

Final test contents:

Problem 1: Confidence interval for an unknown population proportion; Sample size calculation

Problem 2: CI for population mean; sample size calculation

Problem 3: Bivariate discrete distribution

- Marginal distribution
- Marginal distribution mean
- Marginal distribution variance
- Covariance
- Correlation
- Independence

Problem 4: Expected value of $C_1X_1 + C_2X_2$
Variance of $C_1X_1 + C_2X_2$

Problem 5: Distribution of \bar{X} (random sample mean) using central limit theorem

$E(\bar{X})$

$\text{Var}(\bar{X})$

Percentiles of Normal distribution

problem 6: Maximum likelihood estimator derivation
Bias, Mean squared error (bonus)
Central limit theorem.]

Exercise set for chapter 5-8

1 Chapter 5

Problem 1. Two random variables have a joint probability mass function in Table 1. Show that the correlation between X and Y is zero but X and Y are not independent.

Table 1: Joint probability mass function

x	-1	0	0	1
y	0	-1	1	0
$f_{X,Y}(x,y)$	1/4	1/4	1/4	1/4

① Marginal distribution of X

x	-1	0	1
$f(x)$	1/4	1/2	1/4

Marginal distribution of Y

y	-1	0	1
$f(y)$	1/4	1/2	1/4

② $E(X) = (-1) \cdot \frac{1}{4} + 0 \cdot \frac{1}{2} + 1 \cdot \frac{1}{4} = 0$

$E(Y) = 0$

$Var(X) = E[X^2] - 0^2 = (-1)^2 \cdot \frac{1}{4} + 0^2 \cdot \frac{1}{2} + 1^2 \cdot \frac{1}{4} = \frac{1}{2}$

$Var(Y) = \frac{1}{2}$

③ $E(XY) = (-1) \cdot 0 \cdot \frac{1}{4} + 0 \cdot (-1) \cdot \frac{1}{4} + 0 \cdot 1 \cdot \frac{1}{4} + 1 \cdot 0 \cdot \frac{1}{4} = 0$

④ $Cov(X, Y) = E(XY) - E(X) \cdot E(Y) = 0$

⑤ $Corr(X, Y) = \frac{Cov(X, Y)}{\sqrt{Var(X) \cdot Var(Y)}} = 0$

$f_{X,Y}(1,1) = 0$

$f_X(1) = \frac{1}{4}$

$f_Y(1) = \frac{1}{4}$

$f_{X,Y}(1,1) \neq f_X(1) f_Y(1)$

Independence of x and $y \iff f_{X,Y}(x,y) = f_X(x) f_Y(y)$ for all x and y .

Problem 2. A random variable X has the probability distribution $f_X(x) = x/18, 0 \leq x \leq 6$.

Determine the following:

(a) The expected value of X . $= \int_0^6 x \cdot \frac{x}{18} dx = \frac{x^3}{54} \Big|_0^6 = \frac{36}{9} = 4$

(b) The variance of X . $= E[X^2] - (E[X])^2 = \int_0^6 x^2 \cdot \frac{x}{18} dx - 4^2$

(c) The cumulative distribution of X .

$= \frac{x^4}{72} \Big|_0^6 - 16 = 18 - 16 = 2$

$F(x) = \begin{cases} 0 & x < 0 \\ \int_0^x u/18 dx = \frac{u^2}{36} \Big|_0^x = \frac{x^2}{36} & 0 \leq x \leq 6 \\ 1 & x > 6 \end{cases}$

X and Y are not independent

$$\textcircled{1} \quad E(C_1 X_1 + C_2 X_2) = C_1 \cdot E(X_1) + C_2 E(X_2)$$

$$\textcircled{2} \quad \text{Var}(C_1 X_1 + C_2 X_2) = C_1^2 \cdot \text{Var}(X_1) + C_2^2 \text{Var}(X_2) + 2 \cdot C_1 \cdot C_2 \text{COV}(X_1, X_2)$$

Problem 3. X and Y are independent, normal random variables with $\underline{E(X) = 2}$, $\text{Var}(X) = 5$, $E(Y) = 6$, $\text{Var}(Y) = 8$.

(a) $\underline{E(3X + 2Y)} = 3 \cdot E(X) + 2 \cdot E(Y) = 3 \cdot 2 + 2 \cdot 6 = 18$

(b) $\text{Var}(3X + 2Y) = 3^2 \cdot \text{Var}(X) + 2^2 \cdot \text{Var}(Y) = 3^2 \cdot 5 + 2^2 \cdot 8 = 45 + 32 = 77$

(c) Note that the sum of two normal random variables is still a normal random variable. What is the distribution of $3X + 2Y$? Please specify its mean and variance.

$$3X + 2Y \sim N(18, 77)$$

2 Chapter 6

Problem 4. Consider the following two samples: Sample 1: 10, 9, 8, 7, 8, 6, 10, 6; Sample 2: 10, 6, 10, 6, 8, 10, 8, 6.

range for sample 2 : 4

range for sample 1: 4

$$\underline{\underline{sd_1: 1.6}}$$

$$\underline{\underline{2sd_2: 1.85}}$$

- (a) Calculate the sample range for both samples. Would you conclude that both samples exhibit the same variability? Explain.
- (b) Calculate the sample standard deviations for both samples. Do these quantities indicate that both samples have the same variability? Explain.

