

Name Jerry

Student Number _____

STA 302 f2014 Quiz 2A

1. (5 points) Let $Y = X_1 + X_2$, where X_1 and X_2 are independent, $X_1 \sim \chi^2(\nu_1)$ and $Y \sim \chi^2(\nu_1 + \nu_2)$, where ν_1 and ν_2 are both positive. Show $X_2 \sim \chi^2(\nu_2)$.

$$M_Y(t) \stackrel{\text{ind.}}{=} M_{X_1}(t) M_{X_2}(t)$$

$$\Rightarrow (1-2t)^{-\frac{(\nu_1+\nu_2)}{2}} = (1-2t)^{-\nu_1/2} M_{X_2}(t)$$

$$\Rightarrow (1-2t)^{-\nu_1/2} (1-2t)^{-\nu_2/2} = (1-2t)^{-\nu_1/2} M_{X_2}(t)$$

$$\Rightarrow (1-2t)^{-\nu_2/2} = M_{X_2}(t)$$

$$\text{So } X_2 \sim \chi^2(\nu_2)$$

2. ~~(2 points)~~ Let $Y_i = \beta x_i + \epsilon_i$ for $i = 1, \dots, n$, where $\epsilon_1, \dots, \epsilon_n$ are a random sample from a distribution with expected value zero and variance σ^2 , and β and σ^2 are unknown constants. The numbers x_1, \dots, x_n are known, observed constants. In homework, you derived the least squares estimator $\hat{\beta} = \frac{\sum_{i=1}^n x_i Y_i}{\sum_{i=1}^n x_i^2}$. Just use this; don't derive it again.

For these questions, you may use well known properties of the expected value and variance without proof.

- (a) (1 point) What is $E(Y_i)$? Show some work.

$$\begin{aligned} E(Y_i) &= E(\beta x_i + \epsilon_i) = \beta x_i + E(\epsilon_i) = \beta x_i + 0 \\ &= \beta x_i \end{aligned}$$

- (b) (1 point) What is $\text{Var}(Y_i)$? Show some work.

$$\text{Var}(Y_i) = \text{Var}(\beta x_i + \epsilon_i) = \text{Var}(\epsilon_i) = \sigma^2$$

- (c) (2 points) Recall that a statistic is an *unbiased estimator* of a parameter if the expected value of the statistic is equal to the parameter. Is $\hat{\beta}$ an unbiased estimator of β ? Answer Yes or No and show your work.

$$\begin{aligned} E(\hat{\beta}) &= E\left(\frac{\sum_{i=1}^n x_i Y_i}{\sum_{i=1}^n x_i^2}\right) = \frac{1}{\sum_{i=1}^n x_i^2} E\left(\sum_{i=1}^n x_i Y_i\right) \\ &= \frac{1}{\sum_{i=1}^n x_i^2} \sum_{i=1}^n x_i E(Y_i) = \frac{1}{\sum_{i=1}^n x_i^2} \sum_{i=1}^n x_i \beta x_i \\ &= \beta \frac{\sum_{i=1}^n x_i^2}{\sum_{i=1}^n x_i^2} = \beta \quad \text{Yes, unbiased} \end{aligned}$$

- (d) (2 points) What is $\text{Var}(\hat{\beta})$? Show your work. Circle your answer.

$$\begin{aligned} \text{Var}(\hat{\beta}) &= \text{Var}\left(\frac{\sum_{i=1}^n x_i Y_i}{\sum_{i=1}^n x_i^2}\right) = \frac{1}{\left(\sum_{i=1}^n x_i^2\right)^2} \text{Var}\left(\sum_{i=1}^n x_i Y_i\right) \\ &= \frac{1}{\left(\sum_{i=1}^n x_i^2\right)^2} \sum_{i=1}^n x_i^2 \text{Var}(Y_i) = \frac{\sum_{i=1}^n x_i^2 \sigma^2}{\left(\sum_{i=1}^n x_i^2\right)^2} \\ &= \frac{\sigma^2}{\sum_{i=1}^n x_i^2} \end{aligned}$$

Name Jenny

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STA 302 f2014 Quiz 2B

1. (5 points) Let X_1, \dots, X_n be random sample from a $N(\mu, \sigma^2)$ distribution. Find the distribution of the sample mean \bar{X} . Zero marks for the correct answer without a proof.

$$\begin{aligned}M_{\bar{X}}(t) &= M_{\frac{1}{n}\sum X_i}(t) = M_{\sum X_i}(t/n) \\&= \prod_{i=1}^n M_{X_i}(t/n) = \prod_{i=1}^n e^{\mu t/n + \frac{1}{2}\sigma^2(t/n)^2} \\&= e^{n(\mu t/n + \frac{1}{2}\sigma^2 t^2/n^2)} \\&= e^{\mu t + \frac{1}{2}(\frac{\sigma^2}{n})t^2}\end{aligned}$$

So $\bar{X} \sim \mathcal{N}(\mu, \frac{\sigma^2}{n})$

2. (5 points) Let $Y_i = \beta x_i + \epsilon_i$ for $i = 1, \dots, n$, where $\epsilon_1, \dots, \epsilon_n$ are a random sample from a distribution with expected value zero and variance σ^2 , and β and σ^2 are unknown constants. The numbers x_1, \dots, x_n are known, observed constants.

(a) Find the Least Squares estimate of β by minimizing the function

$$Q(\beta) = \sum_{i=1}^n (Y_i - \beta x_i)^2$$

over all values of β . Let $\hat{\beta}$ denote the point at which $Q(\beta)$ is minimal. **Circle your formula for $\hat{\beta}$.** Don't bother with the second derivative test.

$$\begin{aligned} \frac{dQ}{d\beta} &= \frac{d}{d\beta} \sum_{i=1}^n (Y_i - \beta x_i)^2 = \sum_{i=1}^n \frac{d}{d\beta} (Y_i - \beta x_i)^2 \\ &= \sum_{i=1}^n 2(Y_i - \beta x_i)(-x_i) \stackrel{\text{set}}{=} 0 \end{aligned}$$

$$\Rightarrow \sum_{i=1}^n (x_i Y_i - \beta x_i^2) = 0 \Rightarrow \sum_{i=1}^n x_i Y_i = \beta \sum_{i=1}^n x_i^2$$

$$\Rightarrow \beta = \frac{\sum_{i=1}^n x_i Y_i}{\sum_{i=1}^n x_i^2} \quad \text{so} \quad \hat{\beta} = \frac{\sum_{i=1}^n x_i Y_i}{\sum_{i=1}^n x_i^2}$$

- (b) Calculate $\hat{\beta}$ for the following data set. ^{The} ~~Your~~ answer is a number. Show some work. Circle your answer.

x^2	9	25	36	16	4
x	3	5	6	4	2
y	-6	-4	-22	12	7

$$\sum x^2 = 90$$

$$\sum xy = -108$$

$$\hat{\beta} = \frac{-108}{90} = -1.2$$