

lec 13 MATH 240/640

$$Z \sim N(0,1), Y = Z^2 \sim \chi_1^2 := \frac{1}{\sqrt{2\pi}} y^{-\frac{1}{2}} e^{-\frac{1}{2}y} \mathbb{1}_{y \in (0, \infty)}$$

Investigate Hyper-Duplex...

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$$1 = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} y^{-\frac{1}{2}} e^{-\frac{1}{2}y} dy \quad \text{by integral rule}$$

$$\Rightarrow \sqrt{2\pi} = \int_0^{\infty} y^{\frac{1}{2}-1} e^{-\frac{1}{2}y} dy = \frac{\Gamma(\frac{1}{2})}{(\frac{1}{2})^{\frac{1}{2}}}$$

$$\Rightarrow \frac{(\frac{1}{2})^{\frac{1}{2}}}{\Gamma(\frac{1}{2})} = \frac{1}{\sqrt{2\pi}} \Rightarrow \Gamma(\frac{1}{2}) = \sqrt{\frac{1}{2}} \sqrt{2\pi} = \sqrt{\pi} \quad \text{nice value of gamma function}$$

$$\Rightarrow Y \sim \chi^2_1 = \frac{1}{\Gamma(\frac{1}{2})} (\frac{1}{2})^{\frac{1}{2}} y^{\frac{1}{2}-1} e^{-\frac{1}{2}y} \mathbb{1}_{y \in (0, \infty)} = \text{Gamma}(\frac{1}{2}, \frac{1}{2})$$

$$\underline{X \sim \text{Gamma}(\alpha, \beta) \quad \frac{1}{\Gamma(\alpha)} \beta^\alpha x^{\alpha-1} e^{-\beta x} \mathbb{1}_{x \in (0, \infty)}} \quad \uparrow$$

$$\text{let } Z_1, \dots, Z_k \stackrel{\text{iid}}{\sim} N(0, 1)$$

$$\text{let } Y = Z_1^2 + \dots + Z_k^2 \sim \chi^2_k \quad \text{"Chi-Squared with } k \text{ degrees of freedom"}$$

$$f_Y(y) = ?$$

$$Z_1^2, \dots, Z_k^2 \stackrel{\text{iid}}{\sim} \text{Gamma}(\frac{1}{2}, \frac{1}{2}) \quad \text{sum of gammas?}$$

let's use ch.f.s!

$$X \sim \text{gamma}(\alpha, \beta), \quad \phi_X(t) = E[e^{itX}] = \int_0^{\infty} e^{itx} \frac{1}{\Gamma(\alpha)} \beta^{\alpha} e^{-\beta x} x^{\alpha-1} dx$$

$$= \frac{\beta^{\alpha}}{\Gamma(\alpha)} \int_0^{\infty} x^{\alpha-1} e^{-(\beta-it)x} dx = \frac{\beta^{\alpha}}{\Gamma(\alpha)} \frac{\Gamma(\alpha)}{(\beta-it)^{\alpha}} = \left(\frac{\beta}{\beta-it}\right)^{\alpha}$$

$$X_1, \dots, X_n \stackrel{iid}{\sim} \text{gamma}(\alpha, \beta)$$

$$T = X_1 + \dots + X_n \sim f_T(t) = ?$$

$$\phi_T(t) = (\phi_X(t))^n = \left(\left(\frac{\beta}{\beta-it}\right)^{\alpha}\right)^n = \left(\frac{\beta}{\beta-it}\right)^{n\alpha} \Rightarrow T \sim \text{gamma}(n\alpha, \beta)$$

\Rightarrow sums of gammas are gammas. This is totally expected!

$$X_1, \dots, X_n \stackrel{iid}{\sim} \text{Erlang}(k, \lambda) \Rightarrow T = X_1 + \dots + X_n \sim \text{Erlang}(nk, \lambda)$$

Since each Erlang is the sum of k $\text{Exp}(\lambda)$'s. So this is the grand sum of nk $\text{Exp}(\lambda)$'s!

