

Deriving the distribution of test statistic for one sample t-test.

Let $x_1, x_2, \dots, x_n \text{ iid } N(\mu, \sigma^2)$. We want to test the hypotheses:

$$H_0: \mu = \mu_0 \quad \text{vs} \quad H_1: \mu \neq \mu_0$$

under H_0 ;

$$T = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

By Theorem, $\bar{x} \sim N(\mu, \sigma^2/n)$

Proof

If $x_1, x_2, \dots, x_n \sim N(\mu, \sigma^2)$

Then $z = \frac{x - \mu}{\sigma} \sim N(0, 1)$

$$f_z(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}, \quad -\infty < z < \infty$$

$$\begin{aligned} M_z(t) &= E e^{tz} = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} \cdot e^{tz} dz \\ &= e^{\frac{1}{2}t^2} \underbrace{\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(z-t)^2} dz}_{z \sim N(t, 1)} \\ &= e^{\frac{1}{2}t^2} \end{aligned}$$

Now $x = \mu + z\sigma$

$$\begin{aligned} \Rightarrow M_x(t) &= M_{\mu + z\sigma}(t) = M_\mu(t) \cdot M_z(\sigma t) \\ &= e^{\mu t} \cdot e^{\frac{1}{2}\sigma^2 t^2} \\ &= e^{\mu t + \frac{1}{2}\sigma^2 t^2} \end{aligned}$$

But $\bar{x} = \frac{1}{n} \sum x_i$

$$\begin{aligned} \Rightarrow M_{\bar{x}}(t) &= \left[M_x\left(\frac{t}{n}\right) \right]^n = \left[e^{\frac{\mu}{n}t + \frac{\sigma^2}{2n^2}t^2} \right]^n \\ &= e^{\mu t + \frac{\sigma^2}{2n}t^2} \end{aligned}$$

which is MGF of $N(\mu, \frac{\sigma^2}{n})$

$$\therefore \bar{x} \sim N(\mu, \frac{\sigma^2}{n})$$

Next, $S^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$

$$\Rightarrow \frac{(n-1)S^2}{\sigma^2} = \sum_{i=1}^n \frac{(x_i - \bar{x})^2}{\sigma^2} =$$

But $\sum_{i=1}^n \frac{(x_i - \mu)^2}{\sigma^2} = \sum_{i=1}^n \frac{(x_i - \bar{x} + \bar{x} - \mu)^2}{\sigma^2} = \frac{\sum (x_i - \bar{x})^2 + \sum (\bar{x} - \mu)^2}{\sigma^2} = \frac{\sum (x_i - \bar{x})^2 + n(\bar{x} - \mu)^2}{\sigma^2}$

$$\Rightarrow \sum_{i=1}^n \frac{(x_i - \mu)^2}{\sigma^2} = \sum_{i=1}^n \frac{(x_i - \bar{x})^2}{\sigma^2} + \frac{n(\bar{x} - \mu)^2}{\sigma^2}$$

But $\frac{\sum (x_i - \mu)^2}{\sigma^2} = Z = \sum_{i=1}^n \left(\frac{x_i - \mu}{\sigma} \right)^2 = \sum_{i=1}^n z_i^2 \sim \chi_n^2$

Also, $\frac{n(\bar{x} - \mu)^2}{\sigma^2} = \left(\frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \right)^2 = Z^2 \sim \chi_1^2$

$$\Rightarrow \chi_n^2 = \sum_{i=1}^n \frac{(x_i - \bar{x})^2}{\sigma^2} + \chi_1^2$$

Let $W = \chi_n^2$, $U = \chi_1^2$ and $V = \frac{(n-1)S^2}{\sigma^2} = \frac{\sum (x_i - \bar{x})^2}{\sigma^2}$

$$\Rightarrow M_W(t) = M_V(t) \cdot M_U(t)$$

$$\Rightarrow M_V(t) = \frac{M_W(t)}{M_U(t)} = \frac{(1-2t)^{-n/2}}{(1-2t)^{-1/2}} = (1-2t)^{-\frac{(n-1)}{2}}$$

Which is MGF of $\chi_{(n-1)}^2$

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

Next, we want to show that \bar{x} and S^2 are independent.
A number of options. Let's use transformation.

