}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                \beta = P_{\theta_1}(X > 0.5) = 1 - \int_0^{0.5} \theta x^{\theta - 1} dx = 1 - 0.5^{\theta}
H_0: \theta = 1, H_A: \theta > 1; \Lambda \nearrow P_{H_0}(-\ln y \le C) = \alpha
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           \begin{aligned} & \text{Hans and or } \text{MI}_{M}(i) &= \text{MI}_{X}(i) &= \text{MI}_{
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                      e^{-(x-\theta)}, n = 10; H_0: \theta = 0, H_1: \theta > 0; T = X_{(1)} \ge 0
 \lim_{n\to\infty} (1+\frac{\pi}{n})^n = e^{\pm a}, \sum_{n=0}^{\infty} \frac{(tx)^n}{n!} = e^{tx}, \\ \Gamma(n) = (n-1)! = (n-1)\Gamma(n-1), \Gamma(1/2) = \sqrt{\pi}, \\ \Gamma(-1/2) = -2\Gamma(1/2), \Gamma(0) = \Gamma(-1) = \infty; \\ \Gamma(t) = \int_0^\infty x^{t-1} e^{-x} dx, \int_0^1 x^{a-1} (1-x)^{b-1} dx = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}; 
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                Let Y = X - \theta, Y_{(1)} \sim Expo(10), \Gamma(1, \frac{1}{10}); 20Y_{(1)} \sim \chi_2^2;
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             P_{H_0}(X_{(1)} \ge C) = P_{H_0}(20Y_{(1)} \ge 20C) \stackrel{\text{set}}{=} 0.05, 20C = 5.991;
P_{H_1}(X_{(1)} \ge 0.2996) = P_{H_1}(Y_{(1)} \ge -0.7) = 1, \min \beta \text{ at } H_0 \text{ is unbia}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             VX = E[Var(X|P)] + Var[E(X|P)] = E[npq] + Var[np]
n\alpha\beta(\alpha+\beta+n)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                          \begin{array}{l} n\!\to\!\infty & (n)^j & n\!\to\!\infty \\ Z2.3 \  \, \text{Bayes estimator.} \\ X_n \  \, \text{iid} Expo(\theta), \pi(\theta) \sim Expo(\lambda) \\ \pi(\theta|\vec{x}) = \frac{g(\vec{x}|\theta)}{m(\vec{x})} = \frac{f(\vec{x}|\theta)h(\theta)}{\int g(\vec{x}|\theta)d\theta} \propto L(\theta)h(\theta) \sim \Gamma(n+1,\frac{1}{\lambda+\sum x_i}) \\ \mathcal{L}_{\theta}(\hat{\theta}) = (\hat{\theta}-\theta)^2; \  \, \text{minPosterior risk} \  \, E[\mathcal{L}_{\theta}(\hat{\theta}|\vec{x})], \  \, \text{Posterior mea} \end{array}
\sum_{n=0}^{\infty} a^n = \frac{a}{1-r}, |r| < 1; \sum_{k=0}^{n-1} a^n = \frac{a(1-r^n)}{1-r}; Finite Binomia \sum_{k=1}^{n} k = \frac{n(n+1)}{2}; \sum_{k=0}^{n} {n \choose k} = 2^n
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                   Expo(\frac{1}{\mu}); H_0: \mu = 5, H_1: \mu \neq 5
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                     (\alpha+\beta)^2(\alpha+\beta+1)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        \begin{array}{l} \left(\lambda - \frac{L(\mu_0)}{L(\mu_1)} = (\frac{\mu_1}{5})^n e^{(\frac{1}{\mu_1} - \frac{1}{5})} \sum x_i \\ = \frac{L(\mu_0)}{L(\mu_1)} = (\frac{\mu_1}{5})^n e^{(\frac{1}{\mu_1} - \frac{1}{5})} \sum x_i \\ \leq c \\ \left(\mu_1 < 5.\Lambda \nearrow \sum x_i \leq c'; \mu_1 > 5.\Lambda \searrow \sum x_i \geq c'; \\ T = \sum x_i \sim Gamma(10, \mu), \frac{\mu}{\mu} T \sim \chi_{20}^2; \\ P_{H_0}(\frac{2T}{5} \leq \chi_{20,0.975}^2(9.59)) + P_{H_0}(\frac{5T}{5} \geq \chi_{20,0.025}^2(34.2)) = 0.05; \\ = c.2 \end{array}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                          \begin{array}{l} P \sim Beta(\frac{1}{2},\frac{1}{2}), EX = \frac{n}{2}; VX = \frac{n(n+1)}{2}; \\ P \sim Beta(\frac{1}{2},1), X \sim Unif(0,n+1), P|X \sim Beta(x+1,n-x+1) \\ \text{Pois-Bino.} \end{array}
                                                                                                                                                                                                                                 \sum_{k=0}^{n} {r+k \choose k} = {r+n+1 \choose n}
\sum_{k=1}^{n} (2k-1) = n^2
