

From last time:

Stat 571

10-29-13

The denominator was maximized

①

when $\hat{\mu} = \bar{x}$ and $\hat{\Sigma} = S_n$

Evaluate the denominator at its max.

$$L(\bar{x}, S_n) = \frac{1}{(2\pi)^{\frac{np}{2}} |S_n|^{\frac{n}{2}}} e^{-\frac{1}{2}(n-1)\text{tr}(S_n^{-1}S)}$$

$$\text{Note } S_n = \frac{(n-1)S}{n}, \quad S_n^{-1} = \frac{n}{(n-1)} S^{-1}$$

$$S_n^{-1}S = \frac{n}{n-1} S^{-1}S = \frac{n}{n-1} I$$

②

$$\text{tr}(S_n^{-1}S) = \frac{n}{n-1} p$$

$$L(\bar{x}, S_n) = \frac{1}{(2\pi)^{\frac{np}{2}} |S_n|^{\frac{n}{2}}} e^{-\frac{1}{2}(n-1)\frac{n}{n-1}p}$$

$$= \frac{1}{(2\pi)^{\frac{np}{2}} |S_n|^{\frac{n}{2}}} e^{-\frac{np}{2}}$$

③

Maximize the numerator:

$$L(\theta) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma|^{1/2}} e^{-\frac{1}{2} \sum_{i=1}^n (\vec{x}_i - \vec{\mu}_0)' \Sigma^{-1} (\vec{x}_i - \vec{\mu}_0)}$$

$$\sum_{i=1}^n (\vec{x}_i - \vec{\mu}_0)' \Sigma^{-1} (\vec{x}_i - \vec{\mu}_0)$$

$$= \sum_{i=1}^n \text{tr} [(\vec{x}_i - \vec{\mu}_0)' \Sigma^{-1} (\vec{x}_i - \vec{\mu}_0)]$$

$$= \sum_{i=1}^n \text{tr} [\Sigma^{-1} (\vec{x}_i - \vec{\mu}_0) (\vec{x}_i - \vec{\mu}_0)']$$

$$= \text{tr} \left[\Sigma^{-1} \underbrace{\sum_{i=1}^n (\vec{x}_i - \vec{\mu}_0) (\vec{x}_i - \vec{\mu}_0)'}_B \right]$$

④

Now

$$L(\theta) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma|^{1/2}} e^{-\frac{1}{2} \text{tr}(\Sigma^{-1} B)}$$

let $b = n/2$ + use 2nd lemma from last time.

⑤

Result is that $L(\theta)$ is maximized

$$\text{when } \hat{\Sigma}_0 = \frac{B}{2b}$$

$$= \frac{1}{n} \sum_{i=1}^n (\vec{x}_i - \mu_0)(\vec{x}_i - \mu_0)'$$

Evaluate the numerator:

$$L(\vec{\mu}_0, \hat{\Sigma}_0) = \frac{1}{(2\pi)^{np/2} |\hat{\Sigma}_0|^{n/2}} e^{-\frac{1}{2} \text{tr}(\hat{\Sigma}_0^{-1} B)}$$

$$\text{But } B = n \hat{\Sigma}_0, \text{ or } \hat{\Sigma}_0 = \frac{B}{n}$$

$$\hat{\Sigma}_0^{-1} = n B^{-1}$$

⑥

$$\hat{\Sigma}_0^{-1} B = n B^{-1} B = n I$$

$$\text{tr}(\hat{\Sigma}_0^{-1} B) = \text{tr}(n I) = np$$

$$L(\vec{\mu}_0, \hat{\Sigma}_0) = \frac{1}{(2\pi)^{np/2} |\hat{\Sigma}_0|^{n/2}} e^{-\frac{1}{2} np}$$

$$\Lambda = \frac{\frac{1}{(2\pi)^{np/2} |\hat{\Sigma}_0|^{n/2}} e^{-np/2}}{\frac{1}{(2\pi)^{np/2} |S_n|^{n/2}} e^{-\frac{np}{2}}} = \frac{|S_n|^{n/2}}{|\hat{\Sigma}_0|^{n/2}}$$

(7)

The likelihood ratio test says to reject H_0 when $\Delta < c$.

Equivalently, reject H_0 when

$$\Delta^{2n} = \frac{|S_n|}{|E_0|} < c'$$

Defn: Δ^{2n} is called Wilks' Lambda

Lemma: let $A = \left[\begin{array}{c|c} A_{11} & A_{12} \\ \hline A_{21} & A_{22} \end{array} \right]$, then

$$\begin{aligned} |A| &= |A_{11}| \cdot |A_{22} - A_{21} A_{11}^{-1} A_{12}| \\ &= |A_{22}| \cdot |A_{11} - A_{12} A_{22}^{-1} A_{21}| \end{aligned}$$

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$$\begin{aligned} \text{Proof: } & \left[\begin{array}{c|c} I & -A_{12} A_{22}^{-1} \\ \hline 0 & I \end{array} \right] \left[\begin{array}{c|c} A_{11} & A_{12} \\ \hline A_{21} & A_{22} \end{array} \right] \left[\begin{array}{c|c} I & 0 \\ \hline -A_{22}^{-1} A_{21} & I \end{array} \right] \\ &= \left[\begin{array}{c|c} A_{11} - A_{12} A_{22}^{-1} A_{21} & 0 \\ \hline 0 & A_{22} \end{array} \right] \end{aligned}$$

$$\begin{array}{c|c} \text{Det. of L.U.S.} & \text{Det. of R.U.S.} \\ \hline = |A| & = |A_{11} - A_{12} A_{22}^{-1} A_{21}| \cdot |A_{22}| \end{array} \quad (9)$$

We have proven the 2nd equality. The 1st equality is done similarly.

