

## Stat 342 Example 3

Suppose the entries of  $\underline{x} = (x_1, x_2, \dots, x_n)$  are iid Poisson  $\lambda$  and the inference problem concerns  $\lambda$ . The joint pmf of  $\underline{x}$  is

$$f(\underline{x}|\lambda) = \prod_{i=1}^n \left( \frac{e^{-\lambda} \lambda^{x_i}}{x_i!} \right)$$

Then treated as a function of  $\lambda$  for data  $\underline{x}$  plugged in, a log-likelihood function is

$$\ln f(\underline{x}|\lambda) = -n\lambda + \left( \sum_{i=1}^n x_i \right) \ln \lambda - \sum_{i=1}^n \ln(x_i!)$$

Then

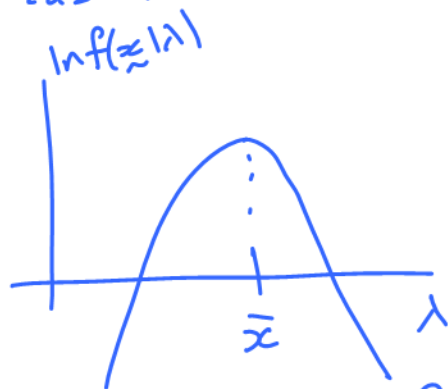
$$\frac{d}{d\lambda} \ln f(\underline{x}|\lambda) = -n + \frac{1}{\lambda} \left( \sum x_i \right)$$

For  $\sum x_i = 0$  this is negative and so the smallest possible value of  $\lambda$ , namely  $\lambda = 0$  maximizes the likelihood.

For other  $\sum x_i$  one may set

$$\frac{d}{d\lambda} \ln f(\underline{x}|\lambda) = 0$$

and solve for  $\lambda$ , obtaining  $\lambda = \bar{x}$ . A cartoon for the  $\sum x_i > 0$  case is below.



Thus in general the likelihood is maximized at  $\bar{x}$  and

$$\hat{\lambda}^{MLE}(\underline{x}) = \bar{x}$$