# Homework 2

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### Question 1

### Question 9

If  $X_1, \ldots, X_n \stackrel{\text{iid}}{\sim} N(\mu, \sigma^2)$  with unknown  $\mu$  and  $\sigma^2$ , then a  $\gamma$ % confidence interval for  $\mu$  is given by

$$\mathcal{I} = \left( \bar{X} - t_{\gamma}(n) \cdot S / \sqrt{n} , \bar{X} + t_{\gamma}(n) \cdot S / \sqrt{n} \right),$$

where  $t_{\gamma}(n) = T_{n-1}^{-1}((1+\gamma)/2)$  is the  $(1+\gamma)/2$ th quantile of the t distribution with df = n-1 and S is the sample standard deviation. The length of this confidence interval is given by

$$\Delta = \max(\mathcal{I}) - \min(\mathcal{I}) = \left(\bar{X} + t_{\gamma}(n) \cdot S / \sqrt{n}\right) - \left(\bar{X} - t_{\gamma}(n) \cdot S / \sqrt{n}\right) = 2t_{\gamma}(n) \cdot S / \sqrt{n}.$$

The squared length is then given by  $\Delta^2 = 4t_\gamma^2(n) \cdot S^2/n$ . Because the sample variance is an unbiased estimator for  $\sigma^2$ , we have  $\mathbb{E}[\Delta^2] = \mathbb{E}\left[4t_\gamma^2(n) \cdot S^2/n\right] = 4t_\gamma^2(n) \cdot \sigma^2/n$ . We now set  $\mathbb{E}[\Delta^2] < \sigma^2/2$ , and after some cancellations, we see that we need  $t_\gamma^2(n)/n < 1/8$ . There is no way to find a closed-form expression for this, so we will have to check the value of  $t_\gamma^2(n)/n$  for increasing values of n. I set up a while loop in R to solve for it, and when  $\gamma=0.9$ , we find that n=24 is the smallest value of n such that  $\mathbb{E}[\Delta^2] < \sigma^2/2$ .

#### Question 10

Let  $\boldsymbol{X} \stackrel{\text{iid}}{\sim} N(\theta, \sigma^2)$ , where  $\theta$  is unknown and  $\sigma^2$  is known, and we assume prior that  $\theta \sim N(\mu, \nu^2)$ , where both  $\mu$  and  $\nu^2$  are known.

(a) Since normal distributions are are conjugate to normal sampling, it follows that  $\theta \mid \mathbf{x} \sim N(\tilde{\mu}, \tilde{\sigma}^2)$ , where

$$\tilde{\mu} = \frac{\sigma^2 \mu + n \nu^2 \bar{x}}{\sigma^2 n \nu^2}$$
 and  $\tilde{\sigma}^2 = \frac{\sigma^2 \nu^2}{\sigma^2 + n \mu^2}$ .

We also know that  $(\theta \mid \mathbf{x} - \tilde{\mu})/\tilde{\sigma} \sim N(0,1)$ , and so a 95% confidence interval for  $\theta \mid \mathbf{x}$  is given by

$$\mathcal{I} = \left( \tilde{\mu} - \Phi^{-1}(0.975) \cdot \tilde{\sigma} , \tilde{\mu} + \Phi^{-1}(0.975) \cdot \tilde{\sigma} \right).$$

(b) We can think of our interval  $\mathcal{I}$  as a function of  $\nu^2$ . To examine what happens to  $\mathcal{I}(\mu^2)$  as  $\nu^2 \to \infty$ , we will first look at  $\tilde{\mu}$  and  $\tilde{\sigma}$ . Using L'Hopital's rule, we have

$$\begin{split} &\lim_{\nu^2 \to \infty} \tilde{\mu} = \lim_{\nu^2 \to \infty} \frac{\sigma^2 \mu + n \nu^2 \bar{x}}{\sigma^2 n \nu^2} = \lim_{\nu^2 \to \infty} \frac{n \bar{x}}{n} = \bar{x}, \\ &\lim_{\nu^2 \to \infty} \tilde{\sigma} = \lim_{\nu^2 \to \infty} \sqrt{\frac{\sigma^2 \nu^2}{\sigma^2 + n \mu^2}} = \sqrt{\lim_{\nu^2 \to \infty} \frac{\sigma^2 \nu^2}{\sigma^2 + n \mu^2}} = \sqrt{\lim_{\nu^2 \to \infty} \frac{\sigma^2}{n}} = \frac{\sigma}{\sqrt{n}}, \end{split}$$

and so  $\mathcal{I}(\nu^2) \to (\bar{x} - \Phi^{-1}(0.975) \cdot \sigma/\sqrt{n}, \bar{x} + \Phi^{-1}(0.975) \cdot \sigma/\sqrt{n})$ , which is a 95% confidence interval for  $\theta$ .

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## Question 11

Let  $\boldsymbol{X} \stackrel{\text{iid}}{\sim} \operatorname{Unif}(0, \theta)$ , where  $\theta$  is unknown, and let  $Y = X_{(n)}$  be the nth order statistic.

- (a) Let  $F_i(x) = \Pr(X_i \leq x) = x/\theta$  for all  $i \in \{1, \dots, n\}$ . Then the cdf of Y is  $G(y) = \prod_{i=1}^n F_i(y) = (y/\theta)^n$ , since all of the  $X_i$ 's are independent. The density of Y is then given by  $g(y) = ny^{n-1}/\theta^n$  for  $y \in [0, \theta]$ . By letting  $W = Y/\theta$ , we have  $Y = \theta W$  and  $\partial Y/\partial W = \theta$ , so the density of  $Y/\theta$  is given by  $h(w) = g(y(w)) \cdot \theta = nw^{n-1}$  for  $w \in [0, 1]$ . The cdf is then seen to be  $H(w) = w^n$ , and so the quantile is given by  $H^{-1}(w) = \sqrt[n]{w}$ .
- (b) Naturally, since we must have  $y \leq \theta$ , it is natural that Y will underestimate  $\theta$ , and is therefore biased. We have

$$\mathbb{E}[Y] = \int_0^\theta y \cdot \frac{ny^{n-1}}{\theta^n} \, \partial y = \frac{n}{\theta^n} \int_0^\theta y^n \, \partial y = \frac{n}{\theta^n} \cdot \frac{y^{n+1}}{n+1} \bigg|_0^\theta = \frac{n}{\theta^n} \cdot \frac{\theta^{n+1}}{n+1} = \frac{n\theta}{n+1},$$

and so the bias is  $bias(Y) = \mathbb{E}[Y] - \theta = -\theta/(n+1)$ .

(d) Using the cdf of  $Y/\theta$ , for any interval involving  $Y/\theta$ , we have  $\Pr(a \le Y/\theta \le b) = b^n - a^n$ , where  $a, b \in (0, 1]$ . Rearranging the terms inside the interval gives us  $\Pr(Y/b \le \theta \le Y/a) = b^2 - a^2 \stackrel{\text{set}}{=} \gamma$ . That is, as long as we impose the constraint that  $b^2 - a^2 = \gamma$ , any interval (Y/b, Y/a) is a  $\gamma\%$  confidence interval for  $\theta$ .