# Probability and Stochastic Processes: A Friendly Introduction for Electrical and Computer Engineers Roy D. Yates and David J. Goodman

**Problem Solutions**: Yates and Goodman, 3.1.6 3.1.7 3.2.5 3.3.5 3.4.8 3.5.1 3.6.7 3.6.8 3.6.9 3.7.8 3.8.5 and 3.8.6

#### Problem 3.1.6

The joint PMF of X and N is  $P_{N,X}(n,x) = P[N = n, X = x]$ , which is the probability that N = n and X = x. This means that both events must be satisfied. The approach we use is similar to that used in finding the Pascal PMF in Example 2.15. Since X can take on only the two values 0 and 1, let's consider each in turn. When X = 0 that means that a rejection occurred on the last test and that the other n - 1 rejections must have occurred in the previous r - 1 tests. Thus,

$$P_{N,X}(n,0) = \binom{r-1}{n-1} (1-p)^{n-1} p^{r-1-(n-1)} (1-p) \qquad n = 1, \dots, r$$

When X = 1 the last test was acceptable and therefore we know that the  $N = n \le r - 1$  tails must have occurred in the previous r - 1 tests. In this case,

$$P_{N,X}(n,1) = \binom{r-1}{n} (1-p)^n p^{r-1-n} p \qquad n = 0, \dots, r-1$$

We can combine these cases into a single complete expression for the joint PMF.

$$P_{X,N}(x,n) = \begin{cases} \binom{r-1}{n-1}(1-p)^n p^{r-n} & x = 0, n = 1, 2, \dots, r \\ \binom{r-1}{n}(1-p)^n p^{r-n} & x = 1, n = 0, 1, \dots, r-1 \\ 0 & \text{otherwise} \end{cases}$$

#### Problem 3.1.7

Each circuit test produces an acceptable circuit with probability *p*. Let *N* denote the number of rejected circuits that occur in *r* tests and *X* is the number of acceptable circuits before the first reject. The joint PMF,  $P_{N,X}(n,x) = P[N = n, X = x]$  can be found by realizing that  $\{N = n, X = x\}$  occurs if and only if the following events occur:

- A The first *x* tests must be acceptable.
- B Test x + 1 must be a rejection since otherwise we would have x + 1 acceptable at the beginning.
- C The remaining r x 1 tests must contain n 1 rejections.

Since the events *A*, *B* and *C* are independent, the joint PMF for  $x + n \le r$ ,  $x \ge 0$  and  $n \ge 0$  is

$$P_{N,X}(n,x) = \underbrace{p^{x}}_{P[A]} \underbrace{(1-p)}_{P[B]} \underbrace{\binom{r-x-1}{n-1}(1-p)^{n-1}p^{r-x-1-(n-1)}}_{P[C]}$$

After simplifying, a complete expression for the joint PMF is

$$P_{N,X}(n,x) = \begin{cases} \binom{r-x-1}{n-1} p^{r-n} (1-p)^n & x+n \le r, x \ge 0, n \ge 0\\ 0 & \text{otherwise} \end{cases}$$

### Problem 3.2.5

N is

$$P_N(n) = \sum_k P_{N,K}(n,k) = \sum_{k=0}^n \frac{100^n e^{-100}}{(n+1)!} = \frac{100^n e^{-100}}{n!}$$

For  $k = 0, 1, \ldots$ , the marginal PMF of K is

$$P_K(k) = \sum_{n=k}^{\infty} \frac{100^n e^{-100}}{(n+1)!}$$
$$= \frac{1}{100} \sum_{n=k}^{\infty} \frac{100^{n+1} e^{-100}}{(n+1)!}$$
$$= \frac{1}{100} \sum_{n=k}^{\infty} P_N(n+1)$$
$$= P[N > k]/100$$

## Problem 3.3.5

The x, y pairs with nonzero probability are shown in the figure at right. From the figure, we observe that for  $w = 0, 1, \ldots, 10$ ,

$$P[W > w] = P[\min(X, Y) > w]$$
  
=  $P[X > w, Y > w]$   
=  $0.01(10 - w)^2$ 



To find the PMF of *W*, we observe that for w = 1, ..., 10,

$$P_W(w) = P[W > w - 1] - P[W > w]$$
  
= 0.01[(10 - w - 1)<sup>2</sup> - (10 - w)<sup>2</sup>]  
= 0.01(21 - 2w)