$$Z_1, \dots, Z_k \stackrel{iid}{\sim} N(0,1)$$

$$Y = Z_1^2 + \dots + Z_k^2 \sim \text{gamma}\left(\frac{k}{2}, \frac{1}{2}\right) = \chi_k^2 = \frac{1}{\Gamma(\frac{k}{2})} \left(\frac{1}{2}\right)^{\frac{k}{2}} x^{\frac{k}{2}-1} e^{-\frac{x}{2}} \mathbb{1}_{x \in (0, \infty)}$$

$$U \sim \chi_{k_1}^2, V \sim \chi_{k_2}^2 \quad U+V \sim \chi_{k_1+k_2}^2 \quad \text{just add} \quad \text{considering two gammas}$$

$$\text{let } X \sim \chi_k^2, Y = \sqrt{X} \sim ? \Rightarrow X = Y^2 = g^{-1}(Y) \text{ which is 1:1 on } \mathcal{X} = (0, \infty)$$

$$\left| \frac{d}{dy} [g^{-1}(y)] \right| = 2|y|$$

$$f_Y(y) = \frac{1}{\Gamma(\frac{k}{2})} \left(\frac{1}{2}\right)^{\frac{k}{2}} (y^2)^{\frac{k}{2}-1} e^{-\frac{y^2}{2}} \mathbb{1}_{\substack{y^2 \in (0, \infty) \\ y \in (0, \infty)}} \quad 2M$$

$$f_Y(y) = \frac{1}{\Gamma(\frac{k}{2})} \frac{1}{2^{\frac{k}{2}}} y^{\frac{k}{2}-1} e^{-\frac{y}{2}} \mathbb{I}_{y \in (0, \infty)}$$

$$= \frac{1}{2^{\frac{k}{2}-1}} \frac{1}{\Gamma(\frac{k}{2})} y^{\frac{k}{2}-1} e^{-\frac{y}{2}} \mathbb{I}_{y \in (0, \infty)} = \chi_k$$

The chi distr. with k degrees of freedom
 Not so many applications...

$X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2) \Rightarrow Z_i := \frac{X_i - \mu}{\sigma} \sim N(0, 1), Z_2 := \frac{X_2 - \mu}{\sigma} \sim N(0, 1)$

$Z_n := \frac{X_n - \mu}{\sigma} \sim N(0, 1)$

$\Rightarrow Z_1, \dots, Z_n \stackrel{iid}{\sim} N(0, 1)$

$\Rightarrow Z_1^2, \dots, Z_n^2 \stackrel{iid}{\sim} \chi_1^2$

$\Rightarrow \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1) \Rightarrow \left(\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \right)^2 = n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2 \sim \chi_1^2$

$\Rightarrow Z_1^2 + \dots + Z_n^2 \sim \chi_n^2$, let $\vec{Z} = \begin{bmatrix} Z_1 \\ \vdots \\ Z_n \end{bmatrix} \Rightarrow \vec{Z}^T \vec{Z} = Z_1^2 + \dots + Z_n^2 \sim \chi_n^2$

Recall $Y = aX + b, X \sim N(\mu, \sigma^2) \Rightarrow f_Y(y) = \frac{1}{|a|} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2} \left(\frac{y-b}{a} - \mu \right)^2} = \frac{1}{\sqrt{2\pi a^2 \sigma^2}} e^{-\frac{1}{2a^2 \sigma^2} (y - (b + a\mu))^2}$

$a > 0, b \in \mathbb{R}$

check work just as well for proving this $\Rightarrow Y \sim N(a\mu + b, a^2 \sigma^2)$

$\sum_{i=1}^n Z_i = Z_1 + \dots + Z_n \sim N(0, \sqrt{n^2})$

$\Rightarrow \bar{Z} = \frac{1}{n} (Z_1 + \dots + Z_n) \Rightarrow \bar{Z} = \frac{1}{n} \left(\frac{Z_1 + \dots + Z_n}{\sqrt{n}} \right) \Rightarrow \sqrt{n} \bar{Z} = \frac{Z_1 + \dots + Z_n}{\sqrt{n}} \Rightarrow \sqrt{n} \bar{Z} \sim N(0, 1)$

$\bar{Z} = \frac{\frac{X_1 - \mu}{\sigma} + \dots + \frac{X_n - \mu}{\sigma}}{n} = \frac{\sum X_i - n\mu}{n\sigma} = \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \Rightarrow \bar{Z}^2 = \left(\frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \right)^2 \Rightarrow n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2 \sim \chi_1^2$

$\sum (Z_i^2 - \bar{Z} + \bar{Z})^2 = \sum (Z_i^2 - \bar{Z})^2 + 2 \sum (Z_i - \bar{Z}) \bar{Z} + \sum \bar{Z}^2$

