

**Yau College Math Competition 2021**  
**Final Probability and Statistics**  
**Individual Exam Problems (May 29, 2021)**

**Problem 1.** Let  $\{X_n\}_{n \geq 1}$  be a sequence of real valued, nonnegative random variables. Assume that there are constants  $C > 0$  and  $\lambda > 0$  such that  $\mathbb{E}X_n \leq Ce^{-\lambda n}$ ,  $\forall n \geq 1$ . Prove that

$$P\left(\limsup_{n \rightarrow \infty} \frac{1}{n} \ln X_n \leq -\lambda\right) = 1.$$

**Solution**

For any  $\lambda_0 \in (0, \lambda)$ , define the events

$$A_n = \{\omega \in \Omega : X_n(\omega) > e^{-\lambda_0 n}\}, \quad n \geq 1.$$

By Chebyshev's inequality,

$$\mathbb{P}(A_n) \leq e^{\lambda_0 n} \mathbb{E}X_n \leq Ce^{(\lambda_0 - \lambda)n}, \quad \forall n \geq 1.$$

Since  $\lambda_0 < \lambda$ , we have

$$\sum_{n=1}^{\infty} \mathbb{P}(A_n) \leq \sum_{n=1}^{\infty} Ce^{(\lambda_0 - \lambda)n} < +\infty.$$

Borel-Cantelli's lemma implies that for  $\mathbb{P}$ -a.s.  $\omega \in \Omega$ , there exists  $n(\omega) \in \mathbb{N}$  such that for all  $n \geq n(\omega)$ , we have  $\omega \in A_n^c$ , that is  $X_n(\omega) \leq e^{-\lambda_0 n}$ . Therefore,

$$\frac{1}{n} \ln X_n(\omega) \leq -\lambda_0, \quad \forall n \geq n(\omega).$$

This implies the desired result since  $\lambda_0$  is an arbitrary number less than  $\lambda$ .

**Problem 2.** Assume that  $X_1, \dots, X_n \sim U[0, 1]$  (uniform distribution) are i.i.d. Denote  $X_{(1)} = \min_{1 \leq k \leq n} X_k$  and  $X_{(n)} = \max_{1 \leq k \leq n} X_k$ . Let  $R = X_{(n)} - X_{(1)}$  be the sample range and  $V = (X_{(1)} + X_{(n)})/2$  be the sample midvalue.

- (1). Find the joint density of  $(X_{(1)}, X_{(n)})$ .
- (2). Find the joint density of  $(R, V)$ .
- (3). Find the density of  $R$  and the density of  $V$ .

**Solution**

(1). Denote  $F(x_1, x_n) = P(X_{(1)} \leq x_1, X_{(n)} \leq x_n)$ , then  $F(x_1, x_n) = 0$  for  $x_1 \notin [0, 1]$  or  $x_n \notin [0, 1]$ . If  $x_1 \geq x_n$ , then  $\{X_{(n)} \leq x_n\} \subset \{X_{(1)} \leq x_1\}$ , and therefore

$$F(x_1, x_n) = P(X_{(n)} \leq x_n).$$

If  $0 \leq x_1 \leq x_n \leq 1$ , then

$$\begin{aligned} P(X_{(1)} \geq x_1, X_{(n)} \leq x_n) &= P(\cup_{k=1}^n \{x_1 \leq X_k \leq x_n\}) \\ &= \prod_{k=1}^n P(x_1 \leq X_k \leq x_n) \\ &= (x_n - x_1)^n, \end{aligned}$$

which implies that

$$\begin{aligned} F(x_1, x_n) &= P(X_{(n)} \leq x_n) - P(X_{(1)} \geq x_1, X_{(n)} \leq x_n) \\ &= P(X_{(n)} \leq x_n) - (x_n - x_1)^n. \end{aligned}$$

Thus,

$$\begin{aligned} f(x_1, x_n) &= \frac{\partial^2 F(x_1, x_n)}{\partial x_1 \partial x_n} \\ &= \begin{cases} n(n-1)(x_n - x_1)^{n-2}, & \text{if } 0 \leq x_1 \leq x_n \leq 1, \\ 0, & \text{elsewhere.} \end{cases} \end{aligned}$$

(2). Note that

$$\begin{pmatrix} X_{(1)} \\ X_{(n)} \end{pmatrix} = \begin{pmatrix} -\frac{1}{2} & 1 \\ \frac{1}{2} & 1 \end{pmatrix} \begin{pmatrix} R \\ V \end{pmatrix} \equiv A \begin{pmatrix} R \\ V \end{pmatrix},$$

thus the joint density of  $(R, V)$  is

$$\begin{aligned} f_{R,V}(r, v) &= f(x_1, x_n) \times |\det A| \\ &= f\left(v - \frac{r}{2}, v + \frac{r}{2}\right) \\ &= n(n-1)r^{n-2}, \end{aligned}$$

where  $(r, v) \in D \equiv \{(r, v) : 0 \leq v - \frac{r}{2} \leq v + \frac{r}{2} \leq 1\}$  and

$$f_{R,V}(r, v) = 0,$$

if  $(r, v) \notin D$ .