The complete expression for the PMF of W is

$$P_W(w) = \begin{cases} 0.01(21 - 2w) & w = 1, 2, \dots, 10 \\ 0 & \text{otherwise} \end{cases}$$

# Problem 3.4.8

For this problem, calculating the marginal PMF of K is not easy. However, the marginal PMF of N is easy to find. For n = 1, 2, ...,

$$P_N(n) = \sum_{k=1}^n \frac{(1-p)^{n-1}p}{n} = (1-p)^{n-1}p$$

That is, N has a geometric PMF. From Appendix A, we note that

$$E[N] = \frac{1}{p}$$
  $\operatorname{Var}[N] = \frac{1-p}{p^2}$ 

We can use these facts to find the second moment of N.

$$E[N^2] = \operatorname{Var}[N] + (E[N])^2 = \frac{2-p}{p^2}$$

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Now we can calculate the moments of K.

$$E[K] = \sum_{n=1}^{\infty} \sum_{k=1}^{n} k \frac{(1-p)^{n-1}p}{n} = \sum_{n=1}^{\infty} \frac{(1-p)^{n-1}p}{n} \sum_{k=1}^{n} k$$

Since  $\sum_{k=1}^{n} k = n(n+1)/2$ ,

$$E[K] = \sum_{n=1}^{\infty} \frac{n+1}{2} (1-p)^{n-1} p = E\left[\frac{N+1}{2}\right] = \frac{1}{2p} + \frac{1}{2}$$

We now can calculate the sum of the moments.

$$E[N+K] = E[N] + E[K] = \frac{3}{2p} + \frac{1}{2}$$

The second moment of K is

$$E[K^{2}] = \sum_{n=1}^{\infty} \sum_{k=1}^{n} k^{2} \frac{(1-p)^{n-1}p}{n} = \sum_{n=1}^{\infty} \frac{(1-p)^{n-1}p}{n} \sum_{k=1}^{n} k^{2}$$

Using the identity  $\sum_{k=1}^{n} k^2 = n(n+1)(2n+1)/6$ , we obtain

$$E[K^{2}] = \sum_{n=1}^{\infty} \frac{(n+1)(2n+1)}{6} (1-p)^{n-1}p = E\left[\frac{(N+1)(2N+1)}{6}\right]$$

Applying the values of E[N] and  $E[N^2]$  found above, we find that

$$E[K^{2}] = \frac{E[N^{2}]}{3} + \frac{E[N]}{2} + \frac{1}{6} = \frac{2}{3p^{2}} + \frac{1}{6p} + \frac{1}{6}$$

Thus, we can calculate the variance of *K*.

Var 
$$[K] = E[K^2] - (E[K])^2 = \frac{5}{12p^2} - \frac{1}{3p} + \frac{5}{12}$$

To find the correlation of N and K,

$$E[NK] = \sum_{n=1}^{\infty} \sum_{k=1}^{n} nk \frac{(1-p)^{n-1}p}{n} = \sum_{n=1}^{\infty} (1-p)^{n-1}p \sum_{k=1}^{n} k$$

Since  $\sum_{k=1}^{n} k = n(n+1)/2$ ,

$$E[NK] = \sum_{n=1}^{\infty} \frac{n(n+1)}{2} (1-p)^{n-1} p = E\left[\frac{N(N+1)}{2}\right] = \frac{1}{p^2}$$

Finally, the covariance is

$$Cov[N,K] = E[NK] - E[N]E[K] = \frac{1}{2p^2} - \frac{1}{2p}$$

## Problem 3.5.1

The event *A* occurs iff X > 5 and Y > 5 and has probability

$$P[A] = P[X > 5, Y > 5] = \sum_{x=6}^{10} \sum_{y=6}^{10} 0.01 = 0.25$$

From Theorem 3.11,

$$P_{X,Y|A}(x,y) = \begin{cases} \frac{P_{X,Y}(x,y)}{P[A]} & (x,y) \in A\\ 0 & \text{otherwise} \end{cases} = \begin{cases} 0.04 & x = 6, \dots, 10; y = 6, \dots, 20\\ 0 & \text{otherwise} \end{cases}$$

