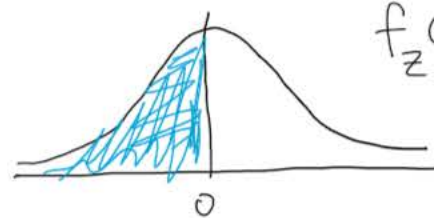


Lecture 18

$Z \sim N(0,1)$, $Y = Z^2 \sim f_Y(y) = ?$ Not 1:1



$$F_Y(y) = P(Y \leq y) = P(Z^2 \leq y) = P(Z \in [-\sqrt{y}, \sqrt{y}]) = 2P(Z \in [0, \sqrt{y}])$$

$$= 2(F_Z(\sqrt{y}) - F_Z(0)) = 2F_Z(y) - 1$$

$$f_Y(y) = \frac{d}{dy} [2F_Z(\sqrt{y}) - 1] = 2 \left(\frac{1}{2} y^{-1/2} \right) f_Z(\sqrt{y}) = y^{-1/2} \frac{1}{\sqrt{2\pi}} e^{-\frac{(\sqrt{y})^2}{2}} \mathbb{1}_{\sqrt{y} \in \mathbb{R}} \mathbb{1}_{y \geq 0}$$

$$\propto y^{\frac{1}{2}-1} e^{-\frac{1}{2}y} \mathbb{1}_{y \geq 0} \propto \text{Gamma}(\frac{1}{2}, \frac{1}{2})$$

$Z_1, \dots, Z_k \stackrel{\text{iid}}{\sim} N(0,1)$, $Y = Z_1^2 + \dots + Z_k^2 \sim \text{Gamma}(\frac{k}{2}, \frac{1}{2})$

Note the beta is always $\frac{1}{2}$ and the alpha is always $k/2$ so k is the only parameter. And because this is a common situation, we give it a special name:

$$\text{Gamma}(\frac{k}{2}, \frac{1}{2}) = \chi_k^2$$

the "chi squared dist. with k degrees of freedom" ($k \in \mathbb{N}$).

$$E[Y] = k \underbrace{E[Z^2]}_1 = k$$

PDF $\hookrightarrow \chi_k^2 = \frac{(\frac{1}{2})^{k/2}}{\Gamma(\frac{k}{2})} y^{\frac{k}{2}-1} e^{-\frac{y}{2}} \mathbb{1}_{y \geq 0}$

$k=1, \Gamma(\frac{1}{2}) = \sqrt{\pi}$
 \downarrow
 $= \frac{1}{\sqrt{2\pi}} y^{-1/2} e^{-\frac{y}{2}} \mathbb{1}_{y \geq 0} = \chi_1^2$

$$X \sim \chi_k^2, Y = \sqrt{X} \Rightarrow x = y^2 = g^{-1}(y), \quad \left| \frac{d}{dy} [g^{-1}(y)] \right| = |2y| = 2y$$

$$f_Y(y) = f_X(y^2) \cdot 2y = \frac{(\frac{1}{2})^{k/2}}{\Gamma(\frac{k}{2})} y^{k-2} e^{-\frac{y^2}{2}} \cdot 2y \mathbb{1}_{\underbrace{y^2 \geq 0}_{y \geq 0}}$$

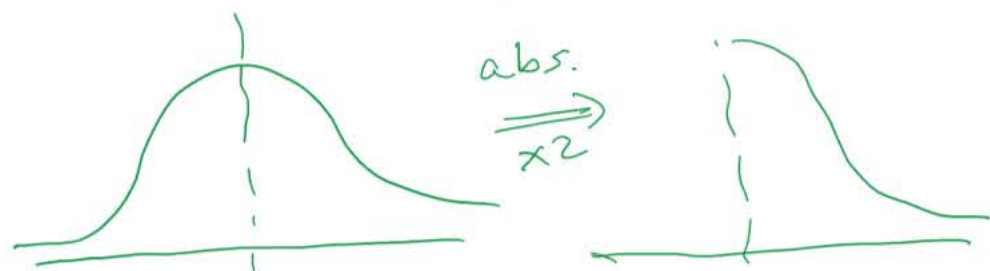
$$= \frac{(\frac{1}{2})^{k/2-1}}{\Gamma(\frac{k}{2})} y^{k-1} e^{-\frac{y^2}{2}} \mathbb{1}_{y \geq 0} = \chi_k$$

The chi distribution with k degrees of freedom

$$= \frac{\Gamma(\frac{k}{2}-1)}{\Gamma(\frac{k}{2})} y^{k-1} e^{-y/2} \mathbb{1}_{y \geq 0} = \chi_k \quad \text{The chi distribution with } k \text{ degrees of freedom.}$$

