HW 5 M362M, Fall 2024

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Problem 1

We always have $Y \ge 0$ and it only takes integral values. Thus $\mathbb{P}[Y < 0] = 0$. For $i = 0, 1, 2, \dots$, we have

$$\mathbb{P}[Y=i] = \sum_{n=1}^{\infty} \mathbb{P}[\xi_1 + \dots + \xi_N = i \mid N=n] \cdot \mathbb{P}[N=n]$$
$$= \sum_{n=1}^{\infty} \mathbb{P}[\xi_1 + \dots + \xi_n = i] \cdot \mathbb{P}[N=n]$$

Note that $\xi_1 + \cdots + \xi_n \sim \text{Binom}(n, p_B)$. If i = 0, then

$$\mathbb{P}[Y=0] = \sum_{n=1}^{\infty} \binom{n}{0} p_B^0 (1-p_B)^n \cdot p_g (1-p_g)^{n-1}$$

$$= (1-p_B) p_g \cdot \sum_{n=0}^{\infty} \left[(1-p_B)(1-p_g) \right]^n$$

$$= \frac{(1-p_B) p_g}{1-(1-p_B)(1-p_g)}$$

If $i = 1, 2, \dots, \mathbb{P}[\xi_1 + \dots + \xi_n = i] = 0$ for n < i, thus

$$\mathbb{P}[Y=i] = \sum_{n=i}^{\infty} \binom{n}{i} p_B^i (1-p_B)^{n-i} \cdot p_g (1-p_g)^{n-1}$$
$$= p_g (1-p_g)^{i-1} p_B^i \cdot \sum_{n=i}^{\infty} \binom{n}{i} [(1-p_B)(1-p_g)]^{n-i}$$

Now we prove the identity, for all $i \in \mathbb{N}_0$, |x| < 1, we have

$$S_i := \sum_{n=i}^{\infty} \binom{n}{i} x^{n-i} = \frac{1}{(1-x)^{i+1}}.$$

We proceed with induction on i. When i = 0, it is a geometric series, and the result holds. Suppose it holds when i = k - 1, then for i = k,

$$S_k - xS_k = 1 + \sum_{n=k+1}^{\infty} \left[\binom{n}{k} - \binom{n-1}{k} \right] x^{n-k}$$

$$= 1 + \sum_{n=k+1}^{\infty} \binom{n-1}{k-1} x^{n-k}$$

$$= S_{k-1}$$

$$= \frac{1}{(1-x)^k}$$

Therefore

$$S_k = \frac{1}{(1-x)^{k+1}}.$$

Thus the statement holds for all $i \in \mathbb{N}_0$. Back to the original problem, we have

$$\mathbb{P}[Y=i] = \begin{cases} \frac{p_g(1-p_g)^{i-1}p_B^i}{[1-(1-p_B)(1-p_g)]^{i+1}}, & i=1,2,\cdots\\ \\ \frac{(1-p_B)p_g}{1-(1-p_B)(1-p_g)} & , & i=0,\\ \\ 0 & , & \text{else.} \end{cases}$$

Problem 2

2.1

$$\begin{split} M_{X_n}(t) &= M_{\sum_{i=1}^n \delta_i}(t) \\ &= \prod_{i=1}^n M_{\delta_i}(t) \\ &= [M_{\delta_1}(t)]^n \\ &= \left[\frac{1}{2}(e^t + e^{-t})\right]^n \end{split} \qquad (\delta_1, \cdots, \delta_n \text{ identically distributed}) \end{split}$$

2.2

We first calculate the first two derivatives of $M_{X_n}(t)$.

$$M'_{X_n}(t) = n \left[\frac{1}{2} (e^t + e^{-t}) \right]^{n-1} \cdot \frac{1}{2} (e^t - e^{-t}) = \frac{n}{2^n} (e^t + e^{-t})^{n-1} (e^t - e^{-t}).$$

$$M''_{X_n}(t) = \frac{n}{2^n} \left[(n-1)(e^t + e^{-t})^{n-2} (e^t - e^{-t})^2 + (e^t + e^{-t})^n \right].$$

We are now in a position to calculate $\mathbb{E}[X_n]$ and $\operatorname{Var}(X_n)$.

