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附录

A.1 常用积分

特殊函数的性质 伽马函数与贝塔函数.

$$\text{伽马函数与递推式: } \Gamma(x) = \int_0^{+\infty} e^{-t} t^{x-1} dt = (x-1)\Gamma(x-1) \quad (x > 0)$$

$$\text{贝塔函数与关系式: } B(x, y) = \int_0^1 t^{x-1} (1-t)^{y-1} dt = \frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)} \quad (x, y > 0)$$

$$\text{勒让德倍量公式: } \Gamma(s)\Gamma\left(s + \frac{1}{2}\right) = \frac{\sqrt{\pi}}{2^{2s-1}}\Gamma(2s) \quad (s > 0)$$

$$\text{余元公式: } B(s, 1-s) = \Gamma(s)\Gamma(1-s) = \frac{\pi}{\sin \pi s} \quad (0 < s < 1)$$

$$\begin{cases} \Gamma(n) = (n-1)!, & n \in \mathbb{N}^+, \\ \Gamma\left(\frac{n}{2}\right) = \frac{(n-2)!!}{2^{(n-1)/2}}\sqrt{\pi}, & n \text{ 为正奇数.} \end{cases}$$

$$B(s, s) = \frac{1}{2^{2s-1}}B\left(\frac{1}{2}, s\right) \quad (s > 0)$$

更多内容可以参考 [Euler 积分笔记](#).

特殊函数的应用

一般的

$$\int_0^1 x^a (1-x^b)^c dx = \frac{1}{b} B\left(\frac{a+1}{b}, c+1\right) \quad (a > -1, b > 0, c > -1)$$

$$\int_0^{+\infty} \frac{x^a dx}{(1+x^b)^c} = \frac{1}{|b|} B\left(c - \frac{a+1}{b}, \frac{a+1}{b}\right) \quad \left(\begin{array}{l} a > -1, b > 0, c > \frac{a+1}{b} \text{ 或} \\ a < -1, b < 0, c > \frac{a+1}{b} \end{array} \right)$$

$$\int_0^{+\infty} x^n e^{-ax^p} dx = \frac{\Gamma\left(\frac{n+1}{p}\right)}{|p|a^{\frac{n+1}{p}}} \quad \left(\begin{array}{l} a > 0, p > 0, n > -1 \text{ 或} \\ a > 0, p < 0, n < -1 \end{array} \right)$$

特殊的

$$\begin{aligned} \int_{-\infty}^{+\infty} \frac{x^m dx}{(1+x^2)^n} &= B\left(n - \frac{m+1}{2}, \frac{m+1}{2}\right) \quad (\text{注意积分限}) \\ \int_0^{+\infty} \frac{x^m dx}{(1+x)^n} &= B(n-m-1, m+1) \end{aligned}$$

$$\int_0^{+\infty} e^{-ax^p} dx = \frac{\Gamma\left(\frac{1}{p}\right)}{pa^{\frac{1}{p}}}$$

$$\int_0^{+\infty} e^{-x^p} dx = \frac{1}{p}\Gamma\left(\frac{1}{p}\right)$$

$$\int_{-\infty}^{+\infty} e^{-ax^2} dx = \sqrt{\frac{\pi}{a}} \quad (\text{注意积分限})$$

$$\int_0^{+\infty} x^n e^{-ax} dx = \frac{\Gamma(n+1)}{a^{n+1}} = \frac{n!}{a^{n+1}}$$

$$\int_0^{+\infty} x^n e^{-ax^2} dx = \frac{\Gamma\left(\frac{n+1}{2}\right)}{2a^{\frac{n+1}{2}}}$$

$$\int_0^{+\infty} x^{2n} e^{-ax^2} dx = \frac{(2n-1)!!}{2(2a)^n} \sqrt{\frac{\pi}{a}}$$

$$\int_0^{+\infty} x^{2n+1} e^{-ax^2} dx = \frac{(2n)!!}{(2a)^{n+1}}$$

$$\int_0^{+\infty} x^2 e^{-ax^2} dx = \frac{1}{4a} \sqrt{\frac{\pi}{a}}$$

[有关 Catalan 常数的积分.](#)

A.2 常用分布

定义说明

- 期望 $\mu := E(X)$.
- 方差 $\sigma^2 := E[(X - \mu)^2]$.
- k 阶原点矩 $\alpha_k := E(X^k)$.
- k 阶中心矩 $\mu_k := E[(X - \mu)^k]$.
- 偏度系数 $\beta_1 = \frac{\mu_3}{\sigma^3}$.
- 峰度系数 $\beta_2 := \frac{\mu_4}{\sigma^4} = \frac{\mu_4}{\mu_2^2}$.
- 变异系数 $v_c := \frac{\sigma}{\mu}$.

A.2.1 一维离散型

1 二项分布

1.1 基础概念

- $X \sim B(n, p)$.
- 理解: 事件发生的概率为 p , 则重复 n 次试验, 事件发生的次数为 x .
- 概率分布: $P(X = i) = b(i; n, p) = \binom{n}{i} p^i (1-p)^{n-i}$.

1.2 数字特征

- 最可能数: $x = \lfloor (n+1)p \rfloor$.
- 期望: $E(X) = np$.
- 方差: $\text{Var}(X) = np(1-p)$.
- 母函数: $G(s) = (ps + q)^n, q = 1 - p$.
- 特征函数: $g(t) = (pe^{it} + q)^n$.

1.3 其它性质

- 二项分布和的函数

$$X_1 \sim B(n_1, p), X_2 \sim B(n_2, p) \Rightarrow X_1 + X_2 \sim B(n_1 + n_2, p).$$

- 发生偶数次的概率为 $p_n = \frac{1}{2}[1 + (1-2p)^n]$.
- 记 $f(p) = P(X \leq k)$, 则 $f'(p) < 0$, 并且

$$f(p) = \frac{n!}{k!(n-k-1)!} \int_0^{1-p} t^k (1-t)^{n-k-1} dt.$$

1.4 参数估计

- 矩估计: $p = m/n$. (MVU 估计)
- 极大似然估计: $p = m/n$. (MVU 估计)
- 贝叶斯估计

- 同等无知原则: $p = \frac{X+1}{n+2}$.

- 若先验密度 $h(p) = p^{a-1}(1-p)^{b-a-1}$, 则 $\tilde{p} = \frac{X+c}{n+d}$.

- 区间估计

- 大样本法: 近似取枢轴变量 $\frac{X - np}{\sqrt{np(1-p)}} \sim N(0, 1)$, 解不等式得

$$\left(1 + \frac{z_{\alpha/2}^2}{n}\right)p^2 - \left(2X + \frac{z_{\alpha/2}^2}{n}\right)p + X^2 < 0$$

$$\begin{aligned} \hat{p}_1, \hat{p}_2 &= \frac{n}{n + z_{\alpha/2}^2} \left(\bar{X} + \frac{z_{\alpha/2}^2}{2n} \pm z_{\alpha/2} \sqrt{\frac{\bar{X}(1-\bar{X})}{n} + \frac{z_{\alpha/2}^2}{4n^2}} \right) \\ &\approx \bar{X} \pm z_{\alpha/2} \sqrt{\frac{\bar{X}(1-\bar{X})}{n}}. \end{aligned}$$

- p^k ($k \leq n$) 的无偏估计是 $\frac{X^k}{n^k}$. (下降阶乘幂)

2 泊松分布

2.1 基础概念

- $X \sim P(\lambda)$.
- 理解: 单位时间内事件平均发生 λ 次, 则某一段单位时间内发生的次数为 x .
- 概率分布: $P(X = i) = \lim_{n \rightarrow \infty} b(i; n, \frac{\lambda}{n}) = \frac{e^{-\lambda} \lambda^i}{i!}$.
- 当二项分布满足 $n > 50, p < 0.1, np < 5$ 时, 用泊松分布近似效果较好.

2.2 数字特征

- 最可能数: $k = \lfloor \lambda \rfloor$.
- 期望: $E(X) = \lambda$.
- 方差: $\text{Var}(X) = \lambda$.
- 中位数: $m_e = \frac{\ln 2 \lambda}{\lambda}$.
- $E|X - m_e| = m_e$.
- 母函数: $G(s) = e^{\lambda(s-1)}, s \in (-\infty, +\infty)$.
- 特征函数: $g(t) = e^{\lambda(e^{it}-1)}$.

2.3 其它性质

- 泊松分布和的函数 (可加性)
 $X_1 \sim P(\lambda_1), X_2 \sim P(\lambda_2) \Rightarrow X_1 + X_2 \sim P(\lambda_1 + \lambda_2)$.
- 记 $f(\lambda) = P(X \leq k)$, 则 $f'(\lambda) < 0$, 并且 $f(\lambda) = \frac{1}{k!} \int_{\lambda}^{+\infty} t^k e^{-t} dt$.
注: 上式可用于参数检验.
- $P_{\lambda}(X \leq k) = \sum_{i=0}^k \frac{e^{-\lambda} \lambda^i}{i!} = \int_{\lambda}^{+\infty} \frac{e^{-t} t^k}{k!} dt = K_{2k+2}(2\lambda)$. (卡方分布函数)
- 若 $X \sim P(\lambda), Y \sim B(X, p)$, 则 $Y \sim P(\lambda p)$.
- 若有一批零件寿命服从指数分布, 固定一个时间 $T > 0$, 让一个元件从时刻 0 开始工作, 每当这个元件坏了的时候马上用一个新的替换, 则到 T 时替换的次数 $X \sim P(\lambda T)$, 即 $P(X = n) = \frac{e^{-\lambda T} (\lambda T)^n}{n!}$.
- 泊松分布的一个应用见特殊函数笔记中的 Dobinski 公式.

2.4 参数估计

- 矩估计
 - $\lambda = m$. (MVU 估计)
 - $\lambda = m_2$ 或 S^2 .
- 极大似然估计: $\lambda = \bar{X}$.
- 贝叶斯估计: 见第四章第五题.
- 区间估计
 - 大样本法: 近似地取 $(Y_n - n\lambda)/\sqrt{n\lambda} \sim N(0, 1)$, 则

$$A, B = \bar{X} + u_{\alpha/2}^2/(2n) \pm u_{\alpha/2} \sqrt{u_{\alpha/2}^2/(4n^2) + \bar{X}/n}, \quad \bar{X} = Y_n/n.$$

3 超几何分布

3.1 基础概念

- $X \sim H(N, n, M)$.
- 理解: N 件产品中有 M 件次品, 从总体中抽 n 件时次品的数量 m .
- 概率分布: $P(X = m) = \frac{\binom{M}{m} \binom{N-M}{n-m}}{\binom{N}{n}}$.

3.2 数字特征

- 期望: $E(X) = \frac{nM}{N}$.
- 方差: $\text{Var}(X) = \frac{nM(N-n)(N-M)}{N^2(N-1)} = \frac{nM}{N} \frac{N-n}{N-1} \left(1 - \frac{M}{N}\right)$.

3.3 其它性质

3.4 参数估计

已知 N, n 估计 M .

- 贝叶斯估计: 采用同等无知原则, 则 $M = \frac{N+2}{n+2}(X+1) - 1$.

4 负二项分布

4.1 基础概念

- $X \sim NB(r, p)$, 又称为正整数形式帕斯卡分布.
- 理解: 合格率为 p , 抽取到 r 个合格产品时, 抽到的不合格产品的个数 x .
- 概率分布: $P(X = i) = d(i; r, p) = \binom{i+r-1}{r-1} p^r (1-p)^i$.

4.2 数字特征

- 数学期望: $E(X) = \frac{r(1-p)}{p}$.
- 方差: $\text{Var}(X) = \frac{r(1-p)}{p^2}$.

4.3 其它性质

4.4 参数估计

注: $m_e := (X_1 + X_2 + \cdots + X_n)/n$.

- 矩估计: $p = \frac{r}{m_e + r}$.
- 极大似然估计: $p = \frac{r}{m_e + r}$.

- 贝叶斯估计: $p = \frac{nr + 1}{nr + nm_e + 1}$.

5 几何分布

5.1 基础概念

- $X \sim GE(p)$.
- 理解: 合格率为 p , 抽取到第一个合格产品时, 抽到的不合格产品的个数 x .
- 概率分布: $P(X = i) = p(1 - p)^i$.
- 累积分布函数: $P(X \leq k) = 1 - (1 - p)^{k+1}$.
- 互补累积分布函数: $P(X \geq k) = (1 - p)^k$.

5.2 数字特征

- 数学期望: $E(X) = \frac{1 - p}{p}$.
- 方差: $\text{Var}(X) = \frac{1 - p}{p^2}$.
- 母函数: $G(s) = \frac{ps}{1 - qs} - 1, s \in \left(-\frac{1}{q}, \frac{1}{q}\right)$.
- 特征函数: $g(t) = \frac{pe^{it}}{1 - qe^{it}} - 1$.

5.3 其它性质

- 几何分布具有无记忆性.
- 若 X_1, X_2, \dots, X_r 独立同分布 $GE(p)$, 则 $X_1 + X_2 + \dots + X_r \sim NB(r, p)$.
- 若 $X_1 \sim GE(1 - p_1)$ 和 $X_2 \sim GE(1 - p_2)$ 独立, 则

$$\begin{aligned}\min(X_1, X_2) &\sim GE(1 - p_1 p_2) \\ \max(X_1, X_2) &\sim P(X = k) = p_1^k(1 - p_1) + p_2^k(1 - p_2) + p_1^k p_2^k(p_1 p_2 - 1)\end{aligned}$$

更一般的, 若 $X_i \sim GE(1 - p_i)$, 则 $\min_i(X_i) \sim GE(1 - \prod_i p_i)$.

5' 几何分布

5'.1 基础概念

- $X \sim G(p)$.
- 理解: 合格率为 p , 抽取到第一个合格产品时, 抽取的总产品的个数 x .
- 概率分布: $P(X = i) = p(1 - p)^{i-1}$.

5'.2 数字特征

- 数学期望: $E(X) = \frac{1}{p}$.
- 方差: $\text{Var}(X) = \frac{1 - p}{p^2}$.
- 母函数

- $G(s) = \frac{ps}{1-qs}, s \in \left(-\frac{1}{q}, \frac{1}{q}\right).$
- $G^{(n)}(1) = \frac{(1-p)^{n-1}}{p^n} n!$.
- 特征函数: $g(t) = \frac{pe^{it}}{1-qe^{it}}.$

5.3 其它性质

- 几何分布具有无记忆性.
- 若 X_1, X_2, \dots, X_r 独立同分布 $G(p)$, 则 $X_1 + X_2 + \dots + X_r - r \sim NB(r, p).$

6 泽塔分布

6.1 基础概念

- $X \sim \text{Zeta}(s).$
- Riemann Zeta 函数: $\zeta(s) = \sum_{n=1}^{\infty} \frac{1}{n^s}.$
- 概率密度函数: $P(X = k) = \frac{1}{\zeta(s)k^s}, k = 1, 2, \dots.$

6.2 数字特征

- k 阶矩: $E(X^k) = \frac{\zeta(s-k)}{\zeta(s)}, s > k + 1.$
- 对数期望: $E(\ln X) = -\frac{\zeta'(s)}{\zeta(s)}, s > 1.$
- 信息熵: $H(X) = E(-\ln(\text{Zeta}(s))) = -\sum_{k=1}^{\infty} \frac{\ln \frac{1}{\zeta(s)k^s}}{\zeta(s)k^s} = \ln \zeta(s) - s \frac{\zeta'(s)}{\zeta(s)}.$

6.3 其它性质

• 问题 1 (最大熵分布)

对于取值为正整数的概率分布, 求给定对数期望的条件下熵最大的分布, 即

$$\begin{aligned} \max = & -\sum_k p_k \ln p_k \\ & \begin{cases} \sum_k p_k = 1, \\ \sum_k p_k \ln k = a. \end{cases} \end{aligned}$$

由 Lagrange 乘数法解得此分布即为 Zeta 分布.

• 性质 1

设 $X \sim \text{Zeta}(s)$, 则素因数分解中素数 p 的指数满足:

$$\nu_p(X) \sim \text{GE}(1 - p^{-s}).$$

证明

$$P(\nu_p(X) \geq k) = \frac{1}{\zeta(s)} \sum_{n=1}^{\infty} \frac{1}{(p^k n)^s} = \frac{1}{p^{ks}}$$

$$P(\nu_p(X) = k) = \frac{1}{p^{ks}} - \frac{1}{p^{(k+1)s}} = (1 - p^{-s})p^{-ks}$$

• 性质 2

设 $X \sim \text{Zeta}(s)$, 若 p 和 q 是两个互素的素数, 则 $\nu_p(X)$ 和 $\nu_q(X)$ 独立.

证明

$$P(\nu_p(X) \geq k, \nu_q(X) \geq l) = \frac{1}{(p^k q^l)^s} = \frac{1}{p^{ks} q^{ls}} = P(\nu_p(X) \geq k)P(\nu_q(X) \geq l).$$

• 性质 3

设 \mathbb{P} 是全体素数的集合, $\{X_p\}_{p \in \mathbb{P}}$ 是一组相互独立的随机变量, 其中 $X_p \sim \text{GE}(1 - p^{-s})$, 则

$$Z = \prod_{p \in \mathbb{P}} p^{X(1-p^{-s})} \sim Z(s).$$

证明 由 Euler 乘积公式 $\frac{1}{\zeta(s)} = \prod_{p \in \mathbb{P}} \left(1 - \frac{1}{p^s}\right)$ 得:

$$P(Z = k) = \prod_{p \in \mathbb{P}} \left(1 - \frac{1}{p^s}\right) \frac{1}{p^{\nu_p(k)s}} = \frac{1}{\zeta(s)} \prod_{p \in \mathbb{P}} \frac{1}{p^{\nu_p(k)s}} = \frac{1}{\zeta(s)k^s}.$$

• 性质 4

若 $X_1 \sim \text{Zeta}(s_1)$ 和 $X_2 \sim \text{Zeta}(s_2)$ 独立, 则

$$\gcd(X_1, X_2) \sim \text{Zeta}(s_1 + s_2).$$

证明

$$\nu_p(\gcd(X_1, X_2)) = \min\{\nu_p(X_1), \nu_p(X_2)\} \sim \text{GE}\left(1 - p^{-(s_1+s_2)}\right).$$

• 问题 2 (两个随机的正整数互素的概率)

X 和 Y 互素 $\Leftrightarrow \gcd(X, Y) = 1$.

正整数集上的均匀分布 $\sim \lim_{s \rightarrow 1^+} \text{Zeta}(s)$.

$$\lim_{s_1, s_2 \rightarrow 1^+} P(\gcd(X, Y) = 1) = \lim_{s_1, s_2 \rightarrow 1^+} \frac{1}{\zeta(s_1 + s_2)} = \frac{1}{\zeta(2)} = \frac{6}{\pi^2}.$$

注: 这并非严格的证明.

A.2.2 一维连续型

1 正态分布

点击查看 Geogebra 图像

或直接打开 [网页链接](#)

1.1 基础概念

- 正态分布又称高斯分布.
- $X \sim N(\mu, \sigma^2)$.
- 概率密度函数: $f(x) = (\sqrt{2\pi}\sigma)^{-1} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$.
- 标准正态分布: $Y = (X - \mu)/\sigma \sim N(0, 1)$.
- 3σ 原则: 0.6826, 0.9544, 0.9974.
- 上 α 分位数: $\Phi(z_\alpha) = 1 - \alpha$.
- $z_{1-\alpha}$

1.2 数字特征

- 期望: μ .
- 方差: σ^2 .
- 二阶原点矩: $\alpha_2 = E(X^2) = \sigma^2 + \mu^2$.
- k 阶中心矩: $\mu_k = \begin{cases} \sigma^k (k-1)!! & k \text{ 为偶数,} \\ 0 & k \text{ 为奇数.} \end{cases}$
- 偏度系数: $\beta_1 = 0$.
- 峰度系数: $\beta_2 = 3$.
- 特征函数: $g(t) = e^{i\mu t - \frac{\sigma^2}{2} t^2}$.
- 矩母函数: $\phi_X(t) = e^{\frac{t^2 \sigma^2}{2n} + \mu t}$.

