

# 6.041/6.431 Spring 2007 Quiz 2 Solutions

Wednesday, April 18, 7:30 - 9:30 PM

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YOU ARE TOLD TO DO SO

Name: \_\_\_\_\_

Recitation Instructor: \_\_\_\_\_

TA: \_\_\_\_\_

Question		Score	Out of
0			3
1a	(i)		5
	(ii)		5
	(iii)		6
	(iv)		6
1b			9
1c			9
1d			9
1e			9
2a	(i)		6
	(ii)		6
	(iii)		6
2b	(i)		7
	(ii)		7
2c			7
Your Grade		100	100

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**Problem 1:** Xavier and Wasima are participating in the 6.041 MIT marathon, where race times are defined by random variables<sup>1</sup>. Let  $X$  and  $W$  denote the race time of Xavier and Wasima respectively. All race times are in hours. Assume the race times for Xavier and Wasima are independent (i.e.  $X$  and  $W$  are independent). Xavier's race time,  $X$ , is defined by the following density

$$f_X(x) = \begin{cases} 2c, & \text{if } 2 \leq x < 3, \\ c, & \text{if } 3 \leq x \leq 4, \\ 0, & \text{otherwise,} \end{cases}$$

where  $c$  is an unknown constant. Wasima's race time,  $W$ , is uniformly distributed between 2 and 4 hours. The density of  $W$  is then

$$f_W(w) = \begin{cases} \frac{1}{2}, & \text{if } 2 \leq w \leq 4, \\ 0, & \text{otherwise.} \end{cases}$$

- (a) (i) (5 pts) Find the constant  $c$

*Soln:* The plot for the PDF of  $X$  is shown in Figure 1. The PDF has to integrate to 1, so the area under  $f_X(x)$  is  $2c+c$ , which must equal 1. Therefore  $c = 1/3$ .

Integration of the PDF:

$$\begin{aligned} \int_2^4 f_X(x) dx &= 1 \\ \text{which breaks up to } \int_2^3 2c dx + \int_3^4 c dx &= 1 \\ &= 2c + c = 1 \\ \text{and } c &= 1/3. \end{aligned}$$

- (ii) (5 pts) Compute  $\mathbf{E}[X]$

*Soln:*

$$\begin{aligned} \mathbf{E}[X] &= \int_2^4 x f_X(x) dx = \int_2^3 x \cdot 2/3 dx + \int_3^4 x \cdot 1/3 dx \\ &= 1/3 \cdot (3^2 - 2^2) + 1/6 \cdot (4^2 - 3^2) = 5/3 + 17/6 \\ &= 17/6. \end{aligned}$$

- (iii) (6 pts) Compute  $\mathbf{E}[X^2]$

*Soln:*

$$\begin{aligned} \mathbf{E}[X^2] &= \int_2^4 x^2 f_X(x) dx = \int_2^3 x^2 \cdot 2/3 dx + \int_3^4 x^2 \cdot 1/3 dx \\ &= 2/9 \cdot (3^3 - 2^3) + 1/9 \cdot (4^3 - 3^3) = 38/9 + 37/9 \\ &= 25/3. \end{aligned}$$

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<sup>1</sup>A runner's race time is defined as the time required for a given runner to complete the marathon.

