

## HOMEWORK 5 SOLUTIONS

### Problem 1

a) We can show this immediately using Corollary 6.4 from Chapter 1 of Durrett.

b) Let  $X = c$  and for any  $\epsilon > 0$  define the open interval  $A_\epsilon = (c - \epsilon, c + \epsilon)$ . The boundary  $\partial A_\epsilon = \{c - \epsilon, c + \epsilon\}$ , therefore  $\mathbb{P}\{X \in \partial A_\epsilon\} = 0$ . By the Portmanteau Theorem

$$\begin{aligned} 1 = \mathbb{P}\{X \in A_\epsilon\} &= \lim_{n \rightarrow \infty} \mathbb{P}\{X_n \in A_\epsilon\} \\ &= \lim_{n \rightarrow \infty} \mathbb{P}\{|X_n - c| < \epsilon\} \end{aligned}$$

**Problem 2:** Observe that  $f_n$  is always nonnegative. To show that is a density function we need to compute the integral:

$$\int_0^1 1 - \cos(2\pi nx) dx = \left[ x - \frac{\sin(2\pi nx)}{2\pi n} \right]_{x=0}^1 = 1.$$

For  $x \in [0, 1]$ , define

$$F_n(x) = x - \frac{\sin(2\pi nx)}{2\pi n}.$$

Observe that for all  $x \in [0, 1]$ ,

$$(1) \quad \lim_{n \rightarrow \infty} F_n(x) = F(x) = x$$

which is the distribution function of a uniform  $[0, 1]$  r.v.

We actually proved that for any point of continuity of  $F(x)$  the convergence (1) holds, therefore the  $\mu_n$  converge weakly to the Lebesgue measure.

**Problem 3:** Let  $\sigma^2 = \text{Var}(X_1)$ . Classical CLT implies that  $\frac{S_n}{\sigma\sqrt{n}} \implies X \sim N(0, 1)$ . Let  $M > 0$ . The set  $(\sigma M, +\infty)$  is open and so by the Portmanteau Theorem,

$$(2) \quad \liminf_{n \rightarrow \infty} \mathbb{P}\left\{\frac{S_n}{\sigma\sqrt{n}} > M\right\} \geq \mathbb{P}\{X > M\} = 1 - \Phi(M) > 0.$$

Hence there exists an  $\epsilon > 0$  and  $N = N(\epsilon)$  such that for all  $n > N$ ,

$$\mathbb{P}\left\{\frac{S_n}{\sqrt{n}} > \sigma M\right\} = \epsilon > 0.$$

An easy lemma from measure theory shows that if  $\mathbb{P}\{A_n\} \geq \epsilon$  for all  $n$ , then  $\mathbb{P}\{A_n \text{ i.o.}\} \geq \epsilon$ . (Why?). Therefore

$$(3) \quad \mathbb{P}\left\{\frac{S_n}{\sqrt{n}} > \sigma M \text{ i.o.}\right\} \geq \epsilon > 0.$$

To finish the proof, observe that by Kolmogorov's (or Hewitt-Savage) 0-1 law we have

$$(4) \quad \mathbb{P} \left\{ \limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{n}} > \sigma M \right\} \in \{0, 1\}.$$

and since

$$(5) \quad \mathbb{P} \left\{ \limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{n}} > \sigma M \right\} = \mathbb{P} \left\{ \frac{S_n}{\sqrt{n}} > \sigma M \text{ i.o.} \right\} > 0,$$

it has to be that  $\limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{n}} > \sigma M$  a.s. Since  $M$  was arbitrary take  $M$  to infinity.

**Problem 4:** Define  $X_{n,j}$  to be the number of trials required to collect  $j$  coupons ( $j \geq 1$ ), if we already collected  $j-1$  coupons. Then we have that

$$(1) \quad \tau_k^n = \sum_{j=1}^k X_{n,j}.$$

$$(2) \quad X_{n,1} \equiv 1, \mathbb{P}\{X_{n,j} = l\} = \left(1 - \frac{j-1}{n}\right) \left(\frac{j-1}{n}\right)^{l-1} \text{ for } l \geq 1$$

$$(3) \quad \mathbb{E}X_{n,j} = \frac{n}{n-j+1}$$

$$(4) \quad \text{Var}(X_{n,j}) = \frac{j-1}{n} \left(1 - \frac{j-1}{n}\right)^{-2}.$$

