

# Homework 2

February 28, 2019

## 1 problem 2.1.1(a)

$$\begin{aligned} P(X = i) &= \sum_{k=n}^N P(X = i|M = k)P(M = k) \\ &= \sum_{k=n}^N \binom{k}{i} \pi^i (1 - \pi)^{k-i} \binom{N}{k} p^k (1 - p)^{N-k} \\ &= \sum_{k=i}^N \frac{k!}{i!(k-i)!} \pi^i (1 - \pi)^{k-i} \frac{N!}{k!(N-k)!} p^k (1 - p)^{N-k} \\ &= \frac{N!}{i!} (1 - p)^N \frac{(\pi p)^i}{(1 - p)^i} \sum_{k=i}^N \frac{1}{(k-i)!(N-k)!} \left( \frac{(1 - \pi)p}{1 - p} \right)^{k-i} \\ &= \frac{N!}{i!(N-i)!} (1 - p)^N \frac{(\pi p)^i}{(1 - p)^i} \sum_{k=i}^N \frac{(N-i)!}{(k-i)!(N-k)!} \left( \frac{(1 - \pi)p}{1 - p} \right)^{k-i} \\ &= \frac{N!}{i!(N-i)!} (1 - p)^N \frac{(\pi p)^i}{(1 - p)^i} \sum_{k=0}^{N-i} \frac{(N-i)!}{k!(N-i-k)!} \left( \frac{(1 - \pi)p}{1 - p} \right)^k \\ &= \frac{N!}{i!(N-i)!} (1 - p)^{N-i} (\pi p)^i \left( 1 + \frac{(1 - \pi)p}{1 - p} \right)^{N-i} \\ &= \frac{N!}{i!(N-i)!} (\pi p)^i (1 - \pi p)^{N-i} \end{aligned} \tag{1}$$

So X is has marginal probability as binomial distribution with parameter  $(N, \pi p)$ .

## 2 problem 2.1.2

- (i) If  $y > n$ , then  $P(X = n|Y = y) = 0$ .
- (ii) If  $y \leq n$ , then

$$P(X = n, Y = y) = P(Y = y|X = n)P(X = n) = \frac{1}{nN} \tag{2}$$

$$\begin{aligned}
P(Y = y) &= \sum_{x=y}^N P(Y = y, X = x) \\
&= \sum_{x=y}^N P(Y = y|X = x)P(X = x) \\
&= \sum_{x=y}^N \frac{1}{Nx}
\end{aligned} \tag{3}$$

$$\begin{aligned}
P(X = n|Y = y) &= \frac{P(X = n, Y = y)}{P(Y = y)} \\
&= \frac{1}{n \sum_{x=y}^N \frac{1}{x}}
\end{aligned} \tag{4}$$

### 3 2.1.4

$$\begin{aligned}
E[X] &= \sum_{n=0}^{20} E[X|N = n]P(N = n) \\
&= \sum_{n=0}^{20} \sum_{i=0}^n iP(X = i|N = n)P(N = n) \\
&= \sum_{n=0}^{20} \sum_{i=0}^n i \binom{n}{i} \frac{1}{2^n} \binom{20}{n} \frac{1}{4^n} \left(\frac{3}{4}\right)^{20-n} \\
&= \sum_{n=0}^{20} \frac{n}{2} \binom{20}{n} \frac{1}{4^n} \left(\frac{3}{4}\right)^{20-n} \\
&= \frac{1}{2} E[N] \\
&= \frac{1}{2} \times \frac{20}{4} \\
&= \frac{5}{2}
\end{aligned} \tag{5}$$

## 4 problem 2.1.8

Let  $A = \{\text{first ball is red}\}$ , and  $B = \{\text{second ball is red}\}$ .

$$\begin{aligned} P(A|B) &= \frac{P(A, B)}{P(B)} \\ &= \frac{P(A, B)}{P(B, A) + P(B, A^c)} \\ &= \frac{\frac{1}{2} \times \frac{2}{3}}{\frac{1}{2} \times \frac{2}{3} + \frac{1}{2} \times \frac{1}{3}} \\ &= \frac{2}{3} \end{aligned} \tag{6}$$

## 5 problem 2.3.1

Let  $\{X_i\}$  be i.i.d random variables with Bernoulli distribution.  
Then  $Z = \sum_{i=1}^N X_i$ .

$$\begin{aligned} f_Z(z) &= P(Z = z) = \sum_{n=0}^{\infty} P\left(\sum_{i=1}^N X_i = z | N = n\right) P(N = n) \\ &= \sum_{n=0}^{\infty} \binom{n}{z} p^z (1-p)^{n-z} e^{-\lambda} \frac{\lambda^n}{n!} \\ &= \frac{e^{-\lambda p} (\lambda p)^z}{z!} \end{aligned} \tag{7}$$