(3) The density of  $R$  is

$$\begin{aligned} f_R(r) &= \int_{-\infty}^{+\infty} f_{R,V}(r, v) dv \\ &= \int_{r/2}^{1-r/2} f_{R,V}(r, v) dv = n(n-1)r^{n-2}(1-r), \quad 0 \leq r \leq 1. \end{aligned}$$

For the density of  $V$ , if  $v \in [0, 1/2]$ , then

$$f_V(v) = \int_{-\infty}^{+\infty} f_{R,V}(r, v) dr = \int_0^{2v} n(n-1)r^{n-2} dr = n(2v)^{n-1},$$

if  $v \in [1/2, 1]$ , then

$$f_V(v) = \int_{-\infty}^{+\infty} f_{R,V}(r, v) dr = \int_0^{2(1-v)} n(n-1)r^{n-2} dr = n(2(1-v))^{n-1}.$$

**Problem 3.** A binary tree is a tree in which each node has exactly two descendants. Suppose that each node of the tree is coloured black with probability  $p$ , and white otherwise, independently of all other nodes. For any path  $\pi$  containing  $n$  nodes beginning at the root of the tree, let  $B(\pi)$  be the number of black nodes in  $\pi$ , and let  $X_n(k)$  be the number of such paths  $\pi$  for which  $B(\pi) \geq k$ .

(1) Show that there exists  $\beta_c$  such that

$$\lim_{n \rightarrow \infty} \mathbb{E}(X_n(\beta n)) = \begin{cases} 0, & \text{if } \beta > \beta_c, \\ \infty, & \text{if } \beta < \beta_c. \end{cases}$$

How to determine the value of  $\beta_c$ ?

(2) For  $\beta \neq \beta_c$ , find the limit  $\lim_{n \rightarrow \infty} \mathbb{P}(X_n(\beta n) \geq 1)$ .

### Solution

The number of paths  $\pi$  containing exactly  $n$  nodes is  $2^{n-1}$ , and each such  $\pi$  satisfies  $\mathbb{P}(B(\pi) \geq k) = \mathbb{P}(S_n \geq k)$  where  $S_n = Y_1 + Y_2 + \cdots + Y_n$  is the sum of  $n$  independent Bernoulli variables having parameter  $p$ . Therefore  $\mathbb{E}(X_n(k)) = 2^{n-1} \mathbb{P}(S_n \geq k)$ . We set  $k = n\beta$ , and need to estimate  $\mathbb{P}(S_n \geq n\beta)$ . It is a consequence of the large deviation theorem that, if  $p \leq \beta < 1$ ,

$$\mathbb{P}(S_n \geq n\beta)^{1/n} \xrightarrow{n \rightarrow \infty} \inf_{t > 0} \left\{ e^{-t\beta} M(t) \right\}$$

where  $M(t) = \mathbb{E}(e^{tY_1}) = q + pe^t$ ,  $q = 1 - p$ . With some calculus, we find that

$$\mathbb{P}(S_n \geq n\beta)^{1/n} \xrightarrow{n \rightarrow \infty} \left( \frac{p}{\beta} \right)^\beta \left( \frac{1-p}{1-\beta} \right)^{1-\beta}, \quad p \leq \beta < 1$$

Hence

$$\mathbb{E}(X_n(\beta n)) \xrightarrow{n \rightarrow \infty} \begin{cases} 0, & \text{if } \gamma(\beta) < 1 \\ \infty, & \text{if } \gamma(\beta) > 1 \end{cases}$$

where

$$\gamma(\beta) = 2 \left( \frac{p}{\beta} \right)^\beta \left( \frac{1-p}{1-\beta} \right)^{1-\beta}$$

is a decreasing function of  $\beta$ . If  $p < \frac{1}{2}$ , there is a unique  $\beta_c \in [p, 1)$  such that  $\gamma(\beta_c) = 1$ ; if  $p \geq \frac{1}{2}$  then  $\gamma(\beta) > 1$  for all  $\beta \in [p, 1)$  so that we may take  $\beta_c = 1$ .

Turning to the final part,

$$\mathbb{P}(X_n(\beta n) \geq 1) \leq \mathbb{E}(X_n(\beta n)) \xrightarrow{n \rightarrow \infty} 0, \quad \text{if } \beta > \beta_c.$$

As for the other case, we will use the Payley-Zygmund inequality

$$\mathbb{P}(N \neq 0) \geq \frac{\mathbb{E}(N)^2}{\mathbb{E}(N^2)}$$

for nonnegative random variable  $N$ .