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                \begin{aligned} & \textit{Pois-Bino} \\ & \textit{Pois}(\lambda).S_{10} = \sum_{i=1}^{10} X_i \sim \textit{Pois}(10\lambda). \\ & \textit{M}_{S_{10}}(t) = \prod_{i=1}^{10} M_X(t) = [e^{\lambda(e^t-1)}]^{10} = e^{10\lambda(e^t-1)} \end{aligned}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             \hat{\theta}_{L_2Bayes} = E[\pi(\theta|\vec{x})] = \frac{n+1}{\lambda + \sum x_i} = \frac{\lambda(\frac{1}{\lambda}, E[\theta])}{\lambda + n\bar{x}} + \frac{n\bar{x}(\frac{1}{\lambda}, \theta_{MLE})}{\lambda + n\bar{x}}
\sum_{k=1}^{n} k^2 = \frac{n(n+1)(2n+1)}{6} \sum_{k=0}^{n} {n \choose m} = {n+1 \choose m+1}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             \sum_{k=1}^{n} k^3 = (\frac{n(n+1)}{2})^2
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             \begin{split} &M_{S_{10}}(t) = 11 \underset{t=-}{\overset{\leftarrow}{\longrightarrow}} M_X(t) = [e^{-x} - 1] \\ &S_{10} = s = S_4 + S_6, P(S_4|S_{10}) = \frac{P(S_4, S_6 = s - S_{10})}{P(S_{10})} \sim Bino(s, \frac{2}{5}) \\ &Pois(\lambda), Pois(3\lambda) P(x|x+y=5) = \frac{P(x)P(y=5-x)}{P(x+y=5)} \sim Bino(5, \frac{1}{3}); \end{split}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        \begin{aligned} & H_0\left(\frac{r_5}{S} \leq \lambda \tilde{z}_{20,0.075}(9.59)\right) + l^*H_0\left(\frac{r_5}{S} \leq \lambda \tilde{z}_{20,0.025}(34.2)\right) = 0.05; \\ & P_H\left(\frac{2T_1}{I_1} \leq \frac{5\lambda^2}{I_1}\right) + P_H\left(\frac{2T_1}{I_1} \geq \frac{5\lambda^2}{I_2}\right) - \frac{5\lambda^2}{I_1} + \frac{5\lambda^2}{I_1} + \frac{5\lambda^2}{I_2} + \frac{5
                                                                                                                                                                                                                           \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}
1 Basic
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                  \begin{aligned} & \underbrace{\sum_{k} = \underbrace{\sum(X_{j} - \bar{X})^{2}}_{k}; \underbrace{\frac{\partial MSE}{\partial k}}_{e} \underbrace{\frac{ef}{\partial s} \cdot OS_{n+1}^{2} \min(MSE)}_{n+1} \\ & \underbrace{\frac{\partial}{\partial s} \cdot CRLB}_{n} \underbrace{\frac{\partial}{\partial \theta} \cdot D_{n} \cdot f(X|\theta)^{2}}_{n} = \underbrace{\frac{[\tau'(\theta)]^{2}}{n!\theta}}_{n!\theta} \end{aligned}}_{n} \end{aligned}
                                                                                                                                                                                                                        \sum_{k=0}^{n} {n \choose k} a^{n-k} b^k = (a+b)^n
1 Basic P(X \le y|X \ge c) = \frac{P(X \le y, X \ge c)}{P(X \ge c)} = \frac{P(c \le X \le y)}{1 - P(X \le c)} = \frac{F(y) - F(c)}{1 - F(c)}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        \begin{aligned} & P(x+y=5) \\ & + S. \text{Cay } Corr \\ & Cov(X,Y) = \sigma_{XY} = E(X - \mu_X)(Y - \mu_Y)) = E[XY] - \mu_X \mu_Y \\ & Cov(x,X) = \sigma_{XY} = E(X - \mu_X)(Y - \mu_Y)) = E[XY] - \mu_X \mu_Y \\ & Cov(x,X) + V = \sigma_{XY} + V = Cov(X,Y) + Cov(X,Y) \\ & Cov(X,X) = 0; Cov(X,X) = E[X^2] - \mu_X^2 \cdot X \perp Y, Cov(X+Y,X-Y) = Cov(X,X) - Cov(Y,Y) - V[Y] - V[Y] \\ & Cov(X,Y) = \frac{Cov(X,Y)}{\sqrt{Var(X)Var(Y)}}, = \rho_X = \frac{\sigma_{XY}}{\sigma_{X}^2 O_Y} \\ & Cov[X,X_n] = \frac{Cov[X,X_n]}{E_{X}^2 - E_{X}^2} = \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} \\ & Cov[X,X_n] = \frac{Cov[X,X_n]}{E_{X}^2 - E_{X}^2} = \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} \\ & \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} = \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} \\ & \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} = \frac{\sigma_X^2}{\sigma_X^2} \\ & \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} = \frac{\sigma_X^2}{\sigma_X^2} \\ & \frac{\sigma_X^2}{\sigma_X^2 - E_{X}^2} = \frac{\sigma_X^2}{\sigma_X^2} \\ & \frac{\sigma_X^2}{\sigma_X^2} = \frac{\sigma_X^2}{\sigma_X^2} \\ &
\begin{array}{l} 5*(9.9) \\ 5*(9.9) + 995*(0.02) \\ \hline 2*(3), E[30 - 3X - X^2] = 30 - 3EX - EX^2 = 9; \end{array}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             *Expo(\theta),median0.5 = \int_0^{\tilde{x}} f()dx, \tilde{x} = \frac{\ln 2}{\theta};
