

★ PDF of Gamma

$$g(x) = cX$$
$$g^{-1}(y) = \frac{y}{c}$$

$$\therefore \frac{\left(\frac{B}{c}\right)^d}{\Gamma(d)} y^{\alpha-1} e^{-\left(\frac{B}{c}y\right)} \perp y \geq 0$$

$$\frac{da}{dy} = \frac{1}{c}$$

$$f \circ (g^{-1}(y)) \frac{1}{c} =$$

$$f_Y(y) = f_X\left(\frac{y}{c}\right) \cdot \frac{1}{c}$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)} \left(\frac{y}{c}\right)^{\alpha-1} e^{-\beta\left(\frac{y}{c}\right)} \mathbb{1}_{\frac{y}{c} > 0} \cdot \frac{1}{c}$$

$$= \frac{1}{c^{\alpha-1}} \cdot \frac{1}{c} \cdot \frac{\beta^\alpha}{\Gamma(\alpha)} y^{\alpha-1} e^{-\frac{\beta}{c} y} \mathbb{1}_{y \geq 0}$$

$$= \frac{\left(\frac{B}{c}\right)^d}{\Gamma(d)} y^{d-1} e^{-\frac{B}{c}y} \mathbb{I}_{y \geq 0}$$

• CLT  $X_1 \dots X_n \stackrel{iid}{\sim}$  with  $\mu, \sigma^2$

$$\Rightarrow \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} \xrightarrow{d} z \sim N(0, 1) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$$

$$X = \sigma Z + \mu \sim N(\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$

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$$Z_i^2 \sim X_i^2 := \text{Gamma}(\frac{1}{2}, \frac{1}{2})$$

$$\underbrace{z_1^2 + \dots + z_m^2}_{X_n} + \underbrace{z_{m+1}^2 + \dots + z_n^2}_{X_{n-m}} \sim X_n^2 = \text{Gamma}(\frac{n}{2}, \frac{1}{2})$$

- $\chi \sim \mathcal{K} k^2$

$$\Rightarrow \frac{x}{k} \sim \text{Gamma}\left(\frac{k}{2}, \frac{k}{2}\right)$$

$$X_1 \sim \chi_{k_1}, \dots, X_n \sim \chi_{k_n}$$

$$R = \frac{X_1/k_1}{X_2/k_2} \sim f_R(r) = ?$$

⇒ so it becomes  $\frac{U}{V} \sim \int_{\text{supp}(U)} f_U(r) f_V(t) \mathbb{1}_{|t|dt}$   
 $\int_{\text{supp}(U)}$   $\int_{\text{supp}(V)}$

Let  $U = \frac{X_1}{k_1} \sim \text{Gamma}(\frac{k_1}{2}, \frac{k_1}{2})$   
 (call a) (call b)

Let  $V = \frac{X_2}{k_2} \sim \text{Gamma}(\frac{k_2}{2}, \frac{k_2}{2})$

$$= \int_0^\infty \left( \frac{a^a}{\Gamma(a)} (rt)^{a-1} e^{-art} \right) \left( \frac{b^b}{\Gamma(b)} t^{b-1} e^{-bt} \right) dt$$

$$= \frac{a^a b^b}{\Gamma(a)\Gamma(b)} r^{a-1} \int_0^\infty t^{(a+b)-1} e^{-(a+b)t} dt$$