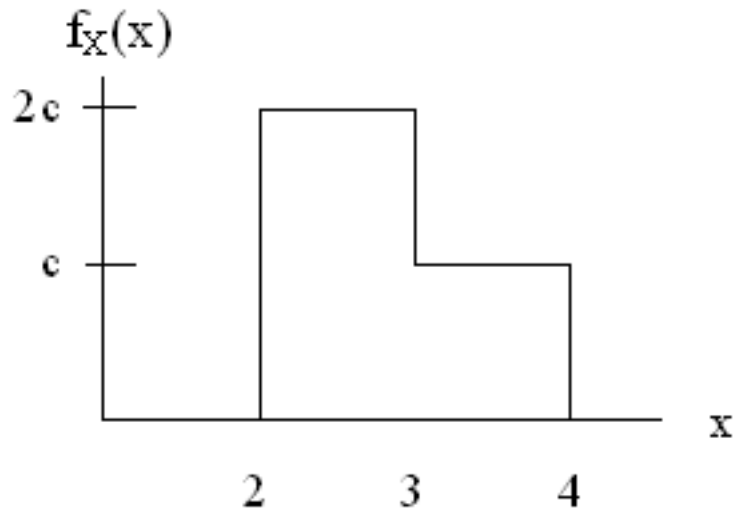


Figure 1: PDF of  $X$

- (iv) (6 pts) Provide a fully labeled sketch of the PDF of  $2X + 1$

*Soln:* Let  $Y = 2X + 1$ . The range of  $Y$  is not from 2 to 4, but now  $5 \leq y \leq 9$ . The shape of the PDF of  $Y$  should look like the PDF of  $X$ , but scaled by a factor such that it normalizes to 1. The range of  $Y$  is double the range of  $X$ , so the density is half. Plot shown below in Figure 2.

Since  $Y = g(X)$  is a linear function of  $X$ , we can use the formula for the derived distribution for a linear function.  $Y = 2X + 1$ , so  $f_Y(y) = \frac{1}{2}f_X(\frac{y-1}{2})$  for  $5 \leq y \leq 9$ . Figure 2 matches this distribution.

- (b) (9 pts) Compute  $\mathbf{P}(X \leq W)$ .

*Soln:*

First we calculate the joint PDF. It should have a non-zero joint density for the region,  $2 \leq x \leq 4$  and  $2 \leq w \leq 4$ . However, it is not uniform within this entire square, as we have seen often in class. Due to the piece-wise uniform density of  $X$ , the square is partitioned into two rectangles of uniform joint densities.  $X$  and  $W$  are independent, so the joint density is just the product of the marginals.

$$\begin{aligned}
 f_{X,W}(x,w) &= f_X(x)f_W(w) \\
 &= f_X(x) \cdot 1/2 \\
 &= \begin{cases} c1 = 2/3 \cdot 1/2 = 1/3 & , \quad 2 \leq x \leq 3, 2 \leq w \leq 4. \\ c2 = 1/3 \cdot 1/2 = 1/6 & , \quad 3 \leq x \leq 4, 2 \leq w \leq 4. \end{cases}
 \end{aligned}$$

Variables  $c1$  and  $c2$  are used to denote the different joint densities, and are shown in the joint plot.

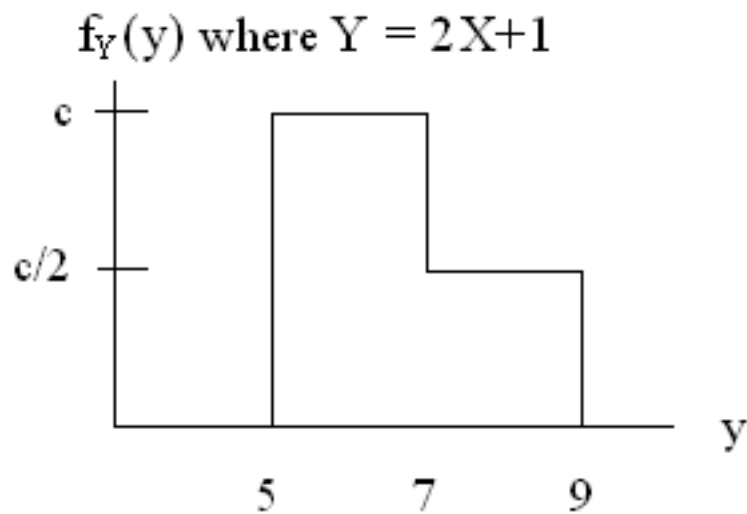


Figure 2: PDF of  $Y = 2X + 1$

As a check, the joint PDF should be normalized to 1, which it is.  
 The joint PDF for  $X$  and  $W$  is shown in Figure 3

