

10.99A)  $P_{\text{Poisson}}(\lambda) = \frac{e^{-\lambda} \lambda^y}{y!}$

~~$L(\lambda) = \prod_{i=1}^n \frac{e^{-\lambda} \lambda^{y_i}}{y_i!} = \frac{e^{-n\lambda} \lambda^{\sum y_i}}{\prod y_i!}$~~

$L(y_1, \dots, y_n | \lambda) = \prod_{i=1}^n \frac{e^{-\lambda} \lambda^{y_i}}{y_i!} = \frac{e^{-n\lambda} \lambda^{\sum y_i}}{\prod y_i!}$

B)  $P_{\text{Poisson}}(k) = \frac{e^{-\lambda} \lambda^k}{k!}$

$\frac{L(\lambda_0)}{L(\lambda_a)} < k \rightarrow \frac{e^{-n\lambda_0} \lambda_0^{\sum y_i}}{e^{-n\lambda_a} \lambda_a^{\sum y_i}} < k \rightarrow e^{-n(\lambda_0 - \lambda_a)} \left(\frac{\lambda_0}{\lambda_a}\right)^{\sum y_i} < k$

$\ln\left(\frac{\lambda_0}{\lambda_a}\right)^{\sum y_i} < \ln(k e^{n(\lambda_0 - \lambda_a)}) \rightarrow \sum y_i \ln\left(\frac{\lambda_0}{\lambda_a}\right) < \ln k + n(\lambda_0 - \lambda_a)$

$\sum y_i > \frac{\ln k + n(\lambda_0 - \lambda_a)}{\ln \lambda_0 - \ln \lambda_a} \rightarrow \sum y_i > k' \rightarrow n\bar{y} > k' \rightarrow n > \frac{k'}{\bar{y}}$

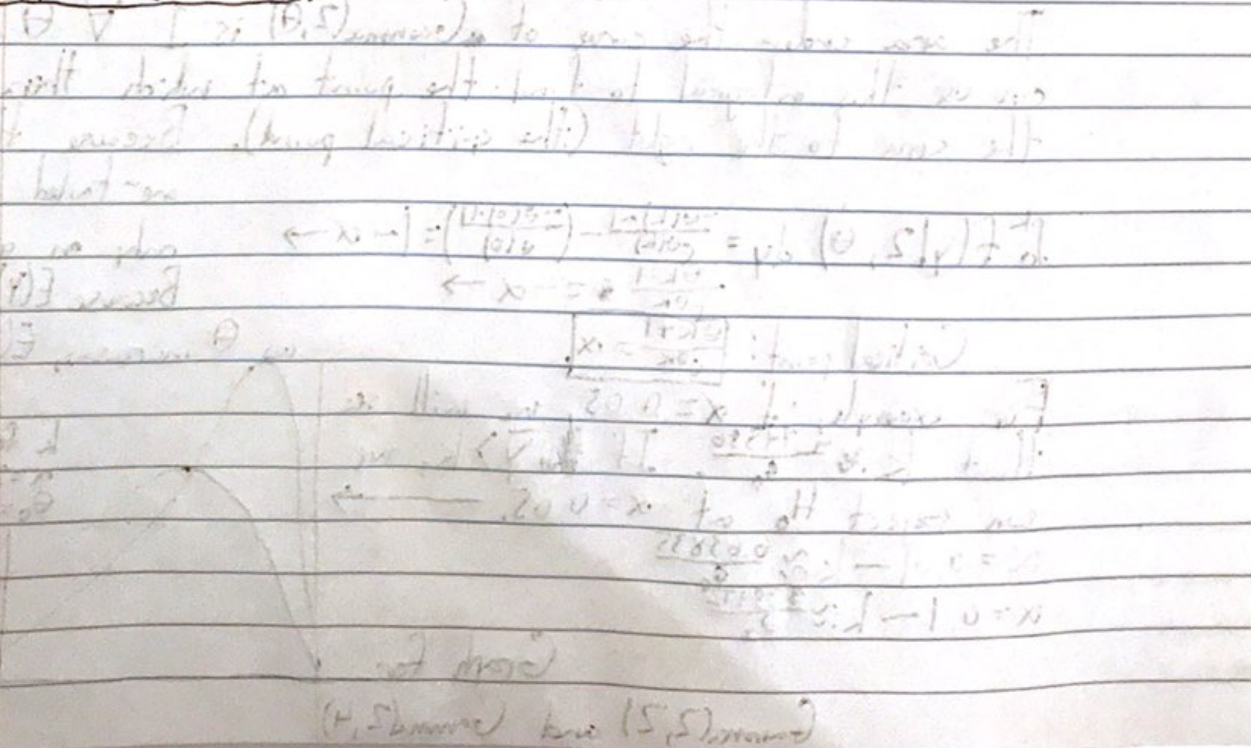
$\alpha = P\left(\sum y_i > k' | H_0 \text{ is true}\right) \rightarrow \alpha = 1 - P\left(\sum y_i \leq k'\right)$

$\alpha = 1 - \sum_{y=0}^{k'} \frac{e^{-n\lambda_0} (n\lambda_0)^y}{y!} \rightarrow \beta = 1 - \alpha = \sum_{y=0}^{k'} \frac{e^{-n\lambda_a} (n\lambda_a)^y}{y!}$

c) Yes, because of Theorem 10.1 and the fact that the same rejection region would have been obtained for any  $\lambda_0$  and  $\lambda_a \ni \lambda_a > \lambda_0$ .

d) We can see from Problem 10.99A that all steps except the last would be the same, and the resulting rejection region is:

$\sum y_i \ln\left(\frac{\lambda_0}{\lambda_a}\right) < \ln k + n(\lambda_0 - \lambda_a) \rightarrow \sum y_i > \frac{\ln k + n(\lambda_0 - \lambda_a)}{\ln \lambda_0 - \ln \lambda_a}, \lambda_a < \lambda_0 \rightarrow \ln\left(\frac{\lambda_0}{\lambda_a}\right) > 0$





$$f(y|2, \theta) = \frac{\theta^2}{\Gamma(2)} y^{2-1} e^{-\theta y} = \theta^2 y e^{-\theta y} \rightarrow L(y_1, \dots, y_n | 2, \theta) = \prod_{i=1}^n \theta^2 y_i e^{-\theta y_i} =$$

$$\frac{L(2, \theta)}{L(2, \hat{\theta})} = \frac{\theta^{2n} \prod y_i e^{-\theta \sum y_i}}{\hat{\theta}^{2n} \prod y_i e^{-\hat{\theta} \sum y_i}} = \left(\frac{\theta}{\hat{\theta}}\right)^{2n} e^{-(\theta - \hat{\theta}) \sum y_i} \quad \ln(L) = \ln(\theta^{2n} \prod y_i e^{-\theta \sum y_i}) =$$

$$\begin{aligned} & \ln(\theta^{2n}) + \ln(\prod y_i) - \sum \theta y_i = \\ & 2n \ln \theta + \ln(\prod y_i) - \sum \theta y_i \\ & \frac{d(\ln L)}{d\theta} = 2n \left(\frac{1}{\theta}\right) + 0 - \sum y_i = 0 \rightarrow \\ & \frac{2n}{\theta} = \sum y_i \rightarrow \hat{\theta} = \frac{2n}{\sum y_i} \rightarrow \hat{\theta} = \frac{2n}{n\bar{y}} \Rightarrow \hat{\theta} = \frac{2}{\bar{y}} \rightarrow \bar{y} = \frac{2}{\hat{\theta}} \end{aligned}$$

$$\begin{aligned} & \ln(\theta^{2n}) + \ln(\prod y_i) - \sum \theta y_i = \\ & 2n \ln \theta + \ln(\prod y_i) - \sum \theta y_i \\ & \frac{d(\ln L)}{d\theta} = 2n \left(\frac{1}{\theta}\right) + 0 - \sum y_i = 0 \rightarrow \\ & \frac{2n}{\theta} = \sum y_i \rightarrow \hat{\theta} = \frac{2n}{\sum y_i} \rightarrow \hat{\theta} = \frac{2}{\bar{y}} \rightarrow \bar{y} = \frac{2}{\hat{\theta}} \end{aligned}$$

Unbiased  $E(\bar{Y}) = E\left(\frac{\sum y_i}{n}\right) = \frac{1}{n} E\left(\sum y_i\right) = \frac{1}{n} \sum E(y_i) = \frac{1}{n} \sum \frac{2}{\theta} = \frac{1}{n} \sum \frac{2}{\theta} = \frac{1}{n} \left(\frac{2n}{\theta}\right) = \frac{2}{\theta} = \bar{y} \leftarrow$

Consistent  $Var(\bar{Y}) = Var\left(\frac{\sum y_i}{n}\right) = \frac{1}{n^2} Var\left(\sum y_i\right) = \frac{1}{n^2} \sum Var(y_i) = \frac{1}{n^2} \sum \left(\frac{2}{\theta^2}\right) = \frac{1}{n^2} \sum \frac{2}{\theta^2} = \frac{1}{n^2} \left(\frac{2n}{\theta^2}\right) = \frac{2}{n\theta^2}$   
 $\lim_{n \rightarrow \infty} \left(\frac{2}{n\theta^2}\right) = 0$

Sufficiently  $g(\bar{y}, \theta_0) = \theta_0^{2n} e^{-\theta_0 n \bar{y}}$ ,  $h(y_1, \dots, y_n) = \prod y_i$   $H_0: \theta = \theta_0$   
 $H_A: \theta < \theta_0$

$$\int_0^{\infty} f(y|2, \theta) \cdot dy = \int_0^{\infty} \theta^2 y e^{-\theta y} dy = \left[ -(\theta y + 1) e^{-\theta y} \right]_0^{\infty} = \frac{-\theta(\infty) - 1}{e^{-\theta(\infty)}} - \left( \frac{-\theta(0) + 1}{e^{-\theta(0)}} \right) = 0 + \frac{\theta + 1}{1} = 1$$

The area under the curve of Gamma(2, θ) is 1  $\forall \theta > 0$ . We can use this integral to find the point at which there is α of the curve to the right (the critical point). Because this is an α-tailed test, α is only on one side.

$$\int_0^k f(y|2, \theta) dy = \frac{-\theta(k) - 1}{e^{-\theta(k)}} - \left( \frac{-\theta(0) + 1}{e^{-\theta(0)}} \right) = 1 - \alpha \rightarrow \frac{-\theta k - 1}{e^{-\theta k}} = -\alpha \rightarrow$$

Because  $E(Y) = \frac{2}{\theta}$ , we see that as θ increases, E(Y) decreases.

Critical point:  $\frac{\theta k + 1}{e^{\theta k}} = \alpha$   
 For example, if  $\alpha = 0.05$ , we will see that  $k \approx \frac{4.4396}{\theta_0}$ . If  $\bar{y} > k$ , we can reject  $H_0$  at  $\alpha = 0.05$ .  
 $\alpha = 0.01 \rightarrow k \approx \frac{6.63835}{\theta_0}$   
 $\alpha = 0.1 \rightarrow k \approx \frac{3.88972}{\theta_0}$