$= \sum (Z_i - \bar{Z})^2 + 2 \sum (Z_i \bar{Z} - \bar{Z}^2) + \sum \bar{Z}^2$

$= \sum (Z_i - \bar{Z})^2 + 2(n \bar{Z}^2 - n \bar{Z}^2) + n \bar{Z}^2$

$$= \sum \left(\frac{X_i - \mu}{\sigma} - \frac{\bar{X} - \mu}{\sigma} \right)^2 + n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2$$

$$= \sum \frac{(X_i - \bar{X})^2}{\sigma^2} + n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2$$

$$= \frac{1}{\sigma^2} \sum (X_i - \bar{X})^2 + n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2$$

$$= \frac{n-1}{\sigma^2} \frac{1}{n-1} \sum (X_i - \bar{X})^2 + n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2$$

$$= \frac{n-1}{\sigma^2} S_n^2 + \underbrace{n \left(\frac{\bar{X} - \mu}{\sigma} \right)^2}_{\sim \chi_1^2} \sim \chi_n^2$$

$$U \sim \chi_{n-1}^2, V \sim \chi_1^2 \Rightarrow U+V \sim \chi_n^2$$

Consequence: $\frac{n-1}{\sigma^2} S^2 \sim \chi_{n-1}^2$ and S^2 independent of \bar{X} .

This is true and published by Cochran in 1939 and the whole proof is called "Cochran's Theorem". Involves more linear algebra than you have so far in MATH 231. So I won't prove it. Wikipedia has a good proof. Sooner of them:
so $\vec{z} := \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$

If $z_1, z_2, \dots, z_n \stackrel{iid}{\sim} N(0,1)$, B_1, B_2, \dots, B_n are symmetric $n \times n$ matrices

where $B_1 + B_2 + \dots + B_n = I_n$ then the following 3 statements are equivalent:

$$\text{rank}[B_1] + \dots + \text{rank}[B_n] = n \iff \text{all } \vec{z}^T B_i \vec{z} \text{ 's are independent} \iff \vec{z}^T B_i \vec{z} \sim \chi_{\text{rank}[B_i]}^2$$

let's see some examples... let $n=3$, $z_1, z_2, z_3 \stackrel{\text{iid}}{\sim} N(0,1)$ (6)

$$b_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad b_2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow b_1 + b_2 = I_3, \text{ both } b_1, b_2 \text{ symmetric}$$

\Rightarrow we can use Cochran's thm!

$$\Rightarrow \vec{z}^T b_1 \vec{z} \sim \chi^2_1 \text{ let's see}$$

$$[z_1, z_2, z_3] \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = [z_1, z_2, z_3] \begin{bmatrix} z_1 \\ 0 \\ 0 \end{bmatrix} = z_1^2 \sim \chi^2_1$$

$$\Rightarrow \vec{z}^T b_2 \vec{z} \sim \chi^2_2 \text{ let's see}$$

$$[z_1, z_2, z_3] \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = [z_1, z_2, z_3] \begin{bmatrix} 0 \\ z_2 \\ z_3 \end{bmatrix} = z_2^2 + z_3^2 \sim \chi^2_2$$

$\Rightarrow \vec{z}^T b_1 \vec{z}$ is independent of $\vec{z}^T b_2 \vec{z}$. let's see

z_1^2 is definitely independent of $z_2^2 + z_3^2$

why do we care??

$$\text{Recall } \vec{z}^T \vec{z} = \sum (z_i - \bar{z} + \bar{z})^2 = \sum (z_i - \bar{z})^2 + n \bar{z}^2$$

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

$$\text{If we can show } \sum (z_i - \bar{z})^2 = \vec{z}^T b_1 \vec{z} \text{ and } n \bar{z}^2 = \vec{z}^T b_2 \vec{z}$$

where b_1, b_2 symmetric, $I = b_1 + b_2$, $\text{rank}(b_1) + \text{rank}(b_2) = n$, then we

know $\frac{n-1}{\sigma^2} S_n^2 \sim \chi^2_{\text{rank}(b_1)}$ and S_n^2 independent of \bar{X} .

let's do it. First, b_2 ...

$$\text{let } J_n := \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \text{ --- } \text{has all ones}$$

$$n \bar{z}^2 = n \left(\frac{z_1 + \dots + z_n}{n} \right) \left(\frac{z_1 + \dots + z_n}{n} \right) = \frac{1}{n} (\vec{z}^T \vec{1}_n) (\vec{1}_n^T \vec{z}) = \vec{z}^T \left(\frac{1}{n} \vec{1}_n \vec{1}_n^T \right) \vec{z} = \vec{z}^T \left(\frac{1}{n} J_n \right) \vec{z}$$

$$b_2 = \begin{bmatrix} \frac{1}{n} & \frac{1}{n} & \dots & \frac{1}{n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{1}{n} & \dots & \dots & \frac{1}{n} \end{bmatrix}$$

symmetric? Yes!