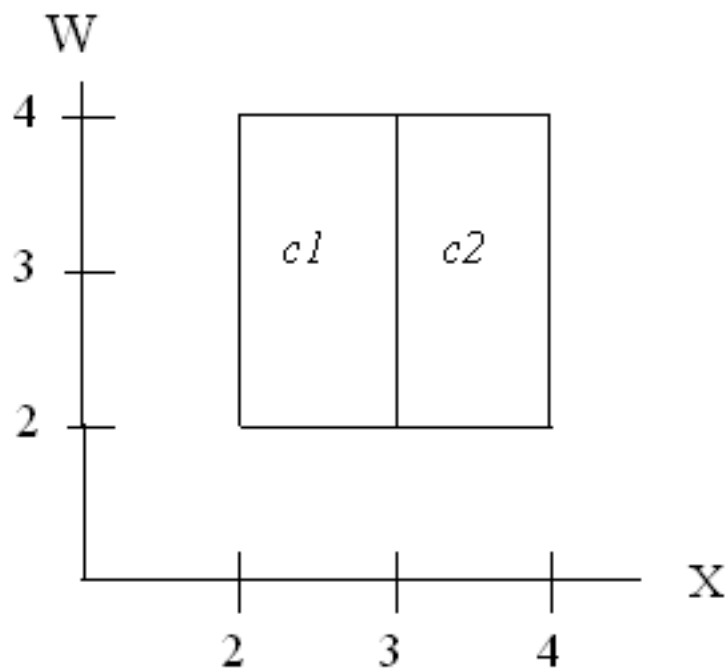


Figure 3: Joint PDF of  $X$  and  $W$

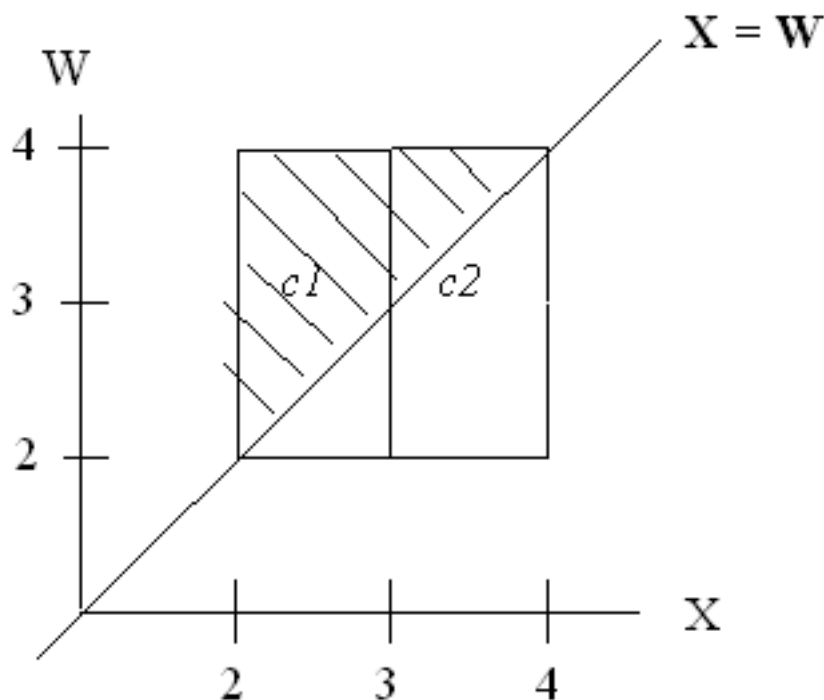


Figure 4:  $\mathbf{P}(X \leq W)$

Looking at the plot of the joint PDF,  $\mathbf{P}(X \leq W)$  is the region above the  $X = W$  line. See Figure 4.

We calculate the weighted area of the shaded region to be:

$$\begin{aligned}\mathbf{P}(X \leq W) &= 1/2 \cdot 1/6 + 3/2 \cdot 1/3 = 1/12 + 1/2 \\ &= 7/12.\end{aligned}$$

The graphical way is the easy solution. Of course, one can integrate:

$$\begin{aligned}\mathbf{P}(X \leq W) &= \int_2^3 \int_x^4 1/3 \, dw dx + \int_3^4 \int_x^4 1/6 \, dw dx \\ &= \frac{1}{3} \int_2^3 (4 - x) \, dx + \frac{1}{6} \int_3^4 (4 - x) \, dx \\ &= 7/12\end{aligned}$$

- (c) (9 pts) Wasima is using a stopwatch to time herself. However, the stopwatch is faulty; it over-estimates her race time by an amount that is uniformly distributed between 0 and  $\frac{1}{10}$

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hours, which is independent of the actual race time. Thus, if  $T$  is the time measured by the stopwatch, then we have

$$f_{T|W}(t|w) = \begin{cases} 10, & \text{if } w \leq t \leq w + \frac{1}{10} \text{ and } 2 \leq w \leq 4, \\ 0, & \text{otherwise.} \end{cases}$$

Find  $f_{W|T}(w|t)$ , when  $t = 3$ .

