

**Probability Theory**  
**2016/17 Semester IIb**  
**Instructor: Daniel Valesin**  
**Final Exam**  
**22/6/2017**  
**Duration: 3 hours**

**Name:** \_\_\_\_\_  
**Student number:** \_\_\_\_\_

---

This exam contains 10 pages (including this cover page) and 7 problems. Enter all requested information on the top of this page.

**Your answers should be written in this booklet. Avoid handing in extra paper.**

You are allowed to have two hand-written sheets of paper and a calculator.

You are required to show your work on each problem.

Do not write on the table below.

Problem	Points	Score
1	14	
2	14	
3	14	
4	14	
5	10	
6	14	
7	10	
Total:	90	



1. (a) (7 points) In how many ways can we distribute  $m$  indistinguishable candies to  $k$  distinguishable children, assuming  $m \geq k$  and each child gets at least one candy?
- (b) (7 points) Half the lamps produced by a factory are of good quality and half of bad quality. The lifetime (in months) of a good-quality lamp is an exponential random variable with parameter  $\beta = 2$ , and the lifetime of a bad-quality lamp is an exponential random variable with  $\beta = 1$ . Assume that you pick a lamp at random in the factory, install it and, after  $t$  months, it is still working. Let  $p(t)$  be the probability that you now attribute to having picked a good-quality lamp. Find the limit of  $p(t)$  as  $t \rightarrow \infty$ .

**Solution.**

- (a) A valid assignment of candies to children is uniquely described by a permutation of symbols: 'o' ( $m - k$  times) and '|' ( $k - 1$  times). The number of such permutations is

$$\frac{(m - k + k - 1)!}{(m - k)!(k - 1)!} = \frac{(m - 1)!}{(m - k)!(k - 1)!}$$

- (b)

$$A = \{\text{Picked good lamp}\}, \quad B_t = \{\text{Lamp survived for time } > t\}.$$

$$\begin{aligned} p(t) = \mathbb{P}(A|B_t) &= \frac{\mathbb{P}(B_t|A) \cdot \mathbb{P}(A)}{\mathbb{P}(B_t|A) \cdot \mathbb{P}(A) + \mathbb{P}(B_t|A^c) \cdot \mathbb{P}(A^c)} \\ &= \frac{e^{-t/2} \cdot \frac{1}{2}}{e^{-t/2} \cdot \frac{1}{2} + e^{-t} \cdot \frac{1}{2}} = \frac{1}{1 + e^{-t/2}} \xrightarrow{t \rightarrow \infty} 1. \end{aligned}$$

2. (a) (7 points) The discrete random variables  $X$  and  $Y$  are jointly distributed as follows.  $X$  follows the Poisson( $\lambda$ ) distribution, and

$$f_{Y|X}(y|x) = \binom{x}{y} p^y (1-p)^{x-y}, \quad x \in \{0, 1, \dots\}, y \in \{0, \dots, x\},$$

where  $p \in (0, 1)$ . Show that the distribution of  $Y$  is Poisson( $\lambda p$ ).

- (b) (7 points) The continuous random variables  $Z$  and  $W$  are independent, with  $Z$  following the exponential distribution with parameter 1 and  $W$  following the (continuous) uniform distribution on  $(0, 1)$ . Find  $\mathbb{P}(Z < W < 3Z)$ .

**Solution.**

- (a) For  $y \in \{0, 1, 2, \dots\}$ ,

$$\begin{aligned} f_Y(y) &= \sum_{x=y}^{\infty} f_X(x) \cdot f_{Y|X}(y|x) = \sum_{x=y}^{\infty} \frac{\lambda^x}{x!} e^{-\lambda} \cdot \frac{x!}{y!(x-y)!} \cdot p^y (1-p)^{x-y} \\ &= \frac{\lambda^y}{y!} \cdot e^{-\lambda} \cdot p^y \cdot \underbrace{\sum_{x=y}^{\infty} \frac{\lambda^{x-y}}{(x-y)!} \cdot (1-p)^{x-y}}_{e^{\lambda(1-p)}} = \frac{(\lambda p)^y}{y!} \cdot e^{-\lambda p}. \end{aligned}$$

- (b)

$$\begin{aligned} \mathbb{P}(Z < W < 3Z) &= \int_0^{\frac{1}{3}} \int_z^{3z} e^{-z} dw dz + \int_{\frac{1}{3}}^1 \int_z^1 e^{-z} dw dz \\ &= \int_0^{\frac{1}{3}} 2z e^{-z} dz + \int_{\frac{1}{3}}^1 (1-z) e^{-z} dz \\ &= -3e^{-\frac{1}{3}} + 2 + e^{-1}. \end{aligned}$$

3. Let  $A$  be a set with  $n$  elements. Assume that we choose a subset of  $A$  at random according to the rule that the probability that a certain subset  $A'$  is chosen is proportional to the number of elements of  $A'$ . Let  $X$  be the number of elements of the subset we choose.