1.3 其它性质

- $aN(\mu, \sigma^2) + b \sim N(a\mu + b, a^2\sigma^2)$.
- 若 X 和 Y 独立同分布 $N(0, 1)$, 则将 (X, Y) 化为极坐标 (R, Θ) 后, R 与 Θ 独立.
- 相互独立的正态分布的函数
 - 分布之和
 - 若 $X_1 \sim N(\mu_1, \sigma_1^2)$, $X_2 \sim N(\mu_2, \sigma_2^2)$ 相互独立, 则 $X_1 + X_2 \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$.
 - 若 $X_i \sim N(\mu_i, \sigma_i^2)$ 且相互独立, 则 $X_1 + \cdots + X_n \sim N(\mu_1 + \cdots + \mu_n, \sigma_1^2 + \cdots + \sigma_n^2)$.
 - 分布之差
 - 若 $X_1 \sim N(\mu_1, \sigma_1^2)$, $X_2 \sim N(\mu_2, \sigma_2^2)$ 相互独立, 则 $X_1 - X_2 \sim N(\mu_1 - \mu_2, \sigma_1^2 + \sigma_2^2)$.
 - 分布之商
 - 若 X_1 和 X_2 独立同分布 $N(0, 1)$, 则 $X_1/X_2 \sim C(1, 0)$ (柯西分布).
 - 分布之积

若 $X_1 \sim N(0, \sigma_1^2)$, $X_2 \sim N(0, \sigma_2^2)$, 则 $X_1 X_2 \sim \frac{1}{\pi \sigma_1 \sigma_2} K_0 \left(\frac{|z|}{\sigma_1 \sigma_2} \right)$ (修正贝塞尔函数; 暂时未学)

◦ 平方之和

若 X_1, X_2, \dots, X_n 独立同分布 $N(0, 1)$, 则 $Y = X_1^2 + X_2^2 + \dots + X_n^2 \sim \chi_n^2$.

• 统计量的分布

◦ \bar{X} 与 $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$ 独立.

◦ 均值已知, 标准差已知

- $\bar{X} \sim N \left(\mu, \frac{\sigma^2}{n} \right)$.
- $\frac{\sqrt{n}(\bar{X} - \mu)}{\sigma} \sim N(0, 1)$.
- $\sum_{i=1}^n \left(\frac{X_i - \mu}{\sigma} \right)^2 \sim \chi_n^2$.

◦ 均值已知, 标准差未知

- $\frac{\sqrt{n}(\bar{X} - \mu)}{S} \sim t_{n-1}$.

◦ 均值未知, 标准差已知

- $\frac{SS}{\sigma^2} = \frac{(n-1)S^2}{\sigma^2} = \sum_{i=1}^n \left(\frac{X_i - \bar{X}}{\sigma} \right)^2 \sim \chi_{n-1}^2$.

◦ 两份相互独立的样本

X_1, X_2, \dots, X_{n_1} , iid, $\sim N(\mu_1, \sigma_1^2)$.

Y_1, Y_2, \dots, Y_{n_2} , iid, $\sim N(\mu_2, \sigma_2^2)$.

- $\bar{X} - \bar{Y} \sim N \left(\mu_1 - \mu_2, \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2} \right)$.
- $\frac{S_1^2}{\sigma_1^2} \bigg/ \frac{S_2^2}{\sigma_2^2} \sim f(n_1 - 1, n_2 - 1)$.
- 当 $\sigma_1^2 = \sigma_2^2 = \sigma^2$ 时,

$$S_\omega := \sqrt{\frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}},$$

$$\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{S_\omega \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t_{n_1 + n_2 - 2}.$$

注: 利用 $\frac{\sqrt{n}(\bar{X} - \mu)}{\sigma}$ 和 $\frac{SS}{\sigma^2}$, 由 t 分布的定义即得.

1.4 参数估计

- 单个正态总体 $N(\mu, \sigma^2)$ 均值 μ 与方差 σ^2 的估计.

- 已知 σ^2 , 估计 μ .

- 矩估计: $\mu = m$.

注: 无论 σ^2 是否已知, 均为 MVU 估计.

- 区间估计 (枢轴变量法)

根据 $\frac{\sqrt{n}(\bar{X} - \mu)}{\sigma} \sim N(0, 1)$, 知

$$[\hat{\mu}_1, \hat{\mu}_2] = \left[\bar{X} - \frac{\sigma}{\sqrt{n}} u_{\alpha/2}, \bar{X} + \frac{\sigma}{\sqrt{n}} u_{\alpha/2} \right].$$

- 已知 μ , 估计 σ^2 .

- 矩估计: $\hat{\sigma}^2 = m_2$.

注: 这是 μ 已知时的 MVU 估计, 且此时均方误差为 $\frac{2}{n} \sigma^4$.

- 区间估计 (枢轴变量法)

根据 $\frac{1}{\sigma^2} \sum_{i=1}^n (X_i - \mu)^2 \sim \chi_n^2$ 知

$$[\hat{\sigma}_1^2, \hat{\sigma}_2^2] = \left[\frac{\sum_{i=1}^n (X_i - \mu)^2}{\chi_n^2 \left(\frac{\alpha}{2} \right)}, \frac{\sum_{i=1}^n (X_i - \mu)^2}{\chi_n^2 \left(\frac{\alpha}{2} \right)} \right].$$

- 估计 μ 和 σ^2 .

- 矩估计

- $\mu = m$.

注: 无论 σ^2 是否已知, 均为 MVU 估计.

- $\sigma^2 = S^2$.

注: 这是 μ 未知时的 MVU 估计.

- 极大似然估计

- $\mu = m$. (MVU 估计)

- $\sigma^2 = m_2$. (非 MVU 估计)

- 区间估计 (枢轴变量法)

- 根据 $\frac{\sqrt{n}(\bar{X} - \mu)}{S} \sim t_{n-1}$, 知一样本 t 区间估计为

$$[\hat{\mu}_1, \hat{\mu}_2] = \left[\bar{X} - \frac{S}{\sqrt{n}} t_{n-1} \left(\frac{\alpha}{2} \right), \bar{X} + \frac{S}{\sqrt{n}} t_{n-1} \left(\frac{\alpha}{2} \right) \right].$$

- 根据 $\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$, 知

$$[\hat{\sigma}_1^2, \hat{\sigma}_2^2] = \left[\frac{(n-1)S^2}{\chi_{n-1}^2\left(\frac{\alpha}{2}\right)}, \frac{(n-1)S^2}{\chi_{n-1}^2\left(1-\frac{\alpha}{2}\right)} \right].$$

- 无偏估计 (通过调整系数而得)

$$\tilde{\sigma} = \sqrt{\frac{n-1}{2}} \frac{\Gamma\left(\frac{n-1}{2}\right)}{\Gamma\left(\frac{n}{2}\right)} S.$$

- 两个正态总体 $N(\mu_1, \sigma_1^2)$ 和 $N(\mu_2, \sigma_2^2)$ 的均值差 $\mu_1 - \mu_2$ 与方差比 σ_1^2/σ_2^2 的区间估计.
 - 估计 $\delta = \mu_1 - \mu_2$.
 - 方差 σ_1^2 和 σ_2^2 已知.

$$U = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1),$$

$$[\hat{\delta}_1, \hat{\delta}_2] = \left[\bar{X} - \bar{Y} - z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}, \bar{X} - \bar{Y} + z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} \right].$$

- 方差 $\sigma_1^2 = \sigma_2^2 = \sigma^2$ 未知.

$$S_\omega^2 := \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2},$$

$$\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{S_\omega \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t_{n_1+n_2-2},$$

$$[\hat{\delta}_1, \hat{\delta}_2] = \left[\bar{X} - \bar{Y} - t_{n_1+n_2-2} \left(\frac{\alpha}{2}\right) S_\omega \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}, \bar{X} - \bar{Y} + t_{n_1+n_2-2} \left(\frac{\alpha}{2}\right) S_\omega \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} \right].$$

- 方差 σ_1^2 和 σ_2^2 未知.

即贝伦斯 - 费歇尔问题, 目前还没有较好的处理方法.

不过可以利用大样本法, 近似同方差已知的情况处理.

$$N(0, 1) \sim \left[(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2) \right] / \sqrt{\sigma_1^2/n + \sigma_2^2/m} \quad (\text{严格的})$$

$$\sim \left[(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2) \right] / \sqrt{S_1^2/n + S_2^2/m} \quad (\text{近似的})$$

- 估计 $\lambda = \sigma_1^2/\sigma_2^2$.
 - 均值 μ_1 和 μ_2 已知.

$$F = \frac{\sum_{i=1}^{n_2} \frac{(Y_i - \mu_2)^2}{n_2 \sigma_2^2}}{\sum_{i=1}^{n_1} \frac{(X_i - \mu_1)^2}{n_1 \sigma_1^2}} \sim F_{n_2, n_1},$$

$$[\hat{\lambda}_1, \hat{\lambda}_2] = \left[\frac{\sum_{i=1}^{n_1} \frac{(X_i - \mu_1)^2}{n_1}}{\sum_{i=1}^{n_2} \frac{(Y_i - \mu_2)^2}{n_2}} F_{n_2, n_1} \left(1 - \frac{\alpha}{2} \right), \frac{\sum_{i=1}^{n_1} \frac{(X_i - \mu_1)^2}{n_1}}{\sum_{i=1}^{n_2} \frac{(Y_i - \mu_2)^2}{n_2}} F_{n_2, n_1} \left(\frac{\alpha}{2} \right) \right].$$

- 均值 μ_1 和 μ_2 未知.

$$\frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \sim F_{n_1-1, n_2-1},$$

$$[\hat{\lambda}_1, \hat{\lambda}_2] = \left[\frac{S_1^2}{S_2^2} F_{n_2-1, n_1-1} (1 - \alpha), \frac{S_1^2}{S_2^2} F_{n_2-1, n_1-1} \left(\frac{\alpha}{2} \right) \right].$$

- 估计变异系数 σ/μ .
 - 矩估计: $\sqrt{m_2}/m$ 或 S/m .
- 估计 $N(\theta, 1)$ 的 θ .
 - 贝叶斯估计: 先验密度 $h(\theta) \sim N(\mu, \sigma^2)$, 则

$$\tilde{\theta} = \frac{n}{n+1/\sigma^2} \bar{X} + \frac{1/\sigma^2}{n+1/\sigma^2} \mu.$$

2 指数分布

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2.1 基础概念

- 指数分布又称为负指数分布.
- $X \sim E(\lambda)$.
- 概率密度函数: $f(x) = \begin{cases} \lambda e^{-\lambda x}, & x > 0, \\ 0, & x \leq 0. \end{cases}$
- 分布函数: $F(x) = \begin{cases} 1 - e^{-\lambda x}, & x > 0, \\ 0, & x \leq 0. \end{cases}$

2.2 数字特征

- 数学期望: $E[X] = \lambda^{-1}$.
- 方差: $\text{Var}[X] = \lambda^{-2}$.
- k 阶原点矩: $E(X^k) = \frac{k!}{\lambda^k}$.
- 特征函数: $g(t) = \frac{\lambda}{\lambda - it}$.
- 矩量母函数: $m_X(t) = \frac{\lambda}{\lambda - t}, t < \lambda$.

2.3 其它性质

- $aE(\lambda) = E\left(\frac{\lambda}{a}\right)$.
- 指数分布具有**无记忆性**, 即 $P(X > m + t | X > m) = P(X > t)$.
- 若有一批元件寿命 $X \sim E(\lambda)$, 让一个元件开始工作, 每当这个元件坏了就用一个新的替换, 则到经历时间 T 后替换的次数 $Y \sim P(\lambda T)$.
- 若 X_1, X_2, \dots, X_n 独立同分布 $E(\lambda)$, 则

$$Y = 2\lambda(X_1 + X_2 + \dots + X_n) \sim \chi_{2n}^2.$$

- 若 $X_i \sim E(\lambda_i)$ 相互独立, 则

$$Y = \min(X_1, X_2, \dots, X_n) \sim E(\lambda_1 + \lambda_2 + \dots + \lambda_n).$$

2.4 参数估计

- 矩估计: $1/\lambda = m$. (MVU 估计)
- 极大似然估计: $\lambda = 1/m$.
- 贝叶斯估计: 若先验密度为 $h(\lambda) = \lambda e^{-\lambda} (\lambda > 0)$, 其它值为零, 则 $\lambda = \frac{n+2}{n\bar{X}+1}$.
- 区间估计
 - 枢轴变量法
 - 估计 λ .
由 $2n\lambda\bar{X} \sim \chi_{2n}^2$, 知
$$[\hat{\theta}_1, \hat{\theta}_2] = \left[\chi_{2n}^2(1-\alpha/2)/(2n\bar{X}), \chi_{2n}^2(\alpha/2)/(2n\bar{X}) \right].$$
 - 估计 $1/\lambda$.
由 $2n\lambda\bar{X} \sim \chi_{2n}^2$, 知
$$[\hat{\theta}_1, \hat{\theta}_2] = \left[(2n\bar{X})/\chi_{2n}^2(1-\alpha/2), (2n\bar{X})/\chi_{2n}^2(\alpha/2) \right].$$
 - 区间估计 (枢轴变量法)
 - 利用 $\frac{4\lambda_1 n \bar{X}}{4\lambda_2 m \bar{Y}} \sim \frac{2n\chi_{2n}^2}{2m\chi_{2m}^2} \sim F_{2n, 2m}$.

3 混合指数分布

3.1 基础概念

- 混合指数分布又称为 **超指数分布** (Hyperexponential Distribution).
- 理解: 设有 m 个平行的服务台 $X_i \sim E(\lambda_i)$, 若顾客有 p_i 的概率选取第 i 个服务台, 则这样顾客的服务时间分布服从 m 阶超指数分布.

- 概率密度函数 $f(x) = \sum_{i=1}^m p_i \lambda_i e^{-\lambda_i x}, \quad (x > 0).$

其中 $\sum_{i=1}^m p_i = 1.$

- 累积分布函数 $F(x) = \sum_{i=1}^m p_i (1 - e^{-\lambda_i x}), \quad (x > 0).$

3.2 数字特征

- 和指数分布一样, 不再赘述. 如

$$k \text{ 阶原点矩 } E(X^k) = \sum_{i=1}^m \frac{k!}{\lambda_i^k} p_i.$$

3.3 其它性质

- 无记忆性.
- 若 $p_i = \frac{1}{m}, i = 1, 2, \dots, m$, 则 $X \sim \chi_{2m}^2$.

3.4 参数估计

- [矩估计](#).
- [优化估计](#).

4 均匀分布

4.1 基础概念

- $X \sim R(a, b).$
- 概率密度函数: $f(x) = \begin{cases} 1/(b-a), & a \leq x \leq b, \\ 0, & x < a \text{ 或 } x > b. \end{cases}$
- 分布函数: $F(x) = \begin{cases} 0, & x \leq a, \\ (x-a)/(b-a), & a < x < b, \\ 1, & x \geq b. \end{cases}$

4.2 数字特征

- 数学期望: $\frac{a+b}{2}.$
- 方差: $\frac{(b-a)^2}{12}.$
- k 阶原点矩: $\alpha_k = \frac{1}{k+1} \frac{b^{k+1} - a^{k+1}}{b-a}.$

- k 阶中心距: $\mu_k = \begin{cases} \frac{1}{k+1} \left(\frac{b-a}{2} \right)^k, & k \text{ 为偶数,} \\ 0, & k \text{ 为奇数.} \end{cases}$
- 偏度系数: $\beta_1 = 0$.
- 峰度系数: $\beta_2 = \frac{9}{5}$.
- 特征函数: $g(t) = \begin{cases} \frac{e^{ibt} - e^{iat}}{it(b-a)}, & t \neq 0, \\ 1, & t = 0. \end{cases}$

4.3 其它性质

- $cR(a, b) + d \sim R(ac + d, bc + d) \ (c > 0)$.
- 若 X_1, X_2, \dots, X_n 独立同分布 $U(a, b)$, 则

$$\begin{aligned} \max(X_1 + X_2 + \dots + X_n) &\sim f(x) = \frac{n(x-a)^{n-1}}{(b-a)^n} \\ \min(X_1 + X_2 + \dots + X_n) &\sim f(x) = \frac{n(b-x)^{n-1}}{(b-a)^n} \end{aligned}$$

- 若 $X \sim U\left(-\frac{\pi}{2}, \frac{\pi}{2}\right)$, 则 $\tan X \sim C(1, 0)$.

4.4 参数估计

- 估计 $R(\theta_1, \theta_2)$ 的参数.
 - 矩估计: $\theta_1 = m - \sqrt{3m_2}, \theta_2 = m + \sqrt{3m_2}$.
 - 极大似然估计: $\theta_1 = \min_i(X_i), \theta_2 = \max_i(X_i)$.
- 估计 $R(0, \theta)$ 的参数.
 - 极大似然估计: $\hat{\theta} = \max_i(X_i)$.
 - 无偏估计
 - $\hat{\theta} = \frac{n+1}{n} \max_i(X_i)$. (MVU 估计, 也是相合估计)
 - $\hat{\theta} = (n+1) \min_i(X_i)$. (方差很大)
 - $\hat{\theta} = \max_i(X_i) + \min_i(X_i)$.
 - 区间估计
 - 由 $\hat{\theta}_1 := \max_i(X_i) \sim F_{\hat{\theta}_1}(x) = \frac{nx^{n-1}}{\theta^n}$,
 $[\max(X_i), (1-\alpha)^{-\frac{1}{n}} \max(X_i)]$ 的置信系数为 $1-\alpha$.

5 对数正态分布

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5.1 基础概念

- $\ln X \sim N(\mu, \sigma^2)$.
- 概率密度函数:
$$f(x, \mu, \sigma) = \begin{cases} \left(x\sqrt{2\pi\sigma}\right)^{-1} \exp\left[-\frac{(\ln x - \mu)^2}{2\sigma^2}\right], & x > 0, \\ 0, & x \leq 0. \end{cases}$$

5.2 数字特征

- 期望: $E(X) = e^{\mu + \sigma^2/2}$.
- 方差: $\text{Var}(X) = (e^{\sigma^2} - 1)e^{2\mu + \sigma^2} = (e^{\sigma^2} - 1)E(X)^2$.
- k 阶原点矩: $\alpha_k = e^{\mu k + k^2 \sigma^2/2}$.
- 偏度系数: $\beta_1 = \frac{\mu_3}{\sigma^3} = \frac{e^{2\sigma^2} - 3e^{\sigma^2} + 1}{(e^{2\sigma^2} - 1)^{3/2}}$.
- 峰度系数: $\beta_2 = \frac{\mu_4}{\sigma^4} = (e^{\sigma^2} - 1)(e^{4\sigma^2} + 2e^{3\sigma^2} + 3e^{2\sigma^2} - 3) > 0$.

5.3 其它性质

- $\ln bX^a \sim N(a\mu + \ln b, a^2\sigma^2)$.
- 对数正态分布总是右偏的.
- 对数正态分布的期望和方差都是两个参数的增函数.
而在正态分布中, 期望与 σ 无关, 方差与 μ 无关.
- $\lim_{\sigma \rightarrow 0+} E(X) = e^\mu$.
 $\lim_{\sigma \rightarrow 0+} \text{Var}(X) = 0$.
当 $\mu = 0$ 时, $E(X^k) = E(X)^{k^2}$.

5.4 参数估计

- 矩估计
 - $\hat{\sigma}^2 = \ln\left(1 + \frac{S^2}{\bar{X}^2}\right)$.
 - $\hat{\mu} = 2 \ln \bar{X} - \frac{1}{2} \ln(\bar{X}^2 + S^2)$.

6 柯西分布

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6.1 基础概念

- $X \sim C(\gamma, x_0)$.
- 概率密度函数: $f(x; x_0, \gamma) = \frac{1}{\pi} \cdot \frac{\gamma}{(x - x_0)^2 + \gamma^2} \quad (-\infty < x < +\infty)$.
- 累积分布函数: $F(x; x_0, \gamma) = \frac{1}{\pi} \arctan \frac{x - x_0}{\gamma} + \frac{1}{2}$.
- 标准柯西分布: $C(1, 0) \sim t_1$.

- 广义柯西分布: $X_k \sim f_m(X_k | \sigma_X) = \frac{a_m}{1 + \left(\frac{X_k^2}{2\sigma_k^2}\right)^m} \quad (a_m > 0.5).$

6.2 数字特征

- 数学期望 **不存在**. (仅 Cauchy 主值积分存在)
- 方差不存在.
- 高阶矩不存在.

