

# ECON 147 MIDTERM EXAM

*YOUR NAME:*

Spring 2018, May 7th  
12:30pm - 1:45pm

## **Instruction**

- This is a closed book and closed note exam. All necessary information will be provided, so no cheat sheet is needed.
- Try to answer all questions and write all answers **within this exam sheets**. You will hand in this exam sheets. Please write legibly.
- *Examination Rules from Department Policy will be strictly followed.*

1. **(30 pts)** Consider the following joint distribution of X and Y:

| Bivariate pmf |   |     |     |     |
|---------------|---|-----|-----|-----|
|               |   | Y   |     |     |
|               |   | 0   | 1   |     |
| X             | 0 | 1/4 | 1/8 | 3/8 |
|               | 1 | 1/8 | 1/8 | 2/8 |
|               | 2 | 1/8 | 1/4 | 3/8 |
|               |   | 1/2 | 1/2 |     |

- 1.1. Find the marginal probability mass functions of X and Y. Using these distributions, compute  $E[X]$ ,  $Var(X)$ ,  $E[Y]$  and  $Var(Y)$ . **[8 pts]**

$$E[X] = 0 \times \frac{3}{8} + 1 \times \frac{2}{8} + 2 \times \frac{3}{8} = 1; \text{ (2pts)}$$

$$E[X^2] = 0^2 \times \frac{3}{8} + 1^2 \times \frac{2}{8} + 2^2 \times \frac{3}{8} = \frac{7}{4};$$

$$Var(X) = E[X^2] - (E[X])^2 = \frac{3}{4}; \text{ (2pts)}$$

$$E[Y] = 0 \times \frac{1}{2} + 1 \times \frac{1}{2} = \frac{1}{2}; \text{ (2pts)}$$

$$E[Y^2] = 0^2 \times \frac{1}{2} + 1^2 \times \frac{1}{2} = \frac{1}{2};$$

$$Var(Y) = E[Y^2] - (E[Y])^2 = \frac{1}{4}. \text{ (2pts)}$$

- 1.2. Compute the conditional pmf  $f(x|Y = 0)$  and  $f(x|Y = 1)$ , and also write down the marginal pmf  $f(x)$ . **[5 pts]**

$$f(x|Y = 0) = \begin{cases} 1/2, & x = 0 \\ 1/4, & x = 1 \\ 1/4, & x = 2 \end{cases}; \text{ (2 pts)}$$

$$f(x|Y = 1) = \begin{cases} 1/4, & x = 0 \\ 1/4, & x = 1 \\ 1/2, & x = 2 \end{cases}; \text{ (2 pts)}$$

$$f(x) = \begin{cases} 3/8, & x = 0 \\ 2/8, & x = 1 \\ 3/8, & x = 2 \end{cases}. \text{ (1 pts)}$$

1.3. Are  $X$  and  $Y$  independent? Fully justify your answer. **[5 pts]**

They are not independent since:  
 $f(x|Y = 0) \neq f(x)$  and  $f(x|Y = 1) \neq f(x)$ .

1.4. Compute  $E[X|Y = 0]$ ,  $Var[X|Y = 0]$  and  $Cov(X, Y)$ . **[6 pts]**

$$\begin{aligned} E[X|Y = 0] &= \sum_{x \in S_x} x f(x|Y = 0) \\ &= 0 \times \frac{1}{2} + 1 \times \frac{1}{4} + 2 \times \frac{1}{4} = \frac{3}{4}; \quad \textbf{(2 pts)} \\ E[X^2|Y = 0] &= \sum_{x \in S_x} x^2 f(x|Y = 0) \\ &= 0^2 \times \frac{1}{2} + 1^2 \times \frac{1}{4} + 2^2 \times \frac{1}{4} = \frac{5}{4}; \\ Var[X|Y = 0] &= E[X^2|Y = 0] - (E[X|Y = 0])^2 = \frac{5}{4} - \frac{9}{16} = \frac{11}{16}; \quad \textbf{(2 pts)} \\ E[XY] &= \sum_{x \in S_x} \sum_{y \in S_y} xy f(x, y) = \frac{5}{8}; \\ Cov(X, Y) &= E[XY] - E[X]E[Y] = \frac{5}{8} - 1 \times \frac{1}{2} = \frac{1}{8}. \quad \textbf{(2 pts)} \end{aligned}$$

