

ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE  
School of Computer and Communication Sciences

Principles of Digital Communications:  
Summer Semester 2012

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Due date: Feb 29, 2012

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## Solution of Homework 2

**Problem 1.** (*Conditioning Technique*)

1. We have

$$\begin{aligned} E\{Y|N = k\} &= E\{E\{\sum_{i=1}^k X_i|N = k\}\} \\ &= E\{\sum_{i=1}^k E\{X_i|N = k\}\} \\ &= E\{\sum_{i=1}^k E\{X_i\}\} \\ &= \frac{k}{2}, \end{aligned}$$

where we used the independence of  $N$  and  $X_i$ ,  $i = 1, 2, \dots, n$ . Hence  $E\{Y|N\} = \frac{N}{2}$  and using the conditioning we have

$$\begin{aligned} E\{Y\} &= E\{E\{Y|N\}\} \\ &= E\{\frac{N}{2}\} \\ &= \frac{1}{2} \times \frac{n+1}{2} \\ &= \frac{n+1}{4}. \end{aligned}$$

2. Similar to the previous part we have

$$\begin{aligned}
E\{Y^2|N = k\} &= E\left\{\left(\sum_{i=1}^k X_i\right)^2|N = k\right\} \\
&= E\left\{\left(\sum_{i=1}^k X_i\right)^2\right\} \\
&= E\{X_1^2 + X_1X_2 + \dots + X_2^2 + X_1X_2 + \dots\} \\
&= E\left\{\sum_{i=1}^k X_i^2 + \sum_{i=1}^k \sum_{j=1, j \neq i}^k (X_iX_j)\right\} \\
&= kE\{X_1^2\} + k(k-1)E\{X_1\}^2 \\
&= \frac{k}{3} + \frac{k(k-1)}{4} \\
&= \frac{k^2}{4} + \frac{k}{12}.
\end{aligned}$$

Hence  $E\{Y^2|N\} = \frac{N^2}{4} + \frac{N}{12}$ . Taking the expectation with respect to  $N$  and using the formulas

$$\begin{aligned}
\sum_{i=1}^n i &= \frac{n(n+1)}{2}, \\
\sum_{i=1}^n i^2 &= \frac{n(n+1)(2n+1)}{6},
\end{aligned}$$

we obtain

$$\begin{aligned}
E\{Y^2\} &= \frac{(n+1)(2n+1)}{24} + \frac{n+1}{24} \\
&= \frac{(n+1)^2}{12},
\end{aligned}$$

and

$$\begin{aligned}
\text{var}(Y) &= E\{Y^2\} - E\{Y\}^2 \\
&= \frac{(n+1)^2}{12} - \frac{(n+1)^2}{16} \\
&= \frac{(n+1)^2}{48}.
\end{aligned}$$

**Problem 2.** (*Conditioning Technique*)

1. Given  $X = x$ ,  $Y$  is uniformly distributed between 0 and  $x$  hence

$$f_{Y|X}(y|x) = \begin{cases} \frac{1}{x} & 0 \leq y \leq x, 0 \leq x \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

2. We use the Bayes rule to find the marginal distribution of  $Y$ ,  $f_Y(y) = \int_y^1 f_{Y|X}(y|x)f_X(x)dx$ . Notice that for a specific  $Y = y$  the value of  $X$  is always greater than  $y$  hence we have

$$\begin{aligned} f_Y(y) &= \int_y^1 f_{Y|X}(y|x)f_X(x)dx \\ &= \int_y^1 \frac{1}{x}dx \\ &= -\log(y), \end{aligned}$$

where  $y \in [0, 1]$  and 0 otherwise. Using the marginal distribution of  $Y$  we can obtain the  $E\{Y\}$  as follows

$$\begin{aligned} E\{Y\} &= \int y f_Y(y)dy \\ &= -\int_0^1 y \log(y)dy \\ &= -\frac{y^2}{2}(\log(y) - \frac{1}{2})\Big|_0^1 \\ &= \frac{1}{4}. \end{aligned}$$

3. We have  $E\{Y|X = x\} = \frac{x}{2}$ . Using the conditioning on  $X = x$  we have

$$\begin{aligned} E\{Y\} &= E\{E\{Y|X\}\} \\ &= E\{\frac{X}{2}\} \\ &= \frac{1}{2} \times \frac{1}{2} = \frac{1}{4}, \end{aligned}$$

which is the same as the previous part.

**Problem 3.** (*Conditioning Technique*) By Symmetry

$$\begin{aligned} E\{Y|X + Y = z\} &= E\{X|X + Y = z\} \\ &= \frac{1}{2}E\{X + Y|X + Y = z\} \\ &= \frac{z}{2}, \end{aligned}$$

which implies that  $E\{X|X + Y\} = \frac{X+Y}{2}$ .