(a) (7 points) Find the probability mass function of  $X$ .

(b) (7 points) Prove that

$$M_X(t) = e^t \left( \frac{e^t + 1}{2} \right)^{n-1}, \quad t \in \mathbb{R}.$$

**Solution.**

(a) For  $A' \subset A$ ,

$$\mathbb{P}(A' \text{ is chosen}) = C \cdot \#A'.$$

To find  $C$ , we compute

$$1 = \sum_{A' \subset A} C \cdot \#A' = C \sum_{k=0}^n \sum_{A': \#A'=k} \#A' = C \sum_{k=0}^n \binom{n}{k} \cdot k = 2^n C \underbrace{\sum_{k=0}^n \binom{n}{k} \cdot k \cdot \frac{1}{2^n}}_{\mathbb{E}(Y) \text{ for } Y \sim \text{Bin}(n, 1/2)} = n2^{n-1}C,$$

so  $C = 1/(n2^{n-1})$ . Hence,

$$f_X(k) = \sum_{A': \#A'=k} \mathbb{P}(A' \text{ is chosen}) = \frac{1}{n2^{n-1}} \sum_{A': \#A'=k} k = \frac{1}{n2^{n-1}} \cdot \binom{n}{k} \cdot k, \quad k \in \{1, \dots, n\}.$$

(b)

$$\begin{aligned} M_X(t) &= \sum_{k=1}^n e^{tk} \cdot f_X(k) = \frac{1}{n2^{n-1}} \sum_{k=1}^n e^{tk} \cdot \binom{n}{k} \cdot k \\ &= \frac{1}{n2^{n-1}} \sum_{k=1}^n e^{tk} \cdot \frac{n!}{(k-1)!(n-k)!} \\ &= \frac{1}{n2^{n-1}} \cdot ne^t \cdot \sum_{k=1}^n e^{t(k-1)} \cdot \frac{(n-1)!}{(k-1)!(n-k)!} \\ &= \frac{e^t}{2^{n-1}} \cdot \underbrace{\sum_{\ell=0}^{n-1} e^{t\ell} \cdot \frac{(n-1)!}{\ell!(n-1-\ell)!}}_{(e^t+1)^{n-1}} = e^t \left( \frac{e^t + 1}{2} \right)^{n-1}. \end{aligned}$$

4. (a) (7 points) A device has lifetime denoted by  $T$ , which follows an exponential distribution with parameter  $\beta = 1.5$ . The device has value  $V = 5$  if it fails before  $t = 3$ ; otherwise, it has value  $V = 2T$ . Find the cumulative distribution function of  $V$ .
- (b) (7 points) Let  $X_1, X_2, \dots, X_n$  be independent random variables, all following an exponential distribution with parameter  $\beta$ . Let

$$Y = \max\{X_1, \dots, X_n\},$$

that is,  $Y$  is equal to the largest among the values  $X_1, \dots, X_n$ . Find the cumulative distribution function of  $Y$ .

- (a) The values that  $V$  can attain are 5 and any number in the interval  $[6, \infty)$ . We have

$$\mathbb{P}(V = 5) = \mathbb{P}(T \leq 3) = \int_0^3 \frac{1}{1.5} e^{-x/1.5} dx = 1 - e^{-2}.$$

If  $x \in [6, \infty)$ , we have

$$\mathbb{P}(V \leq x) = \mathbb{P}(T \leq x/2) = \int_0^{x/2} \frac{1}{1.5} e^{-x/1.5} dx = 1 - e^{-x/3}.$$

Putting these facts together, we obtain

$$F_V(x) = \begin{cases} 0 & \text{if } x < 5; \\ 1 - e^{-2} & \text{if } 5 \leq x < 6; \\ 1 - e^{-x/3} & \text{if } x \geq 6. \end{cases}$$

- (b)

$$F_Y(y) = \mathbb{P}(Y \leq y) = \mathbb{P}(\max\{X_1, \dots, X_n\} \leq y) = \mathbb{P}(\cap\{X_i \leq y\}) = (F_{X_1}(y))^n = \left(1 - e^{-y/\beta}\right)^n.$$

5. Let  $X$  be a random variable with expectation  $\mu_X$  and variance  $\sigma_X^2$ , and  $Y$  a random variable with expectation  $\mu_Y$  and variance  $\sigma_Y^2$ . Let  $\rho_{X,Y}$  be the correlation between  $X$  and  $Y$ . Express the following quantities in terms of  $\mu_X$ ,  $\mu_Y$ ,  $\sigma_X$ ,  $\sigma_Y$ , and  $\rho_{X,Y}$ .