6.3 其它性质

- 可加性: 若 X_i 独立同分布 $C(\gamma, x_0)$, 则 $X_1 + X_2 + \cdots + X_n \sim C(n\gamma, nx_0)$.
- 若 X_1 和 X_2 独立同分布 $N(0, 1)$, 则 $\frac{X_1}{X_2} \sim C(1, 0)$.
- 若 $X \sim U\left(-\frac{\pi}{2}, \frac{\pi}{2}\right)$, 则 $\tan X \sim C(1, 0)$.

6.4 参数估计

- 参数估计: 可使用样本中位数 \tilde{m} 估计.

7 拉普拉斯分布

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7.1 基础概念

- $X \sim \text{La}(\mu, \lambda)$.
- 概率密度函数: $f(x) = \frac{1}{2\lambda} e^{-\frac{|x-\mu|}{\lambda}}$.
- 累积分布函数: $F(x) = \begin{cases} \frac{1}{2} e^{\frac{x-\mu}{\lambda}}, & x < \mu. \\ 1 - \frac{1}{2} e^{\frac{\mu-x}{\lambda}}, & x \geq \mu. \end{cases}$
- 参数说明
 - μ 是位置参数.
 - γ 是尺度参数, 越小曲线越陡.
 - 当 $\mu = 0$ 时, 正半部分是指数分布 $E(\lambda^{-1})$ 概率密度的一半.

7.2 数字特征

- 矩及相关量
 - k 阶中心距: $\mu_k = E[(X - \mu)^k] = \begin{cases} 0, & k \text{ 为奇数,} \\ k! \lambda^k, & k \text{ 为偶数.} \end{cases}$
 - 期望 $E(X) = \mu$.
 - 方差 $\text{Var}(X) = 2\lambda^2$.
- 同指数分布 (注意与 k 阶中心距、期望和方差比较)
 - $E(|X - \mu|^k) = k! \lambda^k$.
 - $E(|X - \mu|) = \lambda$.

- $\text{Var}(|X - \mu|) = \lambda^2$.
- 一些系数
 - 偏度系数 $\beta_1 = \frac{\mu_3}{\sigma^3} = 0$.
 - 峰度系数 $\beta_2 = \frac{\mu_4}{\sigma^4} = 6$.
- 相关函数
 - 矩量母函数 $m(t) = \frac{e^{\frac{\mu t}{\lambda}}}{1 - \lambda^2 t^2}$.
 - 特征函数 $g(t) = E(e^{itX}) = \frac{e^{\frac{i\mu t}{\lambda}}}{1 + \lambda^2 t^2}$.

7.3 其它性质

- $a\text{La}(\mu, \lambda) + b \sim \text{La}(a\mu + b, a\lambda)$.
 $\text{La}(\mu, \lambda) \sim \lambda\text{La}(0, 1) + \mu$.
- 注意到 Laplace 分布与指数分布的关系, 可以立即得到如下平凡的结论
 - 对于 $X \sim \text{La}(\alpha, \beta)$, 有 $\frac{2}{\beta}|X - \alpha| \sim \chi_2^2 \sim \Gamma(1, 2)$.
 - 若 X 和 Y 独立同分布于 $\text{La}(\alpha, \beta)$, 则 $\frac{|X - \alpha|}{|Y - \alpha|} \sim F_{2,2}$.
 - 若 $X_{11}, X_{12}, X_{21}, X_{22}$ 独立同分布于 $N(0, 1)$, 则 $D = \begin{vmatrix} X & Y \\ Z & W \end{vmatrix} \sim \text{La}(0, 2)$.
- 与稳健性的联系
 古典回归分析中, 用偏差平方和的大小作为标准, 这种回归不具有稳健性.
 而改成偏差的绝对值和作为标准, 却具有稳健性 (尽管求解更加困难).
- 标准 Laplace 分布
 - 概率密度: $\text{La}(x; 0, 1) = \frac{e^{-|x|}}{2}$.
 - 特征函数: $\phi_L(t) = \frac{1}{1 + t^2}$.

7.4 参数估计

- 估计 μ .
 - 矩估计: $\hat{\mu} = \bar{X}$.
 - 极大似然估计: $\hat{\mu} = m_e$ (中位数).
- 估计 λ .
 - 矩估计: $\hat{\lambda} = \frac{\sqrt{2}}{2} S$ (标准差).
 - 类似矩估计的估计:

$$\hat{\lambda} = \frac{1}{n} \sum_{i=1}^n |X_i - \mu| \approx \frac{1}{n} \sum_{i=1}^n |X_i - \hat{\mu}|.$$

8 卡方分布

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8.1 基础概念

- 自由度为 n 的皮尔逊卡方密度与卡方分布 $X \sim \chi_n^2$.
- 概率密度函数

$$k_n(x) = \begin{cases} \frac{e^{-\frac{x}{2}} x^{\frac{n}{2}-1}}{\Gamma\left(\frac{n}{2}\right) 2^{\frac{n}{2}}}, & x > 0, \\ 0, & x \leq 0. \end{cases}$$

- 例子 (以下 $x > 0$)

$$\begin{aligned} k_1(x) &= \frac{e^{-\frac{x}{2}}}{\sqrt{2\pi x}} & k_2(x) &= \frac{1}{2} e^{-\frac{x}{2}} \\ k_3(x) &= \frac{\sqrt{x} e^{-\frac{x}{2}}}{\sqrt{2\pi}} & k_4(x) &= \frac{x}{4} e^{-\frac{x}{2}} \end{aligned}$$

- 上 α 分位数 $\chi_\alpha^2(n)$.
- 由中心极限定理近似求值 $X \sim \chi_n^2$,

$$\frac{X - n}{\sqrt{2n}} \sim N(0, 1) \Rightarrow \frac{\chi_\alpha^2(n) - n}{\sqrt{2n}} \approx z_\alpha \Rightarrow \chi_\alpha^2(n) \approx n + z_\alpha \sqrt{2n}.$$

8.2 数字特征

- $E(X) = n$.
- $\text{Var}(X) = 2n$.

注意到方差是均值的两倍, 可以以此检验是否为卡方分布.

- $E(X^{-1}) = \frac{1}{n-2}$.
- $E(X^k) = \frac{2^k \Gamma\left(\frac{n}{2} + k\right)}{\Gamma\left(\frac{n}{2}\right)} = 2^k \left(\frac{n}{2}\right)^{(k)}$.
- 特征函数: $g(t) = \frac{1}{(1 - 2it)^{\frac{n}{2}}}$.
- 矩量母函数: $m_X(t) = \frac{1}{(1 - 2t)^{\frac{n}{2}}}, \quad t < \frac{1}{2}$.

8.3 其它性质

- 若 X_1, X_2, \dots, X_n 独立同分布 $N(0, 1)$, 则

$$Y = X_1^2 + X_2^2 + \dots + X_n^2 \sim \chi_n^2.$$

- 若 $X_1 \sim \chi_m^2$ 与 $X_2 \sim \chi_n^2$ 独立, 则

$$X_1 + X_2 \sim \chi_{m+n}^2.$$

- 若 X_1, X_2, \dots, X_n 独立同分布 $E(\lambda)$, 则

$$X = 2\lambda(X_1 + X_2 + \dots + X_n) \sim \chi_{2n}^2.$$

- 若 X_1, X_2, \dots, X_n 独立同分布 $N(\mu, \sigma^2)$, 则

$$\frac{SS}{\sigma^2} = \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2.$$

9 t 分布

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9.1 基础概念

- 自由度为 n 的 t 分布 $X \sim t_n$.
- 概率密度函数

$$t_n(x) = \frac{\Gamma\left(\frac{n+1}{2}\right)}{\sqrt{n\pi} \Gamma\left(\frac{n}{2}\right)} \left(1 + \frac{x^2}{n}\right)^{-\frac{n+1}{2}} = \frac{\left(1 + \frac{x^2}{n}\right)^{-\frac{n+1}{2}}}{B\left(\frac{n}{2}, \frac{1}{2}\right) \sqrt{n}}, \quad -\infty < x < +\infty.$$

- 上 α 分位数 $t_\alpha(n) \approx z_\alpha$. (正态分布的上 α 分位数)
- 由对称性知: $t_{1-\alpha}(n) = -t_\alpha(n)$.

9.2 数字特征

- $E(t_n) = 0$ ($n > 1$).
- $\text{Var}(t_n) = \frac{n}{n-2}$ ($n > 2$).
- $E(X^k) = \frac{B\left(\frac{n-k}{2}, \frac{k+1}{2}\right)}{B\left(\frac{n}{2}, \frac{1}{2}\right)} n^{\frac{k}{2}}$ ($-1 < k < n$).

9.3 其它性质

- 若 $X \sim t_n$, 则 $X^2 \sim F_{1,n}$.
- 若 $X \sim N(0, 1)$ 与 $Y \sim \chi_n^2$ 独立, 则

$$\frac{X}{\sqrt{Y/n}} \sim t_n.$$

- 若 X_1, X_2, \dots, X_n 独立同分布 $N(\mu, \sigma^2)$, 则

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{n-1}.$$

- 设 X_1, X_2, \dots, X_n 独立同分布 $N(\mu_1, \sigma^2)$, Y_1, Y_2, \dots, Y_m 独立同分布 $N(\mu_2, \sigma^2)$, 且 X_i, Y_j 独立, 则

$$\frac{\sqrt{\frac{nm(n+m-2)}{n+m}} \left[(\bar{X} + \bar{Y}) - (\mu_1 - \mu_2) \right]}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2 + \sum_{j=1}^m (Y_j - \bar{Y})^2}} \sim t_{n+m-2}.$$

10 F 分布

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10.1 基础概念

- 自由度为 (m, n) 的 F 分布 $X \sim F_{m,n}$.
- 概率密度函数

$$\begin{aligned} f_{m,n}(x) &= \frac{m^{\frac{m}{2}} n^{\frac{n}{2}} x^{\frac{m}{2}-1} I_{\{x \geq 0\}}}{B\left(\frac{m}{2}, \frac{n}{2}\right) (mx + n)^{\frac{m+n}{2}}} \\ &= \frac{\left(\frac{m}{n}\right)^{\frac{m}{2}} x^{\frac{m}{2}-1} I_{\{x \geq 0\}}}{B\left(\frac{m}{2}, \frac{n}{2}\right) \left(1 + \frac{m}{n}x\right)^{\frac{m+n}{2}}}. \end{aligned}$$

- 若 $F \sim F(m, n)$, 则 $F^{-1} \sim F(n, m)$.
- 第一自由度为 n_1 , 第二自由度为 n_2 的 F 分布的上 α 分位数 $F_\alpha(n_1, n_2)$.
- $F_\alpha(m, n) \cdot F_{1-\alpha}(n, m) = 1$.

10.2 数字特征

- $E(f_{m,n}) = \frac{n}{n-2} \ (n > 2)$.
- $\text{Var}(f_{m,n}) = \frac{2n^2(m+n-2)}{m(n-2)^2(n-4)}$.
- $E(X^k) = \frac{B\left(\frac{m}{2} + k, \frac{n}{2} - k\right)}{\left(\frac{m}{n}\right)^k B\left(\frac{m}{2}, \frac{n}{2}\right)}$.

10.3 其它性质

- 设 X_1, X_2 独立, $X_1 \sim \chi_m^2$, $X_2 \sim \chi_n^2$, 则

$$\frac{X_1}{m} \bigg/ \frac{X_2}{n} \sim F_{m,n}.$$

- 设 X_1, X_2, \dots, X_n 独立同分布 $N(\mu_1, \sigma_1^2)$, Y_1, Y_2, \dots, Y_m 独立同分布 $N(\mu_2, \sigma_2^2)$, 且 X_i, Y_j 独立, 则

$$\frac{S_Y}{\sigma_2^2} \bigg/ \frac{S_X}{\sigma_1^2} \sim F_{m-1, n-1}.$$

- $\forall k, n, \in \mathbb{N}, a \in (0, 1) : kF_{k,n}(a) \geq F_{1,n}(a)$.

11 贝塔分布

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11.1 基础概念

- $X \sim \text{Be}(\alpha, \beta)$ ($\alpha, \beta > 0$).
- 概率密度函数 $f(x; \alpha, \beta) = \frac{x^{\alpha-1}(1-x)^{\beta-1}}{\text{B}(\alpha, \beta)}$ ($0 < x < 1$).
- 累积分布函数 $F(x; \alpha, \beta) = \frac{\text{B}_x(\alpha, \beta)}{\text{B}(\alpha, \beta)} = I_x(\alpha, \beta)$.
 - 不完全 Beta 函数 $\text{B}_x(\alpha, \beta)$.
 - 正则不完全 Beta 函数 $I_x(\alpha, \beta)$.

11.2 数字特征

- 常用统计量
 - 众数 $M_0 = \frac{\alpha - 1}{\alpha + \beta - 2}$. (伯努利分布参数的极大似然估计)
 - 期望 $E(X) = \frac{\alpha}{\alpha + \beta}$. (伯努利分布参数的贝叶斯估计 & 同等无知原则)
 - 方差 $\text{Var}(X) = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$.
- 矩及相关量
 - k 阶矩 $E(X^k) = \frac{\text{B}(\alpha + k, \beta)}{\text{B}(\alpha, \beta)} = \frac{(\alpha)^{(k)}}{(\alpha + \beta)^{(k)}} = \frac{\alpha + k - 1}{\alpha + \beta + k - 1} E(X^{k-1})$.
 - 偏度 $\beta_1 = \frac{\mu_3}{\sigma^3} = \frac{2(\beta - \alpha)\sqrt{\alpha + \beta + 1}}{(\alpha + \beta + 2)\sqrt{\alpha\beta}}$.
 - 峰度 $\beta_2 = \frac{\mu_4}{\mu_2^2} - 3 = \frac{6[(\alpha - \beta)^2(\alpha + \beta + 1) - \alpha\beta(\alpha + \beta + 2)]}{\alpha\beta(\alpha + \beta + 2)(\alpha + \beta + 3)}$.

11.3 其它性质

- Beta 分布即伯努利分布的共轭先验分布.
- $E(\ln X) = \psi(\alpha) - \psi(\alpha + \beta)$.
 - $\text{B}_{p,q}(\alpha, \beta) := \frac{\partial^{p+q} \text{B}(x, y)}{\partial^p x \partial^q y} = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} \ln^p x \ln^q (1-x) dx$.
 - $\text{B}_{1,0}(x, y) = \text{B}(x, y)(\psi(x) - \psi(x + y))$.
 - $\text{B}_{0,1}(x, y) = \text{B}(x, y)(\psi(y) - \psi(x + y))$.
- Beta 分布与 Gamma 分布的关系.

12 伽马分布

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12.1 基础概念

- $X \sim \text{Ga}(\alpha, \beta) \sim \Gamma\left(\alpha, \frac{1}{\beta}\right)$.
- 概率密度函数 $f(x; \alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \quad (x > 0)$.
 - α 称为形状参数.
 - β 称为逆尺度参数.
- 累积分布函数 $F(x; \alpha, \beta) = \frac{\gamma(\alpha, \beta x)}{\Gamma(\alpha)}$.
 - 其中 $\gamma(s, x) = \int_0^x t^{s-1} e^{-t} dt$ 为下不完全 Gamma 函数.
 - 此外 $\Gamma(s, x) = \int_x^{+\infty} t^{s-1} e^{-t} dt$ 为上不完全 Gamma 函数.
- 注意区分
 - Gamma 分布 $\text{Ga}(\alpha, \beta)$ 或 $\text{Gamma}(\alpha, \beta)$, 其累积分布函数如上所示.
 - Gamma 分布的另一种定义 $\Gamma(\alpha, \beta)$, 其累积分布函数为 $\frac{\gamma(\alpha, x/\beta)}{\Gamma(\alpha)}$.
 - 上述定义的密度函数 $\Gamma(x; \alpha, \beta)$ 或 $\Gamma(X | \alpha, \beta)$.
 - 上不完全 Gamma 函数 $\Gamma(s, x)$ 和 Gamma 函数 $\Gamma(s)$.
- 当 $\alpha \in \mathbb{N}$ 时, 退化为埃尔朗分布.

12.2 数字特征

- 有量纲参数
 - 众数 $M_0 = \frac{\alpha - 1}{\beta} \quad (\alpha > 1)$.
 - k 阶原点矩 $\alpha_k = E(X^k) = \frac{\Gamma(\alpha + k)}{\beta^k \Gamma(\alpha)} = \frac{(\alpha)^{(k)}}{\beta^k}$. (上升阶乘幂)
 - 期望 $\mu = E(X) = \frac{\alpha}{\beta}$.
 - 方差 $\sigma^2 = \text{Var}(X) = \frac{\alpha}{\beta^2}$.
- 无量纲参数
 - 偏度系数 $\beta_1 = \frac{\mu_3}{\sigma^3} = \frac{2}{\sqrt{\alpha}}$.
 - 峰度系数 $\beta_2 = \frac{\mu_4}{\sigma^4} = \frac{6}{\alpha}$.
 - 变异系数 $c_v = \frac{\sigma}{\mu} = \frac{1}{\sqrt{\alpha}}$.
- 特征函数 $g(t) = \left(1 - \frac{it}{\beta}\right)^{-\alpha}$.
- 矩量母函数 $m_X(t) = m_X(t) = \left(1 - \frac{t}{\beta}\right)^{-\alpha}, t < \beta$.

12.3 其它性质

- 变化趋势
 - 当 $\alpha \in (0, 1]$ 时, $f(x; \alpha, \beta)$ 递减.
 - 当 $\alpha \in (1, +\infty)$ 时, $f(x; \alpha, \beta)$ 先增后减, 为单峰函数.
 - 无量纲参数与图像形状仅与 α 有关, 故 α 称为形状参数.
- 特殊情况
 - 指数分布 $\text{Ga}(1, \lambda) \sim E(\lambda)$.
另一定义 $\Gamma\left(1, \frac{1}{\lambda}\right) \sim E(\lambda)$. (此外也有一种定义, 使得 $\Gamma(1, \lambda) \sim E(\lambda)$, 问就是别用)
 - 卡方分布 $\text{Ga}\left(\frac{n}{2}, \frac{1}{2}\right) \sim \chi_n^2$.
另一定义 $\Gamma\left(\frac{n}{2}, 2\right) \sim \chi_n^2$.
- 函数运算
 - 数乘
 - 若 $X \sim \text{Ga}(\alpha, \beta)$, 则 $\lambda X \sim \text{Ga}\left(\alpha, \frac{\beta}{\lambda}\right)$.
 - 因此 β 称为尺度参数或逆尺度参数, 即 β 越大, 曲线越窄, 图像越接近 y 轴.
 - 可加性
 - 若 $X_1 \sim \text{Ga}(\alpha_1, \beta)$ 和 $X_2 \sim \text{Ga}(\alpha_2, \beta)$ 独立, 则 $X + Y \sim \text{Ga}(\alpha_1 + \alpha_2, \beta)$.
 - 特例 1 (卡方分布): 若 $X_m \sim \chi_m^2$, $X_n \sim \chi_n^2$ 独立, 则 $\chi_m^2 + \chi_n^2 \sim \chi_{m+n}^2$.
 - 特例 2 (正态分布): X_i 独立同分布 $N(0, 1)$, 则 $X_1^2 + X_2^2 + \cdots + X_n^2 \sim \chi_n^2$.
 - 特例 3 (指数分布): 若 X_i 独立同分布 $E(\lambda)$, 则 $2\lambda(X_1 + X_2 + \cdots + X_n) \sim \chi_{2n}^2$.

13 威布尔分布

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13.1 基础概念

- 概率密度函数: $f(x; \lambda, k) = \begin{cases} \frac{k}{\lambda} \left(\frac{x}{\lambda}\right)^{k-1} e^{-\left(\frac{x}{\lambda}\right)^k}, & x \geq 0, \\ 0, & x < 0. \end{cases}$
- 累积分布函数: $F(x) = \begin{cases} 1 - e^{-\left(\frac{x}{\lambda}\right)^k}, & x \geq 0, \\ 0, & x \leq 0. \end{cases}$

13.2 数字特征

- n 阶原点矩: $E(X^n) = \lambda^n \Gamma\left(1 + \frac{n}{k}\right)$.
- 期望、方差、偏度、峰度等可由原点矩直接得到, 形式复杂故不再列出.

