

2018 Engineering Mathematics II Quiz 02 Solutions

Candidate Avg.		11.378
Max		19
Median		12
Min		0
Total Avg.		11.378

EM2 TAs: Onyu Kang, Jiho Choi

If you have any question regarding with your grade, please contact the following TA.

1. [2, 2, 2, 2, 2 points]

1-a) [2 points] Onyu Kang

1-b) [2 points] Onyu Kang

1-c) [2 points] Onyu Kang

1-d) [2 points] Onyu Kang

1-e) [2 points] Onyu Kang

2. [2, 2 points]

2-A) [2 points] Jiho Choi

2-B) [2 points] Jiho Choi

3. [3, 3 points]

3-A) [3 points] Jiho Choi

3-B) [3 points] Jiho Choi

Total grade points **[20 points]**

- surname ㄱ ~ ㄷ: Onyu Kang,

- surname ㄹ ~ ㅎ: Jiho Choi

1. [2 points each]

1-a) [2 points] - Onyu Kang

Let x be a Geometric random variable. Prove that $E[X] = \sum_{i=1}^{\infty} \Pr(X \geq i)$. Does the equality hold for random variables whose values are integers? Describe the conditions that make the equality holds true.

Solution)

$$\begin{aligned}\sum_{i=1}^{\infty} \Pr(X \geq i) &= \sum_{i=1}^{\infty} \sum_{j=i}^{\infty} \Pr(X = j) \\ &= \sum_{j=1}^{\infty} \sum_{i=1}^j \Pr(X = j) \\ &= \sum_{j=1}^{\infty} j \Pr(X = j) \\ &= E[X].\end{aligned}$$

-----[1 pts]

The condition that makes the equality hold true is that X be a discrete random variable that takes on only nonnegative integer values.

-----[1 pts]

■

Scores were specified for each.

1-b) [2 points] - Onyu Kang

Prove or give a counter example for the following claim; if X and Y are independent random variables, then for any functions g(·) and f(·) $E[g(X) \cdot f(Y)] = E[g(X)] \cdot E[f(Y)]$

Solution)

In continuous independent random variable case,

$$\begin{aligned} E[g(X) \cdot f(Y)] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x)f(y)f_x(x)f_y(y)dx dy \\ &= \int_{-\infty}^{\infty} f_y(y)f(y) \int_{-\infty}^{\infty} g(x)f_x(x)dx dy \\ &= \int_{-\infty}^{\infty} f_y(y)f(y)E[g(X)]dy \\ &= E[g(X)] \int_{-\infty}^{\infty} f_y(y)f(y)dy \\ &= E[g(X)] \cdot E[f(Y)] \end{aligned}$$

In discrete independent random variable case,

$$\begin{aligned} E[g(X) \cdot f(Y)] &= \sum_x \sum_y g(x)f(y)f_x(x)f_y(y) \\ &= \sum_x g(x)f_x(x) \sum_y f(y)f_y(y) \\ &= E[g(X)] \cdot E[f(Y)] \end{aligned}$$

■

If you prove one of the above cases, you will get 2 points.

1-c) [2 points] - Onyu Kang

The MGF of Bernoulli random variable with parameter p is $p \cdot e^t + 1 - p$. Using Theorem 4.3, show that the MGF of Binomial random variable $X \sim B(n, p)$ is $(p \cdot e^t + 1 - p)^n$. Using the above MGF and the propositions, $E[X] = M_X^{(1)}(0)$ and $E[X^2] = M_X^{(2)}(0)$, derive expectation and 2nd moment of X .

Solution)

Since $\{x_1, x_2, \dots, x_i\}$ are independent, $X = x_1 + x_2 + \dots + x_i$.

$$\begin{aligned} M_X(t) &= \prod_{i=1}^n M_{x_i}(t) \\ &= \prod_{i=1}^n (p \cdot e^t + 1 - p) \\ &= (p \cdot e^t + 1 - p)^n \quad \text{-----[1 pt]} \end{aligned}$$

$$\begin{aligned} M_X^{(1)}(t) &= n(p \cdot e^t + 1 - p)^{n-1} (p \cdot e^t) \\ \therefore E[X] &= M_X^{(1)}(0) = np \quad \text{-----[0.5 pts]} \end{aligned}$$

$$\begin{aligned} M_X^{(2)}(t) &= np \cdot e^t (p(n e^t - 1) + 1) (1 + (-1 + e^t)p)^{n-2} \\ \therefore E[X^2] &= M_X^{(2)}(0) = n^2 p^2 - np^2 + np \quad \text{-----[0.5 pts]} \end{aligned}$$

■

Scores were specified for each.