Theorem: $\Lambda^{2ln} = \frac{1}{1 + (\bar{x} - \bar{\mu}_0)' S_n^{-1} (\bar{x} - \bar{\mu}_0)}$

Proof: Let $A = \left[\begin{array}{c|c} n S_n & \sqrt{n}(\bar{x} - \bar{\mu}_0) \\ \hline \sqrt{n}(\bar{x} - \bar{\mu}_0)' & -1 \end{array} \right]_{(p+1) \times (p+1)} \quad (10)$

Compute $|A|$ two ways, using the lemma.

$$\begin{aligned} |A| &= |n S_n| \cdot \left| -1 - \sqrt{n}(\bar{x} - \bar{\mu}_0)' (n S_n)^{-1} \sqrt{n}(\bar{x} - \bar{\mu}_0) \right| \\ &= n^p |S_n| (-1 - (\bar{x} - \bar{\mu}_0)' S_n^{-1} (\bar{x} - \bar{\mu}_0)) \end{aligned}$$

Also,

(11)

$$|A| = |-1| \cdot |n S_n - \sqrt{n}(\bar{x} - \vec{\mu}_0)(-1)^{-1} \sqrt{n}(\bar{x} - \vec{\mu}_0)'|$$

$$= -n^p \left| S_n + \underbrace{(\bar{x} - \vec{\mu}_0)(\bar{x} - \vec{\mu}_0)'} \right|$$

Using page (4) of Thursday's notes,

$$n \left[S_n + (\bar{x} - \vec{\mu})(\bar{x} - \vec{\mu})' \right] = \sum_{i=1}^n (\vec{x}_i - \vec{\mu})(\vec{x}_i - \vec{\mu})'$$

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$$\begin{aligned} \sum S_n + (\bar{x} - \vec{\mu}_0)(\bar{x} - \vec{\mu}_0)' &= \frac{1}{n} \sum_{i=1}^n (\vec{x}_i - \vec{\mu}_0)(\vec{x}_i - \vec{\mu}_0)' \\ &= \hat{\Phi}_0 \end{aligned}$$

Now, Equate the 2 expressions for $|A|$:

$$\begin{aligned} -n^p |S_n| (1 + (\bar{x} - \vec{\mu}_0)' S_n^{-1} (\bar{x} - \vec{\mu}_0)) \\ = -n^p |\hat{\Phi}_0| \end{aligned}$$

$$\therefore \frac{|S_n|}{|\hat{\Sigma}_0|} = \frac{1}{1 + (\bar{x} - \bar{\mu}_0)' S_n^{-1} (\bar{x} - \bar{\mu}_0)} //$$

Defn: $T^2 = n(\bar{x} - \bar{\mu}_0)' S^{-1} (\bar{x} - \bar{\mu}_0)$

is Hotelling's T^2 .

How are they related?

$$S_n = \frac{(n-1)S}{n}$$

$$S_n^{-1} = \frac{n}{(n-1)} S^{-1}$$

$$\Lambda^{2/n} = \frac{1}{1 + \frac{T^2}{n-1}}$$

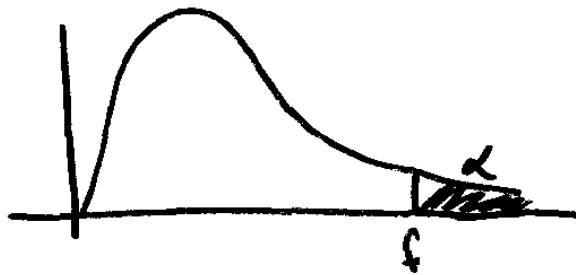
$$\Lambda^{2/n} < c' \iff T^2 > c''$$

(15)

Write

$$T^2 = \underbrace{\sqrt{n}(\bar{x} - \vec{\mu}_0)'}_{[N_p(\vec{0}, \Sigma)]'} \underbrace{\left[\frac{(n-1)\Sigma}{(n-1)} \right]^{-1}}_{\left[\frac{\text{Wishart}}{df} \right]^{-1}} \underbrace{\sqrt{n}(\bar{x} - \vec{\mu}_0)}_{N_p(\vec{0}, \Sigma)}$$

Hotelling showed that $T^2 \cdot \frac{n-p}{p(n-1)} \sim F_{p, n-p}$



Compute $f \cdot \frac{p(n-1)}{n-p}$

Reject H_0 when

$$T^2 > f \cdot \frac{p(n-1)}{n-p}.$$