- 1.5. Now suppose  $X$  and  $Y$  represent outcomes of two lotteries, and you would like to invest your \$100 to these lotteries. Assume you are extremely risk averse, hence only care about the risk (preferably low variance) of your investment. How do you want to distribute your \$100, i.e., how do you choose your  $\alpha \in [0, 1]$  for  $\alpha X + (1 - \alpha)Y$ ? **[6 pts]**

$$\begin{aligned} Var(\alpha X + (1 - \alpha)Y) &= \alpha^2 \sigma_x^2 + (1 - \alpha)^2 \sigma_y^2 + 2\alpha(1 - \alpha)\sigma_{x,y} \\ &= \frac{3}{4}\alpha^2 + \frac{1}{4}(1 - \alpha)^2 + 2\alpha(1 - \alpha)\frac{1}{8} \end{aligned}$$

which implies that

$$\begin{aligned} \frac{\partial Var(\alpha X + (1 - \alpha)Y)}{\partial \alpha} &= \frac{3}{2}\alpha - \frac{1}{2}(1 - \alpha) + \frac{1}{4}(1 - 2\alpha) \\ &= 2\alpha - \frac{1}{2} - \frac{1}{2}\alpha + \frac{1}{4} \\ &= \frac{3\alpha}{2} - \frac{1}{4}. \end{aligned}$$

Therefore, the portfolio with smallest variance will be  $\alpha = \frac{1}{6}$ . That is, put 100/6 on X and 500/6 on Y.

2. **(25 pts)** Consider a 3-year period and that there are 3 mutual funds. The performance of each mutual fund relative to market is random in the sense that each fund has a 50-50 chance of outperforming the market in any year and that performance is independent from year to year, and across each fund. Let  $t = 1, \dots, 3$  (year) and  $i = 1, \dots, 3$  (mutual funds), and

$$Y_{it} = \begin{cases} 1, & \text{if fund } i \text{ outperforms the market in year } t \\ 0, & \text{otherwise.} \end{cases}$$

- 2.1. What is the probability that at least one fund outperforms the market in all three years? **[8 pts]**

**Answer:** The probability that fund  $i$  outperforms the market in all three years is  $(0.5)^3 = \frac{1}{8}$ . The probability that fund  $i$  fails to outperform the market in all three years is  $1 - \frac{1}{8} = 7/8$ . Therefore, none of the funds outperform the market in all three years is  $(7/8)^3$  and the probability that at least one fund outperforms the market in all three years is

$$1 - (7/8)^3 = 0.33.$$

- 2.2. Find the  $E[Y_{it}]$  and  $Var(Y_{it})$ . **[3 pts]**

**Answer:**

$$\begin{aligned} E[Y_{it}] &= 1 * \frac{1}{2} + 0 * \frac{1}{2} = \frac{1}{2}; \\ Var(Y_{it}) &= E[Y_{it}^2] - (E[Y_{it}])^2 = \frac{1}{2} - \frac{1}{4} = \frac{1}{4} \end{aligned}$$

- 2.3. Now assume that  $P(Y_{it} = 1) = p_i$  is unknown. Given a random sample  $\{Y_{i1}, Y_{i2}, Y_{i3}\}$ , we may want to use  $\hat{p}_i = \frac{1}{3} \sum_{t=1}^3 Y_{it}$  as an estimator for  $p_i$ . Show that  $\hat{p}_i$  is an unbiased estimator for  $p_i$ . **[4 pts]**

**Answer:** By the linearity of the expectation operator:

$$E[\hat{p}_i] = \frac{1}{3} \sum_{t=1}^3 E[Y_{it}] = \frac{1}{3} \sum_{t=1}^3 p_i = p_i.$$

- 2.4. Suppose we are interested in estimating the odds ratio  $\gamma = \frac{p_i}{1-p_i}$ , then a natural estimator is  $\hat{\gamma} = \frac{\hat{p}_i}{1-\hat{p}_i}$ . Is  $\hat{\gamma}$  an unbiased estimator for  $\gamma$ ? Explain why or why not. [6 pts]