1-d) [2 points] - Onyu Kang

Continuation of C). Let X and Y are independent Binomial random variables; $X \sim B(n, p)$ and $Y \sim B(m, p)$. Define a new random variable $Z = X + Y$ and derive the MGF of Z. Also using the uniqueness property of MGF (Theorem 4.2), show that $Z \sim B(n+m, p)$.

Solution)

By theorem 4.3,

$$\begin{aligned} M_Z(t) &= M_{X+Y}(t) \\ &= M_X(t)M_Y(t) \\ &= (p \cdot e^t + 1 - p)^n (p \cdot e^t + 1 - p)^m \\ &= (p \cdot e^t + 1 - p)^{n+m} \end{aligned} \quad \text{-----[1 pt]}$$

$W \sim B(n+m, p)$

By theorem 4.2,

$$M_W(t) = (p \cdot e^t + 1 - p)^{n+m} = M_Z(t)$$

Then W and Z have the same distribution.

$\therefore Z \sim B(n+m, p)$ -----[1 pt]

■

Scores were specified for each.

1-e) [2 points] - Onyu Kang

The MGF of a Geometric random variable with parameter p is $pe^t \div (1 - (1 - p)e^t)$. Let X and Y are two independent Geometric random variables with the same parameter p . Find the MGF of $Z=X+Y$. Also compute the expectation of Z using $E[Z] = M_Z^{(1)}(0)$.

Solution)

$$\begin{aligned} M_Z(t) &= M_{X+Y}(t) \\ &= M_X(t)M_Y(t) \\ &= \left(\frac{pe^t}{(1-(1-p)e^t)}\right)^2 \end{aligned} \quad \text{-----[1 pt]}$$

$$\begin{aligned} M_Z^{(1)}(t) &= \frac{2p^2e^{2t}}{\left((p-1)e^t+1\right)^3} \\ \therefore E[Z] &= M_Z^{(1)}(0) = \frac{2}{p} \end{aligned} \quad \text{----[1 pt]}$$

Scores were specified for each.

2. [2, 2 points] Memoryless property & Other

2-A) [2 points] Jiho Choi

X is a geometric random variable with success probability p $\rightarrow \Pr(x = n) = (1 - p)^{n-1} * p$

(1 point)

$$\Pr(X > 3 | X > 2) = \frac{\Pr(x > 3 \cap x > 2)}{\Pr(x > 2)} = \frac{\Pr(x > 3)}{\Pr(x > 2)} = \frac{\sum_{i=4}^{\infty} (1-p)^{i-1} * p}{\sum_{i=3}^{\infty} (1-p)^{i-1} * p} = \frac{(1-p)^3}{(1-p)^2} = (1-p)$$

(1 point)

$$\Pr(X > 1) = \sum_{i=2}^{\infty} (1-p)^{i-1} p = \frac{1-p}{1-(1-p)} * p = (1-p)$$

2-B) [2 points] Jiho Choi

$$\Pr(Y > 3 | Y > 2) = \frac{\Pr(Y > 3 \cap Y > 2)}{\Pr(Y > 2)} = \frac{\Pr(Y > 3)}{\Pr(Y > 2)}$$

To find Y which satisfies $\Pr(Y > 3 | Y > 2) = \Pr(Y > 3)$, Y with the property $\Pr(Y > 2) = 1$ is needed.

ex.) $Y = \{3\}, \{3, 4\}, \{3, 4, 5\}$

Any Y with the above property will get full points.

3. [3, 3 points]

3-A) [3points] Jiho Choi

$X \sim B(n, p) \rightarrow E[X] = np, \text{Var}[X] = np(1-p), k \text{ is integer } \geq 2$

(1 point) Markov's Bound: $\Pr(x \geq a) \leq \frac{E[X]}{a}$

$$\Pr\left(X \geq \frac{(k-1)n}{k}\right) \leq \frac{E[X]}{\frac{(k-1)n}{k}} = \frac{np}{\frac{(k-1)n}{k}} = \frac{k}{k-1}p$$

(1 point) Chebyshev's Bound: $\Pr(|X - E[X]| \geq a) \leq \frac{\text{Var}[X]}{a^2}$

