

## Second midterm test of Probability Theory and Mathematical Statistics

Date:

Name:

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### THEORETICAL QUESTIONS

1. Define the 2-dimensional random vector.
2. Define the 2-dimensional discrete random vector.  
What is the range and the joint distribution of such a random vector?
3. State the Cauchy–Bunyakovsky–Schwarz inequality.
4. When do we say that  $\xi_1, \xi_2, \dots, \xi_n$  is an independent sample?
5. State the Central Limit Theorem.

### PROBLEMS

Please round the final results to 4 decimal places!

- 1.-2.** The  $(\xi_1, \xi_2)$  two-dimensional discrete random vector variable is given by its joint distribution

$(\xi_1, \xi_2)$	1	3	5
2	0	$2p$	$3p$
4	$3p$	$2p$	0

Determine the parameter  $p$  and the value of  $\mathbb{D}^2(2\xi_1 + 3\xi_2)$ .

3. Let  $\xi$  be a  $\mathcal{N}(\mu = 2, \sigma^2)$  distributed random variable, where  $\sigma^2$  is unknown. Determine the value of  $\sigma$ , if we know the random variable  $\xi$  falls into the interval  $(1.6, 2.4)$  with probability 0.85.
4. Let  $\xi_1, \xi_2, \dots, \xi_n$  be an  $\text{Exp}(\lambda)$  distributed independent sample. Give the maximal likelihood estimation of the unknown parameter  $\lambda$ . What is the estimation of  $\lambda$  using the method of moments?
5. Let us consider the 6 element sample

2.1, 6.3, 4.6, 6.4, 1.2, 3.5

Calculate the notable statistics, namely the sample mean  $(\bar{\xi})$ , the sample variance  $(s_n^2)$ , the corrected sample variance  $(s_n^{*2})$ , the sample median  $(\text{med}(\xi))$ , and the median absolute deviation  $(\text{MAD}(\xi))$ .

## SOLUTION

1.-2.

$(\xi_1, \xi_2)$	1	3	5
2	0	$2p$	$3p$
4	$3p$	$2p$	0

$2p + 3p + 3p + 2p = 10p = 1$ , thus we have that  $p = 0.1$ .

$(\xi_1, \xi_2)$	1	3	5	
2	0	0.2	0.3	0.5
4	0.3	0.2	0	0.5
	0.3	0.4	0.3	

$\xi_1$

$x_i$	2	4
$p_i$	0.5	0.5

$$\mathbb{E}(\xi_1) = 2 \cdot 0.5 + 4 \cdot 0.5 = 3$$

$$\mathbb{E}(\xi_1^2) = 2^2 \cdot 0.5 + 4^2 \cdot 0.5 = 10$$

$$\mathbb{D}^2(\xi_1) = \mathbb{E}(\xi_1^2) - (\mathbb{E}(\xi_1))^2 = 10 - 3^2 = 1$$

$$\mathbb{D}(\xi_1) = \sqrt{1} = 1.$$

$\xi_2$

$y_j$	1	3	5
$p_j$	0.3	0.4	0.3

$$\mathbb{E}(\xi_2) = 1 \cdot 0.3 + 3 \cdot 0.4 + 5 \cdot 0.3 = 3$$

$$\mathbb{E}(\xi_2^2) = 1^2 \cdot 0.3 + 3^2 \cdot 0.4 + 5^2 \cdot 0.3 = 11.4$$

$$\mathbb{D}^2(\xi_2) = \mathbb{E}(\xi_2^2) - (\mathbb{E}(\xi_2))^2 = 11.4 - 3^2 = 2.4$$

$$\mathbb{D}(\xi_2) = \sqrt{2.4} = 1.5492.$$

$$\mathbb{E}(\xi_1, \xi_2) = 2 \cdot 3 \cdot 0.2 + 2 \cdot 5 \cdot 0.3 + 4 \cdot 1 \cdot 0.3 + 4 \cdot 3 \cdot 0.2 = 7.8$$

$$\text{cov}(\xi_1, \xi_2) = \mathbb{E}(\xi_1 \xi_2) - \mathbb{E}(\xi_1)\mathbb{E}(\xi_2) = 7.8 - 3 \cdot 3 = -1.2$$