*Soln:* Be careful here, that  $T$  is the race time measured by the stopwatch, not just the over-estimated race time.

$$f_{W|T}(w|3) = \frac{f_{W,T}(w, 3)}{f_T(3)}$$

where  $f_{W,T}(w, 3) = f_{T|W}(3|w)f_W(w) = 10 \cdot 1/2 = 5$  for  $(3 - 1/10) \leq w \leq 3$ .

$$\text{and } f_T(3) = \int_{3-1/10}^3 f_{W,T}(w, 3) dw = 5 \cdot (1/10) = 1/2.$$

Therefore,

$$f_{W|T}(w|3) = \begin{cases} 10, & \text{if } (3 - 1/10) \leq w \leq 3 \text{ and } t = 3, \\ 0, & \text{otherwise.} \end{cases}$$

- (d) (9 pts) Wasima realizes her stopwatch is faulty and buys a new stopwatch. Unfortunately, the new stopwatch is also faulty; this time, the watch adds random noise  $N$  that is normally distributed with mean  $\mu = \frac{1}{60}$  hours and variance  $\sigma^2 = \frac{4}{3600}$ . Find the probability that the watch over-estimates the actual race time by more than 5 minutes,  $\mathbf{P}(N > \frac{5}{60})$ . For full credit express your final answer as a number.

*Soln:*  $N$  is Normal( $1/60$ ,  $4/3600$ ). We standardize  $N$  to have mean 1 and standard deviation 1 to utilize the Normal table.

$$\begin{aligned} \mathbf{P}(N > \frac{5}{60}) &= 1 - \mathbf{P}(N < \frac{5}{60}) \\ &= 1 - \mathbf{P}(\frac{N - 1/60}{2/60} < \frac{5/60 - 1/60}{2/60}) \\ &= 1 - \Phi(2). \end{aligned}$$

If we looked it up,  $\Phi(2) = 0.9772$ , so  $\mathbf{P}(N > \frac{5}{60}) = 1 - 0.9772 = 0.0028$ .

- (e) (9 pts) Wasima has a sponsor for the marathon! If Wasima finishes the marathon in  $w$  hours, the sponsor pays her  $\frac{24}{w}$  thousand dollars. Define

$$S = \frac{24}{W}$$

Find the PDF of  $S$ .

*Soln:* Use derived distributions to find the CDF of  $S$ , then differentiate with respect to  $s$  to find the PDF of  $S$ .

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The range of  $S$  is determined from the range of  $W$ . Since  $2 \leq w \leq 4$  for a nonzero PDF of  $W$ ,  $24/4 \leq s \leq 24/2$  for a nonzero PDF of  $S$ .

$$\begin{aligned}\mathbf{P}(S \leq s) &= \mathbf{P}(24/W \leq s) = \mathbf{P}(W \geq 24/s) \\ &= 1 - F_W(24/s) = 1 - \int_2^{24/s} f_W(w)dw \\ &= 1 - (12/s - 1) = 2 - 12/s\end{aligned}$$

Taking the derivative with respect to  $s$ ,

$$\begin{aligned}f_S(s) &= \frac{d}{ds}(2 - 12/s) \\ &= \begin{cases} 12/s^2, & \text{if } 6 \leq s \leq 12 \\ 0, & \text{otherwise.} \end{cases}\end{aligned}$$

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**Problem 2.** Consider the following family of **independent** random variables  $N, A_1, B_1, A_2, B_2, \dots$ , where  $N$  is a nonnegative discrete random variable and each  $A_i$  or  $B_i$  is normal with mean 1 and variance 1. Let  $A = \sum_{i=1}^N A_i$  and  $B = \sum_{i=1}^N B_i$ . Recall that the sum of a fixed number of independent normal random variables is normal.

(a) Assume  $N$  is geometrically distributed with a mean of  $1/p$ .

(i) (6 pts) Find the mean,  $\mu_a$ , and the variance,  $\sigma_a^2$ , of  $A$ .