$$\mathbb{E}[X_n] = M'_{X_n}(0)$$
= 0, $(e^0 - e^{-0} = 0)$

and

$$\operatorname{Var}(X_n) = \mathbb{E}[X_n^2] - (\mathbb{E}[X_n])^2$$

$$= \mathbb{E}[X_n^2]$$

$$= M''_{X_n}(0) + M'_{X_n}(0)$$

$$= n$$

$$(\mathbb{E}[X_n] = 0)$$

Therefore

$$Z_n = \frac{X_n}{\sqrt{n}}$$

2.3

$$M_{Z_n}(t) = \left[M_{\frac{\delta_1}{\sqrt{n}}}(t) \right]^n$$
$$= \left[\frac{1}{2} \left(e^{\frac{t}{\sqrt{n}}} + e^{\frac{-t}{\sqrt{n}}} \right) \right]^n$$

2.4

$$M_{Z_{\infty}}(t) = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \cdot e^{tx} dx$$

$$= e^{\frac{1}{2}t^2} \cdot \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-t)^2} dx$$

$$= e^{\frac{1}{2}t^2} \cdot \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy \qquad (y = x - t)$$

$$= e^{\frac{1}{2}t^2}$$

2.5

$$\ln\left(\lim_{n\to\infty} M_{Z_n}(t)\right) = \lim_{n\to\infty} \ln M_{Z_n}(t) \qquad (\ln(\cdot) \text{ is continuous})$$

$$= \lim_{u\to 0^+} \frac{\ln\left(\frac{e^{ut} + e^{-ut}}{2}\right)}{u^2} \qquad (u = \frac{1}{\sqrt{n}})$$

$$= \lim_{u\to 0^+} \frac{te^{ut} - te^{-ut}}{2u \cdot \underbrace{(e^{ut} + e^{-ut})}_{\text{Converges to 2}}} \qquad (\text{L'Hospital's Rule})$$

$$= \lim_{u\to 0^+} \frac{e^{ut} - e^{-ut} + 2t^2e^{-ut}}{4} \qquad (\text{L'Hospital's Rule})$$

$$= \frac{1}{2}t^2.$$

Thus $\lim_{n\to\infty} M_{Z_n}(t) = e^{\frac{1}{2}t^2} = M_{Z_{\infty}}(t)$

Problem 3

3.1

$$\mathbb{P}[Z_2 = 4] = \mathbb{P}[Z_{1,1} = 2, Z_{1,2} = 2] \cdot \mathbb{P}[Z_1 = 2]$$
$$= 0.3^2 \times 0.3$$
$$= 0.027$$

$$\mathbb{P}[Z_2 = 3] = \mathbb{P}[Z_{1,1} = 1, Z_{1,2} = 2] \cdot \mathbb{P}[Z_1 = 2]$$

$$+ \mathbb{P}[Z_{1,1} = 2, Z_{1,2} = 1] \cdot \mathbb{P}[Z_1 = 2]$$

$$= 0.6 \times 0.3 \times 0.3 + 0.3 \times 0.6 \times 0.3$$

$$= 0.108$$

$$\begin{split} \mathbb{P}[Z_2 = 2] &= \mathbb{P}[Z_{1,1} = 1, Z_{1,2} = 1] \cdot \mathbb{P}[Z_1 = 2] \\ &+ \mathbb{P}[Z_{1,1} = 0, Z_{1,2} = 2] \cdot \mathbb{P}[Z_1 = 2] \\ &+ \mathbb{P}[Z_{1,1} = 2, Z_{1,2} = 0] \cdot \mathbb{P}[Z_1 = 2] \\ &+ \mathbb{P}[Z_{1,1} = 2] \cdot \mathbb{P}[Z_1 = 1] \\ &= 0.6^2 \times 0.3 + 0.1 \times 0.3 \times 0.3 \\ &+ 0.3 \times 0.1 \times 0.3 + 0.3 \times 0.6 \\ &= 0.306 \end{split}$$

$$\mathbb{P}[Z_2 = 1] = \mathbb{P}[Z_{1,1} = 1, Z_{1,2} = 0] \cdot \mathbb{P}[Z_1 = 2]$$

$$+ \mathbb{P}[Z_{1,1} = 0, Z_{1,2} = 1] \cdot \mathbb{P}[Z_1 = 2]$$

$$+ \mathbb{P}[Z_{1,1} = 1] \cdot \mathbb{P}[Z_1 = 1]$$

$$= 0.6 \times 0.1 \times 0.3 + 0.1 \times 0.6 \times 0.3$$

$$+ 0.6 \times 0.6$$

$$= 0.396$$

$$\mathbb{P}[Z_2 = 0] = \mathbb{P}[Z_{1,1} = 0, Z_{1,2} = 0] \cdot \mathbb{P}[Z_1 = 2]$$