## Problem 3.6.7

The key to solving this problem is to find the joint PMF of *M* and *N*. Note that  $N \ge M$ . For n > m, the joint event  $\{M = m, N = n\}$  has probability

$$P[M = m, N = n] = P[\overrightarrow{dd\cdots dv}\overrightarrow{dd\cdots dv}]$$
$$= (1-p)^{m-1}p(1-p)^{n-m-1}p$$
$$= (1-p)^{n-2}p^2$$

A complete expression for the joint PMF of M and N is

$$P_{M,N}(m,n) = \begin{cases} (1-p)^{n-2}p^2 & m = 1, 2, \dots, n-1; \ n = m+1, m+2, \dots \\ 0 & \text{otherwise} \end{cases}$$

For n = 2, 3, ..., the marginal PMF of N satisfies

$$P_N(n) = \sum_{m=1}^{n-1} (1-p)^{n-2} p^2 = (n-1)(1-p)^{n-2} p^2$$

Similarly, for m = 1, 2, ..., the marginal PMF of *M* satisfies

$$P_M(m) = \sum_{n=m+1}^{\infty} (1-p)^{n-2} p^2$$
  
=  $p^2 [(1-p)^{m-1} + (1-p)^m + \cdots]$   
=  $(1-p)^{m-1} p$ 

The complete expressions for the marginal PMF's are

$$P_M(m) = \begin{cases} (1-p)^{m-1}p & m = 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$
$$P_N(n) = \begin{cases} (n-1)(1-p)^{n-2}p^2 & n = 2, 3, \dots \\ 0 & \text{otherwise} \end{cases}$$

Not surprisingly, if we view each voice call as a successful Bernoulli trial, *M* has a geometric PMF since it is the number of trials up to and including the first success. Also, *N* has a Pascal PMF since it is the number of trials required to see 2 successes. The conditional PMF's are now easy to find.

$$P_{N|M}(n|m) = \frac{P_{M,N}(m,n)}{P_{M}(m)} = \begin{cases} (1-p)^{n-m-1}p & n=m+1, m+2, \dots \\ 0 & \text{otherwise} \end{cases}$$

The interpretation of the conditional PMF of *N* given *M* is that given M = m, N = m + N' where N' has a geometric PMF with mean 1/p. The conditional PMF of *M* given *N* is

$$P_{M|N}(m|n) = \frac{P_{M,N}(m,n)}{P_{N}(n)} = \begin{cases} 1/(n-1) & m = 1, \dots, n-1 \\ 0 & \text{otherwise} \end{cases}$$

Given that call N = n was the second voice call, the first voice call is equally likely to occur in any of the previous n - 1 calls.

# Problem 3.6.8

- (a) The number of buses, *N*, must be greater than zero. Also, the number of minutes that pass cannot be less than the number of buses. Thus, P[N = n, T = t] > 0 for integers *n*, *t* satisfying  $1 \le n \le t$ .
- (b) First, we find the joint PMF of *N* and *T* by carefully considering the possible sample paths. In particular,  $P_{N,T}(n,t) = P[ABC] = P[A]P[B]P[C]$  where the events *A*, *B* and *C* are

 $A = \{n - 1 \text{ buses arrive in the first } t - 1 \text{ minutes}\}$ 

 $B = \{$ none of the first n - 1 buses are boarded $\}$ 

 $C = \{ at time t a bus arrives and is boarded \}$ 

These events are independent since each trial to board a bus is independent of when the buses arrive. These events have probabilities

$$P[A] = {\binom{t-1}{n-1}} p^{n-1} (1-p)^{t-1-(n-1)}$$
$$P[B] = (1-q)^{n-1}$$
$$P[C] = pq$$

Consequently, the joint PMF of N and T is

$$P_{N,T}(n,t) = \begin{cases} \binom{t-1}{n-1} p^{n-1} (1-p)^{t-n} (1-q)^{n-1} p q & n \ge 1, t \ge n \\ 0 & \text{otherwise} \end{cases}$$