*Soln:* This is a random sums problem so the mean and variance of  $A$  is found using the laws of iterated expectations and total variance.

$$\begin{aligned}\mu_a &= \mathbf{E}[A] = \mathbf{E}[\mathbf{E}[A|N]] = \mathbf{E}[N\mathbf{E}[A_i]] = \mathbf{E}[A_i]\mathbf{E}[N] \\ &= 1/p. \\ \sigma_a^2 &= \text{Var}(A) = \mathbf{E}[\text{Var}(A|N)] + \text{Var}(\mathbf{E}[A|N]) = \mathbf{E}[N \text{Var}(A_i)] + \text{Var}(N\mathbf{E}[A_i]) \\ &= \text{Var}(A_i)\mathbf{E}[N] + \mathbf{E}[A_i]^2\text{Var}(N) = 1/p + (1-p)/p^2 \\ &= 1/p^2.\end{aligned}$$

(ii) (6 pts) Find  $c_{ab}$ , defined by  $c_{ab} = \mathbf{E}[AB]$ .

*Soln:* It is important that although  $A_i$  and  $B_i$  are both Normal, they are NOT the same. Having the same distribution does not make two random variables *equal* to one another. For those who set  $A = B$ , it is incorrect.

Also,  $A$  and  $B$  are not independent... they BOTH depend on a random variable  $N$ ! However, if  $N$  is known, then  $A$  and  $B$  become independent, which is what we make use of in this problem. The trick here is to use iterated expectations, because we are working with random sums.

$$\begin{aligned}c_{ab} = \mathbf{E}[AB] &= \mathbf{E}[(A_1 + A_2 + A_3 + \dots A_N)(B_1 + B_2 + B_3 + \dots B_N)] \\ &= \mathbf{E}[\mathbf{E}[(A_1 + A_2 + A_3 + \dots A_N)(B_1 + B_2 + B_3 + \dots B_N)|N]] \\ &= \mathbf{E}[N\mathbf{E}[A_i]N\mathbf{E}[B_i]] = \mathbf{E}[N^2\mathbf{E}[A_i]\mathbf{E}[B_i]] = \mathbf{E}[A_i]\mathbf{E}[B_i]\mathbf{E}[N^2] \\ &= 1 \cdot 1 \cdot (\text{Var}(N) + \mathbf{E}[N]^2) = (1-p)/p^2 + 1/p^2 \\ &= (2-p)/p^2.\end{aligned}$$

Many of you wrote that  $\mathbf{E}[AB] = \text{cov}(A, B) + \mathbf{E}[A]\mathbf{E}[B]$ . This is technically correct, but the calculation was usually incorrect.  $\text{cov}(A, B)$  is not 1 or 0 (common assumption, not to be confused with  $\text{cov}(A_i, B_j) = 0$ ). It falls out that the covariance between  $A$  and  $B$  is  $(1-p)/p^2 = \text{Var}(N)$ . You can derive this from the law of total variance.

(iii) (6 pts) We observe  $B$  (but not  $N$ ) and wish to estimate  $A$ , using a linear estimator of the form  $c + dB$ , where  $c$  and  $d$  are constants. Find values for  $c$  and  $d$  that result in the smallest possible mean squared error. Express your answer in terms of constants such as  $\mu_a$ ,  $c_{ab}$ , etc., without plugging in the values found in parts (i) and (ii).

*Soln:* Given  $B$ , the linear estimator for  $A$  is

$$\hat{A} = \mathbf{E}[A] + \rho_{A,B} \frac{\sigma_A}{\sigma_B} (B - \mathbf{E}[B]).$$



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Therefore the constants  $c$  and  $d$  are:

$$\begin{aligned}c &= \mathbf{E}[A] - \rho_{A,B} \frac{\sigma_A}{\sigma_B} \mathbf{E}[B], \text{ and} \\d &= \rho_{A,B} \frac{\sigma_A}{\sigma_B} \\ \text{where } \rho_{A,B} &= \frac{\text{cov}(A, B)}{\sigma_A \sigma_B} \\ &= \frac{\mathbf{E}[AB] - \mathbf{E}[A]\mathbf{E}[B]}{\sigma_A \sigma_B}.\end{aligned}$$

Plugging everything in,

$$\begin{aligned}c &= \mu_a - \frac{(c_{ab} - \mu_a \mu_b)}{\sigma_b^2} \mu_b. \\d &= (c_{ab} - \mu_a \mu_b) / \sigma_b^2.\end{aligned}$$

(b) Now assume that  $N$  can take only the values 1 (with probability 1/3) and 2 (with probability 2/3).