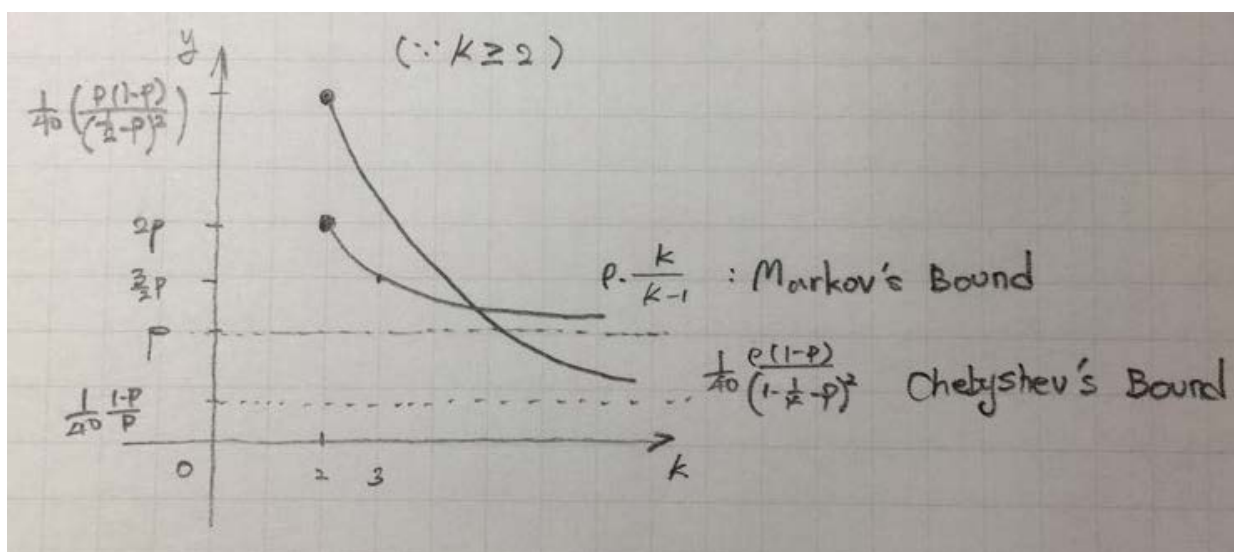
$$\Pr\left(X \geq \frac{(k-1)n}{k}\right)$$

$$\Pr\left(X - E[X] \geq \frac{(k-1)n}{k} - E[X]\right) = \frac{1}{2} * \Pr\left(|X - E[X]| \geq \frac{(k-1)n}{k} - E[X]\right)$$

$$= \frac{1}{2} * \Pr\left(|X - E[X]| \geq \frac{(k-1)n}{k} - np\right) \leq \frac{1}{2} \frac{\text{Var}[X]}{\left(n\left(1 - \frac{1}{k} - p\right)\right)^2} = \frac{1}{2} \frac{np(1-p)}{n^2\left(1 - \frac{1}{k} - p\right)^2} = \frac{1}{2} \frac{p(1-p)}{n\left(1 - \frac{1}{k} - p\right)^2}$$

(1 point)

$n=20$, Markov's Bound: $\frac{k}{k-1}p$, Chebyshev's Bound: $\frac{1}{40} \frac{p(1-p)}{\left(1 - \frac{1}{k} - p\right)^2}$



3-B) [3 points] Jiho Choi

$$\Pr(X - E[X] \geq t * \sigma[X]) \leq \frac{1}{1 + t^2}$$

Using Chebyshev's Bound:

$$\begin{aligned} \Pr(X - E[X] \geq t * \sigma[X]) &= \Pr(X - E[X] + z \geq t * \sigma[X] + z) = \Pr((X - E[X] + z)^2 \geq (t * \sigma[X] + z)^2) \\ &\leq \frac{E[(X - E[X] + z)^2]}{(t\sigma + z)^2} \quad (\because \text{Chebyshev's Inequality}) \\ &= \frac{\sigma^2 + z^2}{(t\sigma + z)^2} \end{aligned}$$

$$f'(z) = 2z(t\sigma + z)^{-2} - 2(\sigma^2 + z^2)(t\sigma + z)^{-2}$$

$$F(z) \text{ is min at } z = \frac{\sigma}{t}, \quad \frac{\sigma^2 + z^2}{(t\sigma + z)^2} = \frac{1}{1 + t^2}$$

Cf) Common Incorrect Answer

Using Chebyshev's Bound: $\Pr(|X - E[X]| \geq a) \leq \frac{\text{Var}[X]}{a^2}$ **not correct for $0 < t < 1$**

↓ need to be symmetric

$$\Pr(X - E[X] \geq t * \sigma[X]) = \frac{1}{2} \Pr(|X - E[X]| \geq t * \sigma[X]) \leq \frac{1}{2} \frac{\text{Var}[X]}{(t * \sigma[X])^2} = \frac{1}{2 * t^2}$$

$$\frac{1}{2 * t^2} \leq \frac{1}{1 + t^2} \quad (\text{for } 1 \leq t)$$