(c) It is possible to find the marginal PMF's by summing the joint PMF. However, it is much easier to obtain the marginal PMFs by consideration of the experiment. Specifically, when a bus arrives, it is boarded with probability q. Moreover, the experiment ends when a bus is boarded. By viewing whether each arriving bus is boarded as an independent trial, N is the number of trials until the first success. Thus, N has the geometric PMF

$$P_N(n) = \begin{cases} (1-q)^{n-1}q & n = 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

To find the PMF of T, suppose we regard each minute as an independent trial in which a success occurs if a bus arrives and that bus is boarded. In this case, the success probability is pq and T is the number of minutes up to and including the first success. The PMF of T is also geometric.

$$P_T(t) = \begin{cases} (1 - pq)^{t-1}pq & t = 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

(d) Once we have the marginal PMFs, the conditional PMFs are easy to find.

$$P_{N|T}(n|t) = \frac{P_{N,T}(n,t)}{P_{T}(t)} = \begin{cases} \binom{t-1}{n-1} \left(\frac{p(1-q)}{1-pq}\right)^{n-1} \left(\frac{1-p}{1-pq}\right)^{t-1-(n-1)} & n = 1, 2, \dots, t \\ 0 & \text{otherwise} \end{cases}$$

That is, given you depart at time T = t, the number of buses that arrive during minutes  $1, \ldots, t - 1$  has a binomial PMF since in each minute a bus arrives with probability p. Similarly, the conditional PMF of T given N is

$$P_{T|N}(t|n) = \frac{P_{N,T}(n,t)}{P_N(n)} = \begin{cases} \binom{t-1}{n-1} p^n (1-p)^{t-n} & t = n, n+1, \dots \\ 0 & \text{otherwise} \end{cases}$$

This result can be explained. Given that you board bus N = n, the time *T* when you leave is the time for *n* buses to arrive. If we view each bus arrival as a success of an independent trial, the time for *n* buses to arrive has the above Pascal PMF.

#### Problem 3.6.9

what type of call (if any) that arrived in any 1 millisecond period, it will be apparent that a fax call arrives with probability  $\alpha = pqr$  or no fax arrives with probability  $1 - \alpha$ . That is, whether a fax message arrives each millisecond is a Bernoulli trial with success probability  $\alpha$ . Thus, the time required for the first success has the geometric PMF

$$P_T(t) = \begin{cases} (1-\alpha)^{t-1}\alpha & t = 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

Note that *N* is the number of trials required to observe 100 successes. Moreover, the number of trials needed to observe 100 successes is N = T + N' where N' is the number of trials needed to observe successes 2 through 100. Since N' is just the number of trials needed to observe 99 successes, it has the Pascal PMF

$$P_{N'}(n) = \begin{cases} \binom{n-1}{98} \alpha^{98} (1-\alpha)^{n-98} & n = 99, 100, \dots \\ 0 & \text{otherwise} \end{cases}$$

Since the trials needed to generate successes 2 though 100 are independent of the trials that yield the first success, N' and T are independent. Hence

$$P_{N|T}(n|t) = P_{N'|T}(n-t|t) = P_{N'}(n-t)$$

Applying the PMF of N' found above, we have

$$P_{N|T}(n|t) = \begin{cases} \binom{n-1}{98} \alpha^{98} (1-\alpha)^{n-t-98} & n = 99+t, 100+t, \dots \\ 0 & \text{otherwise} \end{cases}$$

Finally the joint PMF of N and T is

$$P_{N,T}(n,t) = P_{N|T}(n|t) P_{T}(t) = \begin{cases} \binom{n-t-1}{98} \alpha^{99} (1-\alpha)^{n-99} \alpha & t = 1, 2, \dots; n = 99+t, 100+t, \dots \\ 0 & \text{otherwise} \end{cases}$$