Therefore,  $S_n = \sum_{j=1}^{\lfloor n\theta \rfloor} X_{n,j}$ . We compute

$$\begin{aligned} \mathbb{E}S_n &= n \sum_{j=1}^{\lfloor n\theta \rfloor} \frac{1}{n-j+1} = n \sum_{j=n-\lfloor n\theta \rfloor+1}^n \frac{1}{j} \\ &\sim n(\log n - \log(n - \lfloor n\theta \rfloor + 1)) \sim n \log \frac{1}{1-\theta}. \end{aligned}$$

$$\begin{aligned} \text{Var}(S_n) &= n \sum_{j=1}^{\lfloor n\theta \rfloor} \frac{j-1}{(n-j+1)^2} = n \sum_{j=n-\lfloor n\theta \rfloor+1}^n \frac{n-j}{j^2} \\ &= \sum_{j=n-\lfloor n\theta \rfloor+1}^n \frac{1-j/n}{(j/n)^2} \sim n \int_{1-\theta}^1 \frac{1-x}{x^2} dx \\ &\sim n \left( \frac{\theta}{1-\theta} - \log \frac{1}{1-\theta} \right). \end{aligned}$$

We would like to apply Lindeberg - Feller on the sequence  $\frac{S_n - \mathbb{E}S_n}{\sqrt{n}}$ . We check the two conditions:

$$(i) \quad \sum_{j=1}^{\lfloor n\theta \rfloor} \frac{\text{Var}(X_{n,j})}{n} \longrightarrow \frac{\theta}{1-\theta} - \log \frac{1}{1-\theta} = \sigma^2 > 0 \text{ (why?)}$$

(ii) This requires to show  $\lim_{n \rightarrow \infty} n^{-1} \sum_{j=1}^{\lfloor n\theta \rfloor} \mathbb{E}((X_{n,j} - \mathbb{E}X_{n,j})^2 \mathbf{1}\{|X_{n,j} - \mathbb{E}X_{n,j}| \geq \varepsilon\sqrt{n}\}) = 0$ .

To obtain this, let us show that for the  $X_{n,j}$ 's we care about, all moments are bounded. Note that

$$\mathbb{P}\{X_{n,j} \geq l\} = \sum_{k=l}^{\infty} \left(1 - \frac{j-1}{n}\right) \left(\frac{j-1}{n}\right)^{k-1} = \left(\frac{j-1}{n}\right)^{l-1} \leq \theta^{l-1}$$

Hence

$$\begin{aligned} \mathbb{E}(X_{n,j}^p) &= \int_0^{+\infty} \mathbb{P}\{X_{n,j} > t\} p t^{p-1} dt \leq \int_0^{+\infty} \theta^{\lceil t \rceil - 1} p t^{p-1} dt \\ &= \theta^{-1} \int_0^{\infty} e^{t \log \theta} p t^{p-1} dt = C(p, \theta) < +\infty \end{aligned}$$

since  $0 < \theta < 1$ . Consequently

$$\begin{aligned} \mathbb{E}((X_{n,j} - \mathbb{E}X_{n,j})^2 \mathbf{1}\{|X_{n,j} - \mathbb{E}X_{n,j}| \geq \varepsilon\sqrt{n}\}) &\leq (\mathbb{E}|X_{n,j} - \mathbb{E}X_{n,j}|^4)^{1/2} \mathbb{P}\{|X_{n,j} - \mathbb{E}X_{n,j}| \geq \varepsilon\sqrt{n}\}^{1/2} \\ &\leq C \frac{1}{\varepsilon\sqrt{n}} \mathbb{E}|X_{n,j} - \mathbb{E}X_{n,j}| \\ &\leq \frac{C}{\varepsilon\sqrt{n}}. \end{aligned}$$

Now for (ii):

$$\lim_{n \rightarrow \infty} n^{-1} \sum_{j=1}^{\lfloor n\theta \rfloor} \mathbb{E}((X_{n,j} - \mathbb{E}X_{n,j})^2 \mathbf{1}\{|X_{n,j} - \mathbb{E}X_{n,j}| \geq \varepsilon\sqrt{n}\}) \leq \lim_{n \rightarrow \infty} n^{-1} \lfloor n\theta \rfloor \frac{C}{\varepsilon\sqrt{n}} = 0.$$

We conclude that  $\frac{S_n - \mathbb{E}S_n}{\sqrt{n}} \xrightarrow{w} N(0, \sigma^2)$ .