

ECE534, Spring 2020: Problem Set #3  
Due Mar 27, 2020

1. All Types of Convergence of Random Sequences

Consider a sequence of random variables  $X_1, X_2, \dots$  with its generic term defined as follows:

$$X_1 \sim \text{Ber}(1/2) \text{ and } X_n = (X_{n-1} + 1) \bmod 2.$$

Does this sequence converge almost surely, in probability, in mean square and in distribution? Justify your answer for each type of convergence separately.

2. Convergence in Probability

Prove that  $X_n \xrightarrow[n \rightarrow \infty]{p} X$  if and only if

$$\lim_{n \rightarrow \infty} E \left[ \frac{|X_n - X|}{1 + |X_n - X|} \right] = 0.$$

3. Almost Sure Convergence [Extra Credit]

Let  $X_1, X_2, \dots$  be a sequence of i.i.d. random variables with  $P(X_n > x) = e^{-x}, x \geq 0$ . Show that

$$\frac{\max\{X_1, X_2, \dots, X_n\}}{\log n} \xrightarrow[n \rightarrow \infty]{\text{a.s.}} 1.$$

**Note:** Use the Borel-Cantelli Lemma.

4. Maximum of a Finite Set of sub-Gaussian Random Variables

A random variable  $X \in \mathbb{R}$  is called *sub-Gaussian* with *variance proxy*  $\sigma^2$  if  $E[X] = 0$  and its moment generating function satisfies:

$$m_X(u) = E[e^{uX}] \leq e^{\frac{u^2 \sigma^2}{2}}, \quad \forall u \in \mathbb{R}.$$

We write  $X \sim \text{subG}(\sigma^2)$ .

(a) Use the Chernoff bound to show that

$$P(X > t) \leq e^{-\frac{t^2}{2\sigma^2}}, \quad \forall t > 0$$

when  $X \sim \text{subG}(\sigma^2)$ .

(b) Let  $X_1, X_2, \dots, X_n$  be  $\text{subG}(\sigma^2)$  random variables, not necessarily independent. Show that the expectation of the maximum can be bounded as

$$E[\max_{1 \leq i \leq n} X_i] \leq \sigma \sqrt{2 \log n}.$$

**Hint:** Start your derivation by noting that

$$E[\max_{1 \leq i \leq n} X_i] = \frac{1}{\lambda} E \left[ \log \left( e^{\lambda \max_{1 \leq i \leq n} X_i} \right) \right], \quad \forall \lambda > 0.$$

(c) With the same assumptions as in the previous part, show that

$$P\left(\max_{1 \leq i \leq n} X_i > t\right) \leq ne^{-\frac{t^2}{2\sigma^2}}, \quad \forall t > 0.$$

## 5. Dominated Convergence

Suppose that  $|X_n| \leq Y, \forall n \geq 1$  and  $E[Y] < \infty$ . Show that if  $X_n \xrightarrow[n \rightarrow \infty]{p} X$ , then  $E[|X_n - X|] \xrightarrow[n \rightarrow \infty]{} 0$  (convergence of  $X_n$  in mean to  $X$  or in the  $L^1$  norm).

**Notes:**

- For a random variable  $Z$  the following holds:  $E[|Z|] < \infty$  if and only if  $\forall \epsilon > 0, \exists \delta > 0$  such that  $E[|Z|\mathbb{1}_A] < \epsilon$  for all events  $A$  such that  $P(A) < \delta$ .
- Define  $V_n = |X_n - X|$ . Start your derivation using Note (b) in Problem 7 for  $C = \{V_n \leq \epsilon\}$ , where  $\epsilon > 0$  is fixed. Conclude your derivation using part (a) by allowing  $\epsilon$  to decrease to zero.

## 6. Sequences of Random Variables and Convergence Sets

Let  $(X_n)_{n \geq 1}$  be a sequence of random variables on  $(\Omega, \mathcal{F}, P)$ . Consider the set  $A = \{\omega \in \Omega : (X_n(\omega))_{n \geq 1} \text{ converges}\}$ . Show that  $A \in \mathcal{F}$  (i.e.,  $A$  is an event) and that there exists a random variable  $X : \Omega \rightarrow \mathbb{R}$ , i.e., an  $\mathcal{F}$ -measurable mapping, such that  $X_n(\omega) \rightarrow X(\omega), \forall \omega \in A$ .

**Hint:** To show that  $A \in \mathcal{F}$ , try to express it in terms of countable unions and intersections. Based on this result, identify  $X$ .

## 7. Polynomial Approximation of Continuous Functions [Extra Credit]

Consider a continuous mapping  $g : [0, 1] \rightarrow \mathbb{R}$ . Show that

$$\lim_{n \rightarrow \infty} \sup_{0 \leq x \leq 1} \left| \sum_{k=0}^n g\left(\frac{k}{n}\right) \binom{n}{k} x^k (1-x)^{n-k} - g(x) \right| = 0.$$

**Notes:**

- The above statement corresponds to the so-called *Weierstrass Approximation Theorem*. It states that every continuous function  $g$  on a closed interval can be approximated *uniformly* over this interval by a polynomial.
- Start your derivation by letting  $Y_n \sim \text{Bin}(n, x)$ . Define the random variable  $X_n = g(x) - g\left(\frac{Y_n}{n}\right)$ .
- Note that for every random variable  $V$ ,  $E[V] = E[V\mathbb{1}_C] + E[V\mathbb{1}_{C^c}]$  for an event  $C$ . In the context of this problem, define  $V = X_n$  and  $C = \{|\frac{Y_n}{n} - x| > \delta\}$ . The goal is then to show that  $|E[X_n]| = O(\epsilon)$  for a sufficiently large  $n$ .  $P(C)$  can be bounded using known inequalities.
- Every continuous function  $g$  on a closed interval is *bounded*.
- $g$  uniformly continuous on  $[0, 1]$ :  $\forall \epsilon > 0, \exists \delta > 0$  such that  $|g(x) - g(\tilde{x})| < \epsilon$  for  $|x - \tilde{x}| < \delta$  with  $x, \tilde{x} \in [0, 1]$ .