$$+ \mathbb{P}[Z_{1,1} = 0] \cdot \mathbb{P}[Z_1 = 1]$$

$$+ \mathbb{P}[Z_1 = 0]$$

$$= 0.1^2 \times 0.3 + 0.1 \times 0.6 + 0.1$$

$$= 0.163$$

$$\mathbb{P}[Z_2 = i] = \begin{cases} 0.027, & i = 4 \\ 0.108, & i = 3 \\ 0.306, & i = 2 \\ 0.396, & i = 1 \\ 0.163, & i = 0 \\ 0, & \text{else.} \end{cases}$$

Let Z be a random variable with the following probability mass function

$$P(Z=i) = \begin{cases} 0.3, & i=2\\ 0.6, & i=1\\ 0.1, & i=0\\ 0, & \text{else.} \end{cases}$$

Define $G_Z(s) = \mathbb{E}[s^Z], G_{Z_n}(s) = \mathbb{E}[s^{Z_n}]$ for $0 \le s \le 1$. We claim that

$$G_{Z_n}(s) = \underbrace{G_Z \circ \cdots \circ G_z}_{n \text{ times}}(s).$$

We proceed by induction on n. When n = 1, the results holds as Z_1 Z. Suppose the statement is true for some $n = k_0$. Then

$$G_{Z_{k_0+1}}(s) = \mathbb{E}\left[\mathbb{E}\left[s^{Z_{k_0+1}} \mid Z_{k_0}\right]\right]$$

$$= \mathbb{E}\left[\mathbb{E}\left[s^{\sum_{j=1}^{Z_{k_0}} Z_{k_0,j}} \middle| Z_{k_0}\right]\right]$$

$$= \mathbb{E}\left[\left(\mathbb{E}\left[s^{Z}\right]\right)^{Z_{k_0}}\right]$$

$$= G_{Z_{k_0}}(G_Z(s))$$

$$= G_{Z_{k_0}}(G_Z(s)$$

Thus the statement is true for all $n \in \mathbb{N}$. Note that

$$G_{Z_n}(s) = \sum_{k=0}^{\infty} s^k \cdot \mathbb{P}[Z_n = k],$$

therefore $G_{Z_n}(0) = \mathbb{P}[Z_n = 0]$, and we have

$$\mathbb{P}[Z_n = 0] = \underbrace{G_Z \circ \cdots \circ G_Z}_{n \text{ times}}(0).$$

We know that $G_Z(s) = 0.1 + 0.6s + 0.3s^2$, and using a computer, $\mathbb{P}[Z_{10} = 0] = \underbrace{G_Z \circ \cdots \circ G_Z}_{10 \text{ times}}(0) = 0.3109$.

```
n_sim = 100000
p_0 = 0.1
p_1 = 0.6
p 2 = 0.3
# This function returns Z_n (n = 1,2,3, ...)
branch = function(n,p_0,p_1,p_2){
  if (n==1){
    # The "1" here represents Z_0 = 1
    return (sample(c(0,1,2), 1, replace=TRUE, prob=c(p_0,p_1,p_2)))
  # This is Z_{n-1}
  prev = branch(n-1,p_0,p_1,p_2)
  if (prev == 0){
   return (0)
  # This is (Z_{n-1,1}, \ldots, Z_{n-1,Z_{n-1}})
  offsprings = sample(c(0,1,2), prev, replace=TRUE, prob=c(p_0,p_1,p_2))
  # Return Z_n
  return (sum(offsprings))
}
# These are the results of Z 10 from the 100000 simulations
Z_10 = replicate(n_sim, branch(10,p_0,p_1,p_2))
# Probability of extinction (Theoretical)
the_prob = 0.3109
# Probability of extinction (Simulation)
sim_prob = mean(Z_10==0)
cat(" The probability of extinction by the 10th generation", "\n", "\n",
      "Probability (Theoretical) :", round(the_prob, 4), "\n",
      "Probatility (Simulation) :", round(sim_prob, 4))
  The probability of extinction by the 10th generation
##
##
## Probability (Theoretical): 0.3109
## Probatility (Simulation) : 0.3101
```

End of Homework