$\text{rank}(b_2) = 1$ only 1 linearly dep. column!

Let's do B_1 now...

$$\sum (z_i - \bar{z})^2 = \sum z_i^2 - 2z_i \bar{z} + \bar{z}^2 = \sum z_i^2 - 2n\bar{z}^2 + \bar{z}^2 = \sum z_i^2 - n\bar{z}^2$$

$$= \bar{z}^T \bar{z} - \bar{z}^T \left(\frac{1}{n} J_n \right) \bar{z}$$

$$= \bar{z}^T I_n \bar{z} - \bar{z}^T \left(\frac{1}{n} J_n \right) \bar{z}$$

$$= \bar{z}^T \left(I_n - \frac{1}{n} J_n \right) \bar{z}$$

$$B_1 = I_n - B_2$$

$$B_1 = \begin{bmatrix} 1 - \frac{1}{n} & -\frac{1}{n} & \dots & -\frac{1}{n} \\ -\frac{1}{n} & 1 - \frac{1}{n} & \dots & -\frac{1}{n} \\ \vdots & \vdots & \ddots & \vdots \\ -\frac{1}{n} & -\frac{1}{n} & \dots & 1 - \frac{1}{n} \end{bmatrix}$$

$$\text{Obviously } B_1 + B_2 = I_n - B_2 + B_2 = I$$

How do we prove the rank $[B_1] = n-1$? Not so clear from just looking at it.

Let's first note the following:

$$\begin{aligned} B_1 B_1 &= \left(I_n - \frac{1}{n} J_n \right) \left(I_n - \frac{1}{n} J_n \right) = I_n I_n - \frac{1}{n} I_n J_n - \frac{1}{n} J_n I_n + \frac{1}{n^2} J_n J_n \\ &= I_n - \frac{2}{n} J_n + \frac{1}{n^2} n J_n = I_n - \frac{2}{n} J_n + \frac{1}{n} J_n = I_n - \frac{1}{n} J_n = B_1 \end{aligned}$$

B_1 is called "idempotent".

Then from 231: if A is symmetric and idempotent, the $\text{rank}(A) = \text{tr}(A) = \sum_{i=1}^n \lambda_i$

$$\Rightarrow \text{rank}[B_1] = \text{tr}[B_1] = \left(1 - \frac{1}{n}\right) + \left(1 - \frac{1}{n}\right) + \dots + \left(1 - \frac{1}{n}\right) = n \left(1 - \frac{1}{n}\right) = n-1$$

(Cochran's) Then

$$\Rightarrow \frac{n-1}{\sigma^2} S^2 \sim \chi_{n-1}^2 \quad \text{and}$$

S^2 independent of \bar{X}_n (1930) Fisher proved it, Geary proved some stuff only after also has this feature

This is a cool result. We need one more

core concept to derive the T, F distributions

$$\frac{\bar{X}_n - \mu}{\frac{S}{\sqrt{n}}} = \frac{\bar{X}_n - \mu}{\frac{1}{\sqrt{n}} \sqrt{\frac{\sigma^2}{n-1} S^2}} = \frac{\frac{\bar{X}_n - \mu}{\frac{\sigma}{\sqrt{n}}}}{\sqrt{\frac{\frac{n-1}{\sigma^2} S^2}{n-1}}} \sim \chi_{n-1}^2$$

We need to derive PDFs of ratios of rv's!

$$X \sim \text{Gamma}(\alpha, \beta)$$

$$Y = aX \sim \frac{1}{a} \frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{x}{a}\right)^{\alpha-1} e^{-\beta \left(\frac{x}{a}\right)} \underbrace{\mathbb{1}_{\frac{x}{a} \in (0, \infty)}}_{x \in (0, \infty)} = \frac{\left(\frac{\beta}{a}\right)^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\left(\frac{\beta}{a}\right)x} \mathbb{1}_{x \in (0, \infty)}$$

where $a > 0$

$$\Rightarrow Y \sim \text{Gamma}\left(\alpha, \frac{\beta}{a}\right)$$