

MODEL ANSWERS OF QUIZ II

[5<sup>pnts.</sup>] 1. Show that

$$\lim_{n \rightarrow \infty} e^{-n} \sum_{k=0}^n \frac{n^k}{k!} = \frac{1}{2}.$$

**Solution:** Consider a sequence of *i.i.d.* random variables  $\{X_n\}$ , where  $X_n \sim \text{Poisson}(1)$  for all  $n = 1, 2, \dots$ . Then  $E(X_n) = \text{Var}(X_n) = 1$  and  $\sum_{i=1}^n X_i \sim \text{Poisson}(n)$  for all  $n$ . Now, using central limit theorem,  $\sqrt{n}(\bar{X}_n - 1) \rightarrow Z \sim N(0, 1)$  in distribution.

$$\begin{aligned} \lim_{n \rightarrow \infty} e^{-n} \sum_{k=0}^n \frac{n^k}{k!} &= \lim_{n \rightarrow \infty} P\left(\sum_{i=1}^n X_i \leq n\right) \\ &= \lim_{n \rightarrow \infty} P(\bar{X}_n \leq 1) \\ &= \lim_{n \rightarrow \infty} P(\sqrt{n}(\bar{X}_n - 1) \leq 0) \\ &= \frac{1}{2}. \end{aligned}$$

[5<sup>pnts.</sup>] 2. A probability class takes two exams, Exam 1 and Exam 2. Let  $X$  and  $Y$  denote the scores in Exam 1 and Exam 2, respectively. Suppose  $(X, Y)$  follow a bivariate normal distribution with parameters  $\mu_X = 70, \mu_Y = 60, \sigma_X = 10, \sigma_Y = 15$  and  $\rho = 0.6$ . If we select a student at random, what is the probability (in term of  $\Phi(\cdot)$ , the cumulative distribution function of a  $N(0, 1)$  random variable) that

(a) the student scores more than 75 in Exam 2, given that the student scored 80 in Exam 1?

**Solution:** We need to find  $P(Y > 75 | X = 80)$ . It is known that the conditional distribution of  $Y$  given  $X = x$  is normal with mean  $\mu_Y + \rho \frac{\sigma_Y}{\sigma_X}(x - \mu_X)$  and variance  $\sigma_X^2(1 - \rho^2)$ . Here  $x = 80$ , and  $Y | X = 80 \sim N(69, 144)$ .

Hence the required probability is

$$\begin{aligned} P(Y > 75 | X = 80) &= P\left(\frac{Y - 69}{12} > \frac{75 - 69}{12} \middle| X = 80\right) \\ &= 1 - \Phi(0.5) \text{ or } \Phi(-0.5). \end{aligned}$$

(b) the sum of his/her Exam 1 and Exam 2 scores is over 150?

**Solution:** Here  $X + Y \sim N(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2 + 2\rho\sigma_X\sigma_Y) \equiv N(130, 505)$ . Hence the required probability is

$$\begin{aligned} P(X + Y > 150) &= P\left(\frac{X + Y - 130}{\sqrt{505}} > \frac{20}{\sqrt{505}}\right) \\ &= 1 - \Phi(0.89) \text{ or } \Phi(-0.89). \end{aligned}$$

- [5<sup>pnts.</sup>] 3. Let  $\{X_n\}$  and  $X$  be random variables defined on a probability space  $(\mathcal{S}, \mathcal{F}, P)$ . Suppose  $X_n$  converges to  $X$  in the 1st mean. Define  $f : \mathbb{R} \rightarrow \mathbb{R}$  by  $f(x) = \log(1 + e^x)$ . Show that  $f(X_n)$  converges to  $f(X)$  in the 1st mean.

**Solution:** Here  $f(\cdot)$  is differentiable and  $f'(x) = \frac{e^x}{1+e^x} \implies |f'(x)| \leq 1$ . Fix  $x < y$ . Then, using mean value theorem, there exists a  $c \in (x, y)$  such that

$$\begin{aligned}\frac{f(x) - f(y)}{x - y} &= f'(c) \\ \implies \frac{|f(x) - f(y)|}{|x - y|} &= |f'(c)| \leq 1 \\ \implies |f(x) - f(y)| &\leq |x - y|.\end{aligned}$$

Note that this statement is true for all  $x \in \mathbb{R}$  and  $y \in \mathbb{R}$ .

Hence

$$\begin{aligned}|f(X_n) - f(X)| &\leq |X_n - X| \\ \implies 0 \leq E|f(X_n) - f(X)| &\leq E|X_n - X| \rightarrow 0 \text{ as } n \rightarrow \infty \text{ since } X_n \rightarrow X \text{ in 1st mean} \\ \implies E|f(X_n) - f(X)| &\rightarrow 0 \text{ as } n \rightarrow \infty \\ \implies f(X_n) &\rightarrow f(X) \text{ in 1st mean.}\end{aligned}$$