(a) (5 points)  $\text{Var}(X - 3Y)$ ;

(b) (5 points)  $\mathbb{E}((3X - 5)(2Y + 1))$ .

**Solution.**

(a)

$$\text{Var}(X - 3Y) = \text{Cov}(X - 3Y, X - 3Y) = \sigma_X^2 + 9\sigma_Y^2 - 6\rho_{X,Y}\sigma_X\sigma_Y.$$

(b) Note that

$$\text{Cov}(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X) - \mathbb{E}(Y) \implies \mathbb{E}(XY) = \rho_{X,Y}\sigma_X\sigma_Y - \mu_X\mu_Y.$$

Hence,

$$\mathbb{E}((3X - 5)(2Y + 1)) = 6\mathbb{E}(XY) - 5 + 3\mathbb{E}(X) - 10\mathbb{E}(Y) = 6(\rho_{X,Y}\sigma_X\sigma_Y - \mu_X\mu_Y) - 5 + 3\mu_X - 10\mu_Y.$$

6. (a) (7 points) Let  $X_1, X_2, \dots$  be independent and identically distributed random variables with finite expectation and variance. Let  $Z_i = X_i + X_{i+1}$ , for  $i = 1, 2, \dots$ . Show that  $\bar{Z}_n = \frac{1}{n} \sum_{i=1}^n Z_i$  converges in probability to a constant, and identify this constant.
- (b) (7 points) Let  $Y_1, Y_2, \dots$  be independent and identically distributed *positive* random variables with finite expectation  $\mu$  and finite variance  $\sigma^2$ . For each natural number  $n$ , let  $W_n$  be the largest value of  $k$  for which the following inequality holds:

$$\sum_{i=1}^k Y_i \leq n.$$

Show that  $W_n/n$  converges in probability to  $1/\mu$ .

*Hint.* Use the facts that  $\{W_n > x\} \subseteq \{\sum_{i=1}^{\lfloor x \rfloor} Y_i < n\}$ ,  $\{W_n < x\} \subseteq \{\sum_{i=1}^{\lceil x \rceil} Y_i > n\}$ .

**Solution.**

- (a) For each  $i$ ,  $\mathbb{E}(Z_i) = \mathbb{E}(X_i) + \mathbb{E}(X_{i+1}) = 2\mathbb{E}(X_i) =: \mu_Z$ . Let us show that  $\bar{Z}_n$  converges in probability to  $\mu_Z$  as  $n \rightarrow \infty$ . We start computing

$$\text{Var}(\bar{Z}_n) = \frac{1}{n^2} \cdot \text{Var} \left( \sum_{i=1}^n Z_i \right) = \frac{1}{n^2} \cdot \left( \sum_{i=1}^n \text{Var}(Z_i) + 2 \sum_{1 \leq i < j \leq n} \text{Cov}(Z_i, Z_j) \right). \quad (\star)$$

Note that  $\text{Var}(Z_i) = \text{Var}(X_i) + \text{Var}(X_{i+1}) = 2\text{Var}(X_1)$ . Moreover, if  $j > i + 1$ , then  $\text{Cov}(Z_i, Z_j) = 0$ , and

$$\text{Cov}(Z_i, Z_{i+1}) = \text{Cov}(X_i + X_{i+1}, X_{i+1} + X_{i+2}) = \text{Var}(X_{i+1}) = \text{Var}(X_1).$$

Hence, the right-hand side of  $(\star)$  is equal to

$$\frac{1}{n^2} (n \cdot \text{Var}(Z_1) + 2 \cdot (n-1) \cdot \text{Var}(Z_1)) = \frac{3n-2}{n^2} \text{Var}(Z_1).$$

Now, for any  $\varepsilon > 0$ ,

$$\mathbb{P}(|\bar{Z}_n - \mu_Z| > \varepsilon) \leq \frac{\text{Var}(\bar{Z}_n)}{\varepsilon^2} = \frac{1}{\varepsilon^2} \cdot \frac{3n-2}{n^2} \xrightarrow{n \rightarrow \infty} 0.$$

- (b) Fix  $\varepsilon > 0$ . We have

$$\mathbb{P} \left( \frac{W_n}{n} > \frac{1}{\mu} + \varepsilon \right) = \mathbb{P} \left( W_n > n \left( \frac{1}{\mu} + \varepsilon \right) \right) \leq \mathbb{P} \left( \sum_{i=1}^{\lfloor n(\frac{1}{\mu} + \varepsilon) \rfloor} Y_i < n \right) = \mathbb{P} \left( \frac{\sum_{i=1}^{a_n} Y_i}{a_n} < \frac{n}{a_n} \right),$$

where  $a_n = \lfloor n(\frac{1}{\mu} + \varepsilon) \rfloor$ . We can find  $\varepsilon' > 0$  such that, for  $n$  large enough,  $\frac{n}{a_n} < \mu - \varepsilon'$ , so that

$$\mathbb{P} \left( \frac{\sum_{i=1}^{a_n} Y_i}{a_n} < \frac{n}{a_n} \right) \leq \mathbb{P} \left( \frac{\sum_{i=1}^{a_n} Y_i}{a_n} < \mu - \varepsilon' \right) \xrightarrow{n \rightarrow \infty} 0$$

by the Weak Law of Large Numbers.



