

样本均值与样本方差相互独立的充要条件

定理 设总体 X 的分布函数 $F(x)$ 具有二阶矩, 即 $EX = \mu < \infty, DX = \sigma^2 < \infty$, 若 (X_1, X_2, \dots, X_n) 为来自总体 X 的一个样本, 则样本均值 $\bar{X} = \frac{1}{n} \sum_{j=1}^n X_j$ 与样本方差 $S^2 = \frac{1}{n-1} \sum_{j=1}^n (X_j - \bar{X})^2$ 相互独立的充要条件是总体 $X \sim N(\mu, \sigma^2)$.^a

证明. 充分性可参见一般教材, 此处从略, 下面证明必要性. 设总体 X 的特征函数为 $\alpha(t)$, \bar{X} 的特征函数为 $\varphi_1(t_1)$, S^2 的特征函数为 $\varphi_2(t_2)$, (\bar{X}, S^2) 的特征函数为 $\varphi(t_1, t_2)$. 因为 \bar{X} 与 S^2 相互独立, 故 $\varphi(t_1, t_2) = \varphi_1(t_1)\varphi_2(t_2)$, 而由特征函数的性质知:

$$\varphi_1(t_1) = \left(\alpha \left(\frac{t_1}{n} \right) \right)^n, \quad \frac{\varphi_2'(0)}{i} = ES^2 = \sigma^2, \quad \text{i.e.} \quad \varphi_2'(0) = i\sigma^2.$$

故

$$\left. \frac{\partial \varphi(t_1, t_2)}{\partial t_2} \right|_{t_2=0} = \varphi_1(t_1)\varphi_2'(0) = i\sigma^2 \left(\alpha \left(\frac{t_1}{n} \right) \right)^n. \quad (1)$$

另一方面, $\varphi(t_1, t_2) = E \left[e^{i(t_1\bar{X} + t_2S^2)} \right]$, 故

$$\begin{aligned} \left. \frac{\partial \varphi(t_1, t_2)}{\partial t_2} \right|_{t_2=0} &= E \left(iS^2 e^{i(t_1\bar{X} + t_2S^2)} \right) \Big|_{t_2=0} = E \left(iS^2 e^{it_1\bar{X}} \right) \\ &= iE \left[\left(\frac{1}{n} \sum_{j=1}^n X_j^2 - \bar{X}^2 \right) e^{it_1\bar{X}} \right] \\ &= \frac{i}{n} \sum_{j=1}^n E \left(X_j^2 e^{i\frac{t_1}{n} \sum_{j=1}^n X_j} \right) - iE \left(\bar{X}^2 e^{it_1\bar{X}} \right). \end{aligned}$$

又 (X_1, X_2, \dots, X_n) 的联合分布函数为

$$F^*(x_1, x_2, \dots, x_n) = F(x_1)F(x_2) \cdots F(x_n).$$

故由 Riemann-Stieltjes 积分的性质有

$$\begin{aligned} E \left(\bar{X}^2 e^{it_1\bar{X}} \right) &= \frac{1}{n^2} \left[\left(\sum_{j=1}^n X_j^2 + \sum_{j \neq k} X_j X_k \right) e^{i\frac{t_1}{n} \sum_{j=1}^n X_j} \right] \\ &= \frac{1}{n^2} \sum_{j=1}^n E \left(X_j^2 e^{i\frac{t_1}{n} \sum_{j=1}^n X_j} \right) + \frac{1}{n^2} \sum_{j \neq k} E \left(X_j X_k e^{i\frac{t_1}{n} \sum_{j=1}^n X_j} \right) \\ &= \left(\alpha \left(\frac{t_1}{n} \right) \right)^{n-1} \frac{1}{n^2} \sum_{j=1}^n \int_{-\infty}^{\infty} x^2 e^{i\frac{t_1}{n} x} dF(x) + \left(\alpha \left(\frac{t_1}{n} \right) \right)^{n-2} \frac{1}{n^2} \left[\int_{-\infty}^{\infty} x e^{i\frac{t_1}{n} x} dF(x) \right]^2 \\ &= \frac{1}{n} \left(\alpha \left(\frac{t_1}{n} \right) \right)^{n-1} \int_{-\infty}^{\infty} x^2 e^{i\frac{t_1}{n} x} dF(x) + \frac{n-1}{n} \left(\alpha \left(\frac{t_1}{n} \right) \right)^{n-2} \left[\int_{-\infty}^{\infty} x e^{i\frac{t_1}{n} x} dF(x) \right]^2. \end{aligned}$$