14 瑞利分布

14.1 基础概念

- 概率密度函数: $f(x) = \frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}, \quad x > 0.$
- 累积分布函数: $F(x) = 1 - e^{-\frac{x^2}{2\sigma^2}}, \quad x > 0.$

14.2 数字特征

- k 阶原点矩: $E(X^k) = (2\sigma^2)^k \Gamma\left(1 + \frac{k}{2}\right).$
- 期望: $E(X) = \sqrt{\frac{\pi}{2}}\sigma \approx 1.253\sigma.$
- 方差: $\text{Var}(X) = \frac{4 - \pi}{2}\sigma^2 \approx 0.429\sigma^2.$

15 帕累托分布

15.1 基础概念

- 帕累托分布 (Pareto Distribution) 又称布拉德福分布, 与幂律分布形式相同. 参考[齐夫定律](#).

- 概率密度函数: $f(x) = \begin{cases} 0, & x < x_{\min}, \\ \frac{kx_{\min}^k}{x^{k+1}}, & x \geq x_{\min}. \end{cases}$
- 互补累积分布函数: $P(X > x) = \begin{cases} 0, & x < x_{\min}, \\ \left(\frac{x_{\min}}{x}\right)^k, & x \geq x_{\min}. \end{cases}$

互补累积分布函数又称为生存函数, 残存函数或可靠性函数.

- 大致服从帕累托分布的例子
 - 个人财富或资源的分布.
 - 人类居住区的大小.
 - 对百科条目的访问.
 - 龙卷风带来的灾难的数量.

15.2 数字特征

- 方便起见, 改变记号如下: $f(x) = \frac{\alpha\theta^\alpha}{x^{\alpha+1}}, \quad x \geq \theta.$
- k 阶矩: $E(X^k) = \frac{\alpha\theta^k}{\alpha - k}, \quad (\alpha > k).$
- 期望: $E(X) = \frac{\alpha\theta}{\alpha - 1}, \quad (\alpha > 1).$
- 方差: $\text{Var}(X) = \frac{\alpha\theta^2}{(\alpha - 1)^2(\alpha - 2)}, \quad (\alpha > 2).$

16 逻辑斯蒂分布

16.1 基础概念

- $X \sim L(\mu, \gamma)$.
- Logistic 分布属于位置-尺度参数族.
- 累积分布函数: $F(x) = \frac{1}{1 + e^{-\frac{x-\mu}{\gamma}}} = \frac{1}{2} \left(1 + \tanh \frac{x-\mu}{2\gamma} \right), \quad x \in \mathbb{R}, \gamma > 0.$
- 概率密度函数: $f(x) = \frac{e^{-\frac{x-\mu}{\gamma}}}{\gamma \left(1 + e^{-\frac{x-\mu}{\gamma}} \right)^2}.$
- 参数说明
 - μ 是位置参数, 称为 **散布中心**.
 - γ 是尺度参数, 称为 **散布程度**.
 - 当 $\mu = 0$ 时, $\gamma = \pm\gamma_0$ 的分布相同.
- 标准 Logistic 分布 $L(0, 1)$.
 - 累积分布函数: $F_0(x) = \frac{1}{1 + e^{-x}}.$
 - 概率密度函数: $f_0(x) = \frac{e^{-x}}{(1 + e^{-x})^2}.$

16.2 数字特征

- 期望 $E(X) = \mu.$
- 方差 $\text{Var}(X) = \frac{\gamma^2 \pi^2}{3}.$

16.3 其它性质

- $aL(\mu, \gamma) + b \sim L(a\mu + b, a\gamma).$
 $L(\mu, \gamma) \sim \gamma L(0, 1) + \mu.$
- 图像特征: $F(\mu - x) + F(\mu + x) = 1.$
- 回归模型:

$$P_i = \frac{1}{1 + e^{-(a+bx_i)}} \Rightarrow \ln \left(\frac{P_i}{1 - P_i} \right) = a + bx_i.$$

- 推广: 多元 Logistic 函数 $y = (1 + e^{-\beta x})^{-1}.$

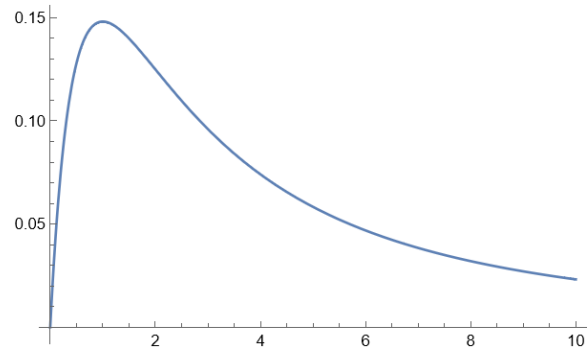
17 广义贝塔分布

不贴 Geogebra 图像的链接了, 给个 mathematica 绘图代码:

```

1  \[Alpha] = 2; \[Beta] = 1; \[Gamma] = 3; \[Lambda] = 0.5;
2  Plot[(
3      \[Lambda]^\[Alpha] \[Beta])
4      x^\[Alpha] - 1
5  ) / (
6      Beta[
7          \[Gamma] - \[Alpha] / \[Beta],
8          \[Alpha] / \[Beta]
9      ] (
10         1 + (\[Lambda] x)^\[Beta]
11     )^\[Gamma]
12 ), {x, 0, 10}]

```



17.1 基础概念

- $X \sim \text{GBeta}(\alpha, \beta, \gamma, \lambda)$. (随便命的名, 不知道有没有人研究过这东西)
- 参数定义域
 - $\lambda > 0$.
 - $\gamma > \frac{\alpha}{\beta} > 0$.
- 概率密度函数

$$f(x; \alpha, \beta, \gamma, \lambda) = \frac{\lambda^\alpha |\beta|}{B\left(\gamma - \frac{\alpha}{\beta}, \frac{\alpha}{\beta}\right)} \frac{x^{\alpha-1}}{[1 + (\lambda x)^\beta]^\gamma} I_{\{x \geq 0\}}.$$

- 其中示性函数 $I_{\{x \geq 0\}} = \begin{cases} 1, & x \geq 0, \\ 0, & x < 0 \end{cases}$
- 作变量代换 $t = \frac{\lambda x}{1 + \lambda x}$, 则可验证 $\int_0^{+\infty} f dx = 1$.

17.2 数字特征

$$E(X^k) = \frac{B\left(\gamma - \frac{\alpha + k}{\beta}, \frac{\alpha + k}{\beta}\right)}{\lambda^k B\left(\gamma - \frac{\alpha}{\beta}, \frac{\alpha}{\beta}\right)} = \lambda^{-k} \left(\gamma - \frac{\alpha}{\beta}\right)^{\left(\frac{k}{\beta}\right)} \left(\frac{\alpha}{\beta}\right)^{\left(\frac{k}{\beta}\right)}.$$

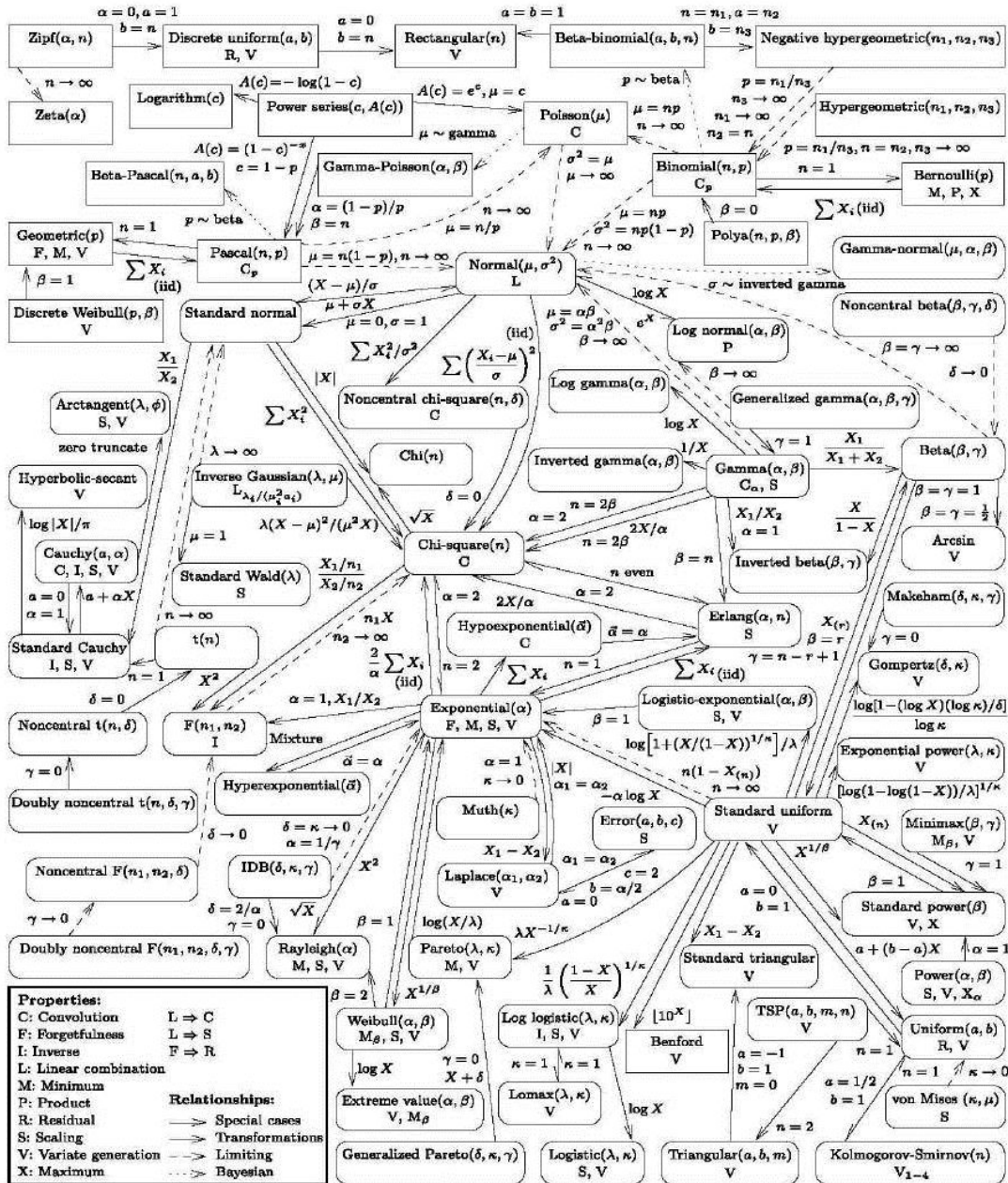
其中上升阶乘幂定义为 $(\alpha)^{(\beta)} = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)}$.

17.3 其它性质

- 特殊情况 (于是可以直接得到 t 分布与 F 分布的 k 阶矩)
 - t 分布: 若 $X \sim \text{GBeta}\left(x; 1, 2, \frac{n+1}{2}, \frac{1}{\sqrt{n}}\right)$, 则 $|X| \sim t_n$.
 - F 分布: 若 $X \sim \text{GBeta}\left(\frac{m}{2}, 1, \frac{m+n}{2}, \frac{m}{n}\right)$, 则 $X \sim F_{m,n}$.

其它分布

- 超指数分布
- Dirichlet 分布
- 广义 Dirichlet 分布
- 组合 Dirichlet 分布
- 刘维尔分布
- 威布尔分布
- 埃尔朗分布
- 帕累托分布



A.2.3 多维离散型

1 多项分布

$$X = (X_1, \dots, X_n) \sim M(N; p_1, \dots, p_n).$$

$$P(X_1 = k_1, X_2 = k_2, \dots, X_n = k_n) = \frac{N!}{k_1! k_2! \dots k_n!} p_1^{k_1} p_2^{k_2} \dots p_n^{k_n}.$$

多项分布的边缘分布是二项分布.

$$(X_1, X_2, \dots, X_n) \sim M(N; p_1, p_2, \dots, p_n) \Rightarrow X_1 + X_2 \sim B(N; p_1 + p_2).$$

A.2.4 多维连续型

1 矩形均匀分布

2 二维正态分布

2.1 基础概念

- $X = (X_1, X_2) \sim N(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2, \rho)$.

$$f(x_1, x_2) = (2\pi\sigma_1\sigma_2\sqrt{1-\rho^2})^{-1} \exp \left[-\frac{1}{2(1-\rho^2)} \left(\frac{(x_1-\mu_1)^2}{\sigma_1^2} - \frac{2\rho(x_1-\mu_1)(x_2-\mu_2)}{\sigma_1\sigma_2} + \frac{(x_2-\mu_2)^2}{\sigma_2^2} \right) \right].$$

- 当且仅当 $\rho = 0$ 时, X_1 和 X_2 独立.
- 边缘分布 $X_1 \sim N(\mu_1, \sigma_1^2)$, $X_2 \sim N(\mu_2, \sigma_2^2)$.

2.2 数字特征

- 相关系数 $\text{Corr}(X_1, X_2) = \rho$.
- 协方差 $\text{Cov}(X_1, X_2) = \rho\sigma_1\sigma_2$.
- 期望 $E(X_1X_2) = \text{Cov}(X_1, X_2) + E(X_1)E(X_2) = \rho\sigma_1\sigma_2 + \mu_1\mu_2$.

2.3 其它性质

- 二维正态分布的[边缘分布](#)是正态分布.
- 二维正态分布的[条件分布](#)是正态分布.

若 $(X, Y) \sim N(a, b, \sigma_1^2, \sigma_2^2, \rho)$, 则给定 $X = x$ 时 Y 的条件分布为

$$N(b + \rho\sigma_2\sigma_1^{-1}(x - a), \sigma_2^2(1 - \rho^2)).$$

- 二维正态分布的[边缘分布的和](#)仍为正态分布
若 $(X_1, X_2) \sim N(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2, \rho)$, 则 $Y = X_1 + X_2 \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2 + 2\rho\sigma_1\sigma_2)$.
- [独立](#)的正态分布的[联合分布](#)是正态分布.
正态分布的联合分布不一定是二维正态分布.
- 若 $Y = X_1 + X_2$ 服从正态分布, X_1, X_2 独立, 则 X_1, X_2 也是正态分布. ★

3 多元正态分布

3.1 基础概念

- 设 (X_1, X_2, \dots, X_n) 为 n 元随机变量, 令

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad \boldsymbol{\mu} = \begin{pmatrix} \mu_1 \\ \mu_2 \\ \vdots \\ \mu_n \end{pmatrix}, \quad \mathbf{C} = \begin{pmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{pmatrix},$$

其中 \mathbf{C} 为[协方差矩阵](#), $c_{ij} = \text{Cov}(X_i, X_j) = \rho_{ij}\sigma_i\sigma_j$.

如果 (X_1, X_2, \dots, X_n) 的概率密度函数为

$$f(x_1, x_2, \dots, x_n) = \frac{e^{-\frac{1}{2}(\mathbf{x}-\boldsymbol{\mu})^T \mathbf{C}^{-1}(\mathbf{x}-\boldsymbol{\mu})}}{(2\pi)^{\frac{n}{2}} |\mathbf{C}|^{\frac{1}{2}}}$$

则称 (X_1, X_2, \dots, X_n) 是参数为 $\boldsymbol{\mu}$, \mathbf{C} 的 n 元正态变量.

3.2 数字特征

- 方差: $\text{Var}(X_i) = c_{ii}$.
- 协方差: $\text{Cov}(X_i, X_j) = c_{ij}$.
- 相关系数: $\text{Corr}(X_i, X_j) = \rho_{ij} = \frac{c_{ij}}{\sqrt{c_{ii}c_{jj}}}$.
- 数学期望: $E(X_i X_j) = c_{ij} + \mu_1 \mu_2$.

3.3 其它性质

- n 维正态分布的[边缘分布](#)是正态分布.
- n 维正态分布的[条件分布](#)是正态分布.
- n 维正态分布的[边缘分布的和](#)是正态分布.
- n 维随机变量 (X_1, X_2, \dots, X_n) 服从 n 维正态分布的充要条件是:

$$\forall l_i \in \mathbb{R} (i = 1, 2, \dots, n) : l_1 X_1 + l_2 X_2 + \dots + l_n X_n \sim N(\mu, \sigma^2).$$

- 若 Y_1, Y_2, \dots, Y_m 都是 n 维正态分布分量 $X_i (i = 1, 2, \dots, n)$ 的[线性函数](#), 则 (Y_1, Y_2, \dots, Y_m) 服从 m 维正态分布.
- n 维正态分布各分量[相互对立](#)充要条件是它们[两两不相关](#).

4 狄利克雷分布

4.1 基础概念

- $\mathbf{X} \sim \text{Dir}(\boldsymbol{\alpha})$.
- Dirichlet 分布又称为多元 Beta 分布, 属于[指数族分布](#).
- 多元 Beta 函数与 Gamma 函数的 **Dirichlet 公式**

$$\begin{aligned} B(\boldsymbol{\alpha}) &= B(\alpha_1, \alpha_2, \dots, \alpha_n) := \int \dots \int \prod_{i=1}^n x_i^{\alpha_i-1} d\mathbf{x} \quad \left(\sum_{i=1}^n x_i = 1 \right) \\ &= \frac{\Gamma(\alpha_1)\Gamma(\alpha_2)\dots\Gamma(\alpha_n)}{\Gamma(\alpha_1 + \alpha_2 + \dots + \alpha_n)} = \frac{\Gamma(\alpha_1)\dots\Gamma(\alpha_n)}{\Gamma(a_0)} \quad \left(a_0 = \sum_{i=1}^n \alpha_i \right) \end{aligned}$$

其中 $d (d \in \mathbb{N}^+)$ 维积分域是一个开放的 $d-1$ 维[正单纯形](#), 由顶点 $(1, 0, \dots, 0), (0, 1, \dots, 0), \dots, (0, 0, \dots, 1)$ 围成.

- 概率密度函数

$$\begin{aligned} \text{Dir}(\mathbf{X} | \boldsymbol{\alpha}) &= \frac{1}{B(\boldsymbol{\alpha})} \prod_{i=1}^d X_i^{\alpha_i-1} = \frac{\Gamma(\alpha_0)}{\prod_{i=1}^d \Gamma(\alpha_i)} \prod_{i=1}^d X_i^{\alpha_i-1} \quad \left(\alpha_0 = \sum_{i=1}^d \alpha_i, d \geq 3 \right) \\ &= \frac{\Gamma(\alpha_0)}{\prod_{i=1}^d \Gamma(\alpha_i)} \left(\prod_{i=1}^{d-1} X_i^{\alpha_i-1} \right) (1 - X_1 - \dots - X_{d-1})^{\alpha_d-1} \quad (\|\mathbf{X}\| = 1) \end{aligned}$$

其中 α 是无量纲的 **分布参数**, $d \geq 3$ 为随机变量的维度.

备注:

- 上式中的范数指 **1-范数** 而非 **2-范数**.
- Dirichlet 分布的 d 维 **支撑集** 同 Dirichlet 公式中的积分域.
- 概率分布记作 $\text{Dir}(\alpha)$, 密度函数记作 $\text{Dir}(\mathbf{X} | \alpha)$.
- 向量 \mathbf{X} 是 $n - 1$ 维, 而 α 是 n 维.
- 对称 Dirichlet 分布
 - 概率密度函数 $\text{Dir}(\mathbf{X} | \alpha) = \frac{\Gamma(d\alpha)}{\Gamma(\alpha)^d} \prod_{i=1}^d X_i^{\alpha_i-1}$.
 - 对称 Dirichlet 分布在每个概率密度相等, 即分布参数 α 在所有维度相同, 取值也被称为 **浓度参数**.
 - 当浓度参数为 1 时, d 维 Dirichlet 分布退化为 $d - 1$ 维正单纯形上的均匀分布, 也被称为 **平 Dirichlet 分布**.
 - 当浓度参数大于 1 时, 对称 Dirichlet 分布是一个 **集中分布**, 此时浓度参数越大, 概率密度越集中.
 - 当浓度参数小于 1 时, 对称 Dirichlet 分布是一个 **稀疏分布**, 此时浓度参数越接近于 0, 概率密度越稀疏.
- 累积分布函数

$$F(\mathbf{b}) = \int_{\mathbb{R}_d \cap [0, \mathbf{b})} \text{Dir}(\mathbf{X} | \alpha) d\mathbf{X}, \quad (\mathbf{b} \in (0, 1])$$

4.2 数字特征

- 众数
 - $M(X_i) = \frac{\alpha_i - 1}{\alpha_i + \alpha_d - 2} (x_d + x_i)$.
注: $x_d + x_i = 1 - x_1 - \cdots - x_{i-1} - x_{i+1} - \cdots - x_{d-1}$ 与 x_i 无关.
 - $M(X_i) = \frac{\alpha_i - 1}{\alpha_0 - d}$.
注: 这是所有分量都取到众数时的取值, 是上式的特例.
- 矩 $E\left(\prod_{i=1}^d X_i^{\beta_i}\right) = \frac{\Gamma(\alpha_0)}{\Gamma(\alpha_0 + \beta_0)} \prod_{i=1}^d \frac{\Gamma(\alpha_i + \beta_i)}{\Gamma(\alpha_i)} = \frac{B(\alpha + \beta)}{B(\alpha)}, \beta_0 = \sum_{i=1}^d \beta_i$.
 - 期望 $E(X_i) = \frac{\alpha_i}{\alpha_0}$.
 - 方差 $\text{Var}(X_i) = E(X_i^2) - E(X_i)^2 = \frac{\alpha_i(\alpha_0 - \alpha_i)}{\alpha_0^2(\alpha_0 + 1)}$.
 - 协方差 $\text{Cov}(X_i, X_j) = E(X_i X_j) - E(X_i)E(X_j) = \frac{\alpha_i(\alpha_0 - \alpha_j)}{\alpha_0^2(\alpha_0 + 1)}$.

