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the pmf for $y_i \sim \text{poi}(\lambda)$
 $f(y_i) = \frac{\lambda^{y_i} e^{-\lambda}}{y_i!}$

$$\begin{aligned} \text{1 } f(y_1, \dots, y_n) &= \prod_{i=1}^n \frac{\lambda^{y_i} e^{-\lambda}}{y_i!} \\ &= \frac{\lambda^{\sum y_i} e^{-n\lambda}}{\prod_{i=1}^n y_i!} \end{aligned}$$

~~state~~ changing variable to

$(y_1, \dots, y_{n-1}, \sum y_i)$ with $t = \sum y_i$
 yields.

$$\begin{aligned} f(y_1, \dots, y_{n-1}, \sum y_i) &= \frac{\lambda^t e^{-n\lambda}}{\left[\prod_{i=1}^{n-1} y_i! \right] (t - \sum_{i=1}^{n-1} y_i)!} \end{aligned}$$

2 the distribution, by the property of iid poisson's
~~to~~ $T = \sum y_i \sim \text{poi}(n\lambda)$
 $\Rightarrow f_T(t) = \frac{(n\lambda)^t e^{-n\lambda}}{t!}$

Combining 1 and 2.

$$f(y_1, \dots, y_n | \sum y_i = t) = \frac{t!}{\left[\prod_{i=1}^{n-1} y_i! \right] (t - \sum_{i=1}^{n-1} y_i)! n^t}$$

which is free from λ .

9.64

a. In class we showed that

\bar{Y} is minimum sufficient for μ .

$\Rightarrow \bar{Y}^2$ is minimum sufficient for μ^2 .

also, we have

$$E\hat{\mu}^2 = E\bar{Y}^2 - \frac{1}{n}$$

$$= \text{Var } \bar{Y} + (E\bar{Y})^2 - \frac{1}{n}$$

$$= \frac{1}{n} + \mu^2 - \frac{1}{n} = \mu^2$$

b. For $X \sim N(\mu, \sigma^2)$

$$E X^4 = \mu^4 + 6\mu^2\sigma^2 + 3\sigma^4$$

$$\Rightarrow \text{Var}(\hat{\mu}) = \text{Var}(\bar{Y}^2 - \frac{1}{n})$$

$$= E\left[\left(\bar{Y}^2 - \frac{1}{n}\right) - \mu^2\right]^2$$

$$= E\bar{Y}^4 - 2\left(\frac{1}{n} + \mu^2\right)E\bar{Y}^2 + \left(\frac{1}{n} + \mu^2\right)^2$$

$$E\bar{Y}^4 = \mu^4 + 6\mu^2\frac{1}{n} + 3\frac{1}{n^2}$$

$$E\bar{Y}^2 = \mu^2 + \frac{1}{n}$$

$$\Rightarrow \text{Var}(\hat{\mu}) = 4\mu^2\frac{1}{n} + 2\frac{1}{n^2}$$