$y^2 \geq 0$
 $y \geq 0$

Note: $Z \sim N(0,1)$, $|Z| = \sqrt{Z^2} \sim \chi_1 = \sqrt{\frac{2}{\pi}} e^{-y/2} \mathbb{1}_{y \geq 0}$



$$= z \left(\underbrace{\frac{1}{\sqrt{2\pi}} e^{-y/2}}_{f_Z} \right) \mathbb{1}_{y \geq 0}$$

Let's prove something about the Gamma:

$$X \sim \text{Gamma}(\alpha, \beta), Y = cX \text{ where } c > 0$$

$$\begin{aligned} f_Y(y) &= \frac{1}{c} f_X\left(\frac{y}{c}\right) = \frac{1}{c} \frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{y}{c}\right)^{\alpha-1} e^{-\beta y/c} \mathbb{1}_{\frac{y}{c} > 0 \Rightarrow y \geq 0} \\ &= \frac{(\beta/c)^\alpha}{\Gamma(\alpha)} y^{\alpha-1} e^{-(\beta/c)y} \mathbb{1}_{y \geq 0} = \text{Gamma}(\alpha, \beta/c) \end{aligned}$$

$$X \sim \chi_k^2, Y = \frac{X}{k} \sim \text{Gamma}\left(\frac{k}{2}, \frac{1/2}{1/k}\right) = \text{Gamma}\left(\frac{k}{2}, \frac{k}{2}\right)$$

Let $X_1 \sim \chi_{k_1}^2$ indep. of $X_2 \sim \chi_{k_2}^2$

$$\text{Let } U = \frac{X_1}{k_1} \sim \text{Gamma}\left(\underbrace{\frac{k_1}{2}}_a, \underbrace{\frac{k_1}{2}}_a\right) \text{ indep of } V = \frac{X_2}{k_2} \sim \text{Gamma}\left(\underbrace{\frac{k_2}{2}}_b, \underbrace{\frac{k_2}{2}}_b\right)$$

$$R = \frac{U}{V} \sim f_R(r) = \int_{\text{supp}[V]} f_U(rt) \mathbb{1}_{rt \in \text{supp}[U]} f_V(t) t dt$$

formula derived previously - check notes.

Now, plug + chng:

$$= \int_0^\infty \frac{a^a}{\Gamma(a)} (rt)^{a-1} e^{-art} \mathbb{1}_{\substack{rt \in [0, \infty) \\ r \in (0, \infty)}} \frac{b^b}{\Gamma(b)} t^{b-1} e^{-bt} t dt$$

↳ The reason why we're doing this is because we want to

the normal family.

↳ The reason why we're doing this is because we want to derive all the distributions in the normal family.

$$= \frac{a^a}{\Gamma(a)} \frac{b^b}{\Gamma(b)} r^{a-1} \mathbb{1}_{r \geq 0} \int_0^b t^{a+b-1} e^{-(ar+b)t} dt$$

$$= a^a b^b r^{a-1} \mathbb{1}_{r \geq 0} \underbrace{\frac{1}{\Gamma(a)\Gamma(b)} \cdot \frac{1}{(ar+b)^{a+b}}}_{\frac{1}{B(a,b)}} = \frac{a^a b^b}{B(a,b)} r^{a-1} \underbrace{(ar+b)^{-(a+b)}}_{b^{-(a+b)} (1 + \frac{a}{b}r)^{-(a+b)}} \mathbb{1}_{r \geq 0}$$

$$= \frac{(a/b)^a}{B(a,b)} r^{a-1} (1 + \frac{a}{b}r)^{-(a+b)} \mathbb{1}_{r \geq 0} = \frac{(k_1/k_2)^{k_1/2}}{B(\frac{k_1}{2}, \frac{k_2}{2})} r^{\frac{k_1}{2}-1} (1 + \frac{k_1}{k_2}r)^{-\frac{k_1+k_2}{2}} \mathbb{1}_{r \geq 0}$$

$= F_{k_1, k_2}$ this is the "F distribution" or Fisher-Snedecor dist. with k_1 numerator degrees of freedom and k_2 denominator degrees of freedom. $k_1, k_2 \in \mathbb{N}$.