4.3 其它性质

- 相关分布
 - 边缘分布 $p(X_i) = \text{Be}(X_i | \alpha_i, \alpha_0 - \alpha_i)$.
- 备注:
- 即 Beta 分布, 或 2 维 Dirichlet 分布.
 - '|' 符号类似分号, 与条件概率 **毫无关系**, 上式可写作 $F(x_i; \alpha_i, \alpha_0 - \alpha_i)$.

- 联合分布

$$p(X_i, X_j) = \text{Dir}(X_i, X_j \mid \boldsymbol{\alpha}), \quad \boldsymbol{\alpha} = [\alpha_i, \alpha_j, \alpha_0], \quad i, j \in \{1, 2, \dots, d\}.$$

即边缘分布 X_i 和 X_j 的联合分布为 3 维 Dirichlet 分布.

- 作为概率分布的性质

- 共轭性: 多项分布的共轭先验是 Dirichlet 分布 (同等无知原则).
- 聚合性: 不懂.
- 中立性

任意的 $(X_1, X_2, \dots, X_s) \in \mathbf{X}$ 都与归一化后的 $(X_{s+1}, \dots, X_d) \in \mathbf{X}$ 相互独立:

$$(X_1, \dots, X_s) \perp \mathbf{X}^*, \quad \mathbf{X}^* = \left(\frac{X_{s+1}}{X_{s+1} + \dots + X_d}, \dots, \frac{X_d}{X_{s+1} + \dots + X_d} \right),$$

$$p(\mathbf{X}^* \mid X_1, X_2, \dots, X_s) = \text{Dir}(\boldsymbol{\alpha}^*), \quad \boldsymbol{\alpha}^* = (\alpha_{s+1}, \alpha_{s+2}, \dots, \alpha_d).$$

- Dirichlet 是服从 Gamma 分布的 d 维 iid 随机变量 $\mathbf{T} = \Gamma(\mathbf{T} \mid \boldsymbol{\alpha}, 1)$ 归一化后的联合分布:

$$T_i = \Gamma(T_i \mid \alpha_i, 1), \quad Z_d = \sum_{i=1}^d T_i$$

$$\mathbf{X} = \frac{1}{Z_d} (T_1, T_2, \dots, T_{d-1}),$$

$$p(\mathbf{X}) = \text{Dir}(\alpha_1, \alpha_2, \dots, \alpha_d).$$

- 信息测度

A.3 常用分布统计表

A.3.1 标准正态分布表

| 标准正态分布表 | | | | | | | | | | |
|---------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| | 0.00 | 0.01 | 0.02 | 0.03 | 0.04 | 0.05 | 0.06 | 0.07 | 0.08 | 0.09 |
| 0.0 | 0.5000 | 0.5040 | 0.5080 | 0.5120 | 0.5160 | 0.5199 | 0.5239 | 0.5279 | 0.5319 | 0.5359 |
| 0.1 | 0.5398 | 0.5438 | 0.5478 | 0.5517 | 0.5557 | 0.5596 | 0.5636 | 0.5675 | 0.5714 | 0.5753 |
| 0.2 | 0.5793 | 0.5832 | 0.5871 | 0.5910 | 0.5948 | 0.5987 | 0.6026 | 0.6064 | 0.6103 | 0.6141 |
| 0.3 | 0.6179 | 0.6217 | 0.6255 | 0.6293 | 0.6331 | 0.6368 | 0.6406 | 0.6443 | 0.6480 | 0.6517 |
| 0.4 | 0.6554 | 0.6591 | 0.6628 | 0.6664 | 0.6700 | 0.6736 | 0.6772 | 0.6808 | 0.6844 | 0.6879 |
| 0.5 | 0.6915 | 0.6950 | 0.6985 | 0.7019 | 0.7054 | 0.7088 | 0.7123 | 0.7157 | 0.7190 | 0.7224 |
| 0.6 | 0.7257 | 0.7291 | 0.7324 | 0.7357 | 0.7389 | 0.7422 | 0.7454 | 0.7486 | 0.7517 | 0.7549 |
| 0.7 | 0.7580 | 0.7611 | 0.7642 | 0.7673 | 0.7704 | 0.7734 | 0.7764 | 0.7794 | 0.7823 | 0.7852 |
| 0.8 | 0.7881 | 0.7910 | 0.7939 | 0.7967 | 0.7995 | 0.8023 | 0.8051 | 0.8078 | 0.8106 | 0.8133 |
| 0.9 | 0.8159 | 0.8186 | 0.8212 | 0.8238 | 0.8264 | 0.8289 | 0.8315 | 0.8340 | 0.8365 | 0.8389 |
| 1.0 | 0.8413 | 0.8438 | 0.8461 | 0.8485 | 0.8508 | 0.8531 | 0.8554 | 0.8577 | 0.8599 | 0.8621 |
| 1.1 | 0.8643 | 0.8665 | 0.8686 | 0.8708 | 0.8729 | 0.8749 | 0.8770 | 0.8790 | 0.8810 | 0.8830 |
| 1.2 | 0.8849 | 0.8869 | 0.8888 | 0.8907 | 0.8925 | 0.8944 | 0.8962 | 0.8980 | 0.8997 | 0.9015 |
| 1.3 | 0.9032 | 0.9049 | 0.9066 | 0.9082 | 0.9099 | 0.9115 | 0.9131 | 0.9147 | 0.9162 | 0.9177 |
| 1.4 | 0.9192 | 0.9207 | 0.9222 | 0.9236 | 0.9251 | 0.9265 | 0.9279 | 0.9292 | 0.9306 | 0.9319 |
| 1.5 | 0.9332 | 0.9345 | 0.9357 | 0.9370 | 0.9382 | 0.9394 | 0.9406 | 0.9418 | 0.9429 | 0.9441 |
| 1.6 | 0.9452 | 0.9463 | 0.9474 | 0.9484 | 0.9495 | 0.9505 | 0.9515 | 0.9525 | 0.9535 | 0.9545 |
| 1.7 | 0.9554 | 0.9564 | 0.9573 | 0.9582 | 0.9591 | 0.9599 | 0.9608 | 0.9616 | 0.9625 | 0.9633 |
| 1.8 | 0.9641 | 0.9649 | 0.9656 | 0.9664 | 0.9671 | 0.9678 | 0.9686 | 0.9693 | 0.9699 | 0.9706 |
| 1.9 | 0.9713 | 0.9719 | 0.9726 | 0.9732 | 0.9738 | 0.9744 | 0.9750 | 0.9756 | 0.9761 | 0.9767 |
| 2.0 | 0.9772 | 0.9778 | 0.9783 | 0.9788 | 0.9793 | 0.9798 | 0.9803 | 0.9808 | 0.9812 | 0.9817 |
| 2.1 | 0.9821 | 0.9826 | 0.9830 | 0.9834 | 0.9838 | 0.9842 | 0.9846 | 0.9850 | 0.9854 | 0.9857 |
| 2.2 | 0.9861 | 0.9864 | 0.9868 | 0.9871 | 0.9875 | 0.9878 | 0.9881 | 0.9884 | 0.9887 | 0.9890 |
| 2.3 | 0.9893 | 0.9896 | 0.9898 | 0.9901 | 0.9904 | 0.9906 | 0.9909 | 0.9911 | 0.9913 | 0.9916 |
| 2.4 | 0.9918 | 0.9920 | 0.9922 | 0.9925 | 0.9927 | 0.9929 | 0.9931 | 0.9932 | 0.9934 | 0.9936 |
| 2.5 | 0.9938 | 0.9940 | 0.9941 | 0.9943 | 0.9945 | 0.9946 | 0.9948 | 0.9949 | 0.9951 | 0.9952 |
| 2.6 | 0.9953 | 0.9955 | 0.9956 | 0.9957 | 0.9959 | 0.9960 | 0.9961 | 0.9962 | 0.9963 | 0.9964 |
| 2.7 | 0.9965 | 0.9966 | 0.9967 | 0.9968 | 0.9969 | 0.9970 | 0.9971 | 0.9972 | 0.9973 | 0.9974 |
| 2.8 | 0.9974 | 0.9975 | 0.9976 | 0.9977 | 0.9977 | 0.9978 | 0.9979 | 0.9979 | 0.9980 | 0.9981 |
| 2.9 | 0.9981 | 0.9982 | 0.9982 | 0.9983 | 0.9984 | 0.9984 | 0.9985 | 0.9985 | 0.9986 | 0.9986 |
| 3.0 | 0.9987 | 0.9987 | 0.9987 | 0.9988 | 0.9988 | 0.9989 | 0.9989 | 0.9989 | 0.9990 | 0.9990 |
| 3.1 | 0.9990 | 0.9991 | 0.9991 | 0.9991 | 0.9992 | 0.9992 | 0.9992 | 0.9992 | 0.9993 | 0.9993 |

A.3.2 卡方分布表

| 卡方分布表（首行为概率值，首列为自由度） | | | | | | | | | | | |
|----------------------|----------|----------|----------|----------|----------|-----------|-----------|-----------|-----------|-----------|-----------|
| | 0.995 | 0.990 | 0.975 | 0.950 | 0.900 | 0.100 | 0.050 | 0.025 | 0.010 | 0.005 | 0.001 |
| 1 | 0.0000 | 0.0002 | 0.0010 | 0.0039 | 0.0158 | 2.7055 | 3.8415 | 5.0239 | 6.6349 | 7.8794 | 10.8276 |
| 2 | 0.0100 | 0.0201 | 0.0506 | 0.1026 | 0.2107 | 4.6052 | 5.9915 | 7.3778 | 9.2103 | 10.5966 | 13.8155 |
| 3 | 0.0717 | 0.1148 | 0.2158 | 0.3518 | 0.5844 | 6.2514 | 7.8147 | 9.3484 | 11.3449 | 12.8382 | 16.2662 |
| 4 | 0.2070 | 0.2971 | 0.4844 | 0.7107 | 1.0636 | 7.7794 | 9.4877 | 11.1433 | 13.2767 | 14.8603 | 18.4668 |
| 5 | 0.4117 | 0.5543 | 0.8312 | 1.1455 | 1.6103 | 9.2364 | 11.0705 | 12.8325 | 15.0863 | 16.7496 | 20.5150 |
| 6 | 0.6757 | 0.8721 | 1.2373 | 1.6354 | 2.2041 | 10.6446 | 12.5916 | 14.4494 | 16.8119 | 18.5476 | 22.4577 |
| 7 | 0.9893 | 1.2390 | 1.6899 | 2.1673 | 2.8331 | 12.0170 | 14.0671 | 16.0128 | 18.4753 | 20.2777 | 24.3219 |
| 8 | 1.3444 | 1.6465 | 2.1797 | 2.7326 | 3.4895 | 13.3616 | 15.5073 | 17.5345 | 20.0902 | 21.9550 | 26.1245 |
| 9 | 1.7349 | 2.0879 | 2.7004 | 3.3251 | 4.1682 | 14.6837 | 16.9190 | 19.0228 | 21.6660 | 23.5894 | 27.8772 |
| 10 | 2.1559 | 2.5582 | 3.2470 | 3.9403 | 4.8652 | 15.9872 | 18.3070 | 20.4832 | 23.2093 | 25.1882 | 29.5883 |
| 11 | 2.6032 | 3.0535 | 3.8157 | 4.5748 | 5.5778 | 17.2750 | 19.6751 | 21.9200 | 24.7250 | 26.7568 | 31.2641 |
| 12 | 3.0738 | 3.5706 | 4.4038 | 5.2260 | 6.3038 | 18.5493 | 21.0261 | 23.3367 | 26.2170 | 28.2995 | 32.9095 |
| 13 | 3.5650 | 4.1069 | 5.0088 | 5.8919 | 7.0415 | 19.8119 | 22.3620 | 24.7356 | 27.6882 | 29.8195 | 34.5282 |
| 14 | 4.0747 | 4.6604 | 5.6287 | 6.5706 | 7.7895 | 21.0641 | 23.6848 | 26.1189 | 29.1412 | 31.3193 | 36.1233 |
| 15 | 4.6009 | 5.2293 | 6.2621 | 7.2609 | 8.5468 | 22.3071 | 24.9958 | 27.4884 | 30.5779 | 32.8013 | 37.6973 |
| 16 | 5.1422 | 5.8122 | 6.9077 | 7.9616 | 9.3122 | 23.5418 | 26.2962 | 28.8454 | 31.9999 | 34.2672 | 39.2524 |
| 17 | 5.6972 | 6.4078 | 7.5642 | 8.6718 | 10.0852 | 24.7690 | 27.5871 | 30.1910 | 33.4087 | 35.7185 | 40.7902 |
| 18 | 6.2648 | 7.0149 | 8.2307 | 9.3905 | 10.8649 | 25.9894 | 28.8693 | 31.5264 | 34.8053 | 37.1565 | 42.3124 |
| 19 | 6.8440 | 7.6327 | 8.9065 | 10.1170 | 11.6509 | 27.2036 | 30.1435 | 32.8523 | 36.1909 | 38.5823 | 43.8202 |
| 20 | 7.4338 | 8.2604 | 9.5908 | 10.8508 | 12.4426 | 28.4120 | 31.4104 | 34.1696 | 37.5662 | 39.9968 | 45.3147 |
| 21 | 8.0337 | 8.8972 | 10.2829 | 11.5913 | 13.2396 | 29.6151 | 32.6706 | 35.4789 | 38.9322 | 41.4011 | 46.7970 |
| 22 | 8.6427 | 9.5425 | 10.9823 | 12.3380 | 14.0415 | 30.8133 | 33.9244 | 36.7807 | 40.2894 | 42.7957 | 48.2679 |
| 23 | 9.2604 | 10.1957 | 11.6886 | 13.0905 | 14.8480 | 32.0069 | 35.1725 | 38.0756 | 41.6384 | 44.1813 | 49.7282 |
| 24 | 9.8862 | 10.8564 | 12.4012 | 13.8484 | 15.6587 | 33.1962 | 36.4150 | 39.3641 | 42.9798 | 45.5585 | 51.1786 |
| 25 | 10.5197 | 11.5240 | 13.1197 | 14.6114 | 16.4734 | 34.3816 | 37.6525 | 40.6465 | 44.3141 | 46.9279 | 52.6197 |
| 26 | 11.1602 | 12.1981 | 13.8439 | 15.3792 | 17.2919 | 35.5632 | 38.8851 | 41.9232 | 45.6417 | 48.2899 | 54.0520 |
| 27 | 11.8076 | 12.8785 | 14.5734 | 16.1514 | 18.1139 | 36.7412 | 40.1133 | 43.1945 | 46.9629 | 49.6449 | 55.4760 |
| 28 | 12.4613 | 13.5647 | 15.3079 | 16.9279 | 18.9392 | 37.9159 | 41.3371 | 44.4608 | 48.2782 | 50.9934 | 56.8923 |
| 29 | 13.1211 | 14.2565 | 16.0471 | 17.7084 | 19.7677 | 39.0875 | 42.5570 | 45.7223 | 49.5879 | 52.3356 | 58.3012 |
| 30 | 13.7867 | 14.9535 | 16.7908 | 18.4927 | 20.5992 | 40.2560 | 43.7730 | 46.9792 | 50.8922 | 53.6720 | 59.7031 |
| 31 | 14.4578 | 15.6555 | 17.5387 | 19.2806 | 21.4336 | 41.4217 | 44.9853 | 48.2319 | 52.1914 | 55.0027 | 61.0983 |
| 32 | 15.1340 | 16.3622 | 18.2908 | 20.0719 | 22.2706 | 42.5847 | 46.1943 | 49.4804 | 53.4858 | 56.3281 | 62.4872 |
| 33 | 15.8153 | 17.0735 | 19.0467 | 20.8665 | 23.1102 | 43.7452 | 47.3999 | 50.7251 | 54.7755 | 57.6484 | 63.8701 |
| 34 | 16.5013 | 17.7891 | 19.8063 | 21.6643 | 23.9523 | 44.9032 | 48.6024 | 51.9660 | 56.0609 | 58.9639 | 65.2472 |
| 35 | 17.1918 | 18.5089 | 20.5694 | 22.4650 | 24.7967 | 46.0588 | 49.8018 | 53.2033 | 57.3421 | 60.2748 | 66.6188 |
| 36 | 17.8867 | 19.2327 | 21.3359 | 23.2686 | 25.6433 | 47.2122 | 50.9985 | 54.4373 | 58.6192 | 61.5812 | 67.9852 |
| 37 | 18.5858 | 19.9602 | 22.1056 | 24.0749 | 26.4921 | 48.3634 | 52.1923 | 55.6680 | 59.8925 | 62.8833 | 69.3465 |
| 38 | 19.2889 | 20.6914 | 22.8785 | 24.8839 | 27.3430 | 49.5126 | 53.3835 | 56.8955 | 61.1621 | 64.1814 | 70.7029 |
| 39 | 19.9959 | 21.4262 | 23.6543 | 25.6954 | 28.1958 | 50.6598 | 54.5722 | 58.1201 | 62.4281 | 65.4756 | 72.0547 |
| 40 | 20.7065 | 22.1643 | 24.4330 | 26.5093 | 29.0505 | 51.8051 | 55.7585 | 59.3417 | 63.6907 | 66.7660 | 73.4020 |
| 41 | 21.4208 | 22.9056 | 25.2145 | 27.3256 | 29.9071 | 52.9485 | 56.9424 | 60.5606 | 64.9501 | 68.0527 | 74.7449 |
| 42 | 22.1385 | 23.6501 | 25.9987 | 28.1440 | 30.7654 | 54.0902 | 58.1240 | 61.7768 | 66.2062 | 69.3360 | 76.0838 |
| 43 | 22.8595 | 24.3976 | 26.7854 | 28.9647 | 31.6255 | 55.2302 | 59.3035 | 62.9904 | 67.4593 | 70.6159 | 77.4186 |
| 44 | 23.5837 | 25.1480 | 27.5746 | 29.7875 | 32.4871 | 56.3685 | 60.4809 | 64.2015 | 68.7095 | 71.8926 | 78.7495 |
| 45 | 24.3110 | 25.9013 | 28.3662 | 30.6123 | 33.3504 | 57.5053 | 61.6562 | 65.4102 | 69.9568 | 73.1661 | 80.0767 |
| 46 | 25.0413 | 26.6572 | 29.1601 | 31.4390 | 34.2152 | 58.6405 | 62.8296 | 66.6165 | 71.2014 | 74.4365 | 81.4003 |
| 47 | 25.7746 | 27.4158 | 29.9562 | 32.2676 | 35.0814 | 59.7743 | 64.0011 | 67.8206 | 72.4433 | 75.7041 | 82.7204 |
| 48 | 26.5106 | 28.1770 | 30.7545 | 33.0981 | 35.9491 | 60.9066 | 65.1708 | 69.0226 | 73.6826 | 76.9688 | 84.0371 |
| 49 | 27.2493 | 28.9406 | 31.5549 | 33.9303 | 36.8182 | 62.0375 | 66.3386 | 70.2224 | 74.9195 | 78.2307 | 85.3506 |
| 50 | 27.9907 | 29.7067 | 32.3574 | 34.7643 | 37.6886 | 63.1671 | 67.5048 | 71.4202 | 76.1539 | 79.4900 | 86.6608 |
| 75 | 47.2060 | 49.4750 | 52.9419 | 56.0541 | 59.7946 | 91.0615 | 96.2167 | 100.8393 | 106.3929 | 110.2856 | 118.5991 |
| 100 | 67.3276 | 70.0649 | 74.2219 | 77.9295 | 82.3581 | 118.4980 | 124.3421 | 129.5612 | 135.8067 | 140.1695 | 149.4493 |
| 150 | 109.1422 | 112.6676 | 117.9845 | 122.6918 | 128.2751 | 172.5812 | 179.5806 | 185.8004 | 193.2077 | 198.3602 | 209.2646 |
| 200 | 152.2410 | 156.4320 | 162.7280 | 168.2786 | 174.8353 | 226.0210 | 233.9943 | 241.0579 | 249.4451 | 255.2642 | 267.5405 |
| 500 | 422.3034 | 429.3875 | 439.9360 | 449.1468 | 459.9261 | 540.9303 | 553.1268 | 563.8515 | 576.4928 | 585.2066 | 603.4460 |
| 750 | 653.9968 | 662.8521 | 676.0026 | 687.4522 | 700.8136 | 800.0428 | 814.8215 | 827.7853 | 843.0290 | 853.5143 | 875.4044 |
| 1000 | 888.5635 | 898.9124 | 914.2572 | 927.5944 | 943.1326 | 1057.7239 | 1074.6794 | 1089.5309 | 1106.9690 | 1118.9481 | 1143.9171 |

