# MATH 505B Homework 4

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### Problem 6.14.1

$$\langle \mathbf{x}, \mathbf{P} \mathbf{y} \rangle = \sum_{k \in \theta} x_k (\mathbf{P} \mathbf{y})_k \pi_k$$

$$= \sum_{k \in \theta} x_k (\sum_j p_{kj} y_j) \pi_k$$

$$= \sum_{k \in \theta} x_k (\sum_j p_{kj} \pi_k y_j)$$

$$= \sum_{k,j} x_k p_{kj} \pi_k y_j$$

$$= \sum_{k,j} x_k (p_{jk} \pi_j y_j) \text{ using reversibility criterion } \pi_j p_{jk} = \pi_k p_{kj}$$

$$= \sum_j p_{jk} x_k \pi_j y_j$$

$$= \sum_j p_{jk} x_k \pi_j y_j$$

$$= \sum_j (\sum_k p_{jk} x_k) \pi_j y_j$$

$$= \langle \mathbf{P} \mathbf{x}, \mathbf{y} \rangle$$

# Problem 6.14.2

For reversibility:  $\pi_i p_{ij} = \pi_j p_{ji}$ :  $p'_{ij} = b_{ij} g_{ij} = \frac{\pi_j g_{ji} g_{ij}}{\pi_i g_{ij} + \pi_j g_{ji}}$  and  $p'_{ji} = \frac{\pi_i g_{ij} g_{ij}}{\pi_i g_{ij} + \pi_j g_{ji}}$ Hence,  $\pi_i p'_{ij} = \pi_j p'_{ji}$  and hence  $b_{ij}$  satisfies reversibility criterion besides  $0 \le b_{ij} \le 1$ .

### Problem 7.2.1

#### 7.2.1(a)

$$\{E(|X+Y|^p)\}^{1/p} = \{E|X^p|\}^{1/p} + \{E|Y^p|^{1/p}\}$$

$$E[|X|] = E[|X_n + X - X_n|]$$

$$\{E[|X|^p]\}^{1/p} = \{E[|X_n + X - X_n|^p]\}^{1/p}$$

$$\leq \{E[|X_n|^p]\}^{1/p} + \{E[|X - X_n|^p]\}^{1/p}$$

$$\implies \lim_{n \to \infty} \inf E[|X|^p] \leq E[|X_n|^p]^{1/p}$$
(1)

Similarly,

$$\begin{aligned}
\{E[|X_n|^p]\}^{1/p} &= \{E[|X_n - X + X|^p]\}^{1/p} \\
&\leq \{E[|X|^p]\}^{1/p} + \{E[|X_n - X|]\}^{1/p} \\
&\implies \lim_{n \to \infty} \sup E[|X_n|^p] \leq E[|X|^p]
\end{aligned} (2)$$

Combining 1, 2:  $E[|X_n|^p] \longrightarrow E[|X|^p] p \ge 1$ 

#### 7.2.1(b)

Using p = 1 in part (a)

#### 7.2.1(c)

Using part (a)  $E[X_n^2] \longrightarrow E[X^2]$   $(X_n \xrightarrow{2} X) \Rightarrow (X_n \xrightarrow{1} X) \implies E[X_n] \longrightarrow E[X]$  and hence  $Var(X_n) \longrightarrow Var(X)$ 

### Problem 7.2.3

Consider  $k \ge 0$ ,  $n \ge 1$ ,  $X_n = k/n \le X < (k+1)/n$ .  $X - 1/n \le X_n \le X$ Define similarly  $Y_n$ .  $Y_n, X_n$  are independent by definition

$$E[X_n] \longrightarrow E[X]$$
 and  $E[Y_n] \longrightarrow E[Y]$ 

Thus, using independence and convergence relations  $E[X_nY_n]=E[X_n]E[Y_n]\longrightarrow E[X]E[Y]$ 

Now,

$$(X - 1/n)(Y - 1/n) \le X_n Y_n \le XY$$

$$\Longrightarrow E[(X - 1/n)(Y - 1/n)] \le E[X_n Y_n] \le E[XY]$$

$$E[(X - 1/n)(Y - 1/n)] = E[XY - \frac{X + Y}{n} + 1/n^2] \implies E[X_n Y_n] \longrightarrow E[XY]$$

Thus combining, the above two results  $E[X_nY_n] \longrightarrow E[XY]$  and  $E[X_nY_n] \longrightarrow E[X][Y]$  we get E[XY] = E[X]E[Y]

### Problem 7.2.10

$$\sum_{r} X_r \sim Poisson(\sum_{r} \lambda_r)$$
 Define  $t = \sum_{r=1}^{n} \lambda_r$ :

$$P(\sum_{r} X_{r} \le x) = \sum_{i=0}^{x} \frac{e^{-t}t^{i}}{i!}$$

$$\lim_{n \to \infty} P(\sum_{r} X_{r} \le x) = \begin{cases} 0 & t \to \infty \\ Poisson(t) & t \text{is finite} \end{cases}$$