WLOG: assume $\mu=0$ and $\sigma^2=1$

Then $S^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{1}{n-1} \left[x_1 - \bar{x} + \sum_{i=2}^n (x_i - \bar{x})^2 \right]$
 $= \frac{1}{n-1} \left[\left[\sum_{i=2}^n (x_i - \bar{x}) \right]^2 + \sum_{i=2}^n (x_i - \bar{x})^2 \right]$ since $\sum (x_i - \bar{x}) = 0$

$\Rightarrow S^2$ is a function of only $x_i - \bar{x}$ for $i=2, 3, \dots$

The joint pdf of x_1, \dots, x_n is

$$f(x_1, \dots, x_n) = \left(\frac{1}{2\pi} \right)^{n/2} e^{-\frac{1}{2} \sum_{i=1}^n x_i^2}, \quad -\infty < x_i < \infty$$

We transform x_1, \dots, x_n into y_1, \dots, y_n as

$$y_1 = \bar{x} \quad (= \frac{1}{n} \sum x_i)$$

$$y_2 = x_2 - \bar{x}$$

$$y_3 = x_3 - \bar{x}$$

$$\vdots$$

$$y_n = x_n - \bar{x}$$

$$\Rightarrow x_1 = y_1 - \sum_{i=2}^n y_i \Rightarrow \frac{\partial x_1}{\partial y_1} = 1, \frac{\partial x_1}{\partial y_2} = -1, \dots, \frac{\partial x_1}{\partial y_n} = -1$$

$$x_2 = y_1 + y_2 \Rightarrow \frac{\partial x_2}{\partial y_1} = 1, \frac{\partial x_2}{\partial y_2} = 1$$

$$\vdots$$

$$x_n = y_1 + y_n \Rightarrow \frac{\partial x_n}{\partial y_1} = 1, \frac{\partial x_n}{\partial y_n} = 1$$

$$\text{So } J = \begin{vmatrix} 1 & -1 & -1 & \dots & -1 \\ 1 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 0 & 0 & \dots & 1 \end{vmatrix} = n$$

$$\Rightarrow f(y_1, \dots, y_n) = \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} \left(y_1 - \sum_{i=2}^n y_i \right)^2 - \frac{1}{2} \sum_{i=2}^n (y_i + y_1)^2}, n$$

$$\left(y_1 - \sum_{i=2}^n y_i \right)^2 = y_1^2 - 2y_1 \sum_{i=2}^n y_i + \left(\sum_{i=2}^n y_i \right)^2 \quad \text{--- (1)}$$

$$\sum_{i=2}^n (y_i + y_1)^2 = \sum_{i=2}^n y_i^2 + 2y_1 \sum_{i=2}^n y_i + (n-1)y_1^2 \quad \text{--- (2)}$$

From (1) and (2) --- (1) + (2) gives

$$= ny_1^2 + \left(\sum_{i=2}^n y_i \right)^2 + \sum_{i=2}^n y_i^2$$

$$\begin{aligned} \Rightarrow f(y_1, \dots, y_n) &= \frac{n}{(2\pi)^{n/2}} e^{-\frac{1}{2} \left[ny_1^2 + \left(\sum_{i=2}^n y_i \right)^2 + \sum_{i=2}^n y_i^2 \right]} \\ &= \left(\frac{n}{2\pi} \right)^{1/2} e^{-\frac{1}{2} ny_1^2} \cdot \frac{n^{1/2}}{(2\pi)^{n-1/2}} e^{-\frac{1}{2} \left[\left(\sum_{i=2}^n y_i \right)^2 + \sum_{i=2}^n y_i^2 \right]}, -\infty < y_i < \infty \end{aligned}$$

Since the joint pdf of y_1, y_2, \dots, y_n factors, y_1 and y_2, y_3, \dots, y_n are independent and hence \bar{x} and s^2 are independent.

$$\begin{aligned} \text{So } \frac{\bar{x} - \mu}{s/\sqrt{n}} &= \frac{\sqrt{n}(\bar{x} - \mu)}{s} = \frac{\sqrt{n}(\bar{x} - \mu)}{\sigma} \bigg/ s/\sigma \\ &= \frac{\sqrt{n}(\bar{x} - \mu)}{\sigma} \bigg/ \sqrt{\frac{(n-1)s^2}{\sigma^2} / (n-1)} \end{aligned}$$

$$\frac{\bar{x} - \mu}{s/\sqrt{n}} = \frac{\frac{\sqrt{n}(\bar{x} - \mu)}{\sigma}}{\sqrt{\frac{(n-1)s^2}{\sigma^2(n-1)}}} = \frac{z}{\sqrt{\chi^2_{(n-1)}/(n-1)}} = \frac{z}{\sqrt{u/v}}$$

