STAT 450: Statistical Theory

Normal samples: Distribution Theory

Theorem: Suppose $X_1,\ldots,X_n$ are independent $N(\mu,\sigma^2)$ random variables. Then

1. $\bar X$ (sample mean)and s^2 (sample variance) independent.
2. $n^{1/2}(\bar X-\mu)/\sigma \sim N(0,1)$.
3. $(n-1)s^2/\sigma^2 \sim \chi^2_{n-1}$.
4. $n^{1/2}(\bar X-\mu)/s \sim t_{n-1}$.

Proof: Let $Z_i=(X_i-\mu)/\sigma$.
Then $Z_1,\ldots,Z_p$ are independent $N(0,1)$.
So $Z=(Z_1,\ldots,Z_p)^t$ is multivariate standard normal.
Note that $\bar X=\sigma\bar Z+\mu$ and $s^2=\sum(X_i-\bar X)^2/(n-1)=\sigma^2\sum(Z_i-\bar Z)^2/(n-1)$ Thus

$$\frac{n^{1/2}(\bar X-\mu)}{\sigma}=n^{1/2}\bar Z$$

$$\frac{(n-1)s^2}{\sigma^2}=\sum(Z_i-\bar Z)^2$$

and

$$T=\frac{n^{1/2}(\bar X-\mu)}{s}=\frac{n^{1/2}\bar Z}{s_Z}$$

where $(n-1)s_Z^2=\sum(Z_i-\bar Z)^2$.
So: reduced to $\mu=0$ and $\sigma=1$.
Step 1: Define

$$Y=(\sqrt{n}\bar Z, Z_1-\bar Z,\ldots,Z_{n-1}-\bar Z)^t\,.$$

(So Y has same dimension as Z.) Now

$$Y=\begin{bmatrix}\frac{1}{\sqrt{n}} & \frac{1}{\sqrt{n}} & \cdots & \frac{1}{\sqrt{n}}\\ 1-\frac{1}{n} & -\frac{1}{n} & \cdots & -\frac{1}{n}\\ -\frac{1}{n} & 1-\frac{1}{n} & \cdots & -\frac{1}{n}\\ \vdots & \vdots & \vdots & \vdots\end{bmatrix}\begin{bmatrix}Z_1\\ Z_2\\ \vdots\\ Z_n\end{bmatrix}$$

or letting M denote the matrix

$$Y=MZ\,.$$

It follows that $Y\sim MVN(0,MM^t)$ so we need to compute MM^t:

$$MM^t=\left[\begin{array}{c|cccc}1 & 0 & 0 & \cdots & 0\\ \hline 0 & 1-\frac{1}{n} & -\frac{1}{n} & \cdots & -\frac{1}{n}\\ \vdots & -\frac{1}{n} & \ddots & \cdots & -\frac{1}{n}\\ 0 & \vdots & \cdots & & 1-\frac{1}{n}\end{array}\right]$$
$$=\left[\begin{array}{c|c}1 & 0\\ \hline 0 & Q\end{array}\right]\,.$$

Solve for Z from Y: $Z_i=n^{-1/2}Y_1+Y_{i+1}$ for $1\le i\le n-1$. Use the identity

$$\sum_{i=1}^n (Z_i-\bar Z)=0$$

to get $Z_n = -\sum_{i=2}^n Y_i + n^{-1/2} Y_1$. So M invertible:

$$\Sigma^{-1} \equiv (MM^t)^{-1} = \left[\begin{array}{c|c} 1 & 0 \\ \hline 0 & Q^{-1} \end{array}\right]$$

Use change of variables to find f_Y. Let $\mathbf{y}_2$ denote vector whose entries are $y_2, \ldots, y_n$. Note that

$$y^t \Sigma^{-1} y = y_1^2 + \mathbf{y}_2^t Q^{-1} \mathbf{y}_2$$

Then

$$\begin{aligned} f_Y(y) =& (2\pi)^{-n/2} \exp[-y^t \Sigma^{-1} y/2] / |\det M| \\ =& \frac{1}{\sqrt{2\pi}} e^{-y_1^2/2} \times \\ & \frac{(2\pi)^{-(n-1)/2} \exp[-\mathbf{y}_2^t Q^{-1} \mathbf{y}_2/2]}{|\det M|} \end{aligned}$$

Note: $f_Y(y)$ is ftn of y_1 times a ftn of $y_2, \ldots, y_n$.

Thus $\sqrt{n}\bar{Z}$ is independent of $Z_1 - \bar{Z}, \ldots, Z_{n-1} - \bar{Z}$.

Since s_Z^2 is a function of $Z_1 - \bar{Z}, \ldots, Z_{n-1} - \bar{Z}$ we see that $\sqrt{n}\bar{Z}$ and s_Z^2 are independent.

Also, density of Y_1 is a multiple of the function of y_1 in the factorization above. But factor is standard normal density so $\sqrt{n}\bar{Z} \sim N(0,1)$.

First 2 parts done. Third part is an exercise.