This solution can also be found a consideration of the sample sequence of Bernoulli trials in which we either observe or do not observe a fax message. To find the conditional PMF  $P_{T|N}(t|n)$ , we first must recognize that N is simply the number of trials needed to observe 100 successes and thus has the Pascal PMF

$$P_N(n) = \begin{cases} \binom{n-1}{99} \alpha^{100} (1-\alpha)^{n-100} & n = 100, 101, \dots \\ 0 & \text{otherwise} \end{cases}$$

Hence the conditional PMF is

$$P_{T|N}(t|n) = \frac{P_{N,T}(n,t)}{P_{N}(n)} = \frac{\binom{n-t-1}{98}}{\binom{n-1}{99}} \frac{1-\alpha}{\alpha}$$

#### Problem 3.7.8

The key to this problem is understanding that "Factory Q" and "Factory R" are synonyms for M = 60 and M = 180. Similarly, "small", "medium", and "large" orders correspond to the events B = 1, B = 2 and B = 3.

(a) The following table given in the problem statement

	Factory Q	Factory R
small order	0.3	0.2
medium order	0.1	0.2
large order	0.1	0.1

can be translated into the following joint PMF for B and M.

$P_{B,M}(b,m)$	m = 60	m = 180
b = 1	0.3	0.2
b=2	0.1	0.2
b=3	0.1	0.1

(b) Before we find E[B], it will prove helpful for the remainder of the problem to find the marginal PMFs  $P_B(b)$  and  $P_M(m)$ . These can be found from the row and column sums of the table of the joint PMF

$P_{B,M}(b,m)$	m = 60	m = 180	$P_B(b)$
b = 1	0.3	0.2	0.5
b = 2	0.1	0.2	0.3
<i>b</i> = 3	0.1	0.1	0.2
$P_M(m)$	0.5	0.5	

The expected number of boxes is

$$E[B] = \sum_{b} bP_B(b) = 1(0.5) + 2(0.3) + 3(0.2) = 1.7$$

(c) From the marginal PMF of *B*, we know that  $P_B(2) = 0.3$ . The conditional PMF of *M* given B = 2 is

$$P_{M|B}(m|2) = \frac{P_{B,M}(2,m)}{P_{B}(2)} = \begin{cases} 1/3 & m = 60\\ 2/3 & m = 180\\ 0 & \text{otherwise} \end{cases}$$

(d) The conditional expectation of M given B = 2 is

$$E[M|B=2] = \sum_{m} mP_{M|B}(m|2) = 60(1/3) + 180(2/3) = 140$$

- (e) From the marginal PMFs we calculated in the table of part (b), we can conclude that *B* and *M* are not independent. since  $P_{B,M}(1, 60) \neq P_B(1)P_M(m)60$ .
- (f) In terms of *M* and *B*, the cost (in cents) of sending a shipment is C = BM. The expected value of *C* is

$$\begin{split} E[C] &= \sum_{b,m} bm P_{B,M}(b,m) \\ &= 1(60)(0.3) + 2(60)(0.1) + 3(60)(0.1) \\ &+ 1(180)(0.2) + 2(180)(0.2) + 3(180)(0.1) = 210 \end{split}$$

# Problem 3.8.5

is added to the jackpot,

$$J_{i-1} = J_i + \frac{N_i}{2}$$

Given  $J_i = j$ ,  $N_i$  has a Poisson distribution with mean j. so that  $E[N_i|J_i = j] = j$  and that  $Var[N_i|J_i = j] = j$ . This implies

$$E[N_i^2|J_i = j] = \operatorname{Var}[N_i|J_i = j] + (E[N_i|J_i = j])^2 = j + j^2$$

In terms of the conditional expectations given  $J_i$ , these facts can be written as

$$E[N_i|J_i] = J_i \qquad E[N_i^2|J_i] = J_i + J_i^2$$

This permits us to evaluate the moments of  $J_{i-1}$  in terms of the moments of  $J_i$ . Specifically,

$$E[J_{i-1}|J_i] = E[J_i|J_i] + \frac{1}{2}E[N_i|J_i] = J_i + \frac{J_i}{2} = \frac{3J_i}{2}$$