U(0,1),E[Y] = E(E[Y|X]) = \alpha + \beta EX^2 = \alpha + \beta/3
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             90^{th} \int_0^{g(\theta)} dx = 0.9; g(\theta) = \frac{\ln 10}{\theta}, VX_{(90)} \ge \frac{g'(\theta)^2}{nI_{\theta}} = \frac{(\ln 10)^2}{n\theta^2}
\begin{cases} f_T(t) = \frac{1}{1.5}e^{-\frac{1}{1.5}t}; 0 \\ P(V=5) = P(t<3) = \int_0^3 dt = 1 - e^{-2} \end{cases}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                          *N(\mu, \sigma^2), \frac{X_{(95)} - \mu}{\sigma} = Z_{(95)} = 1.645; X_{(95)} = \mu + 1.645\sigma;
\begin{cases} r(v = 0) = r(v < 0) = J_0 \text{ at } = 1 - e^{-\frac{v}{2}} & 5 \le v < 6 \\ P(V \le v) = P(2T < v) = \int_0^{v/2} dt = 1 - e^{-\frac{v}{2}} & 6 \le v \\ \frac{v}{6}(x + 1), -1 \le x \le 2, Y = X^2; P(Y \le y) = P(X^2 \le y) = 1 \\ P(1 \le X \le \sqrt{y}) = \int_1^{\sqrt{y}} \int_0^{y} dx dx = \frac{v}{y} + \frac{2\sqrt{y}}{4}, -\frac{1}{3} \quad x \ge 1 \\ \dots \sqrt{y} \end{cases}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                4.7 Inequity;Chebychev;Cauchy-Schwarz;Jensen g(x) \ge 0 \forall r > 0, P(g(X) \ge r) \le \frac{Eg(X)}{r}; P(|x - \mu| < t\sigma) \ge 1 - \frac{1}{t^2};
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                \frac{d}{d\mu}X_{(95)} = 1; V[\hat{\mu}_{(95)}] \ge \frac{\sigma^2}{n}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                      \begin{aligned} &\frac{d\mu}{g(x)} \ge 0 \forall r > 0, P(g(X) \ge r) \le \frac{Eg(X)}{r}; P(|x-\mu| < t\sigma) \ge 1 - \frac{1}{12}; \\ &\sigma = 3, t = 2; P(-4 < x < 8) \ge \frac{3}{4}; \sigma = 2, t = \frac{5}{2}, P(-2 < x < 8) = \frac{3}{25}; \\ &\frac{d\sigma}{g(95)} |\ge \frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} \\ &EXY| \le |EXY| \le |EX|^2 |\ge \frac{1}{2} (E|Y|^2) \frac{1}{2}; Cov(X,Y) \le \sigma_X^2 \sigma_Y^2; \\ &Convex function Eg(X) \ge g(EX) \end{aligned} 
 \begin{aligned} &\frac{d\mu}{d\sigma} X(95) = 1.645; V[\hat{\rho}_{(95)}] \ge \frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} \\ &\frac{d\sigma}{g(95)} |\ge \frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} > \frac{1.645^2 \cdot 2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} > \frac{1.645^2 \cdot 2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{1.645^2 \cdot 2\sigma^4}{n} \\ &\frac{1.645^2 \cdot 2\sigma^4}{n} > \frac{1.645^2 \cdot 2\sigma^4
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                   \lim_{n \to \infty} \frac{CRLB}{Var(\hat{\theta})}; ARE: \lim_{n \to \infty} \frac{Var(\hat{\theta}_2)}{Var(\hat{\theta}_1)}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                Asym effi of MLE: \sqrt{n} [\tau(\hat{\theta}_{MLE}) - \tau(\theta)] \stackrel{D}{\rightarrow} N(0, \frac{1}{I_0})
\begin{cases} P(-\sqrt{y} \le X \le \sqrt{y}) = \int_{-\sqrt{y}}^{\sqrt{y}} f(x) dx = \frac{4\sqrt{y}}{9} \end{cases}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                   *Classical** f(x;\theta) = (\theta+1)x^{\theta}, 0 < x < 1, Beta(\theta+1,1); H_0: \theta = 0.5, H_1: \theta > 0.5
f_Y(y) = \begin{cases} \frac{1}{9} + \frac{1}{9\sqrt{y}} & 1 \le y \le 4\\ \frac{2}{9\sqrt{y}} & 0 \le y \le 1 \end{cases}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             \begin{aligned} f(x;\theta) &= (\theta + 1)x^{n}, 0 < x < 1.6Eta(\theta + 1, 1); H_{0} : \theta = 0.5, H_{1} : \theta : \theta \\ \hat{\theta}_{MOM} &= \frac{2X^{n}}{1 - X^{n}}; \hat{\theta}_{MLE} &= \frac{n}{-\sum \ln x_{i}} - 1; \hat{\theta}_{U} &= \frac{n - 1}{-\sum \ln x_{i}} - 1; \\ \text{CRLB} &= \frac{(\theta + 1)^{n}}{2}; \sqrt{n}(\hat{\theta} - \theta) \sim N(0, (\theta + 1)^{2}) \\ \Lambda &= \frac{L(\theta_{0})}{L(\theta_{1})} = e^{(\theta_{1} - 0.5)(-\sum x_{i})} \le C \nearrow \text{MLR}, \text{T suff} \end{aligned}