A.3.3 t 分布表

| t 分布表 | | | | | | | |
|-------|--------|--------|--------|---------|---------|---------|----------|
| | 0.25 | 0.1 | 0.05 | 0.025 | 0.01 | 0.005 | 0.0025 |
| 1 | 1.0000 | 3.0777 | 6.3138 | 12.7062 | 31.8205 | 63.6567 | 127.3213 |
| 2 | 0.8165 | 1.8856 | 2.9200 | 4.3027 | 6.9646 | 9.9248 | 14.0890 |
| 3 | 0.7649 | 1.6377 | 2.3534 | 3.1824 | 4.5407 | 5.8409 | 7.4533 |
| 4 | 0.7407 | 1.5332 | 2.1318 | 2.7764 | 3.7469 | 4.6041 | 5.5976 |
| 5 | 0.7267 | 1.4759 | 2.0150 | 2.5706 | 3.3649 | 4.0321 | 4.7733 |
| 6 | 0.7176 | 1.4398 | 1.9432 | 2.4469 | 3.1427 | 3.7074 | 4.3168 |
| 7 | 0.7111 | 1.4149 | 1.8946 | 2.3646 | 2.9980 | 3.4995 | 4.0293 |
| 8 | 0.7064 | 1.3968 | 1.8595 | 2.3060 | 2.8965 | 3.3554 | 3.8325 |
| 9 | 0.7027 | 1.3830 | 1.8331 | 2.2622 | 2.8214 | 3.2498 | 3.6897 |
| 10 | 0.6998 | 1.3722 | 1.8125 | 2.2281 | 2.7638 | 3.1693 | 3.5814 |
| 11 | 0.6974 | 1.3634 | 1.7959 | 2.2010 | 2.7181 | 3.1058 | 3.4966 |
| 12 | 0.6955 | 1.3562 | 1.7823 | 2.1788 | 2.6810 | 3.0545 | 3.4284 |
| 13 | 0.6938 | 1.3502 | 1.7709 | 2.1604 | 2.6503 | 3.0123 | 3.3725 |
| 14 | 0.6924 | 1.3450 | 1.7613 | 2.1448 | 2.6245 | 2.9768 | 3.3257 |
| 15 | 0.6912 | 1.3406 | 1.7531 | 2.1314 | 2.6025 | 2.9467 | 3.2860 |
| 16 | 0.6901 | 1.3368 | 1.7459 | 2.1199 | 2.5835 | 2.9208 | 3.2520 |
| 17 | 0.6892 | 1.3334 | 1.7396 | 2.1098 | 2.5669 | 2.8982 | 3.2224 |
| 18 | 0.6884 | 1.3304 | 1.7341 | 2.1009 | 2.5524 | 2.8784 | 3.1966 |
| 19 | 0.6876 | 1.3277 | 1.7291 | 2.0930 | 2.5395 | 2.8609 | 3.1737 |
| 20 | 0.6870 | 1.3253 | 1.7247 | 2.0860 | 2.5280 | 2.8453 | 3.1534 |
| 21 | 0.6864 | 1.3232 | 1.7207 | 2.0796 | 2.5176 | 2.8314 | 3.1352 |
| 22 | 0.6858 | 1.3212 | 1.7171 | 2.0739 | 2.5083 | 2.8188 | 3.1188 |
| 23 | 0.6853 | 1.3195 | 1.7139 | 2.0687 | 2.4999 | 2.8073 | 3.1040 |
| 24 | 0.6848 | 1.3178 | 1.7109 | 2.0639 | 2.4922 | 2.7969 | 3.0905 |
| 25 | 0.6844 | 1.3163 | 1.7081 | 2.0595 | 2.4851 | 2.7874 | 3.0782 |
| 26 | 0.6840 | 1.3150 | 1.7056 | 2.0555 | 2.4786 | 2.7787 | 3.0669 |
| 27 | 0.6837 | 1.3137 | 1.7033 | 2.0518 | 2.4727 | 2.7707 | 3.0565 |
| 28 | 0.6834 | 1.3125 | 1.7011 | 2.0484 | 2.4671 | 2.7633 | 3.0469 |
| 29 | 0.6830 | 1.3114 | 1.6991 | 2.0452 | 2.4620 | 2.7564 | 3.0380 |
| 30 | 0.6828 | 1.3104 | 1.6973 | 2.0423 | 2.4573 | 2.7500 | 3.0298 |
| 31 | 0.6825 | 1.3095 | 1.6955 | 2.0395 | 2.4528 | 2.7440 | 3.0221 |
| 32 | 0.6822 | 1.3086 | 1.6939 | 2.0369 | 2.4487 | 2.7385 | 3.0149 |
| 33 | 0.6820 | 1.3077 | 1.6924 | 2.0345 | 2.4448 | 2.7333 | 3.0082 |
| 34 | 0.6818 | 1.3070 | 1.6909 | 2.0322 | 2.4411 | 2.7284 | 3.0020 |
| 35 | 0.6816 | 1.3062 | 1.6896 | 2.0301 | 2.4377 | 2.7238 | 2.9960 |
| 36 | 0.6814 | 1.3055 | 1.6883 | 2.0281 | 2.4345 | 2.7195 | 2.9905 |
| 37 | 0.6812 | 1.3049 | 1.6871 | 2.0262 | 2.4314 | 2.7154 | 2.9852 |
| 38 | 0.6810 | 1.3042 | 1.6860 | 2.0244 | 2.4286 | 2.7116 | 2.9803 |
| 39 | 0.6808 | 1.3036 | 1.6849 | 2.0227 | 2.4258 | 2.7079 | 2.9756 |
| 40 | 0.6807 | 1.3031 | 1.6839 | 2.0211 | 2.4233 | 2.7045 | 2.9712 |
| 41 | 0.6805 | 1.3025 | 1.6829 | 2.0195 | 2.4208 | 2.7012 | 2.9670 |
| 42 | 0.6804 | 1.3020 | 1.6820 | 2.0181 | 2.4185 | 2.6981 | 2.9630 |
| 43 | 0.6802 | 1.3016 | 1.6811 | 2.0167 | 2.4163 | 2.6951 | 2.9592 |
| 44 | 0.6801 | 1.3011 | 1.6802 | 2.0154 | 2.4141 | 2.6923 | 2.9555 |
| 45 | 0.6800 | 1.3006 | 1.6794 | 2.0141 | 2.4121 | 2.6896 | 2.9521 |
| 46 | 0.6799 | 1.3002 | 1.6787 | 2.0129 | 2.4102 | 2.6870 | 2.9488 |
| 47 | 0.6797 | 1.2998 | 1.6779 | 2.0117 | 2.4083 | 2.6846 | 2.9456 |
| 48 | 0.6796 | 1.2994 | 1.6772 | 2.0106 | 2.4066 | 2.6822 | 2.9426 |
| 49 | 0.6795 | 1.2991 | 1.6766 | 2.0096 | 2.4049 | 2.6800 | 2.9397 |
| 50 | 0.6794 | 1.2987 | 1.6759 | 2.0086 | 2.4033 | 2.6778 | 2.9370 |

A.3.4 F 分布表

1 上 0.1 分位数

| F 分布表 (上 0.1 分位数) | | | | | | | | | | | | | | | | | | | | |
|-------------------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|--|--|
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 12 | 15 | 20 | 24 | 30 | 40 | 60 | 120 | | |
| 1 | 39.8635 | 49.5000 | 53.5932 | 55.8330 | 57.2401 | 58.2044 | 58.9060 | 59.4390 | 59.8576 | 60.1950 | 60.7052 | 61.2203 | 61.7403 | 62.0020 | 62.2650 | 62.5291 | 62.7943 | 63.0606 | | |
| 2 | 8.5263 | 9.0000 | 9.1618 | 9.2434 | 9.2926 | 9.3255 | 9.3491 | 9.3668 | 9.3805 | 9.3916 | 9.4081 | 9.4247 | 9.4413 | 9.4496 | 9.4579 | 9.4662 | 9.4746 | 9.4829 | | |
| 3 | 5.5383 | 5.4624 | 5.3908 | 5.3426 | 5.3092 | 5.2847 | 5.2662 | 5.2517 | 5.2400 | 5.2304 | 5.2156 | 5.2003 | 5.1845 | 5.1764 | 5.1681 | 5.1597 | 5.1512 | 5.1425 | | |
| 4 | 4.5448 | 4.3246 | 4.1909 | 4.1072 | 4.0506 | 4.0097 | 3.9790 | 3.9549 | 3.9357 | 3.9199 | 3.8955 | 3.8704 | 3.8443 | 3.8310 | 3.8174 | 3.8036 | 3.7896 | 3.7753 | | |
| 5 | 4.0604 | 3.7797 | 3.6195 | 3.5202 | 3.4530 | 3.4045 | 3.3679 | 3.3393 | 3.3163 | 3.2974 | 3.2682 | 3.2380 | 3.2067 | 3.1905 | 3.1741 | 3.1573 | 3.1402 | 3.1228 | | |
| 6 | 3.7759 | 3.4633 | 3.2888 | 3.1808 | 3.1075 | 3.0546 | 3.0145 | 2.9830 | 2.9577 | 2.9369 | 2.9047 | 2.8712 | 2.8363 | 2.8183 | 2.8000 | 2.7812 | 2.7620 | 2.7423 | | |
| 7 | 3.5894 | 3.2574 | 3.0741 | 2.9605 | 2.8833 | 2.8274 | 2.7849 | 2.7516 | 2.7247 | 2.7025 | 2.6681 | 2.6322 | 2.5947 | 2.5753 | 2.5555 | 2.5351 | 2.5142 | 2.4928 | | |
| 8 | 3.4579 | 3.1131 | 2.9238 | 2.8064 | 2.7264 | 2.6683 | 2.6241 | 2.5893 | 2.5612 | 2.5380 | 2.5020 | 2.4642 | 2.4246 | 2.4041 | 2.3830 | 2.3614 | 2.3391 | 2.3162 | | |
| 9 | 3.3603 | 3.0065 | 2.8129 | 2.6927 | 2.6106 | 2.5509 | 2.5053 | 2.4694 | 2.4403 | 2.4163 | 2.3789 | 2.3396 | 2.2983 | 2.2768 | 2.2547 | 2.2320 | 2.2085 | 2.1843 | | |
| 10 | 3.2850 | 2.9245 | 2.7277 | 2.6053 | 2.5216 | 2.4606 | 2.4140 | 2.3772 | 2.3473 | 2.3226 | 2.2841 | 2.2435 | 2.2007 | 2.1784 | 2.1554 | 2.1317 | 2.1072 | 2.0818 | | |
| 11 | 3.2252 | 2.8595 | 2.6602 | 2.5362 | 2.4512 | 2.3891 | 2.3416 | 2.3040 | 2.2735 | 2.2482 | 2.2087 | 2.1671 | 2.1230 | 2.1000 | 2.0762 | 2.0516 | 2.0261 | 1.9997 | | |
| 12 | 3.1765 | 2.8068 | 2.6055 | 2.4801 | 2.3940 | 2.3310 | 2.2828 | 2.2446 | 2.2135 | 2.1878 | 2.1474 | 2.1049 | 2.0597 | 2.0360 | 2.0115 | 1.9861 | 1.9597 | 1.9323 | | |
| 13 | 3.1362 | 2.7632 | 2.5603 | 2.4337 | 2.3467 | 2.2830 | 2.2341 | 2.1953 | 2.1638 | 2.1376 | 2.0966 | 2.0532 | 2.0070 | 1.9827 | 1.9576 | 1.9315 | 1.9043 | 1.8759 | | |
| 14 | 3.1022 | 2.7265 | 2.5222 | 2.3947 | 2.3069 | 2.2426 | 2.1931 | 2.1539 | 2.1220 | 2.0954 | 2.0537 | 2.0095 | 1.9625 | 1.9377 | 1.9119 | 1.8852 | 1.8572 | 1.8280 | | |
| 15 | 3.0732 | 2.6952 | 2.4898 | 2.3614 | 2.2730 | 2.2081 | 2.1582 | 2.1185 | 2.0862 | 2.0593 | 2.0171 | 1.9722 | 1.9243 | 1.8990 | 1.8728 | 1.8454 | 1.8168 | 1.7867 | | |
| 16 | 3.0481 | 2.6682 | 2.4618 | 2.3327 | 2.2438 | 2.1783 | 2.1280 | 2.0880 | 2.0553 | 2.0281 | 1.9854 | 1.9399 | 1.8913 | 1.8656 | 1.8388 | 1.8108 | 1.7816 | 1.7507 | | |
| 17 | 3.0262 | 2.6446 | 2.4374 | 2.3077 | 2.2183 | 2.1524 | 2.1017 | 2.0613 | 2.0284 | 2.0009 | 1.9577 | 1.9117 | 1.8624 | 1.8362 | 1.8090 | 1.7795 | 1.7487 | 1.7169 | | |
| 18 | 3.0070 | 2.6239 | 2.4160 | 2.2858 | 2.1958 | 2.1296 | 2.0785 | 2.0379 | 2.0047 | 1.9770 | 1.9333 | 1.8868 | 1.8368 | 1.8103 | 1.7827 | 1.7537 | 1.7232 | 1.6910 | | |
| 19 | 2.9899 | 2.6056 | 2.3970 | 2.2663 | 2.1760 | 2.1094 | 2.0580 | 2.0171 | 1.9836 | 1.9557 | 1.9117 | 1.8647 | 1.8142 | 1.7873 | 1.7582 | 1.7279 | 1.6968 | 1.6659 | | |
| 20 | 2.9747 | 2.5893 | 2.3801 | 2.2489 | 2.1582 | 2.0913 | 2.0397 | 1.9985 | 1.9649 | 1.9367 | 1.8924 | 1.8449 | 1.7938 | 1.7667 | 1.7373 | 1.7068 | 1.6758 | 1.6453 | | |
| 21 | 2.9610 | 2.5746 | 2.3649 | 2.2333 | 2.1423 | 2.0751 | 2.0233 | 1.9819 | 1.9480 | 1.9197 | 1.8750 | 1.8271 | 1.7756 | 1.7481 | 1.7193 | 1.6890 | 1.6589 | 1.6288 | | |
| 22 | 2.9486 | 2.5613 | 2.3512 | 2.2193 | 2.1279 | 2.0605 | 2.0084 | 1.9668 | 1.9327 | 1.9043 | 1.8593 | 1.8111 | 1.7590 | 1.7312 | 1.7021 | 1.6714 | 1.6419 | 1.6121 | | |
| 23 | 2.9374 | 2.5493 | 2.3387 | 2.2065 | 2.1149 | 2.0472 | 1.9949 | 1.9531 | 1.9189 | 1.8903 | 1.8450 | 1.7964 | 1.7439 | 1.7159 | 1.6864 | 1.6554 | 1.6254 | 1.5951 | | |
| 24 | 2.9271 | 2.5383 | 2.3274 | 2.1949 | 2.1030 | 2.0351 | 1.9826 | 1.9407 | 1.9063 | 1.8775 | 1.8319 | 1.7831 | 1.7302 | 1.7019 | 1.6721 | 1.6421 | 1.6121 | 1.5815 | | |
| 25 | 2.9177 | 2.5283 | 2.3170 | 2.1842 | 2.0922 | 2.0241 | 1.9714 | 1.9292 | 1.8947 | 1.8658 | 1.8200 | 1.7708 | 1.7175 | 1.6890 | 1.6589 | 1.6282 | 1.5984 | 1.5677 | | |
| 26 | 2.9091 | 2.5191 | 2.3075 | 2.1745 | 2.0822 | 2.0139 | 1.9610 | 1.9188 | 1.8841 | 1.8550 | 1.8090 | 1.7596 | 1.7059 | 1.6771 | 1.6468 | 1.6167 | 1.5868 | 1.5563 | | |
| 27 | 2.9012 | 2.5106 | 2.2987 | 2.1655 | 2.0730 | 2.0045 | 1.9515 | 1.9091 | 1.8743 | 1.8451 | 1.7989 | 1.7492 | 1.6951 | 1.6662 | 1.6356 | 1.6052 | 1.5753 | 1.5453 | | |
| 28 | 2.8938 | 2.5028 | 2.2906 | 2.1571 | 2.0645 | 1.9959 | 1.9427 | 1.9001 | 1.8652 | 1.8359 | 1.7895 | 1.7395 | 1.6852 | 1.6560 | 1.6252 | 1.5952 | 1.5655 | 1.5355 | | |
| 29 | 2.8870 | 2.4955 | 2.2831 | 2.1494 | 2.0566 | 1.9878 | 1.9345 | 1.8918 | 1.8568 | 1.8274 | 1.7808 | 1.7306 | 1.6759 | 1.6465 | 1.6155 | 1.5852 | 1.5557 | 1.5259 | | |
| 30 | 2.8807 | 2.4887 | 2.2761 | 2.1422 | 2.0492 | 1.9803 | 1.9269 | 1.8841 | 1.8490 | 1.8195 | 1.7727 | 1.7223 | 1.6673 | 1.6377 | 1.6065 | 1.5762 | 1.5467 | 1.5171 | | |
| 40 | 2.8354 | 2.4404 | 2.2261 | 2.0909 | 1.9968 | 1.9269 | 1.8725 | 1.8289 | 1.7929 | 1.7627 | 1.7146 | 1.6624 | 1.6052 | 1.5741 | 1.5411 | 1.5056 | 1.4672 | 1.4284 | | |
| 60 | 2.7911 | 2.3933 | 2.1774 | 2.0410 | 1.9457 | 1.8747 | 1.8194 | 1.7748 | 1.7380 | 1.7070 | 1.6574 | 1.6034 | 1.5435 | 1.5107 | 1.4755 | 1.4373 | 1.3952 | 1.3476 | | |
| 120 | 2.7478 | 2.3473 | 2.1300 | 1.9923 | 1.8959 | 1.8238 | 1.7675 | 1.7220 | 1.6842 | 1.6524 | 1.6012 | 1.5450 | 1.4821 | 1.4472 | 1.4094 | 1.3676 | 1.3203 | 1.2646 | | |