### 5.1 part a

$$E(Z) = \lambda p \text{ and } Var(Z) = \lambda p$$

### 5.2 part b

$$Z \sim \text{Poisson}(\lambda p)$$

## 6 problem 2.3.3

### 6.1 part a

$E(\xi_k) = 0$ ,  $Var(\xi_k) = 1$ ,  $E(N) = \frac{1-\alpha}{\alpha}$  and  $Var(N) = \frac{1-\alpha}{\alpha^2}$

$$\begin{aligned} E(Z) &= \sum_{n=0}^{\infty} E(Z|N=n)P(N=n) \\ &= \sum_{n=0}^{\infty} \sum_{i=1}^n E(\xi_k)P(N=n) \\ &= 0 \end{aligned} \tag{8}$$

From textbook,  $Var(Z) = \frac{1-\alpha}{\alpha} \times 1 + 0 = \frac{1-\alpha}{\alpha}$

### 6.2 part b

(i) In  $Z^3 = (\sum_{i=1}^N \xi_i)^3$ , each monomial is not a square of some product of  $\xi_i$ . Then the conditional expectation of each monomial is 0. So we have  $E(Z^3) = 0$ .

(ii)

$$\begin{aligned} E[Z^4] &= \sum_{n=1}^{\infty} E[Z^4|N=n]P(N=n) \\ &= \sum_{n=1}^{\infty} E[(\xi_1 + \dots + \xi_n)^4]P(N=n) \end{aligned} \tag{9}$$

Note that

$$(\xi_1 + \dots + \xi_n)^4 = \sum_{i=1}^n \xi_i^4 + \sum_{i \neq j} a_{ij} \xi_i^3 \xi_j + \sum_{i \neq j} b_{ij} \xi_i^2 \xi_j^2 + \sum_{i \neq j \neq k} c_{ijk} \xi_i^2 \xi_j \xi_k + \sum_{i \neq j \neq k \neq l} d_{ijkl} \xi_i \xi_j \xi_k \xi_l \tag{10}$$

where

$$\sum_{i \neq j} b_{ij} = \binom{n}{2} \binom{4}{2} = 3n(n-1) \tag{11}$$

Since  $\xi_1, \xi_2, \dots$  are iid and their mean are 0, we know that

$$\begin{aligned} E[(\xi_1 + \dots + \xi_n)^4] &= E\left[\sum_{i=1}^n \xi_i^4\right] + E\left[\sum_{i \neq j} b_{ij} \xi_i^2 \xi_j^2\right] \\ &= n + 3n(n-1) = (3n-2)n \end{aligned} \quad (12)$$

Then we could have

$$\begin{aligned} E[Z^4] &= \sum_{n=1}^{\infty} n(3n-2)\alpha(1-\alpha)^n \\ &= \alpha\left[3\sum_{n=1}^{\infty} n(n-1)(1-\alpha)^n + \sum_{n=1}^{\infty} n(1-\alpha)^n\right] \\ &= \alpha\left[3\sum_{n=2}^{\infty} n(n-1)(1-\alpha)^n + \sum_{n=1}^{\infty} n(1-\alpha)^n\right] \end{aligned} \quad (13)$$

Let  $x = 1 - \alpha$ , then

$$\begin{aligned} \sum_{n=0}^{\infty} x^n &= \frac{1}{1-x} \\ \sum_{n=1}^{\infty} nx^n &= x\left(\frac{d}{dx} \frac{1}{1-x}\right) = \frac{x}{(1-x)^2} \\ \sum_{n=2}^{\infty} n(n-1)x^n &= x^2 \frac{d}{dx} \frac{1}{(1-x)^2} = \frac{2x^2}{(1-x)^3} \end{aligned} \quad (14)$$

So we have

$$E[Z^4] = \frac{5\alpha^2 - 11\alpha + 6}{\alpha^2} \quad (15)$$

## 7 problem 2.3.4

### 7.1 part a

$$\begin{aligned} E(S_n) &= \sum_{n=0}^{\infty} E(\xi_1 + \dots + \xi_n | N = n)P(N = n) \\ &= \mu \sum_{n=0}^{\infty} nP(N = n) \\ &= \mu\lambda \end{aligned} \quad (16)$$

and

$$\text{Var}(S_n) = E((S_n - \mu\lambda)^2) = \sigma^2\lambda + \mu^2\lambda \quad (17)$$

## 7.2 part b

$\text{Var}(N) = \frac{1-p}{p^2}$ ,  $E(S_N) = \mu\lambda$  and

$$\text{Var}(S_N) = \frac{\sigma^2(1-p)}{p} + \frac{\mu^2(1-p)}{p^2} \quad (18)$$

## 7.3 part c

Expectations in two parts increase with the same speed since they have the same formula. Variance in (b) increases faster than the one in (a).