2.3 Moment M_X(t) = E[e^tX] M_X(t) = E[e^tX] \frac{e^t}{4-3e^t} \sim Geom(\frac{1}{4}); EX = \frac{1}{p} = 4; VarX = \frac{1-p}{p^2} = 12; \frac{\sigma}{\mu} = \frac{\sqrt{3}}{2}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           \ln x \sim Expo(\theta + 1), -2(\theta + 1) \sum \ln x \sim Gamma(n, 2), \chi^{2}_{2n}, n = 10
P_{H_0}(-3\sum\ln x_i \le \chi^2_{20,0.95}(10.851)) = 0.05
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        P(nY_1 < y) = P(Y_1 < \frac{y}{n}) = \int_0^{1-\frac{\pi}{n}} Y_n dy_1 = 1 - \int_0^{1-\frac{\pi}{n}} Y_n dy_1 = 1 - (1 - \frac{y}{n})^n;
Or transf W_1 = ny_1 f(w) = f(\frac{\pi}{n}) |\frac{dy}{dw}| = n(1 - \frac{w}{n})^{n-1} \frac{1}{n};
W_n = n(1 - Y_n), f(w_n) = f(1 - \frac{w}{n}) |\frac{dy}{dw}| = n(1 - \frac{w}{n})^{n-1} \frac{1}{n}
\lim_{x \to \infty} nY_1 = \lim_{x \to \infty} n(1 - Y_n) = 1 - e^{-\frac{y}{n}} \sim Expo(1)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             X = e^{Z} \sim LN(0,1); (e^{n\mu + n^2\sigma^2/2}) M_Z(t) = e^{\frac{1}{2}t^2};
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             Bino(n, p), Y Compl/Suffi E\left[\frac{Y}{n}\right] = p; E\left[\frac{Y^2 - Y}{n^2 - n}\right] = p^2
E[X^n] = E[(e^Z)^n] = M_Z(n) = e^{\frac{1}{2}n^2}; E[X^{1,2,3}] = e^{\frac{1}{2}}, e^2, e^{\frac{9}{2}} skewness E[X^n] = E[(e^Z)^n] = E[(e^Z)^
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        E(\frac{Y(n-Y)}{n(n-1)}) = pq; E(\frac{Y(n-Y)}{(Y-k)!(k-Y)!k!(n-k)!}) = \binom{n}{k}p^k(1-p)^{n-1}
7.3.4 Loss function, absolute error loss L(\theta,a) = |a-\theta| squared error loss L(\theta,a) = (a-\theta)^2 binary loss L(\theta,a) = \begin{cases} 0 & \theta \in \Theta_0 \\ 1 & \theta \in \Theta_0 \end{cases}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           \begin{array}{ll} & X \to \infty & X \to \infty \\ U(0,\theta) : X_{(n)} = \frac{n}{\theta} \left( \frac{x}{\theta} \right)^{n-1} \sim Beta(n,1); V[X_{II}] = \frac{n\theta^2}{(n+1)^2(n+2)} \\ EX_{(n)} = \frac{n\theta}{n+1}, E[\frac{n+1}{\theta} X_{(n)}] = \theta; E[\frac{n+2}{\theta} X_{(n)}^2] = \theta^2 \text{is UMVUE} \\ Expo(\lambda) : X_{(1)} = n\lambda e^{-n\lambda X} \sim Expo(n\lambda) \\ 5 \in \text{Convergency}. \end{array} 
EXPO.

X_1 \sim \Gamma(\alpha_1, \beta), X_2 \sim \Gamma(\alpha_2, \beta), U = \frac{X_1}{W = X_1 + X_2} \sim Beta(\alpha_1, \alpha_2);

X_1, X_2 \sim Exp(1), \Gamma(1, 1), U \sim \beta(1, 1), U(0, 1); V \sim \Gamma(2, 1), xe^{-X}

X_1, X_2 \sim Exp(\lambda), Y = X_1 - X_2 \sim DExp(0, \lambda), \frac{\lambda}{2}e^{-\lambda|x|} \forall x > 0

X \sim Beta(\theta, 1), -\ln X \sim Exp(\theta, 0), -2\theta \sum \ln X \sim \Gamma(n, 2), \chi_{2n}^2
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             E[\hat{\theta}_{MLE}] = \frac{n\theta}{n-1}, Bias = \frac{\theta}{n-1}; E[\hat{\theta}_{MLE}^2] = \frac{n^2\theta^2}{(n-1)(n-2)}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                8 Hypo testNeyman-Pearson,MLR;Karlin-Rubin. Size: \alpha = P_{H_0}(RR) Power: \beta = 1 - \alpha = P_{H_1}(RR) H_0: \theta = \theta_0, H_1: \theta = \theta_1, \theta_0 < \theta_1, RR \Lambda = \frac{L(\theta_0)}{L(\theta_1)} < C is MP
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           \begin{split} V[\theta_{MLE}] &= \frac{n-1}{n^2\theta^2}, & n-1 \\ V[\theta_{MLE}] &= \frac{n^2\theta^2}{(n-1)^2(n-2)}; & MSE = \frac{n+1}{(n-1)^2}\theta^2; \\ \lim_{n \to \infty} \frac{CRLB}{V\theta_U} &= \lim_{n \to \infty} \frac{n-2}{n} = 1 \\ H_0: \theta = 3.H_1: \theta = 4P_{H_0}\left(6\sum\ln(x+1) < \lambda_{2n,\alpha}^2\right) = \alpha \text{ is UMP} \end{split}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             5.5 Convergence

CLT \frac{\sqrt{n}(X_n - \mu)}{\sigma} \xrightarrow{\mathcal{D}} N(0,1); Stein n(\theta, \sigma^2), E[g(X)(X - \theta)]
DExp,Laplace_
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        Expo family g(t|\theta) = h(t)c(\theta)e^{W(\theta)t} T has MLR.

If free of \theta is Suffi, \Lambda = \frac{L(\theta_0)}{L(\theta_1)} has \nearrow MLR in T,T.