Derivation of the χ_n^2 density:

Suppose $Z_1, \ldots, Z_n$ are independent $N(0,1)$. Define χ_n^2 distribution to be that of $U = Z_1^2 + \cdots + Z_n^2$. Define angles $\theta_1, \ldots, \theta_{n-1}$ by

$$\begin{aligned} Z_1 &= U^{1/2} \cos\theta_1 \\ Z_2 &= U^{1/2} \sin\theta_1 \cos\theta_2 \\ \vdots &= \vdots \\ Z_{n-1} &= U^{1/2} \sin\theta_1 \cdots \sin\theta_{n-2} \cos\theta_{n-1} \\ Z_n &= U^{1/2} \sin\theta_1 \cdots \sin\theta_{n-1} \end{aligned}$$

(Spherical co-ordinates in n dimensions. The θ values run from 0 to π except last θ from 0 to 2π.) Derivative formulas:

$$\frac{\partial Z_i}{\partial U} = \frac{1}{2U} Z_i$$

and

$$\frac{\partial Z_i}{\partial \theta_j} = \begin{cases} 0 & j > i \\ -Z_i \tan\theta_i & j = i \\ Z_i \cot\theta_j & j < i \end{cases}$$

Fix $n = 3$ to clarify the formulas.

Use shorthand $R = \sqrt{U}$.

Matrix of partial derivatives is

$$\left[\begin{array}{ccc} \frac{\cos\theta_1}{2R} & -R\sin\theta_1 & 0 \\ \frac{\sin\theta_1\cos\theta_2}{2R} & R\cos\theta_1\cos\theta_2 & -R\sin\theta_1\sin\theta_2 \\ \frac{\sin\theta_1\sin\theta_2}{2R} & Rcos\theta_1\sin\theta_2 & R\sin\theta_1\cos\theta_2 \end{array}\right]$$

Find determinant by adding $2U\cos\theta_1/\sin\theta_1$ times col 1 to col 2 then $\cos\theta_1\sin\theta_1\sin\theta_2/\cos\theta_2$ times col 2 to col 3 (no change in determinant).

Resulting matrix is lower triangular; diagonal entries $U^{-1/2}\cos\theta_1/2$, $U^{1/2}\cos\theta_2/\cos\theta_1$ and $U^{1/2}\sin\theta_1/\cos\theta_2$. We multiply these together to get the Jacobian

$$U^{1/2}\sin(\theta_1)/2$$

(non-negative for all U and θ_1).

General n: every term in the first column contains a factor $U^{-1/2}/2$ while every other entry has a factor $U^{1/2}$.

FACT: Multiplying a column in a matrix by c multiplies the determinant by c.

SO: Jacobian of the transformation is $u^{(n-1)/2}/2$ times some function, say h, which depends only on the angles.

Thus the joint density of $U, \theta_1, \ldots \theta_{n-1}$ is

$$(2\pi)^{-n/2}\exp(-u/2)u^{(n-2)/2}h(\theta_1, \cdots, \theta_{n-1})/2$$

To compute the density of U we must do an $n-1$ dimensional multiple integral $d\theta_{n-1}\cdots d\theta_1$.

Answer has the form

$$cu^{(n-2)/2}\exp(-u/2)$$

for some c.

Evaluate c by making

$$\int f_U(u)du = c\int u^{(n-2)/2}\exp(-u/2)du = 1$$

Substitute $y = u/2$, $du = 2dy$ to see that

$$c2^{n/2}\int y^{(n-2)/2}e^{-y}dy = c2^{n/2}\Gamma(n/2) = 1$$

CONCLUSION: the χ^2_n density is

$$\frac{1}{2\Gamma(n/2)}\left(\frac{u}{2}\right)^{(n-2)/2}e^{-u/2}1(u>0)\,.$$

Fourth part: consequence of first 3 parts and def'n of t_ν distribution.

Defn: $T \sim t_\nu$ if T has same distribution as

$$Z/\sqrt{U/\nu}$$

where $Z \sim N(0,1)$, $U \sim \chi^2_\nu$ and Z and U are independent.

Derive density of T in this definition:

$$\begin{aligned} P(T \le t) &= P(Z \le t\sqrt{U/\nu}) \\ &= \int_0^\infty \int_{-\infty}^{t\sqrt{u/\nu}} f_Z(z)f_U(u)dzdu \end{aligned}$$

Differentiate wrt t by differentiating inner integral:

$$\frac{\partial}{\partial t}\int_{at}^{bt} f(x)dx = bf(bt) - af(at)$$

by fundamental thm of calculus. Hence

$$\frac{d}{dt}P(T \le t) = \int_0^\infty f_U(u)\left(\frac{u}{\nu}\right)^{1/2}\frac{\exp[-t^2u/(2\nu)]}{\sqrt{2\pi}}du$$

Plug in

$$f_U(u) = \frac{1}{2\Gamma(\nu/2)}(u/2)^{(\nu-2)/2}e^{-u/2}$$

to get

$$f_T(t) = \frac{\int_0^\infty (u/2)^{(\nu-1)/2} \exp[-u(1+t^2/\nu)/2] du}{2\sqrt{\pi\nu}\Gamma(\nu/2)} .$$

Substitute $y = u(1+t^2/\nu)/2$, to get

$$dy = (1+t^2/\nu)du/2$$

$$(u/2)^{(\nu-1)/2} = [y/(1+t^2/\nu)]^{(\nu-1)/2}$$

leading to

$$f_T(t) = \frac{(1+t^2/\nu)^{-(\nu+1)/2}}{\sqrt{\pi\nu}\Gamma(\nu/2)} \int_0^\infty y^{(\nu-1)/2} e^{-y} dy$$

or

$$f_T(t) = \frac{\Gamma((\nu+1)/2)}{\sqrt{\pi\nu}\Gamma(\nu/2)} \frac{1}{(1+t^2/\nu)^{(\nu+1)/2}} \ ,.$$