**Answer:** Biased estimator.

$$\hat{\gamma} = \frac{\hat{p}_i}{1-\hat{p}_i} = \frac{1}{1-\hat{p}_i} - 1.$$

Since  $(1-x)^{-1}$  is a nonlinear function of  $x$ ,

$$E \left[ \frac{1}{1-\hat{p}_i} \right] \neq \frac{1}{1-E[\hat{p}_i]}$$

where  $E[\hat{p}_i] = p_i$  as we show in 2.3. Therefore,

$$E[\hat{\gamma}] = E \left[ \frac{1}{1-\hat{p}_i} \right] - 1 \neq \frac{1}{1-E[\hat{p}_i]} - 1 = p$$

- 2.5. Now assume that  $t = 1, \dots, 100$  (years) which is large enough (sample period is  $T = 100$ ), and that  $\hat{p}_i = 0.5$ . Construct an 95% asymptotic confidence interval for  $p_i$ . (You can use the fact  $SE(\hat{p}_i) = \sqrt{\frac{\hat{p}_i(1-\hat{p}_i)}{100}}$  and  $q_{0.025}^Z = -2$ ). [4 pts]

**Answer:** Using the formula

$$\hat{p}_i \pm q_{0.975}^Z \cdot SE(\hat{p}_i)$$

we get

$$\begin{aligned} 0.5 \pm 2 \sqrt{\frac{0.5(1-0.5)}{100}} &= 0.5 \pm 0.1; \\ \text{or } 0.5 \pm 1.96 \sqrt{\frac{0.5(1-0.5)}{100}} &= 0.5 \pm 0.098. \end{aligned}$$

3. **(30 pts)** Let  $Y_1, Y_2, Y_3, Y_4$  and  $Y_5$  be iid  $(\mu, \sigma^2)$ .
- 3.1. Let  $W_1 = \frac{1}{8}Y_1 + \frac{1}{8}Y_2 + \frac{1}{2}Y_3 + \frac{1}{8}Y_4 + \frac{1}{8}Y_5$ . Find the mean and variance of  $W_1$ ? **[4 pts]**

**Answer:** By definition

$$\begin{aligned} E[W_1] &= \left( \frac{1}{8} + \frac{1}{8} + \frac{1}{2} + \frac{1}{8} + \frac{1}{8} \right) \mu = \mu, \text{ (2 pts)} \\ \text{Var}(W_1) &= \left( \frac{1}{64} + \frac{1}{64} + \frac{1}{4} + \frac{1}{64} + \frac{1}{64} \right) \sigma^2 = \frac{5\sigma^2}{16} \text{ (2 pts)} \end{aligned}$$

- 3.2. Let  $W_2 = \frac{1}{4}Y_1 + \frac{1}{4}Y_2 + \frac{1}{5}Y_3 + \frac{1}{4}Y_4 + \frac{1}{4}Y_5$ . Find the mean and variance of  $W_2$ ? **[4 pts]**

**Answer:** By definition

$$\begin{aligned} E[W_2] &= \left( \frac{1}{4} + \frac{1}{4} + \frac{1}{5} + \frac{1}{4} + \frac{1}{4} \right) \mu = \frac{6}{5}\mu, \text{ (2 pts)} \\ \text{Var}(W_2) &= \left( \frac{1}{16} + \frac{1}{16} + \frac{1}{25} + \frac{1}{16} + \frac{1}{16} \right) \sigma^2 = \frac{29\sigma^2}{100} \text{ (2 pts)} \end{aligned}$$

- 3.3. Which estimator of  $\mu$  ( $W_1$  or  $W_2$ ) do you prefer? Fully justify your answer. **[13 pts]**

**Answer:** By definition

$$\begin{aligned} \text{MSE}(W_1) &= \frac{5}{16}\sigma^2 \text{ (3 pts)} \\ \text{MSE}(W_2) &= \frac{\mu^2}{25} + \frac{29}{100}\sigma^2 \text{ (3 pts)} \end{aligned}$$