DEXPLAPMACE
L(\theta|x) = 2^{-n}e^{-\sum |x_i - \theta|}; \theta_{MLE} = X_{((n+1)/2)} \text{ sample median}
E[X_{((n+1)/2)} = \int_{-\infty}^{\theta} x \frac{1}{2}e^{x - \theta} dx + \int_{\theta}^{\infty} x \frac{1}{2}e^{-x + \theta} dx = \theta
Let it it is the sum of the
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             f(x,y) = 6(y-x), 0 < x < y < 1, Z = \frac{X+Y}{2}, W = Y; z < w < 2z;
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                     r^2E[g'(X)];
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             |J| = 2g(z, w) = 24(w - z)g(z) = \int_{z}^{2z} g(z, w)dw = 12z^{2}, 0 < z < \frac{1}{2}
|J| = 2g(z, w) = 24(w - z)g(z) = \int_{z}^{2z} g(z, w)dw = 12z^{2}, 0 < z < \frac{1}{2}
|J| = 2g(z, w)dw = 12(z - 1)^{2}, \frac{1}{2} < z < \frac{1}{2}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                Delta Method: if \sqrt{n}(Y_n - \theta) \stackrel{D}{\longrightarrow} N(0, \sigma^2),
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             P_{H_0}(RR) = \alpha \text{ is UMP}
\begin{array}{ll} \text{Let}(\mathbf{n}(1)/2) & \mathbf{J} - \infty^{-2} \mathbf{I} & \mathbf{J} = \mathbf{J} & \mathbf{J} = \mathbf{J} \\ \text{Triffing}(\mathbf{n}, 1), \mathbf{Beta}(1, 1); & -\ln F_1 - \ln(1 - F) \sim \mathbf{Exp}(1), \mathbf{Gamma}(1, 1); & \mathbf{J} \in \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J} = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{v} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{v} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{v} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J}, \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} < \mathbf{J}' - \mathbf{J} < \mathbf{J} \\ \text{Hil}(0, 1), \mathbf{J}' = \mathbf{J}' - \mathbf{J} 
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                   Beta(\theta,\theta),H_0: \theta=1, H_1: \theta=2 One Obs X_1<\frac{2}{3}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             f(y_1, y_2) = c(1 - y_2).0 \le y_1 \le y_2 \le 1; \int_0^1 \int_0^{y_2} dy_1 dy_2 = 1, c = 6;
f(y_1) = \int_{y_1}^1 dy_2 = 3(y_1 - 1)^2; f(y_2) = \int_0^{y_2} dy_1 = 6(y_2 - y_2^2)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                      = P_{\theta_0}(x_1 < \frac{2}{3}) = \int_0^{\frac{2}{3}} f(x|\theta_0) = \frac{2}{3}
\frac{J(u,v)}{U(0,1),U=Y_1+Y_2,0} < u-\overline{v} < 1, u-1 < \overline{v} < u, f=1, f(u,v)=1 \\ \frac{J(u)}{v} = 1, f(u,v)=1 \\ \frac{J(u)}{v} = 1, f(u,v) = 1 \\ \frac{J(u)}{v} = 1, f(u,v) = 1, f(u,v) = 1 \\ \frac{J(u)}{v} = 1, f(u,v) = 1, f(u,v) = 1 \\ \frac{J(u)}{v} = 1, f(u,v) = 1, f(u,v) = 1, f(u,v) = 1 \\ \frac{J(u)}{v} = 1, f(u,v) = 1, f(u,
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                             \begin{array}{c} \sigma = 0 \quad \text{or} \quad \sigma = 0 \\ \beta = 1 - P_{\theta_1} \left( x_1 < \frac{2}{3} \right) = 1 - \int_{0}^{2} \int_{0}^{3} f(x|\theta_1) = \frac{7}{27} \\ \times M_m \sim \operatorname{Expo}\left(\frac{1}{\theta_1}\right)_{1} Y_m \sim \operatorname{Expo}\left(\frac{1}{\theta_2}\right)_{1} H_0 : \theta_1 = \theta_2, H_1 : \theta_1 \neq \theta_2; \\ \cdot \theta_{MLE} = \frac{\sum x_1 + \sum y_1}{m + n} : \hat{\theta}_1 = \frac{\sum x_1}{m} : \hat{\theta}_2 = \frac{\sum y_1}{n} \\ 0 = \frac{\sum x_1}{\sum x_1 + \sum y_1} \sim \frac{\Gamma(m, \theta_1)}{\Gamma(m, \theta_1) + \Gamma(n, \theta_2)} \\ \cdot \Lambda = \frac{(m + n)^{m + n}}{m^m n} \tau^m (1 - T)^n \propto \operatorname{Beta}(m + 1, n + 1) \leq C' \\ \times N(\mu, 25)_{1} H_0 : \mu = 50, H_1 : \mu = 55, Z_{0.05} = 1.645, Z_{0.1} = 1.28; \\ P_{H_0} \left( X_n \geq c \right) = 0.05 \, P_{H_1} \left( X_n \geq c \right) = \frac{1}{2} \left( X_n \geq c \right) = \frac{1}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        (Y_1 \leq \frac{1}{2}|Y_2 \leq \frac{3}{4}) = \frac{P(Y_1 \leq \frac{1}{2}, Y_2 \leq \frac{3}{4})}{P(Y_2 \leq 3/4)} = \frac{P(Y_1 \leq \frac{1}{2}, Y_2 \leq \frac{1}{2}) + P(Y_1 \leq \frac{1}{2}, \frac{1}{2} \leq Y_2 \leq \frac{3}{4})}{P(Y_2 \leq 3/4)}