This implies

$$E[J_{i-1}] = \frac{3}{2}E[J_i]$$

We can use this the calculate  $E[J_i]$  for all *i*. Since the jackpot starts at 1 million dollars,  $J_6 = 10^6$  and  $E[J_6] = 10^6$ . This implies

$$E[J_i] = (3/2)^{6-i} 10^6$$

Now we will find the second moment  $E[J_i^2]$ . Since  $J_{i-1}^2 = J_i^2 + N_i J_i + N_i^2/4$ , we have

$$E[J_{i-1}^{2}|J_{i}] = E[J_{i}^{2}|J_{i}] + E[N_{i}J_{i}|J_{i}] + E[N_{i}^{2}|J_{i}]/4$$
  
=  $J_{i}^{2} + J_{i}E[N_{i}|J_{i}] + (J_{i} + J_{i}^{2})/4$   
=  $(3/2)^{2}J_{i}^{2} + J_{i}/4$ 

By taking the expectation over  $J_i$  we have

$$E[J_{i-1}^2] = (3/2)^2 E[J_i^2] + E[J_i]/4$$

This recursion allows us to calculate  $E[J_i^2]$  for i = 6, 5, ..., 0. Since  $J_6 = 10^6$ ,  $E[J_6^2] = 10^{12}$ . From the recursion, we obtain

$$E[J_5^2] = (3/2)^2 E[J_6^2] + E[J_6]/4 = (3/2)^2 10^{12} + \frac{1}{4} 10^6$$
  

$$E[J_4^2] = (3/2)^2 E[J_5^2] + E[J_5]/4 = (3/2)^4 10^{12} + \frac{1}{4} [(3/2)^2 + (3/2)] 10^6$$
  

$$E[J_3^2] = (3/2)^2 E[J_4^2] + E[J_4]/4 = (3/2)^6 10^{12} + \frac{1}{4} [(3/2)^4 + (3/2)^3 + (3/2)^2] 10^6$$

The same recursion will also allow us to show that

$$E[J_2^2] = (3/2)^8 10^{12} + \frac{1}{4} \Big[ (3/2)^6 + (3/2)^5 + (3/2)^4 + (3/2)^3 \Big] 10^6$$
  

$$E[J_1^2] = (3/2)^{10} 10^{12} + \frac{1}{4} \Big[ (3/2)^8 + (3/2)^7 + (3/2)^6 + (3/2)^5 + (3/2)^4 \Big] 10^6$$
  

$$E[J_0^2] = (3/2)^{12} 10^{12} + \frac{1}{4} \Big[ (3/2)^{10} + (3/2)^9 + \dots + (3/2)^5 \Big] 10^6$$

Finally, day 0 is the same as any other day in that  $J = J_0 + N_0/2$  where  $N_0$  is a Poisson random variable with mean  $J_0$ . By the same argument that we used to develop recursions for  $E[J_i]$  and  $E[J_i^2]$ , we can show

$$E[J] = (3/2)E[J_0] = (3/2)^7 10^6 \approx 17 \times 10^6$$

and

$$E[J^{2}] = (3/2)^{2}E[J_{0}^{2}] + E[J_{0}]/4$$
  
=  $(3/2)^{14}10^{12} + \frac{1}{4}[(3/2)^{12} + (3/2)^{11} + \dots + (3/2)^{6}]10^{6}$   
=  $(3/2)^{14}10^{12} + \frac{10^{6}}{2}(3/2)^{6}[(3/2)^{7} - 1]$ 

Finally, the variance of J is

Var 
$$[J] = E[J^2] - (E[J])^2 = \frac{10^6}{2}(3/2)^6[(3/2)^7 - 1]$$

Since the variance is hard to interpret, we note that the standard deviation of *J* is  $\sigma_J \approx 9572$ . Although the expected jackpot grows rapidly, the standard deviation of the jackpot is fairly small.

