

# MTH210a (2023): Quiz 4

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Roll. No. Solutions

Let  $(X_i, Y_i) \stackrel{iid}{\sim} F$ , where  $F$  is some unknown bivariate distribution with mean vector  $(\theta, \theta)^T$  and finite marginal variances. In order to estimate  $\theta$ , you choose the estimator:

$$\hat{\theta} = \frac{1}{n} \sum_{i=1}^n \frac{X_i + Y_i}{2}$$

1. <sup>10</sup> (7 marks) Write the steps of bootstrapping to obtain a confidence interval around  $\hat{\theta}$ . Explain all steps carefully.

Bonus

2. (3 marks) Propose an alternative way (without using bootstrap) to obtain a confidence interval around  $\hat{\theta}$ ?

①  $(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n) \stackrel{iid}{\sim} F$  with mean  $\begin{pmatrix} \theta \\ \theta \end{pmatrix}$ .

$$\hat{\theta} = \frac{1}{n} \sum_{i=1}^n \frac{X_i + Y_i}{2}$$

(-5) if parametric bootstrap done correctly.

In order to obtain C.I around  $\hat{\theta}$ , we will implement non-parametric bootstrap.

For  $b=1, \dots, B$  we draw a sample of size  $n$  from

$\{(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n)\}$  with replacement.

(-2) if details not given about sampling method.

Bootstrap sample 1:  $(X_{11}^*, Y_{11}^*), (X_{12}^*, Y_{12}^*), \dots, (X_{1n}^*, Y_{1n}^*)$ : set  $\hat{\theta}_1^* = \frac{1}{n} \sum_{i=1}^n \frac{X_{1i}^* + Y_{1i}^*}{2}$

Bootstrap sample 2:  $(X_{21}^*, Y_{21}^*), (X_{22}^*, Y_{22}^*), \dots, (X_{2n}^*, Y_{2n}^*)$ : set  $\hat{\theta}_2^* = \frac{1}{n} \sum_{i=1}^n \frac{X_{2i}^* + Y_{2i}^*}{2}$

$\vdots$   
Bootstrap sample B:  $(X_{B1}^*, Y_{B1}^*), (X_{B2}^*, Y_{B2}^*), \dots, (X_{Bn}^*, Y_{Bn}^*)$ : set  $\hat{\theta}_B^* = \frac{1}{n} \sum_{i=1}^n \frac{X_{Bi}^* + Y_{Bi}^*}{2}$

The  $100(1-\alpha)$  confidence interval is then

$$\left( \hat{\theta}_{(L \frac{\alpha}{2} B)}^*, \hat{\theta}_{(U(1-\frac{\alpha}{2}) B)}^* \right)$$

(-2) if order statistic is missing  $\theta(\cdot)$

(-3) if CIs are missing.

② Note that

$$\text{Var}(\hat{\theta}) = \frac{1}{n} \text{Var}\left(\frac{X_1 + Y_1}{2}\right) = \frac{1}{4n} [\text{Var}(X_1) + \text{Var}(Y_1) + 2\text{Cov}(X_1, Y_1)]$$

then by CLT  $\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} N(0, \tau^2)$  where  $\tau^2 = \frac{\text{Var}(X_1) + \text{Var}(Y_1) + 2\text{Cov}(X_1, Y_1)}{4}$ .

Let  $\hat{\tau}^2 = \frac{S_x^2 + S_y^2 + 2S_{xy}}{4}$  where

$$S_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 \quad S_y^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$$

$$S_{xy} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

Then  $\hat{\theta} \pm Z_{\frac{\alpha}{2}} \sqrt{\frac{\hat{\tau}^2}{n}}$  is a potential alternative

for a confidence interval.

Points given only if  
completely correct.