\begin{split} &\sqrt{n}(\frac{1}{X} - \frac{1}{\mu}) \xrightarrow{\gamma} N(0, (\frac{1}{\mu}), r^{uu}, \mu, (\frac{1}{\Lambda}))} \\ \widehat{V}ar[\frac{1}{X}] &\approx (\frac{1}{X})^4 S^2, \frac{\sqrt{n}(\frac{1}{X} - \frac{1}{\mu})}{(1/X)^2 S} \xrightarrow{D} N(0, 1) \\ &\frac{1}{Exp(\lambda)}, \sqrt{n}(R_n - \frac{1}{\lambda}) \xrightarrow{D} N(0, \frac{1}{\lambda^2}), g(u) = u^2, g'(u) = 2u \neq 0 \end{split}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                      \int_{0}^{\frac{1}{2}} \int_{0}^{y_{2}} f(x) dy_{1} dy_{2} + \int_{\frac{1}{2}}^{\frac{3}{4}} \int_{0}^{\frac{1}{2}} f(x) dy_{1} dy_{2}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                =\frac{\frac{1}{2} - \frac{1}{2}}{\int_{0}^{3/4} f_{Y_{2}}(y_{2}) dy_{2}} f(y_{1}, y_{2}) = C = 2.0 \le y_{1} \le y_{2} \le 1; \int_{0}^{1} \int_{0}^{y_{2}}; f_{Y_{1}} = 2 - 2y_{1} f_{Y_{2}} = 2y_{2}
\begin{split} & \frac{V(N-1)}{V(\bar{X})} = \frac{\sigma^2}{n}; E(Z^{2k}) = \frac{(2k)!}{2^K K!}, k = 1, 2, ...; E(Z^{2k+1}) = 0 \\ & V[S^2] > \underbrace{CRLB} > V[S^2_n]; \frac{2\sigma^4}{n-1} \ge \frac{2\sigma^4}{n} > \frac{2(n-1)\sigma^4}{n^2}; \end{split}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           -=\frac{25/32}{27/32}=25/27
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                  \sqrt{n}(\bar{X}_n^2 - (\frac{1}{\lambda})^2) \to n(0, \frac{4}{\lambda^4})
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           \begin{array}{l} \sqrt{n(x_n^{-1}(\frac{\pi}{n})^{-1})^{-n}} \frac{n(y_n^{-1}(\frac{\pi}{n})^{-1})}{\sqrt{n}(e^{-X}n - e^{-X}n}, g(\mu) = e^{-H}, g'(\mu) = -e^{-H} \neq 0 \\ \sqrt{n}(e^{-X}n - e^{-H}) \rightarrow n(0, \frac{[(e^{-H})^{t}]^{2}}{\sqrt{n}(e^{-X}n - e^{-H})}) = n(0, \mu e^{-2\mu})) \\ 6.2.1 \text{ Suffi}, 6.2.2 \text{ Mini Suffi}, 6.2.3 \text{ Ancillary Stat.} \\ \frac{\prod f(x|\theta)}{g(t|\theta)} = h(\vec{x}) \text{ free of } \theta; \frac{f(x|\theta)}{f(y|\theta)} \text{ constant fn of } \theta \text{ if } T(x) = T(y). \end{array} 
\begin{split} E[S] &= \sqrt{\frac{2\sigma^2}{n-1}} \frac{\Gamma(\frac{n}{2})}{\Gamma(\frac{n-1}{2})} \approx \sigma - \frac{\sigma}{4(n-1)}; \\ V[S] &= \sigma^2 - \frac{2\sigma^2}{n-1} \frac{\Gamma^2(\frac{n}{2})}{\Gamma^2(\frac{n-1}{2})} \approx \frac{\sigma^2}{2(n-1)} \end{split}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                     \begin{array}{l} .f_{Y_1} = 2 - 2y_1 j Y_2 = 2y_2 \\ E[Y_1 + Y_2] = \int_0^1 y_1(2 - 2y_1) dy_1 + \int_0^1 y_2(2y_2) dy_2 = \frac{1}{3} + \frac{1}{2} = 1 \\ P(Y_1 \leq \frac{3}{4} | Y_1 > \frac{1}{3}) = \frac{\int_1^{3/4}}{1 - \int_0^1 Y_3} = \frac{55}{64}, \\ .f(x;y) = Cxy, 0 \leq x \leq 2, 0 \leq y \leq 2, x + y \leq 2; \\ xy < x < 2 - y < 2, 0 < y < 2 - x < 2, \int_0^2 \int_0^{2 - x}, c \in \frac{3}{2}; \\ f(x) = \int_0^{2 - x} f(x, y) dy = \frac{3}{4} x(x - 2)^2; f(y) \text{same} \\ P(X < Y) = P(X < Y, Y < 1) + P(X < Y, Y > 1) = P(X < Y < 1) \\ 1 + P(X < 2 - Y, 1 < Y < 2) = \int_0^1 \int_0^y dx dy + \int_1^2 \int_0^{2 - y} dx dy = \frac{1}{2} \end{array}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                      \left(\frac{(Z_{\alpha} + Z_{\beta})\sigma}{\delta = \mu_1 - \mu_0}\right)^2 = \left[\frac{(1.645 + 1.28)5}{55 - 50}\right]^2 = 8.56 \approx 9
