

# Assignment1(stat)

Hoang Giap Vu

August 2023

## Problem 1

The dataset  $x$  is produced by the following code:

```
x <- c(-20, -10, 3, 4, 5, 6, 7, 20, 39)
```

## Problem 2

$$X \stackrel{d}{=} \mu + \sigma * Z$$

$$P(X \leq x) = P(\mu + \sigma * Z \leq x)$$

$$P(X \leq x) = P(Z \leq (x - \mu)/\sigma)$$

So for  $x = c_{\frac{k}{n+1}}$  ( $1 \leq k \leq n$ ),

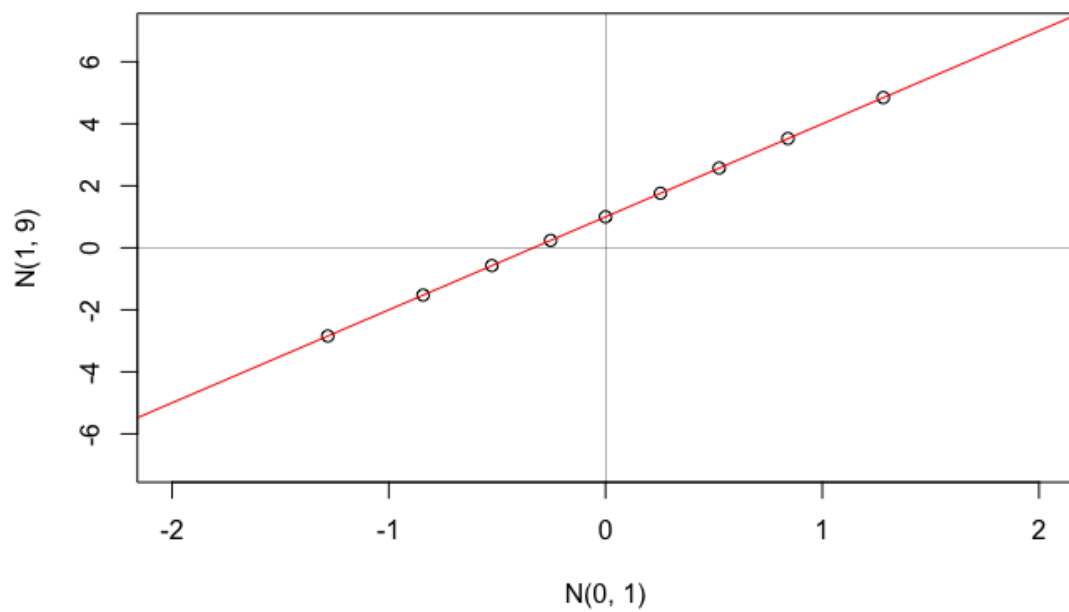
$$P(Z \leq (c_{\frac{k}{n+1}} - \mu)/\sigma) = P(X \leq c_{\frac{k}{n+1}})$$

$$P(Z \leq (c_{\frac{k}{n+1}} - \mu)/\sigma) = \frac{k}{n+1}$$

$$P(Z \leq (c_{\frac{k}{n+1}} - \mu)/\sigma) = P(Z \leq \Phi^{-1}(\frac{k}{n+1}))$$

So for  $k = 1, \dots, n$

$$(c_{\frac{k}{n+1}} - \mu)/\sigma = \Phi^{-1}(\frac{k}{n+1})$$



### Problem 3

a.

$$\sum_{i=1}^n (X_i - \bar{X})^2 = \sum_{i=1}^n X_i^2 - n\bar{X}^2$$

$$E(\hat{\theta}) = aE\left(\sum_{i=1}^n X_i^2 - n\bar{X}^2\right) = a[nE(X^2) - nE(\bar{X}^2)]$$

$$= a[n(\mu^2 + \sigma^2) - n(\sigma^2/n + \mu^2)] = a(n-1)\sigma^2$$

$$bias(\hat{\theta}) = E(\hat{\theta}) - \sigma^2 = [a(n-1) - 1]\sigma^2$$

$$bias(\hat{\theta})^2 = E(\hat{\theta}) - \sigma^2 = [a(n-1) - 1]^2 \sigma^4$$

$$var(\hat{\theta}) = a^2(n-1)^2 var\left(\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2\right)$$

$$= a^2(n-1)^2 \frac{2\sigma^4}{n-1}$$

$$= 2a^2(n-1)\sigma^4$$

$$MSE(\hat{\theta}) = 2a^2(n-1)\sigma^4 + [a(n-1) - 1]^2\sigma^4$$

**b.**

Differentiate:

$$d/da(MSE(\hat{\theta})) = 4a(n-1)\sigma^4 + 2[a(n-1) - 1](n-1)\sigma^4$$

Solve  $d/da(MSE(\hat{\theta})) = 0$ :

$$4a(n-1)\sigma^4 + 2[a(n-1) - 1](n-1)\sigma^4 = 0$$

If  $\sigma \neq 0$ :

$$\Rightarrow 4a(n-1) + 2[a(n-1) - 1](n-1) = 0$$

$$\Rightarrow 4a(n-1) + 2a(n-1)^2 - 2(n-1) = 0$$

$$\Rightarrow a = \frac{2(n-1)}{4(n-1) + 2(n-1)^2}$$

$$\Rightarrow a = \frac{1}{2+n-1} = \frac{1}{n+1}$$

**c.**

The solution from b. does not coincide with the MLE. However, since it is not asymptotically unbiased, it does not refute the optimality property of the MLE.