# Problem 3.8.6

(a) The sample space is

$$S_{X,Y,Z} = \{(x, y, z) | x + y + z = 5, x \ge 0, y \ge 0, z \ge 0, x, y, z \text{ integer} \}$$

$$= \begin{cases} (0, 0, 5), \\ (0, 1, 4), & (1, 0, 4), \\ (0, 2, 3), & (1, 1, 3), & (2, 0, 3), \\ (0, 3, 2), & (1, 2, 2), & (2, 1, 2), & (3, 0, 2), \\ (0, 4, 1), & (1, 3, 1), & (2, 2, 1), & (3, 1, 1), & (4, 0, 1), \\ (0, 5, 0), & (1, 4, 0), & (2, 3, 0), & (3, 2, 0), & (4, 1, 0), & (5, 0, 0) \end{cases}$$

(b) As we see in the above list of elements of  $S_{X,Y,Z}$ , just writing down all the elements is not so easy. Similarly, representing the joint PMF is usually not very straightforward. Here are the probabilities in a list.

(x, y, z)	$P_{X,Y,Z}\left(x,y,z ight)$	$P_{X,Y,Z}(x,y,z)$ (decimal)
(0, 0, 5)	$(1/6)^5$	$1.29 \times 10^{-4}$
(0, 1, 4)	$5(1/2)(1/6)^4$	$1.93 \times 10^{-3}$
(1, 0, 4)	$5(1/3)(1/6)^4$	$1.29  imes 10^{-3}$
(0, 2, 3)	$10(1/2)^2(1/6)^3$	$1.16 \times 10^{-2}$
(1, 1, 3)	$20(1/3)(1/2)(1/6)^3$	$1.54  imes 10^{-2}$
(2, 0, 3)	$10(1/3)^2(1/6)^3$	$5.14 \times 10^{-3}$
(0, 3, 2)	$10(1/2)^3(1/6)^2$	$3.47 \times 10^{-2}$
(1, 2, 2)	$30(1/3)(1/2)^2(1/6)^2$	$6.94 \times 10^{-2}$
(2, 1, 2)	$30(1/3)^2(1/2)(1/6)^2$	$4.63 \times 10^{-2}$
(3, 0, 2)	$10(1/2)^3(1/6)^2$	$1.03 \times 10^{-2}$
(0, 4, 1)	$5(1/2)^4(1/6)$	$5.21 \times 10^{-2}$
(1, 3, 1)	$20(1/3)(1/2)^3(1/6)$	$1.39 \times 10^{-1}$
(2, 2, 1)	$30(1/3)^2(1/2)^2(1/6)$	$1.39 \times 10^{-1}$
(3, 1, 1)	$20(1/3)^3(1/2)(1/6)$	$6.17  imes 10^{-2}$
(4, 0, 1)	$5(1/3)^4(1/6)$	$1.03 \times 10^{-2}$
(0, 5, 0)	$(1/2)^5$	$3.13 \times 10^{-2}$
(1, 4, 0)	$5(1/3)(1/2)^4$	$1.04  imes 10^{-1}$
(2, 3, 0)	$10(1/3)^2(1/2)^3$	$1.39 \times 10^{-1}$
(3, 2, 0)	$10(1/3)^3(1/2)^2$	$9.26  imes 10^{-2}$
(4, 1, 0)	$5(1/3)^4(1/2)$	$3.09  imes 10^{-2}$
(5, 0, 0)	$(1/3)^5$	$4.12 \times 10^{-3}$

(c) Note that *Z* is the number of three page faxes. In principle, we can sum the joint PMF  $P_{X,Y,Z}(x,y,z)$  over all *x*, *y* to find  $P_Z(z)$ . However, it is better to realize that each fax has 3 pages with probability 1/6, independent of any other fax. Thus, *Z* has the binomial PMF

$$P_Z(z) = \begin{cases} \binom{5}{z} (1/6)^z (5/6)^{5-z} & z = 0, 1, \dots, 5\\ 0 & \text{otherwise} \end{cases}$$

- (d) From the properties of the binomial distribution given in Appendix A, we know that E[Z] = 5(1/6).
- (e) We want to find the conditional PMF of the number X of 1-page faxes and number Y of 2-page faxes given Z = 2 3-page faxes. Note that given Z = 2, X + Y = 3. Hence for non-negative integers x, y satisfying x + y = 3,

$$P_{X,Y|Z}(x,y|2) = \frac{P_{X,Y,Z}(x,y,2)}{P_Z(2)} = \frac{\frac{5!}{x!y!2!}(1/3)^x(1/2)^y(1/6)^2}{\binom{5}{2}(1/6)^2(5/6)^3}$$