We have that  $\mathbb{E}(X_n(\beta n)^2) = \sum_{\pi, \rho} \mathbb{E}(I_\pi I_\rho)$ , where the sum is over all such paths  $\pi, \rho$ , and  $I_\pi$  is the indicator function of the event  $\{B(\pi) \geq \beta n\}$ . Hence

$$\mathbb{E}(X_n(\beta n)^2) = \sum_{\pi} \mathbb{E}(I_\pi) + \sum_{\pi \neq \rho} \mathbb{E}(I_\pi I_\rho) = \mathbb{E}(X_n(\beta n)) + 2^{n-1} \sum_{\rho \neq L} \mathbb{E}(I_L I_\rho)$$

where  $L$  is the path which always takes the left fork (there are  $2^{n-1}$  choices for  $\pi$ , and by symmetry each provides the same contribution to the sum). We divide up the last sum according to the number of nodes in common to  $\rho$  and  $L$ , obtaining  $\sum_{m=1}^{n-1} 2^{n-m-1} \mathbb{E}(I_L I_M)$  where  $M$  is a path having exactly  $m$  nodes in common with  $L$ . Now

$$\mathbb{E}(I_L I_M) = \mathbb{E}(I_M \mid I_L = 1) \mathbb{E}(I_L) \leq \mathbb{P}(T_{n-m} \geq \beta n - m) \mathbb{E}(I_L),$$

where  $T_{n-m}$  has the Binomial( $n-m, p$ ) distribution (the 'most value' to  $I_M$  of the event  $\{I_L = 1\}$  is obtained when all  $m$  nodes in  $L \cap M$  are black). However

$$\mathbb{E}(I_M) = \mathbb{P}(T_n \geq \beta n) \geq p^m \mathbb{P}(T_{n-m} \geq \beta n - m),$$

so that  $\mathbb{E}(I_L I_M) \leq p^{-m} \mathbb{E}(I_L) \mathbb{E}(I_M)$ . It follows that  $N = X_n(\beta n)$  satisfies

$$\mathbb{E}(N^2) \leq \mathbb{E}(N) + 2^{n-1} \sum_{m=1}^{n-1} 2^{n-m-1} \cdot \frac{1}{p^m} \mathbb{E}(I_L) \mathbb{E}(I_M) = \mathbb{E}(N) + \frac{1}{2} (\mathbb{E}(N))^2 \sum_{m=1}^{n-1} \left( \frac{1}{2p} \right)^m$$

whence, by the Payley-Zygmund inequality,

$$\mathbb{P}(N \neq 0) \geq \frac{1}{\mathbb{E}(N)^{-1} + \frac{1}{2} \sum_{m=1}^{n-1} (2p)^{-m}}.$$

If  $\beta < \beta_c$  then  $\mathbb{E}(N) \rightarrow \infty$  as  $n \rightarrow \infty$ . It is immediately evident that  $\mathbb{P}(N \neq 0) \rightarrow 1$  if  $p \leq \frac{1}{2}$ . Suppose finally that  $p > \frac{1}{2}$  and  $\beta < \beta_c$ . By the above inequality,

$$\mathbb{P}(X_n(\beta n) > 0) \geq c(\beta), \quad \forall n \quad (0.1)$$

where  $c(\beta)$  is some positive constant. Take  $\epsilon > 0$  such that  $\beta + \epsilon < \beta_c$ . Fix a positive integer  $m$ , and let  $\mathcal{P}_m$  be a collection of  $2^m$  disjoint paths each of length  $n - m$  starting from depth  $m$  in the tree. Now

$$\mathbb{P}(X_n(\beta n) = 0) \leq \mathbb{P}(B(v) < \beta n \text{ for all } v \in \mathcal{P}_m) = \mathbb{P}(B(v) < \beta n)^{2^m},$$

where  $v \in \mathcal{P}_m$ . However

$$\mathbb{P}(B(v) < \beta n) \leq \mathbb{P}(B(v) < (\beta + \epsilon)(n - m))$$

if  $\beta n < (\beta + \epsilon)(n - m)$ , which is to say that  $n \geq (\beta + \epsilon)m/\epsilon$ . Hence, for all large  $n$ ,

$$\mathbb{P}(X_n(\beta n) = 0) \leq (1 - c(\beta + \epsilon))^{2^m}$$

by (0.1). We let  $n \rightarrow \infty$  and  $m \rightarrow \infty$  in that order, to obtain  $\mathbb{P}(X_n(\beta n) = 0) \rightarrow 0$  as  $n \rightarrow \infty$ . In summary,

$$\mathbb{P}(X_n(\beta n) \geq 1) \xrightarrow{n \rightarrow \infty} \begin{cases} 0, & \text{if } \beta > \beta_c, \\ 1, & \text{if } \beta < \beta_c. \end{cases}$$