故

$$\begin{aligned} \frac{\partial \varphi(t_1, t_2)}{t_2} \Big|_{t_2=0} &= \frac{n-1}{n} i \left\{ \left(\alpha \left(\frac{t_1}{n} \right) \right)^{n-1} \int_{-\infty}^{\infty} x^2 e^{i \frac{t_1}{n} x} dF(x) \right. \\ &\quad \left. + \frac{n-1}{n} \left(\alpha \left(\frac{t_1}{n} \right) \right)^{n-2} \left[\int_{-\infty}^{\infty} x e^{i \frac{t_1}{n} x} dF(x) \right]^2 \right\} \end{aligned} \quad (2)$$

由 (1) 和 (2) 式得

$$\alpha \left(\frac{t_1}{n} \right) \int_{-\infty}^{\infty} x^2 e^{i \frac{t_1}{n} x} dF(x) - \left[\int_{-\infty}^{\infty} x e^{i \frac{t_1}{n} x} dF(x) \right]^2 = \left(\alpha \left(\frac{t_1}{n} \right) \right)^2 \sigma^2. \quad (3)$$

又因为 $\alpha(t) = \int_{-\infty}^{\infty} e^{itx} dF(x)$, 故

$$\frac{d\alpha(t)}{dt} = i \int_{-\infty}^{\infty} x e^{itx} dF(x), \quad \frac{d^2\alpha(t)}{dt^2} = i^2 \int_{-\infty}^{\infty} x^2 e^{itx} dF(x).$$

则由 (3) 式可知, 关于 $\alpha(t)$ 有

$$-\alpha \frac{d^2\alpha}{dt^2} + \left(\frac{d\alpha}{dt} \right)^2 = \sigma^2 \alpha^2. \quad (4)$$

下面解此微分方程. 令

$$\frac{d\alpha}{dt} = p, \quad \frac{d^2\alpha}{dt^2} = \frac{dp}{dt} = \frac{dp}{d\alpha} \frac{d\alpha}{dt} = p \frac{dp}{d\alpha}.$$

故 (4) 化为一阶方程:

$$-\alpha p \frac{dp}{d\alpha} + p^2 = -\frac{\alpha}{2} \frac{dp^2}{d\alpha} + p^2 = \sigma^2 \alpha^2.$$

令 $Q = p^2$, 将上述方程化为

$$-\frac{\alpha}{2} \frac{dQ}{d\alpha} + Q = \sigma^2 \alpha^2.$$

对此一阶线性微分方程, 不难得到通解为 $Q = (-2\sigma^2 \ln \alpha + C_1) \alpha^2$. 于是可得

$$p = \alpha \sqrt{-2\sigma^2 \ln \alpha + C_1} = \frac{d\alpha}{dt}.$$

再一步解得此方程通解为

$$-\frac{1}{\sigma^2} \sqrt{-2\sigma^2 \ln \alpha + C_1} = t + C_2,$$

进而得

$$\ln \alpha = A + Bt - \frac{1}{2} \sigma^2 t^2, \quad \alpha(t) = e^{A+Bt-\frac{1}{2}\sigma^2 t^2}.$$

又根据特征函数的性质有: $1 = \alpha(0) = e^A$, $A = 0$, 再由 $\mu = \frac{\alpha'(t)}{i} \Big|_{t=0} = \frac{B}{i}$, 故 $B = i\mu$, 即总体 X 的特征函数为 $\alpha(t) = e^{i\mu t - \frac{1}{2}\sigma^2 t^2}$, 故 $X \sim N(\mu, \sigma^2)$, 证毕. \square

^a证明方法选自 <http://www.doc88.com/p-7156270513349.html>