

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

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

↓

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

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

$$\left(\frac{1}{2} z^2 \right) = x$$

part of $N(0, 1)$

• Except

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

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

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

$$\frac{y}{\beta} = x, \text{ use } \Gamma(\alpha) \text{ definition, etc. . . .}$$

Later. $Y \sim \text{Gamma}(\alpha_1, \beta)$
 $X \sim \text{Gamma}(\alpha_2, \beta)$) indep $\Rightarrow X+Y \sim \text{Gamma}(\alpha_1 + \alpha_2, \beta)$

Γ & Gamma Probability Distribution

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

Shape Rate

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

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

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

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 $\sigma^2 = V(Y) = \alpha\beta^2$.

check!

check!

"Go together"

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

use

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

Recall: Poisson λ : average no. occurrences
 Poisson rate in 1 unit in a given time frame.

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)

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

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

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

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

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 k^{nd} arrival

$P(Y > y) = P(\text{none or one arrival in } [0, y])$

$$= \frac{(\lambda y)^0 e^{-\lambda y}}{0!} + \frac{(\lambda y)^1 e^{-\lambda y}}{1!} = (\lambda y + 1) e^{-\lambda y}$$

$$F(y) = \vdots$$

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

MGF

$(Y_1 \text{ has } m_1(t))$	$(Y_2 \text{ has } m_2(t))$	indep
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Then $Y_1 + Y_2$ has mgf $m_1(t)m_2(t)$.

$$E(e^{(Y_1+Y_2)t}) = E(e^{tY_1}) E(e^{tY_2}) = m_1(t) \cdot m_2(t)$$

\uparrow
 indep

MGF of Gamma (α, β)

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

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

$= 1$

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

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

We can show

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

$m_1(t) = (1-2t)^{-\nu_2/2}$

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

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

Normal Case

$$\begin{bmatrix} Y_1 \sim N(\mu_1, \sigma_1^2) \\ Y_2 \sim N(\mu_2, \sigma_2^2) \end{bmatrix} \text{indep}$$



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

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

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

∴ 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 α th Poisson event with $\lambda = 1/\beta$.

Chi-squared Distribution

ν (integer)

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

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

ν : degrees of freedom \Rightarrow "# of independent components"

$Z \sim N(0, 1)$ then $Z^2 \sim \chi^2(1)$

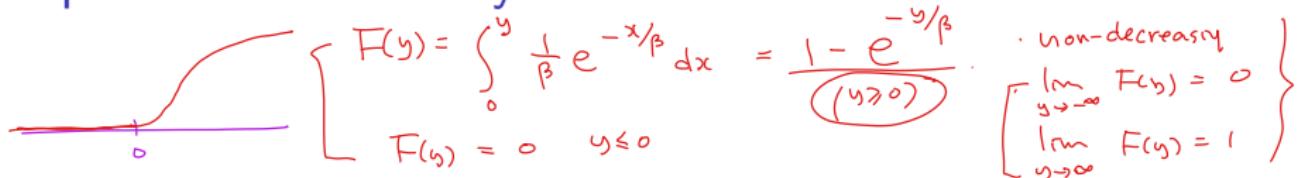
$$E(Y) = \nu$$
$$V(Y) = 2\nu$$

$Z_1, \dots, Z_\nu \sim N(0, 1)$ then $\sum_i Z_i^2 \sim \chi^2(\nu)$
 \downarrow Indep

$$V\left(\sum_i Z_i^2\right) = \dots$$

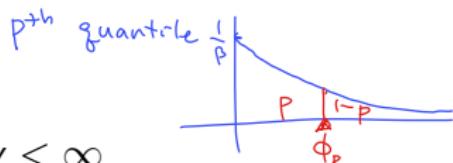
$$E\left(\sum_i Z_i^2\right) = \nu \cdot E(Z_i^2) = \nu$$

Exponential Probability Distribution



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

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



$$\phi_p = -\beta \cdot \log(1-p)$$

- Modeling waiting time for the first Poisson event
- $\mu = E(Y) = \beta$ and $\sigma^2 = V(Y) = \beta^2$
- Memoryless property

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

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