With some algebra, the complete expression of the conditional PMF is

$$P_{X,Y|Z}(x,y|2) = \begin{cases} \frac{3!}{x!y!} (2/5)^x (3/5)^y & x+y=3, x \ge 0, y \ge 0; x, y \text{ integer} \\ 0 & \text{otherwise} \end{cases}$$

To interpret the above expression, we observe that if Z = 2, then Y = 3 - X and

$$P_{X|Z}(x|2) = P_{X,Y|Z}(x,3-x|2) = \begin{cases} \binom{3}{x} (2/5)^x (3/5)^{3-x} & x = 0, 1, 2, 3 \\ 0 & \text{otherwise} \end{cases}$$

That is, given Z = 2, there are 3 faxes left, each of which independently could be a 1-page fax. The conditonal PMF of the number of 1-page faxes is binomial where 2/5 is the conditional probability that a fax has 1 page given that it either has 1 page or 2 pages. Moreover given X = x and Z = 2 we must have Y = 3 - x.

- (f) Given Z = 2, the conditional PMF of X is binomial for 3 trials and success probability 2/5. The conditional expectation of X given Z = 2 is E[X|Z=2] = 3(2/5) = 6/5.
- (g) There are several ways to solve this problem. The most straightforward approach is to realize that for integers  $0 \le x \le 5$  and  $0 \le y \le 5$ , the event  $\{X = x, Y = y\}$  occurs iff  $\{X = x, Y = y, Z = 5 (x + y)\}$ . For the rest of this problem, we assume *x* and *y* are non-negative integers so that

$$P_{X,Y}(x,y) = P_{X,Y,Z}(x,y,5-(x+y))$$
  
= 
$$\begin{cases} \frac{5!}{x!y!(5-x-y)!} \left(\frac{1}{3}\right)^x \left(\frac{1}{2}\right)^y \left(\frac{1}{6}\right)^{5-x-y} & 0 \le x+y \le 5, x \ge 0, y \ge 0\\ 0 & \text{otherwise} \end{cases}$$

Tha above expression may seem unwieldy and it isn't even clear that it will sum to 1. To simplify the expression, we observe that

$$P_{X,Y}(x,y) = P_{X,Y,Z}(x,y,5-x-y) = P_{X,Y|Z}(x,y|5-x+y)P_Z(5-x-y)$$

Using  $P_Z(z)$  found in part (c), we can calculate  $P_{X,Y|Z}(x,y|5-x-y)$  for  $0 \le x+y \le 5$ . integer valued.

$$P_{X,Y|Z}(x,y|5-x+y) = \frac{P_{X,Y,Z}(x,y,5-x-y)}{P_Z(5-x-y)}$$
  
=  $\binom{x+y}{x} \left(\frac{1/3}{1/2+1/3}\right)^x \left(\frac{1/2}{1/2+1/3}\right)^y$   
=  $\binom{x+y}{x} \left(\frac{2}{5}\right)^x \left(\frac{3}{5}\right)^{(x+y)-x}$ 

In the above expression, it is wise to think of x + y as some fixed value. In that case, we see that given x + y is a fixed value, X and Y have a joint PMF given by a binomial distribution in x. This should not be surprising since it is just a generalization of the case when Z = 2. That is, given that there were a fixed number of faxes that had either one or two pages, each of those faxes is a one page fax with probability (1/3)/(1/2 + 1/3) and so the number of one page faxes should have a binomial distribution, Moreover, given the number X of one page faxes, the number Y of two page faxes is completely specified. Finally, by rewriting  $P_{X,Y}(x, y)$  given above, the complete expression for the joint PMF of X and Y is

$$P_{X,Y}(x,y) = \begin{cases} \binom{5}{5-x-y} \left(\frac{1}{6}\right)^{5-x-y} \left(\frac{5}{6}\right)^{x+y} \binom{x+y}{x} \left(\frac{2}{5}\right)^x \left(\frac{3}{5}\right)^y & x+y \le 5, x \ge 0, y \ge 0\\ 0 & \text{otherwise} \end{cases}$$