3 Chapter 7

Problem 5. Describe the population/distribution parameter that the following numerical summaries trying to estimate.

- (a) Sample mean $\xrightarrow{\text{Estimate}}$ population mean
- (b) Sample variance $\xrightarrow{\text{Estimate}}$ population variance
- (c) Sample proportion $\xrightarrow{\text{Estimate}}$ population proportion

$$\hat{\lambda} = \bar{x} \xrightarrow{\text{Estimate}} \lambda$$

Problem 6. A random sample of size $n_1 = 16$ is selected from a normal population with a mean of 75 and a standard deviation of 8. A second random sample of size $n_2 = \overline{9}$ is taken from another normal population with mean 70 and standard deviation 12. Let \bar{X}_1 and \bar{X}_2 be the two sample means. Note the difference of two normal random variables is still a normal random variable. Find:

- (a) The probability that $\bar{X}_1 - \bar{X}_2$ exceeds 4.
- (b) The 95th percentile of the distribution for $\bar{X}_1 - \bar{X}_2$.

\bar{X}_1 follows a Normal distribution $N(75, \frac{8^2}{16})$

\bar{X}_2 follows a Normal distribution $N(70, \frac{12^2}{9})$

$$E(\bar{X}_1 - \bar{X}_2) = E(\bar{X}_1) - E(\bar{X}_2) = 5$$

(Assume the two samples are drawn independently)

$$\text{Var}(\bar{X}_1 - \bar{X}_2) = 4 + 16 = 20$$

$$\bar{X}_1 - \bar{X}_2 \sim N(5, 20)$$

$$a) P(\bar{X}_1 - \bar{X}_2 > 4) = 1 - P(\bar{X}_1 - \bar{X}_2 \leq 4) = 1 - \Phi_{0.1}\left(\frac{4-5}{\sqrt{20}}\right)$$

$$b) \mu + z_{0.95} \cdot \sigma \rightarrow \Phi_{0.1}^{-1}(0.95) = 1 - \Phi_{0.1}(0.274)$$

\rightarrow 95th percentile of standard Normal distribution

$$= 5 + 1.64 \cdot \sqrt{20} = 1 - 0.412$$
$$= 12.37 = 0.588$$

Problem 7. Suppose we have a random sample of size $2n$ from a population denoted by X , and $E(X) = \mu$, $\text{Var}(X) = \sigma^2$. Let $\bar{X}_1 = \frac{1}{2n} \sum_{i=1}^{2n} X_i$ and $\bar{X}_2 = \frac{1}{n} \sum_{i=1}^n X_i$ be two estimators of μ . Which is the better estimator of μ ? Hint, compare the bias and the MSE of the two estimators.

$$\text{Bias for } \bar{X}_1 : E(\bar{X}_1) - \mu = \mu - \mu = 0$$

$$\text{Bias for } \bar{X}_2 : E(\bar{X}_2) - \mu = \mu - \mu = 0$$

$$E\left(\frac{1}{2n} \sum_{i=1}^{2n} X_i\right) = \frac{1}{2n} \cdot \sum_{i=1}^{2n} E(X_i) = \frac{1}{2n} \cdot E(X_1) + \frac{1}{2n} E(X_2) + \dots + \frac{1}{2n} E(X_{2n})$$
$$= \frac{1}{2n} \cdot 2n \cdot \mu = \mu$$

$$\text{MSE for } \bar{X}_1 : E(\bar{X}_1 - \mu)^2 = \text{Var}(\bar{X}_1) = \frac{\sigma^2}{2n}$$

\bar{X}_1 is better than \bar{X}_2 in terms of MSE.

$$\text{MSE for } \bar{X}_2 : E(\bar{X}_2 - \mu)^2 = \text{Var}(\bar{X}_2) = \frac{\sigma^2}{n}$$

4 Chapter 8

An article in the Journal of the American Statistical Association ["Illustration of Bayesian Inference in Normal Data Models Using Gibbs Sampling" (1990, Vol. 85, pp. 972-985)] measured the weight of 30 rats under experiment controls. Suppose that 12 were underweight rats.

(a) Calculate a 95% two-sided confidence interval on the true proportion of rats that would show underweight from the experiment.

$$\left(\hat{p} - 1.96 \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}, \hat{p} + 1.96 \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right)$$
$$\hat{p} = \frac{12}{30} = 0.4, n = 30$$
$$0.4 \pm 1.96 \sqrt{\frac{0.4 \cdot 0.6}{30}}$$
$$\rightarrow [0.225, 0.575]$$

- (b) Using the point estimate of p obtained from the preliminary sample, what sample size is needed to be 95% confident that the error in estimating the true value of p is less than 0.02?
- (c) How large must the sample be if you wish to be at least 95% confident that the error in estimating p is less than 0.02, regardless of the true value of p ?

$$n \geq \left(\frac{z_{\alpha/2}}{E} \right)^2 \cdot p(1-p)$$

\uparrow $z_{0.05} = 1.96$
 \downarrow Margin of error

$p(1-p)$ reaches maximum
at $p = 0.5$

b) $\hat{p} = 0.4$

$$n \geq \left(\frac{1.96}{0.02} \right)^2 \cdot 0.4 \cdot 0.6$$

$$\geq 2304.96$$

c) $n \geq \left(\frac{1.96}{0.02} \right)^2 \cdot 0.5 \cdot 0.5$

$$\geq 2401$$