

M339J: May 6<sup>th</sup>, 2022.

MLE: Bernoulli .

$X \sim \text{Bernoulli}(q)$

↑ the probab. of "success"

Support( $X$ ) = {0, 1}

$$X \sim \begin{cases} 0 & \text{w/ probab. } 1-q \\ 1 & \text{w/ probab. } q \end{cases}$$

Let  $x_1, x_2, \dots, x_n$  be the observations from  $\text{Bernoulli}(q)$ .

They will all be 0 or 1.

We write the pmf as:

$$f(x; q) = \begin{cases} q & \text{if } x=1 \\ 1-q & \text{if } x=0 \end{cases}$$

$$f(x; q) = q^x (1-q)^{1-x}$$

$\Rightarrow$  The likelihood ftn:

$$L(q) = \prod_{j=1}^n f(x_j; q) = \prod_{j=1}^n (q^{x_j} (1-q)^{1-x_j})$$

$$L(q) = \prod_{j=1}^n q^{x_j} \cdot \prod_{j=1}^n (1-q)^{1-x_j} = q^{\sum x_j} (1-q)^{n - \sum x_j}$$

$\Rightarrow$  The log-likelihood:

$$l(q) = (\sum x_j) \cdot \ln(q) + (n - \sum x_j) \cdot \ln(1-q)$$

Differentiating w.r.t.  $q$ , we get

$$l'(q) = (\sum x_j) \cdot \frac{1}{q} + (n - \sum x_j) (-1) \frac{1}{1-q} = 0$$

Solve for  $q$ :

$$(\sum x_j)(1-q) = nq - (\sum x_j) \cdot q$$

$$\hat{q}_{\text{MLE}} = \bar{x} \dots \text{sample mean}$$

## MLE: Poisson.

$X \sim \text{Poisson}(\text{mean} = \lambda)$

pmf:  $f_X(x; \lambda) = e^{-\lambda} \frac{\lambda^x}{x!}$  ;  $x = 0, 1, 2, \dots$

Let  $x_1, x_2, \dots, x_n$  be the observations from  $\text{Poisson}(\lambda)$ .

$\Rightarrow$  The likelihood function is:

$$L(\lambda) = \prod_{j=1}^n f(x_j; \lambda) = \prod_{j=1}^n \left( e^{-\lambda} \cdot \frac{\lambda^{x_j}}{x_j!} \right)$$

$$L(\lambda) = e^{-n\lambda} \cdot \prod_{j=1}^n \lambda^{x_j} \cdot \frac{1}{\prod_{j=1}^n x_j!} \propto e^{-n\lambda} \cdot \lambda^{\sum x_j}$$

$\Rightarrow$  The log-likelihood is up to a constant:

$$l(\lambda) = -n\lambda + (\sum x_j) \ln(\lambda)$$

$$l'(\lambda) = -n + (\sum x_j) \cdot \frac{1}{\lambda} = 0$$

$$\Rightarrow \hat{\lambda}_{\text{MLE}} = \bar{x}$$

Problem.  $X \sim \text{Poisson}(\text{mean} = \lambda)$

Observations other than 0 and 1 have been deleted from the data.

The data contain an equal number of 0s and 1s.

Find the MLE for  $\lambda$ .

$\frac{n}{2}$  zeros and  $\frac{n}{2}$  ones

$\rightarrow$ : Condition on all the remaining data points being  $\leq 1$ .

$$L(\lambda) = \frac{1}{(\text{P}[X \leq 1; \lambda])^n} \cdot \frac{(e^{-\lambda} \cdot \frac{\lambda^0}{0!})^{\frac{n}{2}}}{e^{-\lambda}} \cdot \frac{(e^{-\lambda} \cdot \frac{\lambda^1}{1!})^{\frac{n}{2}}}{e^{-\lambda} \cdot \lambda}$$

$$L(\lambda) = \frac{1}{(e^{-\lambda}(1+\lambda))^n} \cdot e^{-\lambda \cdot \frac{n}{2}} \cdot e^{-\lambda \cdot \frac{n}{2}} \cdot \lambda^{\frac{n}{2}}$$

$$L(\lambda) = \frac{\lambda^{\frac{n}{2}}}{(1+\lambda)^n}$$

$$\Rightarrow l(\lambda) = \frac{n}{2} \cdot \ln(\lambda) - n \cdot \ln(1+\lambda)$$

$$\Rightarrow l'(\lambda) = \frac{n}{2} \cdot \frac{1}{\lambda} - n \cdot \frac{1}{1+\lambda} = 0 \quad | : n$$

$$\frac{1}{2\lambda} = \frac{1}{1+\lambda} \quad \Rightarrow \quad \boxed{\hat{\lambda}_{MLE} = 1}$$

□