### Problem 7.4.1

$$E[X_1] = 0 * (1 - \frac{1}{n \log n} + 0 * \frac{1}{2n \log n}$$

$$= 0$$

$$E[X_1^2] = \frac{2n^2}{2n \log n} + 0 * (1 - \frac{1}{n \log n})$$

$$= \frac{n}{\log n}$$

$$E[(\frac{1}{n}S_n - 0)^2] = \frac{1}{n^2} Var(S_n)$$

$$= \frac{1}{n^2 \log n}$$

$$= \frac{1}{n \log n}$$

$$\longrightarrow 0$$

 $\sum_i P(|X_i| \ge i) \longrightarrow \infty$  Hence, using Borel-Cantelli Lemma(7.3.10b) we have  $P(|X_j| \ge j) = 1$  for some  $j |X_j| = |S_j - S_{j-1}| \ge j$  and hence  $S_j/j$  diverges.

# Problem 7.5.1

Define  $I_i(j)$  as the indicator variable denoting if the  $X_j$  lies in the  $i^{th}$  interval,

$$\log R_m = \sum_{i=1}^n Z_m(i) \log p_i$$
$$= \sum_{i=1}^n \sum_{j=1}^m I_i(j) p_i$$

Define  $\sum_{i=1}^{m} I_i(j) = Y_j$ , then  $\log R_m = \sum_{j=1}^{m} Y_j$  $E[Y_j] = \sum_{i=1}^{n} p_i \log p_i = -h$  Thus, by strong law of convergence  $\frac{1}{m} \sum_{j=1}^{m} Y_j = -h = E[Y_j]$ 

### Problem 7.5.3

Transient  $P(X_n = i | X_0 = i) < 1$ Using strong law  $S_n/n \longrightarrow E[X_1]$  If  $E[X_1] \neq 0$  then  $P[S_n = 0 | S_1 = 0] < 1$  as  $S_n = 0$  happens only finitely often

## Problem 7.7.1

$$\begin{split} E[X_i X_j] &= E[E[X_i X_j | X_0, X_1, \dots, X_{j-1}]] \\ &= E[E[X_i (S_j - S_{j-1}) | X_0, X_1, \dots, X_{j-1}]] \\ &= E[X_i (E[S_j - S_{j-1} | X_0, X_1, \dots, X_{j-1}])] \\ &= E[X_i (E[S_j | X_0, X_1, \dots, X_{j-1}] - S_{j-1})] \\ &= E[X_i (S_{j-1} - S_{j-1})] \\ &= 0 \end{split}$$

# Problem 7.7.3

$$E[X_{n+1}|X_0, X_1, \dots, X_n] = aX_n + X_{n-1}$$

$$E[S_{n+1}|X_0, X_1, \dots, X_n] = E[\alpha X_{n+1} + X_n | X_0, X_1, \dots, X_n]$$

$$= \alpha E[X_{n+1}|X_0, X_1, \dots, X_n] + X_n$$

$$= (\alpha a + 1)X_n + \alpha bX_{n-1}$$

$$= S_n = \alpha X_n + X_{n-1}$$

$$\implies \alpha = \frac{1}{1-a}, b = \frac{1}{\alpha}$$

# Problem 7.7.4

 $X_n$ : Net profit per unit stake on  $n^{th}$  play.  $S_i = S_{i-1} + f_i(X_1, X_2, \dots, X_i)$  such that  $S_1 = X_1 Y$  Thus,  $S_n = \sum_{i=1}^n X_i f_{i-1}(X_1, X_2, \dots, X_{i-1})$ 

$$S_{n+1} = S_n + f_{n+1}(X_1, X_2, \dots, X_n)X_{n+1}$$

$$E[S_{n+1} - S_n | X_1, X_2, \dots, X_n] = E[X_{n+1} f_{n+1}(X_1, X_2, \dots, X_n) | X_1, X_2, \dots, X_n]$$

$$= f_{n+1}(X_1, X_2, \dots, X_n)E[X_{n+1} | X_1, X_2, \dots, X_n]$$

$$= 0$$

$$\implies E[S_{n+1} | X_1, X_2, \dots, X_n] = S_n$$

# Problem 7.8.1

 $E[X_i] = 0$  By Doob-Kolmogorov inquality:

$$\begin{split} P(\max_{i \leq j \leq n} |S_j| > \epsilon) \leq \frac{1}{\epsilon^2} \sum_{j=1}^n E[S_n^2] \\ E[S_n^2] &= Var(S_n) + E[S_n]^2 \\ &= Var(S_n) \\ &= \sum Var(X_i) \\ \Longrightarrow P(\max_{i \leq j \leq n} |S_j| > \epsilon) \leq \frac{1}{\epsilon^2} \sum_{j=1}^n Var(X_j) \end{split}$$

# Problem 7.8.3

By theorem 7.8.1  $S_n$  converges to S almost surely Now, using the above proved fact that  $S_n \longrightarrow S \implies Var(S_n) \longrightarrow Var(S) \implies Var(S) \longrightarrow 0$