(i) (7 pts) Give a formula for the PDF of  $A$ .

*Soln:* If  $N = 1$ ,  $A = A_1$ , which has a Normal distribution with mean 1 and variance 1.

If  $N = 2$ ,  $A = A_1 + A_2$ , which is the sum of two Normals. Therefore the distribution of  $A$  is Normal(1 + 1, 1 + 1) or Normal(2, 2).

Using total probability theorem, we find:

$$\begin{aligned}f_A(a) &= f_{A|N=1}(a)P_N(1) + f_{A|N=2}(a)P_N(2) \\ &= \text{Normal}(1, 1) \cdot 1/3 + \text{Normal}(2, 2) \cdot 2/3 \\ &= \frac{1}{3\sqrt{2\pi}} e^{-(a-1)^2/2} + \frac{2}{3\sqrt{4\pi}} e^{-(a-2)^2/4}.\end{aligned}$$

(ii) (7 pts) Find the conditional probability  $\mathbf{P}(N = 1 \mid A = a)$ .

*Soln:* This is simply:

$$\begin{aligned}\mathbf{P}(N = 1 \mid A = a) &= \frac{\mathbf{P}(A = a, N = 1)\delta}{\mathbf{P}(A = a)\delta} \\ \text{where } \mathbf{P}(A = a)\delta &= f_A(a) \text{ was found in part (a)} \\ \text{and the joint is } P(A = a)P(N = 1)\delta &= f_A(a)P_N(1). \\ \text{Then, } \mathbf{P}(N = 1 \mid A = a) &= \frac{\frac{1}{3\sqrt{2\pi}} e^{-(a-1)^2/2}}{\frac{1}{3\sqrt{2\pi}} e^{-(a-1)^2/2} + \frac{2}{3\sqrt{4\pi}} e^{-(a-2)^2/4}}.\end{aligned}$$

(c) (7 pts) Is it true that  $\mathbf{E}[A \mid N] = \mathbf{E}[A \mid B, N]$ ? Either provide a proof, or an explanation why the equality does not hold.

*Soln:* Yes they are equal.

As a first check, they are both random variables.  $A$  and  $B$  are not independent from one another because they both depend on the RV  $N$  for the random sum. But, if we condition on  $N$ , then  $A$  and  $B$  are independent (hence they are conditionally independent). Is that what the right side of the equation states?

These expectations are equal if the PDFs of  $A|N$  and  $A|(B, N)$  are equal. Once  $N$  is known, knowing  $B$  doesn't change what one knows about  $A$ , so this not only shows that  $A$  and  $B$  are conditionally independent, given  $N$ , but  $A|N$  has the same information as  $A, B|N$ .

Conditional independence of events  $X$  and  $Y$  on  $Z$  is defined as:

$$\begin{aligned}\mathbf{P}(X \cap Y|Z) &= \mathbf{P}(X|Z)\mathbf{P}(Y|Z) \\ &\text{or, equivalently} \\ \mathbf{P}(X|Y \cap Z) &= \mathbf{P}(X|Z)\end{aligned}$$

Therefore, we show that the equality holds here.

$$\begin{aligned}\mathbf{E}[A|N] &= \mathbf{E}[A|B, N] \\ \int a f_{A|N}(a|n) da &= \int a f_{A|B, N}(a|b, n) da\end{aligned}$$

The above statement can be shown to be equal if the probabilities are shown to be the same or the PDFs are derived to be equal:

$$\begin{aligned}f_{A|N}(a|n) &= f_{A|B, N}(a|b, n) = \frac{f_{A, B, N}(a, b, n)}{f_{B, N}(b, n)} \\ &= \frac{f_{A, B|N}(a, b|n)P_N(n)}{f_{B|N}(b|n)P_N(n)} = \frac{f_{A|N}(a|n)f_{B|N}(b|n)}{f_{B|N}(b|n)} \\ &= f_{A|N}(a|n).\end{aligned}$$

So  $\mathbf{E}[A|N] = \mathbf{E}[A|B, N]$  is true.