Therefore

$$\begin{aligned} \text{MSE}(W_1) - \text{MSE}(W_2) &= \frac{125 - 116}{400}\sigma^2 - \frac{\mu^2}{25} \\ &= \frac{9}{400}\sigma^2 - \frac{\mu^2}{25} \text{ (3 pts)} \end{aligned}$$

If  $\frac{9}{400}\sigma^2 > \frac{\mu^2}{25}$ , we prefer  $W_2$ . If  $\frac{9}{400}\sigma^2 < \frac{\mu^2}{25}$ , we prefer  $W_1$ . If  $\frac{9}{400}\sigma^2 = \frac{\mu^2}{25}$ , there is no difference between these two estimators. **(4 pts)**

- 3.4. An estimator  $W_a$  is called a linear estimator of  $\mu$  if it takes the following form

$$W = a_1 Y_1 + a_2 Y_2 + a_3 Y_3 + a_4 Y_4 + a_5 Y_5.$$

Among all possible unbiased linear estimators of  $\mu$  based on  $Y_1, Y_2, Y_3, Y_4$  and  $Y_5$ , does there exist a best (unbiased) estimator which has smallest mean square error? Fully justify your answer. [Hint: the inequality  $(\sum_{i=1}^5 a_i)^2 \leq 5 \sum_{i=1}^5 a_i^2$  will be useful. ] **[9 pts]**

**Answer:** for any linear unbiased estimator,

$$\begin{aligned} Var(W) &= \left( \sum_{i=1}^5 a_i^2 \right) \sigma^2 \text{ (1 pt)} \\ E[W] &= \mu \text{ (1 pt)} \\ MSE(W) &= \left( \sum_{i=1}^5 a_i^2 \right) \sigma^2 \text{ (1 pt, or 3 pts if directly write MSE of W).} \end{aligned}$$

While for the sample average

$$\begin{aligned} Var(\bar{Y}) &= \frac{\sigma^2}{n} \text{ (1 pt)} \\ E[\bar{Y}] &= \mu \text{ (1 pt)} \\ MSE(\bar{Y}) &= \frac{\sigma^2}{n} \text{ (1 pt, or 3 pts if directly write MSE of } \bar{Y} \text{).} \end{aligned}$$

Since  $\frac{\sigma^2}{5} \leq (\sum_{i=1}^5 a_i^2) \sigma^2$  by the inequality in hint,

$$Var(\bar{Y}) \leq Var(W_a)$$

so  $\bar{Y}$  is the *best* in MSE sense **(2 pts)**.



4. **(15 pts, 5pts each)** Indicate whether the following statements are true or false (circle one). Briefly discuss why it is so.
- 4.1. If  $r_t$  is *continuously compounded (cc)* monthly return at month  $t$  for  $t = 1, 2, \dots, 12$ , then the annualized cc return is  $r_A = 12r_t$ .

True

False

Why?

**Answer: False.** The annualized cc return is  $r_A = r_1 + r_2 + \dots + r_{12}$ .

- 4.2. Let  $r_{GS,t}$  and  $r_{AIG,t}$  be *cc* 1-month returns for Goldman Sachs Group (GS) and American International Group (AIG). If we construct a portfolio using the share  $\alpha \in [0, 1]$  for GS, the portfolio cc return is  $r_{p,t} = \alpha r_{GS,t} + (1 - \alpha) r_{AIG,t}$ .

True

False

Why?

**Answer: False.** Because cc return is not additive across portfolio shares.

- 4.3. In 5.2., if 5% quantile of the portfolio *simple return* is given as  $q_{0.05}^{R_p} = -0.5$ , then 5% monthly Value-at-Risk for the \$10,000 investment in this portfolio is  $\$10,000 \times (-0.5) = -\$5,000$ .

**Answer: True.** Because by the definition of VaR.

5. Bonus Questions (**5 pts, 1 puts each**). Briefly explain what are the following R commands:

5.1. colnames

**Answer:** Show the column names of data

5.2. tail

**Answer:** Show the last 6 observations of the data

5.3. rmvnorm

**Answer:** generate multivariate normal random variable

5.4. cumprod

**Answer:** calculate the commulative products

5.5. qnorm

**Answer:** calculate the quantile of normal random variable