$$r(\xi_1, \xi_2) = \frac{\text{cov}(\xi_1, \xi_2)}{\mathbb{D}(\xi_1)\mathbb{D}(\xi_2)} = \frac{-1.2}{1 \cdot 1.5492} = -0.7746.$$

$$\begin{aligned} \mathbb{D}^2(2\xi_1 + 3\xi_2) &= \mathbb{D}^2(2\xi_1) + \mathbb{D}^2(3\xi_2) + 3\text{cov}(2\xi_1 + 3\xi_2) = 4\mathbb{D}^2(\xi_1) + 9\mathbb{D}^2(\xi_2) + 18\text{cov}(\xi_1, \xi_2) = \\ &= 4 \cdot 1 + 9 \cdot 2.4 + 18 \cdot (-1.2) = 4 \end{aligned}$$

3.

$$\begin{aligned} 0.85 &= \mathbb{P}(1.6 < \xi < 2.4) = \mathbb{P}\left(\frac{1.6 - 2}{\sigma} < \eta < \frac{1.6 - 2}{\sigma}\right) = \\ &= \mathbb{P}\left(\frac{-0.4}{\sigma} < \eta < \frac{0.4}{\sigma}\right) = \Phi\left(\frac{0.4}{\sigma}\right) - \Phi\left(-\frac{0.4}{\sigma}\right) = 2\Phi\left(\frac{0.4}{\sigma}\right) - 1 \end{aligned}$$

Thus we have

$$\Phi\left(\frac{0.4}{\sigma}\right) = \frac{1.85}{2} = 0.925.$$

$$\frac{0.4}{\sigma} = 1.44, \quad \Rightarrow \quad \sigma = \frac{0.4}{1.44} = 0.2778.$$

4.  $f(x, \lambda) = \lambda e^{-\lambda \xi_i} \Rightarrow L(\xi, \lambda) = \prod_{i=1}^n \lambda e^{-\lambda \xi_i}$

$$\frac{\partial}{\partial \lambda} l(\lambda) = \frac{\partial}{\partial \lambda} \ln \left( \prod_{i=1}^n \lambda e^{-\lambda \xi_i} \right) = 0$$

$$\frac{\partial}{\partial \lambda} \sum_{i=1}^n (\ln(\lambda) - \lambda \xi_i) = \sum_{i=1}^n \left( \frac{1}{\lambda} - \xi_i \right) = \frac{n}{\lambda} - \sum_{i=1}^n \xi_i = 0$$

Maximum-likelihood estimator:

$$\hat{\xi} = \frac{1}{\frac{1}{n} \sum_{i=1}^n \xi_i} = \frac{1}{\bar{\xi}}$$

Estimator using the method of moments

$$\mathbb{E}(\xi) = \frac{1}{\lambda} \quad \Rightarrow \quad \lambda = \frac{1}{\mathbb{E}(\xi)} \quad \Rightarrow \quad \frac{1}{\lambda} = \frac{1}{\bar{\xi}}$$

5.

$$\begin{aligned} \bar{\xi} &= \frac{1}{n} \sum \xi_i = \frac{2.1 + 6.3 + 4.6 + 6.4 + 1.2 + 3.5}{6} = 4.0167 \\ m_2 &= \frac{1}{n} \sum \xi_i^2 = \frac{2.1^2 + 6.3^2 + 4.6^2 + 6.4^2 + 1.2^2 + 3.5^2}{6} = 19.985 \\ s_n^2 &= m_2 - (\bar{\xi})^2 = 19.985 - 4.0167^2 = 3.8511 \\ s_n^{*2} &= \frac{n}{n-1} s_n^2 = \frac{6}{5} \cdot 3.8511 = 4.6213. \end{aligned}$$

The ordered sample:

1.2, 2.1, 3.5, 4.6, 6.3, 6.4

$$\text{med}(\xi) = \frac{3.5 + 4.6}{2} = 4.05$$

$$|1.2 - 4.05| = 2.85$$

$$|2.1 - 4.05| = 1.95$$

$$|3.5 - 4.05| = 0.55$$

$$|4.6 - 4.05| = 0.55$$

$$|6.3 - 4.05| = 2.25$$

$$|6.4 - 4.05| = 2.35$$

0.55, 0.55, 1.95, 2.25, 2.35, 2.85

$$\text{MAD}(\xi) = \frac{1.95 + 2.25}{2} = 2.1.$$