

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx \quad : \text{generalized factorial}$$

$$\Gamma(\alpha) = (\alpha-1)\Gamma(\alpha-1) \quad : \text{Integration by parts}$$



$$\alpha : \text{positive integer} \Rightarrow \Gamma(\alpha) = (\alpha-1)!$$

Need a computer to evaluate  $\Gamma(\alpha)$ .

Except

$$\Gamma\left(\frac{1}{2}\right) = \int_0^{\infty} x^{-\frac{1}{2}} e^{-x} dx = \dots = \int_0^{\infty} \frac{1}{\sqrt{z}} e^{-\frac{1}{2}z^2} dz = \frac{1}{\sqrt{2}} \cdot \sqrt{2\pi} = \sqrt{\pi}$$

$\frac{1}{2}z^2 = x$       part of  $N(0,1)$

$$\Gamma\left(\frac{5}{2}\right) = \left(\frac{3}{2}\right)\Gamma\left(\frac{3}{2}\right) = \frac{3}{2} \cdot \frac{1}{2} \cdot \Gamma\left(\frac{1}{2}\right) = \dots$$

$$\int_0^{\infty} \frac{1}{\beta^{\alpha} \Gamma(\alpha)} \cdot y^{\alpha-1} e^{-y/\beta} dy = 1$$

↑  
p/s = x, use  $\Gamma(\alpha)$  definition, etc. ....

later.

$$\begin{aligned} Y &\sim \text{Gamma}(\alpha_1, \beta) \\ X &\sim \text{Gamma}(\alpha_2, \beta) \end{aligned} \Bigg) \text{indep} \Rightarrow X+Y \sim \text{Gamma}(\alpha_1+\alpha_2, \beta)$$

# $\Gamma$ $\gamma$ Gamma Probability Distribution

$$\text{Gamma}(\alpha, \lambda) = \text{Gamma}(\alpha, \frac{1}{\beta})$$

Shape      rate

$Y \sim \text{Gamma}(\alpha, \beta)$ : gamma r.v. if

$$f(y) = \frac{y^{\alpha-1} e^{-y/\beta}}{\beta^\alpha \Gamma(\alpha)}, \quad 0 \leq y < \infty$$

$$\int_0^\infty y^{\alpha-1} e^{-y/\beta} dy$$

for  $\alpha > 0$ ,  $\beta > 0$ , and where the gamma function is

$$\Gamma(\alpha) = \int_0^\infty y^{\alpha-1} e^{-y} dy.$$

•  $\mu = E(Y) = \alpha\beta$  and  $\sigma^2 = V(Y) = \alpha\beta^2$ .

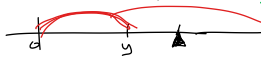
$$\int_0^\infty y \cdot \frac{1}{\beta^\alpha \Gamma(\alpha)} y^{\alpha-1} e^{-y/\beta} dy$$

use  $\begin{cases} \frac{\Gamma(\alpha+1)}{\Gamma(\alpha)} = \alpha \\ \frac{\Gamma(\alpha+2)}{\Gamma(\alpha)} = \alpha(\alpha+1) \end{cases}$

"Go together"

Recall: Poisson ( $\lambda$ ) :  $\lambda$  : average no. occurrences  
 in a given time frame.  
 Poisson rate in 1 unit

Ex 4.168

$Y$ : length of time until the first arrival (bet'n two successive events)  
 # of Poisson events  $\sim \text{Poisson}(\lambda y)$   
 ( $y$  min)

$$P(Y > y) = P(\text{no arrival in the time } \underbrace{[0, y]}_y)$$

$$= \frac{(\lambda y)^0 e^{-\lambda y}}{0!} = e^{-\lambda y}$$

CDF:  $F(y) = 1 - e^{-\lambda y}$

$f(y) = \lambda e^{-\lambda y} : \text{Exp}(\frac{1}{\lambda})$

Ex 4.169

• Calls come at rate  $10/\text{hr}$   
 •  $P(\text{more than 15 minutes elapsed bet'n two calls})$

$Y$  : length of time (hr) bet'n two calls.

$\sim \text{Exp}(\frac{1}{10})$

$$P(Y > .25 \text{ hr}) = \int_{.25}^{\infty} 10 e^{-10y} dy = .082$$

Ex 170

Generalize to Gamma

a.  $Y$ : waiting time until  $2^{\text{nd}}$  arrival  $k$

$$P(Y > y) = P(\text{none or one arrival in } [0, y]) \\ = \frac{(\lambda y)^0 e^{-\lambda y}}{0!} + \frac{(\lambda y)^1 e^{-\lambda y}}{1!} = (\lambda y + 1) e^{-\lambda y}$$

$$F(y) = \int_0^y f(t) dt$$

$$f(y) = \lambda^2 y e^{-\lambda y} \quad ; \text{ Gamma}(2, \frac{1}{\lambda})$$

$k$

MGF

$(Y_1 \text{ has } m_1(t), Y_2 \text{ has } m_2(t)) \text{ indep}$

Then  $Y_1 + Y_2$  has mgf  $m_1(t)m_2(t)$ .