$$\frac{\Gamma(a+b)}{(b+ar)^{a+b}}$$

$$= \frac{a^a b^b}{B(a,b)} r^{a-1} (b+ar)^{-(a+b)}$$

$$\underbrace{(b(1+\frac{a}{b}r))}^{-(a+b)}$$

$$= \frac{a^a b^b}{B(a,b)} r^{a-1} \frac{(1+\frac{a}{b}r)^{-(a+b)}}{b^a b^b}$$

$$= \frac{(\frac{a}{b})^a}{B(a,b)} r^{a-1} (1+\frac{a}{b}r)^{-(a+b)}$$

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

$$= F_{k_1, k_2}$$

F-distribution with  $k_1$  &  $k_2$  degree of freedom

Not:  $k_1, k_2 \in \mathbb{N}$

Name  
 || Fisher -  
 second order... ||

•  $Z \sim N(0,1)$  ind. of  $X \sim \chi_k^2$

$$W = \frac{Z}{\sqrt{X/k}} \sim f_W(w) = ?$$

→ symmetric

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

check

Note:  $f_W(w) = f_W(-w)$

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

take d/dw both sides

$$\Rightarrow 2w f_W(w^2) = f_W(w) - f_W(-w) \quad \text{chain rule}$$

$$= 2f_W(w)$$

$$\Rightarrow f_W(w) = w f_W(w^2)$$

Note that  $k_1 = 1$  &  $k_2 = k$ .

Now, plug this in the formula that we just derived:

$$= w \frac{\left(\frac{1}{k}\right)^{\frac{1}{2}}}{B\left(\frac{1}{2}, \frac{k}{2}\right)} (w^2)^{\frac{1}{2}-1} \left(1 + \frac{1}{k} w^2\right)^{-\left(\frac{1+k}{2}\right)}$$

Note that  $\frac{1}{B\left(\frac{1}{2}, \frac{k}{2}\right)} = \frac{\Gamma\left(\frac{k+1}{2}\right)}{\Gamma\left(\frac{1}{2}\right)\Gamma\left(\frac{k}{2}\right)}$

$$(w^2)^{\frac{1}{2}-1} = w^{2-\frac{1}{2}} = w^{-\frac{1}{2}}$$

$$\frac{1}{\sqrt{\pi}}$$

$$\text{so, } w \cdot w^{-\frac{1}{2}} \frac{\Gamma\left(\frac{k+1}{2}\right)}{\sqrt{k\pi} \Gamma\left(\frac{k}{2}\right)} \left(1 + \frac{w^2}{k}\right)^{-\frac{k+1}{2}} = T_k$$

student's t distribution with k degree of freedom

$Z_1$  &  $Z_2 \stackrel{i.i.d.}{\sim} N(0, 1)$

$$R = \frac{Z_1}{Z_2} \sim \text{Cauchy}(0, 1) = \frac{1}{\pi} \cdot \frac{1}{r^2 + 1}$$

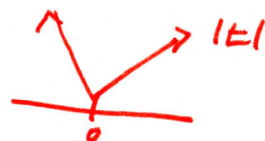
$$Y = c + \sigma R \quad \text{where } \sigma > 0$$

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

Note:  $T_1 = \frac{\Gamma\left(\frac{1+1}{2}\right)}{\sqrt{\pi} \Gamma\left(\frac{1}{2}\right)} \left(1 + \frac{w^2}{1}\right)^{-\frac{1+1}{2}} = \frac{1}{\pi} (1+w^2)^{-1} = \frac{1}{\pi} \cdot \frac{1}{1+w^2} = \text{Cauchy}(0, 1)$

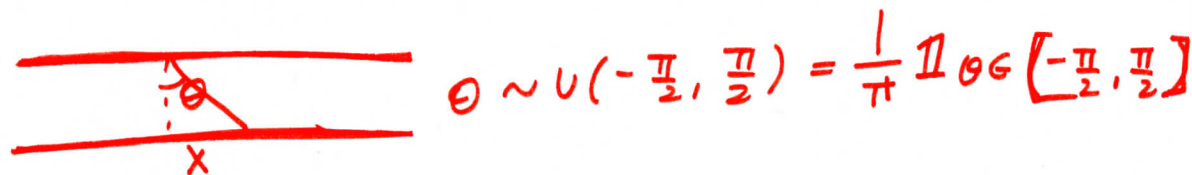
$$\phi_R(t) = E[e^{itR}] = \frac{1}{\pi} \int_{\mathbb{R}} \frac{e^{itr}}{r^2 + 1} dr = \dots = e^{-|t|}$$

Complex analysis



$$\phi_R(t) = \begin{cases} e^{-t} & \text{if } t > 0 \\ e^{-t} & \text{if } t < 0 \\ \text{undefined} & t = 0 \end{cases}$$