Proof

Let $U = \chi^2_v$ and $z = \frac{\sqrt{n}(\bar{x} - \mu)}{\sigma}$

Then $f_z(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$ and $f_u(u) = \frac{u^{v/2-1}}{2^{v/2}\Gamma(v/2)} e^{-u/2}$

Since z and u are independent,

$$\begin{aligned} f(z, u) &= f(z) \cdot f(u) \\ &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} \cdot \frac{u^{v/2-1}}{2^{v/2}\Gamma(v/2)} e^{-u/2} \end{aligned}$$

Let $T = \frac{z}{\sqrt{u/v}}$ and $w = u$

$$\Rightarrow u = w \quad \text{and} \quad z = T\sqrt{\frac{w}{v}} = T\sqrt{\frac{w}{v}}$$

$$\Rightarrow \frac{\partial u}{\partial T} = 0, \quad \frac{\partial u}{\partial w} = 1 \quad \text{and} \quad \frac{\partial z}{\partial T} = \sqrt{\frac{w}{v}}, \quad \frac{\partial z}{\partial w} = \frac{1}{2} T \sqrt{\frac{w}{v}}^{-1/2}$$

$$\Rightarrow |J| = \begin{vmatrix} 0 & 1 \\ \sqrt{\frac{w}{v}} & \frac{1}{2} T \sqrt{\frac{w}{v}}^{-1/2} \end{vmatrix} = \sqrt{\frac{w}{v}}$$

$$\Rightarrow f(T, w) = \sqrt{\frac{w}{v}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(T\sqrt{\frac{w}{v}})^2} \cdot \frac{w^{v/2-1}}{2^{v/2}\Gamma(v/2)} e^{-w/2}$$

$$= \left(\frac{w}{2\pi v}\right)^{1/2} \cdot e^{-\frac{1}{2} \frac{T^2 w}{v}} \cdot \frac{w^{v/2-1}}{2^{v/2}\Gamma(v/2)} e^{-w/2}, \quad w > 0, -\infty < T < \infty$$

Now $f_T(t) = \int_0^\infty f(T, w) dw$

$$= \frac{1}{\sqrt{2\pi v} \cdot 2^{v/2}\Gamma(v/2)} \int_0^\infty e^{-\frac{1}{2}(\frac{T^2}{v} + 1)w} \cdot w^{v/2-1} dw$$

$$= \frac{1}{\sqrt{2\pi v} 2^{v/2} \Gamma(v/2)} \int_0^\infty \underbrace{w^{\frac{v+1}{2}-1} e^{-\frac{w}{\left(\frac{2v}{T^2+v}\right)}}}_{\text{kernel of GAM}\left(\frac{2v}{T^2+v}, \frac{v+1}{2}\right)} dw$$

$$= \frac{\left(\frac{2v}{T^2+v}\right)^{v/2+1} \Gamma\left(\frac{v+1}{2}\right)}{\sqrt{2\pi v} 2^{v/2} \Gamma(v/2)} \int_0^\infty \underbrace{\frac{w^{\frac{v+1}{2}-1}}{\left(\frac{2v}{T^2+v}\right)^{v/2+1} \Gamma\left(\frac{v+1}{2}\right)}}_1 e^{-\frac{w}{\left(\frac{2v}{T^2+v}\right)}} dw$$

$$\Rightarrow f_T(t) = \frac{\Gamma\left(\frac{v+1}{2}\right)}{\Gamma(v/2)} \frac{1}{\sqrt{2\pi v}} \left(\frac{2v+T^2}{T^2}\right)^{v/2}$$

$$= \frac{\Gamma\left(\frac{v+1}{2}\right)}{\Gamma(v/2)} \frac{1}{\sqrt{\pi v}} \left(1 + \frac{T^2}{v}\right)^{v/2}$$

$$\therefore f_T(t) = \frac{\Gamma\left(\frac{v+1}{2}\right)}{\Gamma(v/2)} \frac{1}{\sqrt{\pi v}} \left(1 + \frac{T^2}{v}\right)^{v/2}, \quad -\infty < t < \infty$$

$$\therefore T = \frac{\bar{x} - \mu}{s/\sqrt{n}} \sim t_v, \quad \text{where } v = n-1 \quad \square$$