Let $Z \sim N(0,1)$, $X \sim \chi_k^2$, $W = \frac{Z}{\sqrt{X/k}} \sim f_W(w) = ?$

Consider $W^2 = \frac{Z^2/1}{X/k} \sim F_{1,k}$

↳ symmetric around 0

$$F_{W^2}(w^2) = P(W^2 \leq w^2) = P(W \in [-w, w]) = F_W(w) - F_W(-w)$$

Take derivatives:

$$\frac{d}{dw} [F_{W^2}(w^2)] = \frac{d}{dw} [F_W(w)] - \frac{d}{dw} [F_W(-w)]$$

$$2w f_{W^2}(w^2) = f_W(w) + f_W(-w) \stackrel{?}{=} 2 f_W(w)$$

$$\Rightarrow f_W(w) = w \frac{(\frac{1}{k})^{k/2}}{B(\frac{1}{2}, \frac{k}{2})} \frac{(w^2)^{\frac{k}{2}-1}}{(1 + \frac{w^2}{k})^{-\frac{1+k}{2}}} \mathbb{1}_{w^2 \geq 0}$$

plug into F dist. $\frac{\Gamma(\frac{k+1}{2})}{\Gamma(\frac{1}{2})\Gamma(\frac{k}{2})}$

$z \in \mathbb{R}$ divided by nonneg. is any \mathbb{R}

$$= \frac{\Gamma(\frac{k+1}{2})}{\sqrt{k\pi} \Gamma(\frac{k}{2})} (1 + \frac{w^2}{k})^{-\frac{k+1}{2}} = T_k \text{ Student's T dist. with } k \text{ degrees of freedom.}$$

If $k \rightarrow \infty$ $T_k \rightarrow \mathcal{Z}$

Student's T dist. has the $N(0,1)$ shape but with thicker tails

$$= \frac{\Gamma(\frac{k+1}{2})}{\sqrt{k\pi} \Gamma(\frac{k}{2})} \left(1 + \frac{u^2}{k}\right)^{-\frac{k+1}{2}} = T_k \quad \text{Student's T dist. with } k \text{ degrees of freedom.}$$

$$\text{If } k \rightarrow \infty \quad T_k \rightarrow Z$$

Student's T dist. has the $N(0,1)$ shape but just thicker tails.

$$\begin{aligned} Z_1, Z_2 &\stackrel{iid}{\sim} N(0,1), \quad R = \frac{Z_1}{Z_2} \sim \int_{\mathbb{R}} f(ru) f(u) |u| du = \\ &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{r^2 u^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} |u| du = \frac{1}{2\pi} \left(\int_{-\infty}^0 e^{-\frac{r^2+1}{2} u^2} -|u| du + \int_0^{\infty} e^{-\frac{r^2+1}{2} u^2} |u| du \right) \\ &= \frac{1}{2\pi} \left(\int_0^{\infty} e^{-\frac{r^2+1}{2} u^2} |u| du + \int_0^{\infty} e^{-\frac{r^2+1}{2} u^2} |u| du \right) = \frac{1}{\pi} \int_0^{\infty} e^{-\frac{r^2+1}{2} u^2} u du \end{aligned}$$

$$\text{let } t = u^2 \Rightarrow \frac{dt}{du} = 2u \Rightarrow du = \frac{1}{2u} dt, \quad u=0 \rightarrow t=0, \quad u \rightarrow \infty \Rightarrow t \rightarrow \infty$$

$$= \frac{1}{\pi} \int_0^{\infty} e^{-\frac{r^2+1}{2} t} \frac{1}{2\sqrt{t}} dt = \frac{1}{2\pi} \cdot \frac{1}{\frac{r^2+1}{2}} \int_0^{\infty} \frac{r^2+1}{2} e^{-\frac{r^2+1}{2} t} dt$$

$$\int_0^{\infty} \lambda e^{-\lambda x} dx = 1$$

PDF of exponential r.v.

$$= \frac{1}{\pi} \frac{1}{1+r^2} = \text{Cauchy}(0,1).$$

$$\text{let } X = c + \sigma R, \quad R \sim \text{Cauchy}(0,1), \quad \sigma > 0$$

$$X \sim \text{Cauchy}(c, \sigma) = \frac{1}{\sigma\pi} \cdot \frac{1}{1 + \left(\frac{x-c}{\sigma}\right)^2}$$

Lecture 19

$$X \sim \text{Cauchy}(0,1) = \frac{1}{\pi} \cdot \frac{1}{x^2+1}$$

$$E[X] = \int_{\mathbb{R}} x \frac{1}{\pi} \frac{1}{x^2+1} dx = \infty \rightarrow \text{the expectation does not exist}$$

$$(M_{X^f}) \quad M_X(t) = E[e^{tx}] = \int_{\mathbb{R}} e^{tx} \frac{1}{\pi} \frac{1}{x^2+1} dx = \infty \rightarrow \text{D.N.E.}$$

$$\phi(t) = \int_{\mathbb{R}} e^{itx} \frac{1}{\pi} \frac{1}{x^2+1} dx = \dots = e^{-|t|}, \quad \phi'_X(t) = \frac{-t}{|t|} e^{-|t|}$$