What's the dist. of light brightness of the floor?



$$\theta \sim U(-\frac{\pi}{2}, \frac{\pi}{2}) = \frac{1}{\pi} \mathbb{1}_{\theta \in [-\frac{\pi}{2}, \frac{\pi}{2}]}$$

$$f_X(x) = f_{\theta}(g^{-1}(x)) \frac{d}{dx}[g^{-1}(x)]$$

$$g(\theta) = \tan(\theta) = \frac{x}{1} = x \Rightarrow \theta = \tan^{-1}(x) = g^{-1}(x)$$

$$\text{So, } f_X(x) = \frac{1}{\pi} \mathbb{1}_{\underbrace{\arctan(x) \in [-\frac{\pi}{2}, \frac{\pi}{2}]}_{x \in [\tan(-\frac{\pi}{2}), \tan(\frac{\pi}{2})]}} \cdot \frac{1}{1+x^2} = \frac{1}{\pi} \cdot \frac{1}{1+x^2} = \text{Cauchy}(0,1)$$

• Application: of  $Z, T, X^2, F$  to statistics:

$$\text{Let } X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$$

$$T = X_1 + \dots + X_n \sim N(n\mu, n\sigma^2)$$

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

the estimator for  $\mu$

$$\bar{X} = \frac{1}{n} \sum X_i$$

estimate for  $\mu$

$$S_n^2 = \frac{1}{n-1} \sum (X_i - \bar{X}_n)^2$$

$$S^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2$$

estimator for  $\sigma^2$

estimate for  $\sigma$

• Want to know ①  $S_n^2 \sim ?$

② Relationship b/w  $\bar{X}_n$  &  $S_n^2$



$$\bullet c_1 \dots c_n \sim N(0, 1)$$

$$\vec{z}^T \vec{z} = \sum z_i^2 \sim \chi_n^2 \quad \text{where } \vec{z} = \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$$

$$\text{Note: } \bar{z}_i = \frac{x_i - \mu}{\sigma}$$

$$\Rightarrow \sum \left( \frac{x_i - \mu}{\sigma} \right)^2 \sim \chi_n^2$$

$$= \frac{1}{\sigma^2} \sum (x_i - \mu)^2$$

$$\text{Note: } x_i - \mu = x_i - \bar{x} + \bar{x} - \mu$$

$$(x_i - \mu)^2 = ((x_i - \bar{x}) + (\bar{x} - \mu))^2$$

$$= (x_i - \bar{x})^2 + 2(x_i - \bar{x})(\bar{x} - \mu) + (\bar{x} - \mu)^2$$

$$= \frac{1}{\sigma^2} \left( \sum (x_i - \bar{x})^2 + 2 \underbrace{\sum (x_i - \bar{x})(\bar{x} - \mu)}_{\sum x_i \bar{x} - \bar{x}^2 - x_i \mu + \bar{x} \mu} + (\bar{x} - \mu)^2 \right)$$

Q

$$\underbrace{\sum x_i \bar{x} - \bar{x}^2 - x_i \mu + \bar{x} \mu}_{n \bar{x}}$$

$$= n \bar{x}^2 - n \bar{x}^2 - \mu n \bar{x} + n \bar{x} \mu$$

$$= 0$$

$$= \frac{\sum (x_i - \bar{x})^2}{\sigma^2} + \frac{n(\bar{x} - \mu)^2}{\sigma^2} \sim \chi_n^2$$

$$\underbrace{\frac{(n-1)S^2}{\sigma^2}}$$

$$\underbrace{\left( \frac{\sqrt{n}(\bar{x} - \mu)}{\sigma} \right)^2}$$

$$\underbrace{\left( \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \right)^2}_{Z^2 \sim \chi_1^2}$$

$$\text{Recall } \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}} \sim N(0, 1)$$

$$\text{b/c } \bar{x} \sim N(\mu, \frac{\sigma^2}{n})$$

$$\bullet \text{ Conjecture: } \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$$