## Problem 4

Let

$$s = \sum_{i=1}^n X_i$$

The likelihood function is:

$$L(p) = p^s(1-p)^{n-s}$$

$$g(p) = s \times \ln(p) + (n-s) \times \ln(1-p)$$

$$s(p) = \frac{s}{p} - \frac{n-s}{1-p}$$

$$s(p) = \frac{s - sp - np + sp}{p(1-p)}$$

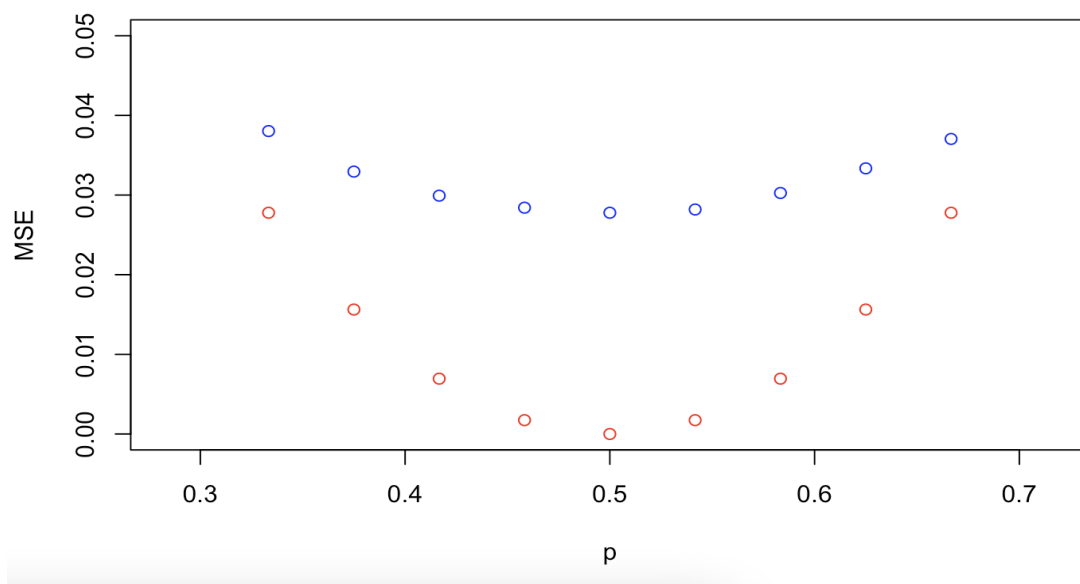
$$s(p) = \frac{s - np}{p(1-p)}$$

Now,

$$\begin{cases} s(p) < 0 & \forall p < \frac{s}{n} \\ s(p) = 0 & \text{if } p = \frac{s}{n} \\ s(p) > 0 & \forall p > \frac{s}{n} \end{cases}$$

So the MLE is

$$\hat{\theta}_{ML} = \begin{cases} \frac{s}{n} & \text{if } 1/3 \leq \frac{s}{n} \leq 2/3 \\ 1/3 & \text{if } \frac{s}{n} < 1/3 \\ 2/3 & \text{if } \frac{s}{n} > 2/3 \end{cases}$$



## Problem 5

### Finding MLE

$$P(X_i = 1) = pr + (1 - p)(1 - r) = 2pr - p - r + 1$$

Let

$$s = \sum_{i=1}^n X_i$$

So

$$L(p) = [2pr - p - r + 1]^s [p + r - 2pr]^{n-s}$$

$$g(p) = s \times \ln(2pr - p - r + 1) + (n - s) \times \ln(p + r - 2pr)$$

$$s(p) = \frac{s(2r - 1)}{2pr - p - r + 1} + \frac{(n - s)(1 - 2r)}{p + r - 2pr}$$

If  $r = 1/2$ ,  $s(p) = 0 \forall p \in [0, 1]$ , so all possible  $p$  values have the same likelihood.

If  $r \neq 1/2$ , we solve  $s(p) = 0$  to find the MLE:

$$\begin{aligned} \frac{s(2r - 1)}{2pr - p - r + 1} + \frac{(n - s)(1 - 2r)}{p + r - 2pr} &= 0 \\ \Leftrightarrow (2r - 1) \left[ \frac{s}{2pr - p - r + 1} + \frac{(s - n)}{p + r - 2pr} \right] &= 0 \\ \Leftrightarrow \frac{s}{2pr - p - r + 1} + \frac{(s - n)}{p + r - 2pr} &= 0 \text{ (since } 2r - 1 \neq 0) \\ \Leftrightarrow sp + sr - 2spr + 2spr - sp - sr + s - 2npr + np + nr - n &= 0 \\ \Leftrightarrow s - 2npr + np + nr - n &= 0 \\ p(2nr - n) &= nr + s - n \\ p &= \frac{nr + s - n}{2nr - n} \end{aligned}$$

So the MLE is

$$\hat{\Theta}_{ML} = \frac{nr + s - n}{2nr - n}$$

where

$$s = \sum_{i=1}^n X_i$$

## Finding MME

$$M_1 = \frac{s}{n}$$

Solve  $E(X) = M_1$ :

$$\begin{aligned} pr + (1 - p)(1 - r) &= \frac{s}{n} \\ 2pr - p - r + 1 &= \frac{s}{n} \\ p(2r - 1) &= \frac{s}{n} + r - 1 \\ p &= \frac{s + rn - n}{n(2r - 1)} \text{ (for } r \neq 1/2) \end{aligned}$$

So

$$\hat{\Theta}_{MM} = \frac{s + rn - n}{n(2r - 1)} \text{ (for } r \neq 1/2)$$