



第43讲 两个正态总体的抽样分布



定理三:设样本 $(X_1,...,X_{n_1})$ 和 $(Y_1,...,Y_{n_2})$ 分别来自总体 $N(\mu_1,\sigma_1^2)$ 和 $N(\mu_2,\sigma_2^2)$,并且它们相互独立. 样本均值分别为 \overline{X} , \overline{Y} ;样本方差分别为 S_1^2 , S_2^2 。则可以得到下面三个抽样分布.





(1)
$$F = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} = \frac{S_1^2}{S_2^2} / \frac{\sigma_1^2}{\sigma_2^2} \sim F(n_1 - 1, n_2 - 1).$$

证明:
$$\chi_1^2 = \frac{(n_1 - 1)S_1^2}{\sigma_1^2} \sim \chi^2(n_1 - 1)$$

 $\chi_2^2 = \frac{(n_2 - 1)S_2^2}{\sigma_2^2} \sim \chi^2(n_2 - 1)$ S_1^2, S_2^2 独立,

$$\therefore F = \frac{S_1^2}{S_2^2} / \frac{\sigma_1^2}{\sigma_2^2} = \frac{\chi_1^2 / (n_1 - 1)}{\chi_2^2 / (n_2 - 1)} \sim F(n_1 - 1, n_2 - 1)$$





(2)
$$\frac{\left(\bar{X} - \bar{Y}\right) - \left(\mu_{1} - \mu_{2}\right)}{\sqrt{\frac{\sigma_{1}^{2}}{n_{1}} + \frac{\sigma_{2}^{2}}{n_{2}}}} \sim N(0,1)$$
证明: $\bar{X} \sim N(\mu_{1}, \frac{\sigma_{1}^{2}}{n_{1}}), \; \bar{Y} \sim N(\mu_{2}, \frac{\sigma_{2}^{2}}{n_{2}}),$

$$\Rightarrow \bar{X} - \bar{Y} \sim N(\mu_{1} - \mu_{2}, \frac{\sigma_{1}^{2}}{n_{1}} + \frac{\sigma_{2}^{2}}{n_{2}}),$$

$$\Rightarrow \left[\left(\bar{X} - \bar{Y}\right) - \left(\mu_{1} - \mu_{2}\right)\right] / \sqrt{\frac{\sigma_{1}^{2}}{n_{1}} + \frac{\sigma_{2}^{2}}{n_{2}}} \sim N(0,1).$$





(3) 当
$$\sigma_1^2 = \sigma_2^2 = \sigma^2$$
时,
$$\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{S_w \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t(n_1 + n_2 - 2)$$
其中 $S_w^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}$, $S_w = \sqrt{S_w^2}$





证明: 当
$$\sigma_1^2 = \sigma_2^2 = \sigma^2$$
时,

由(2)
$$\frac{\left(\bar{X} - \bar{Y}\right) - \left(\mu_{1} - \mu_{2}\right)}{\sigma\sqrt{\frac{1}{n_{1}} + \frac{1}{n_{2}}}} \sim N(0,1)$$
$$\frac{(n_{1} - 1)S_{1}^{2}}{\sigma^{2}} \sim \chi^{2}(n_{1} - 1), \frac{(n_{2} - 1)S_{2}^{2}}{\sigma^{2}} \sim \chi^{2}(n_{2} - 1)$$

$$\therefore \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{\sigma^2} \sim \chi^2(n_1 + n_2 - 2)$$





由
$$t$$
分布定义, $\frac{(X-Y)-(\mu_1-\mu_2)}{c}$

$$S_{w}\sqrt{\frac{1}{n_{1}}+\frac{1}{n_{2}}}$$
 $(\overline{\mathbf{v}},\overline{\mathbf{v}})$

$$= \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{\sigma \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} / \sqrt{\frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{\sigma^2(n_1 + n_2 - 2)}}$$

$$\sim t(n_1 + n_2 - 2)$$

思考: 若 σ^2 未知, 为什么用 S_w^2 来估计 σ^2 , Ω^2



而不用 S_1^2 或 S_2^2 来估计 σ^2 呢?





$$E(S_1^2) = \sigma^2 \qquad E(S_2^2) = \sigma^2$$

$$E(S_2^2) = \sigma^2$$

$$E(S_w^2) = \sigma^2$$

$$D(S_1^2) = \frac{2\sigma^4}{n_1 - 1}$$

$$D(S_1^2) = \frac{2\sigma^4}{n_1 - 1} \quad D(S_2^2) = \frac{2\sigma^4}{n_2 - 1} \quad -\$42 \text{ if } \emptyset 2$$

$$S_{w}^{2} = \frac{(n_{1}-1)S_{1}^{2} + (n_{2}-1)S_{2}^{2}}{n_{1}+n_{2}-2}$$
从直观来看, S_{w}^{2} 比 S_{1}^{2} 或 S_{2}^{2} 包含更多 O^{2} 的信息

$$D(S_w^2) = \frac{2\sigma^4}{n_1 + n_2 - 2}$$
 此 $D(S_1^2), D(S_2^2)$ 更小.





对于单个正态总体 $N(\mu,\sigma^2)$,得到了 \overline{X} , S^2 的分布,用于对 μ,σ^2 进行推断(区间估计,假设检验).

对于两个独立正态总体 $N(\mu_1, \sigma_1^2)$, $N(\mu_2, \sigma_2^2)$, 得到了 $\overline{X} - \overline{Y}, S_1^2/S_2^2$ 的分布,用于对 $\mu_1 - \mu_2, \frac{\sigma_1^2}{\sigma_2^2}$ 进行推断.