We now turn to

$$\mathbb{P}\left(\frac{W_n}{n} < \frac{1}{\mu} - \varepsilon\right) = \mathbb{P}\left(W_n < n\left(\frac{1}{\mu} - \varepsilon\right)\right) \leq \mathbb{P}\left(\sum_{i=1}^{\lceil n(\frac{1}{\mu} - \varepsilon) \rceil} Y_i > n\right) = \mathbb{P}\left(\frac{\sum_{i=1}^{b_n} Y_i}{b_n} > \frac{n}{b_n}\right),$$

where  $b_n = \lceil n(\frac{1}{\mu} - \varepsilon) \rceil$ . We can find  $\varepsilon'' > 0$  such that, for  $n$  large enough,  $\frac{n}{b_n} > \mu + \varepsilon''$ , so that

$$\mathbb{P}\left(\frac{\sum_{i=1}^{b_n} Y_i}{b_n} > \frac{n}{b_n}\right) \leq \mathbb{P}\left(\frac{\sum_{i=1}^{b_n} Y_i}{b_n} > \mu + \varepsilon''\right) \xrightarrow{n \rightarrow \infty} 0,$$

again by the Weak Law of Large Numbers.

7. (10 points) At each time instant  $t = 0, 1, 2, \dots$ , a particle occupies a position in the set  $\mathbb{Z} = \{\dots, -2, -1, 0, 1, 2, \dots\}$ . Assume that the initial position is 0 and, from time  $t$  to time  $t + 1$ , the particle moves:

- one unit to the left with probability  $1/3$ ;
- one unit to the right with probability  $1/6$ ;
- two units to the right with probability  $1/2$ .

Find a value  $k$  such that the probability that the particle is to the left of  $k$  at time 10000 is approximately 70%.

If you don't have a calculator, you may use the approximation:  $\sqrt{65} \approx 8.06$ .

**Solution.** Let  $Y_1, Y_2, \dots$  be independent and identically distributed random variables with probability mass function

$$f(-1) = \frac{1}{3}, \quad f(1) = \frac{1}{6}, \quad f(2) = \frac{1}{2}.$$

Then, let  $W_0 = 0$  and  $W_n = \sum_{i=1}^n Y_i$  for  $n \geq 1$ . The sequence  $W_0, W_1, \dots$  thus represents the successive positions of the particle. Note that the expectation and variance of the  $Y_i$ 's is respectively

$$\begin{aligned} \mu &= \frac{1}{3}(-1) + \frac{1}{6} \cdot 1 + \frac{1}{2} \cdot 2 = \frac{5}{6}, \\ \sigma^2 &= \mathbb{E}(Y_i^2) - \left(\frac{5}{6}\right)^2 = \frac{1}{3} \cdot (-1)^2 + \frac{1}{6} \cdot 1^2 + \frac{1}{2} \cdot 2^2 - \frac{25}{36} = \frac{65}{36}. \end{aligned}$$

We want  $k$  such that

$$\mathbb{P}(W_{10^4} \leq k) \approx 0.7.$$

By the Central Limit Theorem,

$$\frac{W_{10^4} - 10^4 \cdot \mu}{10^2 \cdot \sigma} \approx Z \sim \mathcal{N}(0, 1),$$

so

$$\mathbb{P}(W_{10^4} \leq k) = \mathbb{P}\left(\frac{W_{10^4} - 10^4 \cdot \mu}{10^2 \cdot \sigma} \leq \frac{k - 10^4 \cdot \mu}{10^2 \cdot \sigma}\right) \approx \mathbb{P}\left(Z \leq \frac{k - 10^4 \cdot \mu}{10^2 \cdot \sigma}\right).$$

The table for the cumulative distribution function of  $\mathcal{N}(0, 1)$  gives

$$\mathbb{P}(Z \leq 0.53) \approx 0.7,$$

so we set

$$\frac{k - 10^4 \cdot \mu}{10^2 \cdot \sigma} = 0.53 \quad \implies \quad k = 10^2 \cdot \sqrt{\frac{65}{36}} \cdot 0.53 + 10^4 \cdot \frac{5}{6} \approx 8404.55.$$