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                n = \left[\frac{\sqrt{-K + P'}}{\delta = \mu_1 - \mu_0}\right]^2 = \left[\frac{1.033 + 1.029}{55 - 50}\right]^2 = 8.56 \approx 9
{}^{\diamond}N(\theta, 1), N(3\theta, 1); H_0 : \theta \le 0, H_1 : \theta > 0; T: X + 3Y \sim n(10\theta, \frac{10}{m}), \Lambda \searrow
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        \begin{array}{c} f(y|\theta) & \text{second in to } \theta \text{ if } (X) = I(y), \\ g_{\theta}(\theta) = \theta + a, \ |Y_{n} - \tilde{Y}|(g_{\theta}(\theta)) = Y_{n} + a - \frac{\sum (Y_{j} + a)}{n} = Y_{n} - \frac{1}{n} \\ \text{invariant under location parameter} g(x|\theta) = f(x - \theta) \\ \text{Se.2.4 Compl Suff; Basu} & \text{Eig}(t) = 0 & \text{Valif } F(g(t) = 0) = 1; \\ F(x|\tilde{\theta}) = h(x)c(\tilde{\theta})e^{-\sum_{j=1}^{k} W_{j}(\tilde{\theta})t_{j}(x)}, \text{ three of } \theta \text{ Wo pen set in } \mathbb{R}^{k}; \\ \text{Expo familiy } t/n = \tilde{Y} \text{ is comp/suff, indep with } Y_{n} - \tilde{Y} \end{array}
E[\sigma^2] = \frac{1}{n} \sum E[(x_i - \mu)^2] = \sigma^2, \mu \text{ known};
E[\sigma^2] = \frac{E[\sum (x_i - \bar{y})^2]}{n} = \frac{(n-1)E[S^2]}{n} = \frac{(n-1)\sigma^2}{n}, \mu \text{ unknown}
E[\Phi^2(z)] = V[\Phi(z)] + E[\Phi(z)]^2 = \frac{1}{12} + \frac{1}{4} = \frac{\pi}{3};
E[n\Phi^{n-1}(z)] = \int_0^1 n\Phi^{n-1}\phi(z)dz = \int_0^1 n\Phi^{n-1}d\Phi(z) = \Phi^n(z)|_0^1 = \frac{\pi}{3}
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                \frac{\mathcal{D}}{\frac{10}{10}}(\bar{X}+3\bar{Y}-10\theta) \xrightarrow{\mathcal{D}} n(0,1); P_{\theta_0}(\sqrt{\frac{m}{10}}(\bar{X}+3\bar{Y}-10\theta_0 > Z_{\alpha}))
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                = P_{\theta_1} \left( \sqrt{\frac{m}{10}} (\bar{X} + 3\bar{Y} - 10\theta_1) > Z_{\alpha} - (\theta_1 - \theta_0) \sqrt{10m} \right)
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                     Pois(\lambda) H_0: \lambda \le 1, H_1: \lambda > 1 \Lambda = (\frac{\lambda_0}{\lambda_1})^{\sum x_i} e^{n(\lambda_1 - \lambda_0)} < C
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           =\sum x_i \sim Pois(n\lambda); \Lambda \searrow \alpha = P_{\theta_0}(T > C) = 0.05
```

D	$E; EX^2; V$	$f(x); F(x), P(X \le x)$	MLE; T; I	$M_x(t); M'(t); M''(t); M^n(t)$
Bern(p)	p;p;pq	$p^{x}q^{1-x}, x = 1,0$ $\binom{n}{x}p^{x}q^{n-x}, x = 0,1n;$	\bar{X} ; $\sum x_i \sim Bino(n, p)$; $1/pq$	$pe^t + q$
Bino(n, p)	np;np(np+q);npq	$\binom{n}{x} p^x q^{n-x}, x = 0, 1n;$	$Xork \ge X_{(n)}; \sum x_i \sim Bino(n, p); 1/pq$	$(pe^t + q)^n$
Geom(p)	$1/p;(p+2q)/p^2;q/p^2$	$pq^{x-1}, x = 1, 2,; 1 - q^x$	$1/\bar{X}; \sum x_i;$	$rac{pe^t}{1-qe^t}$, $t<-\ln q$; $rac{pqe^t}{(1-qe^t)^2}$; $rac{2pqe^t}{(1-qe^t)^3}$
NBino(r, p)	r/p ; ; rq/p^2	$\binom{x-1}{r-1} p^r q^{x-r}, x = r, r+1$	\bar{X} ; $\sum x_i$	$\left(\frac{pe^t}{1-qe^t}\right)^r, t < \ln q$
HGeom(N,m,k)	$\left(\frac{km}{N}; \mu \frac{(N-m)(N-k)}{N(N-1)}\right)$	$\binom{m}{x}\binom{N-m}{k-x}/\binom{N}{k}$		
$Pois(\mu)$	μ ; μ ; μ ² + μ	$\frac{\mu^x}{x!}e^{-\mu}$, $x = 0, 1; e^{-\mu} \sum_{i=0}^x \frac{\mu^i}{i!}$	\bar{X} ; $\sum x_i$; $1/\mu$	$e^{\mu(e^t-1)};\mu e^t M(t);\mu e^t (1+\mu e^t) M(t)$
Unif(n)	$\frac{n+1}{2}$; $\frac{(n+1)(2n+1)}{6}$; $\frac{n^2-1}{12}$	$\frac{1}{n}$, $x = 1, 2n, n = b - a + 1; \frac{x - a + 1}{n}$	X_n Comp	$\frac{1}{n} \sum_{i=1}^{n} e^{ti}, \frac{e^{at} - e^{(b+1)t}}{n(1-e^t)}$
Unif(a,b)	$\frac{a+b}{2}$; ; $\frac{(b-a)^2}{12}$	$\frac{1}{b-a}, x \in (a,b); \frac{x-a}{b-a}$	$\min x_{(1)}, x_{(n)}; R$ ancillary	$\frac{e^{tb} - e^{ta}}{t(b-a)}$