$$E(e^{(Y_1 + Y_2)t}) \underset{\substack{\uparrow \\ \text{indep}}}{=} E(e^{tY_1}) E(e^{tY_2}) = m_1(t) \cdot m_2(t)$$

MGF of Gamma  $(\alpha, \beta)$

$$\int e^{ty} \cdot \frac{1}{\Gamma(\alpha) \beta^\alpha} y^{\alpha-1} e^{-y/\beta} dy =$$

$y(t - \frac{1}{\beta})$   
 $(t - \frac{1}{\beta})^{-1} = \beta^*$

$$= \frac{1}{\beta^\alpha} \left[ \underbrace{\frac{1}{\beta^{*\alpha} \Gamma(\alpha)} \cdot y^{\alpha-1} e^{-y/\beta^*}}_{=1} dy \right] \beta^{*\alpha}$$

$$= \left( \frac{\beta^*}{\beta} \right)^\alpha = \underline{\underline{(1 - \beta t)^{-\alpha}}}$$

Use this find  
 $E(Y)$   $E(Y^2)$  ...  
 $E(Y^k)$  ...

• We can show

$m_1(t) = (1-2t)^{-1/2}$

$Y_1 \sim \chi^2(\nu_1), \quad Y_2 \sim \chi^2(\nu_2) \rightarrow (1-2t)^{-\nu_2/2}$

$Y_1 + Y_2 \sim \chi^2(\nu_1 + \nu_2)$

$\text{mgf of } Y_1 + Y_2 = (1-2t)^{-\frac{\nu_1 + \nu_2}{2}} : \text{mgf of } \chi^2(\nu_1 + \nu_2)$

Normal case

$$\left[ \begin{array}{l} Y_1 \sim N(\mu_1, \sigma_1^2) \\ Y_2 \sim N(\mu_2, \sigma_2^2) \end{array} \right]_{\text{indep}}$$



$$\underline{Y_1 + Y_2 \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)}$$

$$\text{mgf of } Y_1 + Y_2 \quad ; \quad \exp\left(\mu_1 t + \frac{1}{2} \sigma_1^2 t^2\right) \times \exp\left(\mu_2 t + \frac{1}{2} \sigma_2^2 t^2\right)$$

$$= \exp\left((\mu_1 + \mu_2)t + \frac{1}{2}(\sigma_1^2 + \sigma_2^2)t^2\right)$$

∴ Normal !!

$$N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$$

# Gamma Probability Distribution

$Y \sim \text{Gamma}(\alpha, \beta)$  is the waiting time for the  $\alpha$ th Poisson event with  $\lambda = 1/\beta$ .

## Chi-square Distribution

$Y \sim \chi^2(\nu)$  if  $Y \sim \text{Gamma}(\nu/2, 2)$   
"Chi"

$\nu$  (integer)

$$\text{Gamma}(3, 2) = \chi^2(6)$$

$\nu$ : degrees of freedom  $\Rightarrow$  "# of independent components"

$$E(Y) = \nu$$

$$V(Y) = 2\nu$$

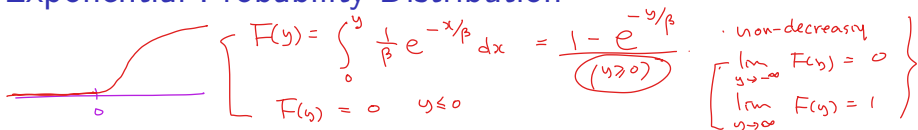
$$Z \sim N(0, 1) \quad \text{then} \quad Z^2 \sim \chi^2(1)$$

$$\left[ \begin{array}{l} Z_1, \dots, Z_\nu \sim N(0, 1) \\ \text{Indep} \end{array} \right] \quad \text{then} \quad \sum_{i=1}^{\nu} Z_i^2 \sim \chi^2(\nu)$$

$$E\left(\sum_{i=1}^{\nu} Z_i^2\right) = \nu \cdot E(Z_i^2) = \nu$$

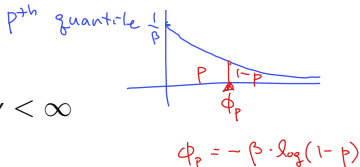
$$V\left(\sum_{i=1}^{\nu} Z_i^2\right) = \dots$$

# Exponential Probability Distribution



$Y \sim \text{Exp}(\beta)$  if  $Y \sim \text{Gamma}(1, \beta)$

$$f(y) = \frac{1}{\beta} e^{-y/\beta}, \quad 0 \leq y < \infty$$



► Modeling waiting time for the first Poisson event

►  $\mu = E(Y) = \underline{\beta}$  and  $\sigma^2 = V(Y) = \underline{\beta^2}$

► Memoryless property

$$\Leftrightarrow P(Y > a+b \mid Y > a) = P(Y > b)$$

$$\Leftrightarrow \frac{1 - F(a+b)}{1 - F(a)} = 1 - F(b) \quad \text{Check!}$$