2 上 0.05 分位数

| F 分布表 (上 0.05 分位数) | | | | | | | | | | | | | | | | | | | |
|--------------------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 12 | 15 | 20 | 24 | 30 | 40 | 60 | 120 | |
| 1 | 161.4476 | 199.5000 | 215.7073 | 224.5832 | 230.1619 | 233.9860 | 236.7684 | 238.8827 | 240.5433 | 241.8817 | 243.9060 | 245.9499 | 248.0131 | 249.0518 | 250.0951 | 251.1432 | 252.1957 | 253.2529 | 253.2529 |
| 2 | 18.5128 | 19.0000 | 19.1643 | 19.2468 | 19.2964 | 19.3295 | 19.3532 | 19.3710 | 19.3848 | 19.3959 | 19.4125 | 19.4291 | 19.4458 | 19.4541 | 19.4624 | 19.4707 | 19.4791 | 19.4874 | 19.4874 |
| 3 | 10.1280 | 9.5521 | 9.2766 | 9.1172 | 9.0135 | 8.9406 | 8.8867 | 8.8452 | 8.8123 | 8.7855 | 8.7446 | 8.7029 | 8.6602 | 8.6385 | 8.6166 | 8.5944 | 8.5720 | 8.5494 | 8.5494 |
| 4 | 7.7086 | 6.9443 | 6.5914 | 6.3882 | 6.2561 | 6.1631 | 6.0942 | 6.0410 | 5.9988 | 5.9644 | 5.9117 | 5.8578 | 5.8025 | 5.7744 | 5.7459 | 5.7170 | 5.6877 | 5.6581 | 5.6581 |
| 5 | 6.6079 | 5.7861 | 5.4095 | 5.1922 | 5.0503 | 4.9503 | 4.8759 | 4.8183 | 4.7725 | 4.7351 | 4.6777 | 4.6188 | 4.5581 | 4.5272 | 4.4957 | 4.4638 | 4.4314 | 4.3985 | 4.3985 |
| 6 | 5.9874 | 5.1433 | 4.7571 | 4.5337 | 4.3874 | 4.2839 | 4.2067 | 4.1468 | 4.0990 | 4.0600 | 3.9999 | 3.9381 | 3.8742 | 3.8415 | 3.8082 | 3.7743 | 3.7398 | 3.7047 | 3.7047 |
| 7 | 5.5914 | 4.7374 | 4.3468 | 4.1203 | 3.9715 | 3.8660 | 3.7870 | 3.7257 | 3.6767 | 3.6365 | 3.5747 | 3.5107 | 3.4445 | 3.4105 | 3.3758 | 3.3404 | 3.3043 | 3.2674 | 3.2674 |
| 8 | 5.3177 | 4.4590 | 4.0662 | 3.8379 | 3.6875 | 3.5806 | 3.5005 | 3.4381 | 3.3881 | 3.3472 | 3.2839 | 3.2184 | 3.1503 | 3.1152 | 3.0794 | 3.0428 | 3.0053 | 2.9669 | 2.9669 |
| 9 | 5.1174 | 4.2565 | 3.8625 | 3.6331 | 3.4817 | 3.3738 | 3.2927 | 3.2296 | 3.1789 | 3.1373 | 3.0729 | 3.0061 | 2.9365 | 2.9005 | 2.8637 | 2.8259 | 2.7872 | 2.7475 | 2.7475 |
| 10 | 4.9646 | 4.1028 | 3.7083 | 3.4780 | 3.3258 | 3.2172 | 3.1355 | 3.0717 | 3.0204 | 2.9782 | 2.9130 | 2.8450 | 2.7740 | 2.7372 | 2.6996 | 2.6609 | 2.6211 | 2.5801 | 2.5801 |
| 11 | 4.8443 | 3.9823 | 3.5874 | 3.3567 | 3.2039 | 3.0946 | 3.0123 | 2.9480 | 2.8962 | 2.8536 | 2.7876 | 2.7186 | 2.6464 | 2.6090 | 2.5705 | 2.5309 | 2.4901 | 2.4480 | 2.4480 |
| 12 | 4.7472 | 3.8853 | 3.4903 | 3.2592 | 3.1059 | 2.9961 | 2.9134 | 2.8486 | 2.7964 | 2.7534 | 2.6866 | 2.6169 | 2.5436 | 2.5055 | 2.4663 | 2.4259 | 2.3842 | 2.3410 | 2.3410 |
| 13 | 4.6672 | 3.8056 | 3.4105 | 3.1791 | 3.0254 | 2.9153 | 2.8321 | 2.7669 | 2.7144 | 2.6710 | 2.6037 | 2.5331 | 2.4589 | 2.4202 | 2.3803 | 2.3392 | 2.2966 | 2.2524 | 2.2524 |
| 14 | 4.6001 | 3.7389 | 3.3439 | 3.1122 | 2.9582 | 2.8477 | 2.7642 | 2.6987 | 2.6458 | 2.6022 | 2.5342 | 2.4630 | 2.3879 | 2.3487 | 2.3082 | 2.2664 | 2.2229 | 2.1778 | 2.1778 |
| 15 | 4.5431 | 3.6823 | 3.2874 | 3.0556 | 2.9013 | 2.7905 | 2.7066 | 2.6408 | 2.5876 | 2.5437 | 2.4753 | 2.4034 | 2.3275 | 2.2878 | 2.2468 | 2.2043 | 2.1601 | 2.1141 | 2.1141 |
| 16 | 4.4940 | 3.6337 | 3.2389 | 3.0069 | 2.8524 | 2.7413 | 2.6572 | 2.5911 | 2.5377 | 2.4935 | 2.4247 | 2.3522 | 2.2756 | 2.2354 | 2.1938 | 2.1507 | 2.1058 | 2.0589 | 2.0589 |
| 17 | 4.4513 | 3.5915 | 3.1968 | 2.9647 | 2.8100 | 2.6987 | 2.6143 | 2.5480 | 2.4943 | 2.4499 | 2.3807 | 2.3077 | 2.2304 | 2.1898 | 2.1477 | 2.1040 | 2.0584 | 2.0107 | 2.0107 |
| 18 | 4.4139 | 3.5546 | 3.1599 | 2.9277 | 2.7729 | 2.6613 | 2.5767 | 2.5102 | 2.4563 | 2.4117 | 2.3421 | 2.2686 | 2.1906 | 2.1497 | 2.1071 | 2.0629 | 2.0166 | 1.9681 | 1.9681 |
| 19 | 4.3807 | 3.5219 | 3.1274 | 2.8951 | 2.7401 | 2.6283 | 2.5435 | 2.4768 | 2.4227 | 2.3779 | 2.3080 | 2.2341 | 2.1555 | 2.1141 | 2.0712 | 2.0264 | 1.9795 | 1.9302 | 1.9302 |
| 20 | 4.3512 | 3.4928 | 3.0984 | 2.8661 | 2.7109 | 2.5990 | 2.5140 | 2.4471 | 2.3928 | 2.3479 | 2.2776 | 2.2033 | 2.1242 | 2.0825 | 2.0391 | 1.9938 | 1.9464 | 1.8963 | 1.8963 |
| 21 | 4.3248 | 3.4668 | 3.0725 | 2.8401 | 2.6848 | 2.5727 | 2.4876 | 2.4205 | 2.3660 | 2.3210 | 2.2504 | 2.1757 | 2.0960 | 2.0540 | 2.0102 | 1.9645 | 1.9165 | 1.8657 | 1.8657 |
| 22 | 4.3009 | 3.4434 | 3.0491 | 2.8167 | 2.6613 | 2.5491 | 2.4638 | 2.3965 | 2.3419 | 2.2967 | 2.2258 | 2.1508 | 2.0707 | 2.0283 | 1.9842 | 1.9380 | 1.8894 | 1.8380 | 1.8380 |
| 23 | 4.2793 | 3.4221 | 3.0280 | 2.7955 | 2.6400 | 2.5277 | 2.4422 | 2.3748 | 2.3201 | 2.2747 | 2.2036 | 2.1282 | 2.0476 | 2.0050 | 1.9605 | 1.9139 | 1.8648 | 1.8128 | 1.8128 |
| 24 | 4.2597 | 3.4028 | 3.0088 | 2.7763 | 2.6207 | 2.5082 | 2.4226 | 2.3551 | 2.3002 | 2.2547 | 2.1834 | 2.1077 | 2.0267 | 1.9838 | 1.9390 | 1.8920 | 1.8424 | 1.7896 | 1.7896 |
| 25 | 4.2417 | 3.3852 | 2.9912 | 2.7587 | 2.6030 | 2.4904 | 2.4047 | 2.3371 | 2.2821 | 2.2365 | 2.1649 | 2.0889 | 2.0075 | 1.9643 | 1.9192 | 1.8718 | 1.8217 | 1.7684 | 1.7684 |
| 26 | 4.2252 | 3.3690 | 2.9752 | 2.7426 | 2.5868 | 2.4741 | 2.3883 | 2.3205 | 2.2655 | 2.2197 | 2.1479 | 2.0716 | 1.9898 | 1.9464 | 1.9010 | 1.8533 | 1.8027 | 1.7488 | 1.7488 |
| 27 | 4.2100 | 3.3541 | 2.9604 | 2.7278 | 2.5719 | 2.4591 | 2.3732 | 2.3053 | 2.2501 | 2.2043 | 2.1323 | 2.0558 | 1.9736 | 1.9299 | 1.8842 | 1.8361 | 1.7851 | 1.7306 | 1.7306 |
| 28 | 4.1960 | 3.3404 | 2.9467 | 2.7141 | 2.5581 | 2.4453 | 2.3593 | 2.2913 | 2.2360 | 2.1900 | 2.1179 | 2.0411 | 1.9586 | 1.9147 | 1.8687 | 1.8203 | 1.7689 | 1.7138 | 1.7138 |
| 29 | 4.1830 | 3.3277 | 2.9340 | 2.7014 | 2.5454 | 2.4326 | 2.3463 | 2.2783 | 2.2229 | 2.1768 | 2.1045 | 2.0275 | 1.9446 | 1.9005 | 1.8543 | 1.8055 | 1.7537 | 1.6981 | 1.6981 |
| 30 | 4.1709 | 3.3158 | 2.9223 | 2.6896 | 2.5336 | 2.4205 | 2.3343 | 2.2662 | 2.2107 | 2.1646 | 2.0921 | 2.0148 | 1.9317 | 1.8874 | 1.8409 | 1.7918 | 1.7396 | 1.6835 | 1.6835 |
| 40 | 4.0847 | 3.2317 | 2.8387 | 2.6060 | 2.4495 | 2.3359 | 2.2490 | 2.1802 | 2.1240 | 2.0772 | 2.0025 | 1.9245 | 1.8399 | 1.7952 | 1.7444 | 1.6928 | 1.6373 | 1.5766 | 1.5766 |
| 60 | 4.0012 | 3.1504 | 2.7581 | 2.5252 | 2.3683 | 2.2541 | 2.1665 | 2.0970 | 2.0401 | 1.9926 | 1.9174 | 1.8364 | 1.7480 | 1.7001 | 1.6491 | 1.5943 | 1.5343 | 1.4673 | 1.4673 |
| 120 | 3.9201 | 3.0718 | 2.6802 | 2.4472 | 2.2899 | 2.1750 | 2.0868 | 2.0164 | 1.9588 | 1.9105 | 1.8337 | 1.7505 | 1.6587 | 1.6084 | 1.5543 | 1.4952 | 1.4290 | 1.3551 | 1.3551 |

| | | F 分布表 (上 0.025 分位数) | | | | | | | | | | | | | | | | | |
|-----|----------|---------------------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|-----------|-----------|-----------|-----------|-----|
| | | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 12 | 15 | 20 | 24 | 30 | 40 | 60 | 120 |
| 1 | 647.7890 | 799.5000 | 864.1630 | 899.5833 | 921.8479 | 937.1111 | 948.2169 | 956.6562 | 963.2846 | 968.6274 | 976.7079 | 984.8668 | 993.1028 | 997.2492 | 1001.4144 | 1005.5981 | 1009.8001 | 1014.0202 | |
| 2 | 38.5063 | 39.0000 | 39.1655 | 39.2484 | 39.2982 | 39.3315 | 39.3552 | 39.3730 | 39.3869 | 39.3980 | 39.4146 | 39.4313 | 39.4479 | 39.4562 | 39.4646 | 39.4729 | 39.4812 | 39.4896 | |
| 3 | 17.4434 | 16.0441 | 15.4392 | 15.1010 | 14.8848 | 14.7347 | 14.6244 | 14.5399 | 14.4731 | 14.4189 | 14.3666 | 14.2527 | 14.1674 | 14.1241 | 14.0805 | 14.0365 | 13.9921 | 13.9473 | |
| 4 | 12.2179 | 10.6491 | 9.9792 | 9.6045 | 9.3645 | 9.1973 | 9.0741 | 8.9796 | 8.9047 | 8.8439 | 8.7512 | 8.6565 | 8.5599 | 8.5109 | 8.4613 | 8.4111 | 8.3604 | 8.3092 | |
| 5 | 10.0070 | 8.4336 | 7.7636 | 7.3879 | 7.1464 | 6.9777 | 6.8531 | 6.7572 | 6.6811 | 6.6192 | 6.5245 | 6.4277 | 6.3286 | 6.2780 | 6.2269 | 6.1750 | 6.1225 | 6.0693 | |
| 6 | 8.8131 | 7.2599 | 6.5988 | 6.2272 | 5.9876 | 5.8198 | 5.6955 | 5.5996 | 5.5234 | 5.4613 | 5.3662 | 5.2687 | 5.1684 | 5.1172 | 5.0652 | 5.0125 | 4.9589 | 4.9044 | |
| 7 | 8.0727 | 6.5415 | 5.8898 | 5.5226 | 5.2852 | 5.1186 | 4.9949 | 4.8993 | 4.8232 | 4.7611 | 4.6658 | 4.5678 | 4.4667 | 4.4150 | 4.3624 | 4.3089 | 4.2544 | 4.1989 | |
| 8 | 7.5709 | 6.0595 | 5.4160 | 5.0526 | 4.8173 | 4.6517 | 4.5286 | 4.4333 | 4.3572 | 4.2951 | 4.1997 | 4.1012 | 3.9995 | 3.9472 | 3.8940 | 3.8398 | 3.7844 | 3.7279 | |
| 9 | 7.2093 | 5.7147 | 5.0781 | 4.7181 | 4.4844 | 4.3197 | 4.1701 | 4.1020 | 4.0260 | 3.9639 | 3.8682 | 3.7694 | 3.6669 | 3.6142 | 3.5604 | 3.5055 | 3.4493 | 3.3918 | |
| 10 | 6.9367 | 5.4564 | 4.8256 | 4.4683 | 4.2361 | 4.0721 | 3.9498 | 3.8549 | 3.7790 | 3.7168 | 3.6209 | 3.5217 | 3.4185 | 3.3654 | 3.3110 | 3.2554 | 3.1984 | 3.1399 | |
| 11 | 6.7241 | 5.2559 | 4.6300 | 4.2751 | 4.0440 | 3.8807 | 3.7586 | 3.6638 | 3.5879 | 3.5257 | 3.4296 | 3.3299 | 3.2261 | 3.1725 | 3.1176 | 3.0613 | 3.0035 | 2.9441 | |
| 12 | 6.5538 | 5.0959 | 4.4742 | 4.1212 | 3.8911 | 3.7283 | 3.6065 | 3.5118 | 3.4358 | 3.3736 | 3.2773 | 3.1772 | 3.0728 | 3.0187 | 2.9633 | 2.9063 | 2.8478 | 2.7874 | |
| 13 | 6.4143 | 4.9653 | 4.3472 | 3.9959 | 3.7667 | 3.6043 | 3.4827 | 3.3880 | 3.3120 | 3.2497 | 3.1532 | 3.0527 | 2.9477 | 2.8932 | 2.8372 | 2.7797 | 2.7204 | 2.6590 | |
| 14 | 6.2979 | 4.8567 | 4.2417 | 3.8919 | 3.6634 | 3.5014 | 3.3799 | 3.2853 | 3.2093 | 3.1469 | 3.0502 | 2.9493 | 2.8437 | 2.7892 | 2.7324 | 2.6742 | 2.6142 | 2.5519 | |
| 15 | 6.1995 | 4.7650 | 4.1528 | 3.8043 | 3.5764 | 3.4147 | 3.2934 | 3.1987 | 3.1227 | 3.0602 | 2.9633 | 2.8621 | 2.7559 | 2.7006 | 2.6437 | 2.5850 | 2.5242 | 2.4611 | |
| 16 | 6.1151 | 4.6867 | 4.0768 | 3.7294 | 3.5021 | 3.3406 | 3.2194 | 3.1248 | 3.0488 | 2.9862 | 2.8890 | 2.7875 | 2.6808 | 2.6252 | 2.5678 | 2.5085 | 2.4471 | 2.3831 | |
| 17 | 6.0420 | 4.6189 | 4.0112 | 3.6648 | 3.4379 | 3.2767 | 3.1556 | 3.0610 | 2.9849 | 2.9222 | 2.8249 | 2.7230 | 2.6158 | 2.5598 | 2.5020 | 2.4422 | 2.3801 | 2.3153 | |
| 18 | 5.9781 | 4.5597 | 3.9539 | 3.6083 | 3.3820 | 3.2209 | 3.0999 | 3.0053 | 2.9291 | 2.8664 | 2.7689 | 2.6667 | 2.5590 | 2.5027 | 2.4445 | 2.3842 | 2.3214 | 2.2558 | |
| 19 | 5.9216 | 4.5075 | 3.9034 | 3.5587 | 3.3327 | 3.1718 | 3.0509 | 2.9563 | 2.8801 | 2.8172 | 2.7196 | 2.6171 | 2.5099 | 2.4532 | 2.3937 | 2.3329 | 2.2696 | 2.2032 | |
| 20 | 5.8715 | 4.4613 | 3.8587 | 3.5147 | 3.2891 | 3.1283 | 3.0074 | 2.9128 | 2.8365 | 2.7737 | 2.6758 | 2.5731 | 2.4645 | 2.4076 | 2.3486 | 2.2873 | 2.2234 | 2.1562 | |
| 21 | 5.8266 | 4.4199 | 3.8188 | 3.4754 | 3.2501 | 3.0895 | 2.9686 | 2.8740 | 2.7977 | 2.7348 | 2.6368 | 2.5338 | 2.4247 | 2.3675 | 2.3082 | 2.2465 | 2.1819 | 2.1141 | |
| 22 | 5.7863 | 4.3828 | 3.7829 | 3.4401 | 3.2151 | 3.0546 | 2.9338 | 2.8392 | 2.7628 | 2.6998 | 2.6017 | 2.4984 | 2.3890 | 2.3315 | 2.2718 | 2.2097 | 2.1446 | 2.0760 | |
| 23 | 5.7498 | 4.3492 | 3.7505 | 3.4083 | 3.1835 | 3.0232 | 2.9023 | 2.8077 | 2.7313 | 2.6682 | 2.5699 | 2.4665 | 2.3567 | 2.2989 | 2.2389 | 2.1763 | 2.1107 | 2.0415 | |
| 24 | 5.7166 | 4.3187 | 3.7211 | 3.3794 | 3.1548 | 2.9946 | 2.8734 | 2.7791 | 2.7027 | 2.6396 | 2.5411 | 2.4374 | 2.3273 | 2.2693 | 2.2090 | 2.1460 | 2.0799 | 2.0099 | |
| 25 | 5.6864 | 4.2909 | 3.6943 | 3.3530 | 3.1287 | 2.9685 | 2.8478 | 2.7531 | 2.6766 | 2.6135 | 2.5149 | 2.4110 | 2.3005 | 2.2422 | 2.1816 | 2.1183 | 2.0516 | 1.9811 | |
| 26 | 5.6586 | 4.2655 | 3.6697 | 3.3289 | 3.1048 | 2.9447 | 2.8240 | 2.7293 | 2.6528 | 2.5896 | 2.4908 | 2.3867 | 2.2759 | 2.2174 | 2.1565 | 2.0928 | 2.0257 | 1.9545 | |
| 27 | 5.6331 | 4.2421 | 3.6472 | 3.3067 | 3.0828 | 2.9228 | 2.8021 | 2.7074 | 2.6309 | 2.5676 | 2.4688 | 2.3644 | 2.2533 | 2.1946 | 2.1334 | 2.0693 | 2.0018 | 1.9299 | |
| 28 | 5.6096 | 4.2205 | 3.6264 | 3.2863 | 3.0626 | 2.9027 | 2.7820 | 2.6872 | 2.6106 | 2.5473 | 2.4484 | 2.3438 | 2.2324 | 2.1735 | 2.1121 | 2.0477 | 1.9797 | 1.9072 | |
| 29 | 5.5878 | 4.2006 | 3.6072 | 3.2674 | 3.0438 | 2.8840 | 2.7633 | 2.6686 | 2.5919 | 2.5286 | 2.4295 | 2.3248 | 2.2131 | 2.1540 | 2.0923 | 2.0276 | 1.9591 | 1.8861 | |
| 30 | 5.5675 | 4.1821 | 3.5894 | 3.2499 | 3.0265 | 2.8667 | 2.7460 | 2.6513 | 2.5746 | 2.5112 | 2.4120 | 2.3072 | 2.1952 | 2.1359 | 2.0739 | 2.0089 | 1.9400 | 1.8664 | |
| 40 | 5.4239 | 4.0510 | 3.4633 | 3.1261 | 2.9037 | 2.7444 | 2.6238 | 2.5289 | 2.4519 | 2.3882 | 2.2882 | 2.1819 | 2.0677 | 2.0069 | 1.9429 | 1.8752 | 1.8028 | 1.7242 | |
| 60 | 5.2856 | 3.9253 | 3.3425 | 3.0077 | 2.7863 | 2.6274 | 2.5068 | 2.4117 | 2.3344 | 2.2702 | 2.1692 | 2.0613 | 1.9445 | 1.8817 | 1.8152 | 1.7440 | 1.6668 | 1.5810 | |
| 120 | 5.1523 | 3.8046 | 3.2269 | 2.8943 | 2.6740 | 2.5154 | 2.3948 | 2.2994 | 2.2217 | 2.1570 | 2.0548 | 1.9450 | 1.8249 | 1.7597 | 1.6899 | 1.6141 | 1.5299 | 1.4327 | |