$Norm(\mu, \sigma^2)$	μ ; $\mu^2 + \sigma^2$; σ^2	$\frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$		$e^{\mu t + \frac{\sigma^2 t^2}{2}}; \ (\mu + \sigma^2 t) M(t); [(\mu + \sigma^2 t)^2 + \sigma^2] M(t)$
SNorm(0,1)	0;1;1	$\frac{1}{\sqrt{2\pi}}e^{-\frac{\chi^2}{2}}$		$\frac{\sigma^2 M(t)}{e^{\frac{t^2}{2}}}$
$LNorm(\mu, \sigma^2)$	$e^{\mu + \frac{\sigma^2}{2}}; e^{2\mu + 2\sigma^2}; E^2 X (e^{\sigma^2} - 1)$	$\frac{1}{x\sigma\sqrt{2\pi}}e^{\frac{-(\ln x - \mu)^2}{2\sigma^2}}$	$\hat{\mu} = \frac{1}{n} \sum \ln x_i, \hat{\sigma}^2 = \frac{1}{n} \sum \ln(x_i - \hat{\mu})^2$	$EX^n = e^{n\mu + n^2\sigma^2/2}$
Cauchy(θ , σ^2)		$\frac{1}{\pi\sigma} \frac{1}{1+(\frac{x-\theta}{\sigma})^2}$ $\frac{1}{2\sigma} e^{- \frac{x-\theta}{\sigma} }$		
$DExpo(\mu, \sigma^2)$	μ ; $\mu^2 + 2\sigma^2$; $2\sigma^2$	$\frac{1}{2\sigma}e^{- \frac{x-\mu}{\sigma} }$	$\hat{\mu} = median, \hat{\sigma} = \frac{1}{n} \sum x_i $	$\frac{e^{\mu t}}{1-\sigma^2 t^2}$
$Expo(\beta)$	β ; ; β ²	$\frac{1}{\beta}e^{-\frac{x}{\beta}}, x \in (0, \infty); 1 - e^{-\lambda x}$	$1/\bar{X}; \sum x_i; \text{mean}^2$	$\frac{\lambda}{\lambda - t}$, $t < \lambda$, $\frac{1}{1 - \beta t}$; $\beta (1 - \beta t)^{-2}$; $2\beta^2 (1 - \beta t)^{-3}$
$Gamm(\alpha, \beta)$	$\alpha\beta$; ; $\alpha\beta^2$;	$\frac{1}{\Gamma(a)\beta^{\alpha}} x^{a-1} e^{-x/\beta}$	$\prod x_i, \sum x_i$	$\left(\frac{1}{1-\beta t}\right)^a, t < \frac{1}{\beta}; EX^n = \frac{\beta^n \Gamma(\alpha+n)}{\Gamma\alpha}$
Beta(a,b)	$\frac{a}{a+b}$; $\frac{a(a+1)}{(a+b)(a+b+1)}$; $\frac{ab}{(a+b)^2(a+b+1)}$	$\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} x^{a-1} (1-x)^{b-1}$		$B(\alpha,\beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}, x \in (0,1); \frac{\Gamma(\alpha+n)\Gamma(\alpha+\beta)}{\Gamma(\alpha+\beta+n)\Gamma(\alpha)}$
χ_p^2	$p;2p+p^2;2p$	$\frac{1}{\Gamma(a)\Gamma(b)} \frac{1}{\Gamma(a)} \frac{1}{2} \frac{p}{2} \frac{p}{2} \frac{p}{2} \frac{p}{2} - 1e^{-\frac{x}{2}}$	$\sum \ln x_i;$	$(1-2t)^{-p/2}, t < \frac{1}{2}$
t_p	$0, p > 1; ; \frac{p}{p-2}, p > 2$	$\frac{\Gamma(\frac{p+1}{2})}{\Gamma(\frac{p}{2})\sqrt{p\pi}} (1 + \frac{x^2}{p})^{-\frac{p+1}{2}} F$	$\left(\frac{q}{q-2}, q > 2;; 2\left(\frac{q}{q-2}\right)^2 \frac{p+q-2}{p(q-4)}, q > 4\right)$	$\frac{\Gamma(\frac{p+q}{2})}{\Gamma(\frac{p}{2})\Gamma(\frac{q}{2})} \left(\frac{p}{q}\right)^{\frac{p}{2}} \frac{x^{\frac{p}{2}-1}}{(1+\frac{p}{q}x)^{\frac{p+q}{2}}}, x > 0$
Arcsine	$\frac{1}{2}$; ; $\frac{1}{8}$	$\frac{1}{\pi\sqrt{x(1-x)}}, x \in [0,1]; \frac{1}{\pi \arcsin\sqrt{x}}$		$Beta(\frac{1}{2},\frac{1}{2})$
Dirichlet	$\frac{a_i}{\sum_k a_k} \sum_{i=1}^k x_i = 1;; \frac{a_i(a_0 - a_i)}{a_0^2(a_0 + 1)}$ $\beta^{1/\gamma} \Gamma(1 + 1/\gamma);;$	$\frac{1}{B(a)} \prod_{i=1}^k x_i^{a_i-1}, x \in (0,1), B(a) = \frac{\prod_{i=1}^k \Gamma(a_i)}{\Gamma(\sum_{i=1}^k a_i)}$		$Cov(X_i, X_j) = \frac{-a_i a_j}{a_0^2 (a_0 + 1)}, a_0 = \sum_{i=1}^k a_i$
Weibull (γ, β)	$\beta^{1/\gamma}\Gamma(1+1/\gamma);;$ $\beta^{2/\gamma}[\Gamma(1+2/\gamma)-\Gamma^2(1+1/\gamma)]$	$\frac{\gamma}{\beta}x^{\gamma-1}e^{-\frac{x^{\gamma}}{\beta}}; 1-e^{-(\frac{x}{\beta})^{\gamma}}$	$x^{\gamma}, \sum x^{\gamma}, \sum \ln x$;; $\beta^{\frac{n}{\gamma}}\Gamma(1+\frac{n}{\gamma})$
$Pareto(\alpha, \beta)$	$\left rac{etalpha}{eta-1},eta>1;;rac{etalpha^2}{(eta-1)^2(eta-2)},eta>2 ight.$	$\left \frac{\beta \alpha \beta}{x\beta + 1}; 1 - \left(\frac{\alpha}{x}\right)^{\beta}, x \ge \alpha \right $	$\hat{\alpha} = \min x_i, \hat{\beta} = \frac{n}{\sum \ln(x_i/x_{(1)})};$ $\sum x_i \text{comp/suff}; 1/\beta^2$	