4 上 0.01 分位数

| | | F 分布表 (上 0.01 分位数) | | | | | | | | | | | | | | | | | |
|-----|-----------|--------------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|--|
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 12 | 15 | 20 | 24 | 30 | 40 | 60 | 120 | |
| 1 | 4052.1807 | 4999.5000 | 5403.3520 | 5624.5833 | 5763.6496 | 5858.9861 | 5928.3557 | 5981.0703 | 6022.4732 | 6055.8467 | 6106.3207 | 6157.2846 | 6208.7302 | 6234.6309 | 6260.6486 | 6286.7821 | 6313.0301 | 6339.3913 | |
| 2 | 98.5025 | 99.0000 | 99.1662 | 99.2494 | 99.2993 | 99.3326 | 99.3564 | 99.3742 | 99.3881 | 99.3992 | 99.4159 | 99.4325 | 99.4492 | 99.4575 | 99.4658 | 99.4742 | 99.4825 | 99.4908 | |
| 3 | 34.1162 | 30.8165 | 29.4567 | 28.7099 | 28.2371 | 27.9107 | 27.6717 | 27.4892 | 27.3452 | 27.2287 | 27.0518 | 26.8722 | 26.6988 | 26.5975 | 26.5045 | 26.4018 | 26.3164 | 26.2211 | |
| 4 | 21.1977 | 18.0000 | 16.6944 | 15.9770 | 15.5219 | 15.2069 | 14.9758 | 14.7989 | 14.6591 | 14.5459 | 14.3736 | 14.1982 | 14.0196 | 13.9291 | 13.8377 | 13.7454 | 13.6522 | 13.5581 | |
| 5 | 16.2582 | 13.2739 | 12.0600 | 11.3919 | 10.9670 | 10.6723 | 10.4555 | 10.2893 | 10.1578 | 10.0510 | 9.8883 | 9.7222 | 9.5526 | 9.4665 | 9.3793 | 9.2912 | 9.2020 | 9.1118 | |
| 6 | 13.7450 | 10.9248 | 9.7795 | 9.1483 | 8.7459 | 8.4661 | 8.2600 | 8.1017 | 7.9761 | 7.8741 | 7.7183 | 7.5590 | 7.3958 | 7.3127 | 7.2285 | 7.1432 | 7.0567 | 6.9690 | |
| 7 | 12.2464 | 9.5466 | 8.4513 | 7.8466 | 7.4604 | 7.1914 | 6.9928 | 6.8400 | 6.7188 | 6.6201 | 6.4691 | 6.3143 | 6.1554 | 6.0743 | 5.9920 | 5.9084 | 5.8236 | 5.7373 | |
| 8 | 11.2586 | 8.6491 | 7.5910 | 7.0061 | 6.6318 | 6.3707 | 6.1776 | 6.0289 | 5.9106 | 5.8143 | 5.6667 | 5.5151 | 5.3591 | 5.2793 | 5.1981 | 5.1156 | 5.0316 | 4.9461 | |
| 9 | 10.5614 | 8.0215 | 6.9919 | 6.4221 | 6.0569 | 5.8018 | 5.6129 | 5.4671 | 5.3511 | 5.2555 | 5.1114 | 4.9621 | 4.8080 | 4.7290 | 4.6486 | 4.5666 | 4.4831 | 4.3978 | |
| 10 | 10.0443 | 7.5594 | 6.5523 | 5.9943 | 5.6393 | 5.3858 | 5.2001 | 5.0567 | 4.9424 | 4.8491 | 4.7059 | 4.5581 | 4.4054 | 4.3269 | 4.2469 | 4.1653 | 4.0819 | 3.9965 | |
| 11 | 9.6460 | 7.2057 | 6.2167 | 5.6683 | 5.3160 | 5.0692 | 4.8861 | 4.7445 | 4.6315 | 4.5393 | 4.3974 | 4.2509 | 4.0990 | 4.0209 | 3.9411 | 3.8596 | 3.7761 | 3.6904 | |
| 12 | 9.3302 | 6.9266 | 5.9265 | 5.4120 | 5.0643 | 4.8206 | 4.6395 | 4.4994 | 4.3875 | 4.2961 | 4.1553 | 4.0096 | 3.8584 | 3.7805 | 3.7008 | 3.6192 | 3.5355 | 3.4494 | |
| 13 | 9.0738 | 6.7109 | 5.7394 | 5.2053 | 4.8616 | 4.6204 | 4.4410 | 4.3021 | 4.1911 | 4.1003 | 3.9603 | 3.8154 | 3.6646 | 3.5868 | 3.5070 | 3.4253 | 3.3413 | 3.2548 | |
| 14 | 8.8616 | 6.5149 | 5.5639 | 5.0354 | 4.6950 | 4.4558 | 4.2779 | 4.1399 | 4.0297 | 3.9394 | 3.8001 | 3.6557 | 3.5052 | 3.4274 | 3.3476 | 3.2656 | 3.1813 | 3.0942 | |
| 15 | 8.6931 | 6.3589 | 5.4170 | 4.8932 | 4.5556 | 4.3183 | 4.1415 | 4.0045 | 3.8948 | 3.8049 | 3.6662 | 3.5222 | 3.3719 | 3.2940 | 3.2141 | 3.1319 | 3.0471 | 2.9595 | |
| 16 | 8.5310 | 6.2262 | 5.2922 | 4.7726 | 4.4374 | 4.2016 | 4.0259 | 3.8906 | 3.7804 | 3.6909 | 3.5527 | 3.4089 | 3.2587 | 3.1808 | 3.1007 | 3.0182 | 2.9330 | 2.8447 | |
| 17 | 8.3997 | 6.1121 | 5.1850 | 4.6680 | 4.3359 | 4.1015 | 3.9267 | 3.7910 | 3.6822 | 3.5921 | 3.4552 | 3.3117 | 3.1615 | 3.0835 | 3.0032 | 2.9205 | 2.8348 | 2.7459 | |
| 18 | 8.2854 | 6.0129 | 5.0919 | 4.5790 | 4.2479 | 4.0146 | 3.8406 | 3.7054 | 3.5971 | 3.5082 | 3.3706 | 3.2273 | 3.0771 | 2.9990 | 2.9185 | 2.8354 | 2.7493 | 2.6597 | |
| 19 | 8.1849 | 5.9259 | 5.0103 | 4.5003 | 4.1708 | 3.9386 | 3.7653 | 3.6305 | 3.5225 | 3.4338 | 3.2965 | 3.1533 | 3.0031 | 2.9249 | 2.8442 | 2.7608 | 2.6742 | 2.5839 | |
| 20 | 8.0960 | 5.8489 | 4.9382 | 4.4307 | 4.1027 | 3.8714 | 3.6987 | 3.5644 | 3.4567 | 3.3682 | 3.2311 | 3.0880 | 2.9377 | 2.8594 | 2.7785 | 2.6947 | 2.6077 | 2.5168 | |
| 21 | 8.0166 | 5.7804 | 4.8740 | 4.3688 | 4.0421 | 3.8117 | 3.6396 | 3.5056 | 3.3981 | 3.3098 | 3.1730 | 3.0300 | 2.8796 | 2.8010 | 2.7200 | 2.6359 | 2.5484 | 2.4568 | |
| 22 | 7.9454 | 5.7190 | 4.8166 | 4.3134 | 3.9880 | 3.7583 | 3.5867 | 3.4530 | 3.3458 | 3.2576 | 3.1209 | 2.9779 | 2.8274 | 2.7488 | 2.6675 | 2.5831 | 2.4951 | 2.4029 | |
| 23 | 7.8811 | 5.6637 | 4.7649 | 4.2636 | 3.9392 | 3.7102 | 3.5390 | 3.4057 | 3.2986 | 3.2086 | 3.0740 | 2.9311 | 2.7805 | 2.7017 | 2.6202 | 2.5355 | 2.4471 | 2.3542 | |
| 24 | 7.8229 | 5.6136 | 4.7181 | 4.2184 | 3.8951 | 3.6667 | 3.4959 | 3.3629 | 3.2560 | 3.1681 | 3.0316 | 2.8887 | 2.7380 | 2.6591 | 2.5773 | 2.4923 | 2.4035 | 2.3100 | |
| 25 | 7.7698 | 5.5680 | 4.6755 | 4.1774 | 3.8550 | 3.6272 | 3.4568 | 3.3232 | 3.2172 | 3.1294 | 2.9931 | 2.8502 | 2.6993 | 2.6203 | 2.5383 | 2.4530 | 2.3637 | 2.2696 | |
| 26 | 7.7213 | 5.5263 | 4.6366 | 4.1400 | 3.8183 | 3.5911 | 3.4210 | 3.2884 | 3.1818 | 3.0941 | 2.9578 | 2.8150 | 2.6640 | 2.5848 | 2.5026 | 2.4170 | 2.3273 | 2.2325 | |
| 27 | 7.6767 | 5.4881 | 4.6009 | 4.1056 | 3.7848 | 3.5580 | 3.3882 | 3.2558 | 3.1494 | 3.0618 | 2.9256 | 2.7827 | 2.6316 | 2.5522 | 2.4699 | 2.3840 | 2.2938 | 2.1985 | |
| 28 | 7.6356 | 5.4529 | 4.5681 | 4.0740 | 3.7539 | 3.5276 | 3.3581 | 3.2259 | 3.1195 | 3.0320 | 2.8959 | 2.7530 | 2.6017 | 2.5223 | 2.4397 | 2.3535 | 2.2629 | 2.1670 | |
| 29 | 7.5977 | 5.4204 | 4.5378 | 4.0449 | 3.7254 | 3.4995 | 3.3303 | 3.1982 | 3.0920 | 3.0045 | 2.8685 | 2.7256 | 2.5742 | 2.4946 | 2.4118 | 2.3253 | 2.2344 | 2.1379 | |
| 30 | 7.5625 | 5.3903 | 4.5097 | 4.0179 | 3.6990 | 3.4735 | 3.3045 | 3.1726 | 3.0665 | 2.9791 | 2.8431 | 2.7002 | 2.5487 | 2.4689 | 2.3860 | 2.2992 | 2.2079 | 2.1108 | |
| 40 | 7.3141 | 5.1785 | 4.3126 | 3.8283 | 3.5138 | 3.2910 | 3.1238 | 2.9930 | 2.8876 | 2.8005 | 2.6648 | 2.5216 | 2.3689 | 2.2880 | 2.2034 | 2.1142 | 2.0194 | 1.9172 | |
| 60 | 7.0771 | 4.9774 | 4.1259 | 3.6490 | 3.3389 | 3.1187 | 2.9530 | 2.8233 | 2.7185 | 2.6318 | 2.4961 | 2.3523 | 2.1978 | 2.1154 | 2.0285 | 1.9360 | 1.8653 | 1.7623 | |
| 120 | 6.8509 | 4.7865 | 3.9491 | 3.4795 | 3.1735 | 2.9559 | 2.7918 | 2.6629 | 2.5586 | 2.4721 | 2.3363 | 2.1915 | 2.0346 | 1.9500 | 1.8600 | 1.7628 | 1.6957 | 1.5933 | |

A.3.5 二项分布表

A.3.6 泊松分布表

A.4 数列和常数

A.4.1 数列

卡特兰数

Catalan 数又称明安图数.

递归定义

$$1. C_0 = C_1 = 1.$$

$$2. C_n = \sum_{k=0}^{n-1} C_k C_{n-1-k}.$$

前几项值: 1, 1, 2, 5, 14, 43, 132, 429, 1439, 4862, 16796...

生成函数

由 $G(x) = \sum_{n=0}^{\infty} C_n x^n$ 知 $G^2(x) = \sum_{n=0}^{\infty} C_{n+1} x^n$, 故

$$\begin{cases} G(x) = \sum_{n=0}^{\infty} C_n x^n \\ C_n = \sum_{k=0}^{n-1} C_k C_{n-1-k} \end{cases} \Rightarrow G^2(x) = \sum_{n=0}^{\infty} C_{n+1} x^n$$
$$\begin{cases} xG^2(x) + 1 = G(x) \\ G(0) = 1 \end{cases} \Rightarrow G(x) = \frac{1 - \sqrt{1 - 4x}}{2x}$$

通项公式

$$1. C_n = \frac{1}{n+1} \binom{2n}{n} = \frac{1}{2n+1} \binom{2n+1}{n}.$$

$$2. C_n = \binom{2n}{n} - \binom{2n}{n-1}.$$

$$3. C_n = \frac{1}{n+1} \sum_{i=0}^n \binom{n}{i}^2.$$

证明

1. 由生成函数泰勒展开即得.

2. 由组合数定义即得.

3. 对比 $(1+x)^n (1+\frac{1}{x})^n = \frac{(1+x)^{2n}}{x^n}$ 两边系数, 即得 $\sum_{i=0}^n \binom{n}{i}^2 = \binom{2n}{n}$.

递推公式: $C_n = \frac{4n-2}{n+1} C_{n-1}$.

证明 由通项公式即得.

例题

1. 满足通项关系 $C_n = \binom{2n}{n} - \binom{2n}{n-1}$ 的场景.
 1. 在 $n \times n$ 网格中, 一开始在 $(0, 0)$ 处, 每次可以向上走一格或向右走一格, 在任一时刻, 向右的次数不少于向上的次数, 则合法的路径有 $\binom{2n}{n} - \binom{2n}{n-1} = C_n$ 种.
 2. 有 n 对括号, 则长度为 $2n$ 的括号序列中合法的序列有 C_n 种. (入栈出栈)
 3. 一个圆周上有 $2n$ 个点, 两两配对并连线, 则所有弦不相交的连接有 C_n 种.
2. 满足递归定义 $C_n = \sum_{k=0}^{n-1} C_k C_{n-k-1}$ 的场景.
 1. 把一个 n 层的矩形阶梯分为 n 个矩形的方法有 C_n 种.
 2. 凸 $n+2$ 边形按顶点连线划分为 n 个三角形的方法有 C_n 种.

A.4.2 常数

卡特兰常数

级数定义与积分定义

$$\begin{aligned} G &= \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n+1)^2} = \int_0^1 \frac{\arctan x}{x} dx \\ &= -\int_0^1 \frac{\ln x}{1+x^2} dx = \int_0^{+\infty} \frac{\ln x}{1+x^2} dx \\ &= -\int_0^{\frac{\pi}{4}} \ln \tan x dx = \int_0^{\frac{\pi}{4}} \ln \cot x dx \\ &= 0.915965594177219015054603515\dots \end{aligned}$$

常用积分

- $\left[0, \frac{\pi}{2}\right]$.
 - 对数与三角
 - 正余弦 (区间再现后相加)
 - $\int_0^{\frac{\pi}{2}} \ln \sin x dx = -\frac{\pi}{2} \ln 2.$
 - $\int_0^{\frac{\pi}{2}} \ln \cos x dx = -\frac{\pi}{2} \ln 2.$
 - 正余切 (卡特兰常数定义)

- $\int_0^{\frac{\pi}{2}} \ln \tan x \, dx = -G.$
 - $\int_0^{\frac{\pi}{2}} \ln \cot x \, dx = G.$
 - $1 \pm \text{正余弦 (由半角公式即得)}$
 - $\int_0^{\frac{\pi}{2}} \ln(1 + \sin x) \, dx = \int_0^{\frac{\pi}{2}} \ln(1 + \cos x) \, dx = 2G - \frac{\pi}{2} \ln 2.$
 - $\int_0^{\frac{\pi}{2}} \ln(1 - \sin x) \, dx = \int_0^{\frac{\pi}{2}} \ln(1 - \cos x) \, dx = -2G - \frac{\pi}{2} \ln 2.$
 - $1 + \text{正余切 (分区间利用结论)}$
 - $\int_0^{\frac{\pi}{2}} \ln(1 + \tan x) \, dx = G + \frac{\pi}{4} \ln 2.$
 - $\int_0^{\frac{\pi}{2}} \ln(1 + \cot x) \, dx = G + \frac{\pi}{4} \ln 2.$
- 幂与三角 (分布积分用结论)
 - $\int_0^{\frac{\pi}{2}} \frac{x \sin x}{1 + \cos x} \, dx = 2G - \frac{\pi}{2} \ln 2.$
 - $\int_0^{\frac{\pi}{2}} \frac{x \sin x}{1 - \cos x} \, dx = 2G + \frac{\pi}{2} \ln 2.$
 - $\int_0^{\frac{\pi}{2}} \frac{x \cos x}{1 + \sin x} \, dx = -2G + \pi \ln 2.$
 - $\int_0^{\frac{\pi}{2}} \frac{x \cos x}{1 - \sin x} \, dx = +\infty.$
- $\left[0, \frac{\pi}{4}\right].$
 - 对数与三角
 - 正余弦 (相加减后解方程)
 - $\int_0^{\frac{\pi}{4}} \ln \sin x \, dx = -\frac{1}{2}G - \frac{\pi}{4} \ln 2.$
 - $\int_0^{\frac{\pi}{4}} \ln \cos x \, dx = \frac{1}{2}G - \frac{\pi}{4} \ln 2.$
 - $1 \pm \text{正余切 (区间再现后展开)}$
 - $\int_0^{\frac{\pi}{4}} \ln(1 + \tan x) \, dx = \frac{\pi}{8} \ln 2.$
 - $\int_0^{\frac{\pi}{4}} \ln(1 - \tan x) \, dx = \frac{\pi}{8} \ln 2 - G.$
 - $\int_0^{\frac{\pi}{4}} \ln(\cot x + 1) \, dx = G + \frac{\pi}{4} \ln 2.$
 - $\int_0^{\frac{\pi}{4}} \ln(\cot x - 1) \, dx = \frac{\pi}{8} \ln 2.$
 - 正余弦和差 (平方之后二倍角)

$$\begin{aligned} \blacksquare \int_0^{\frac{\pi}{4}} \ln(\cos x + \sin x) dx &= \frac{1}{2}G - \frac{\pi}{8}\ln 2. \\ \blacksquare \int_0^{\frac{\pi}{4}} \ln(\cos x - \sin x) dx &= -\frac{1}{2}G - \frac{\pi}{8}\ln 2. \end{aligned}$$

○ 幂与三角

■ $x \cdot$ 正余切 (分布积分用结论)

$$\begin{aligned} \blacksquare \int_0^{\frac{\pi}{4}} x \tan x dx &= \frac{1}{2}G - \frac{\pi}{8}\ln 2. \\ \blacksquare \int_0^{\frac{\pi}{4}} x \cot x dx &= \frac{1}{2}G + \frac{\pi}{8}\ln 2. \end{aligned}$$

• 其它区间

○ 幂与对数 (三角换元用结论)

$$\begin{aligned} \blacksquare \int_0^1 \frac{\ln(1+x^2)}{1+x^2} dx &= -G + \frac{\pi}{2}\ln 2. \\ \blacksquare \int_0^1 \frac{\ln(1-x^2)}{1+x^2} dx &= -G + \frac{\pi}{4}\ln 2. \\ \blacksquare \int_0^{\frac{\sqrt{2}}{2}} \frac{\ln x}{\sqrt{1-x^2}} dx &= -\frac{1}{2}G - \frac{\pi}{4}\ln 2. \end{aligned}$$
