

STAT 205A - Problem Set 04

Walid Krichene (23265217)

October 1, 2013

(4.1. Monte Carlo Integration) Let $f : [0, 1] \rightarrow \mathbb{R}$ be such that $\int_{[0,1]} f^2(x)dx < \infty$. Let U_i be i.i.d. Uniform(0, 1). Let

$$D_n = n^{-1} \sum_{i=1}^n f(U_i) - \int_0^1 f(x)dx$$

1. Use Chebyshev's inequality to bound $P(|D_n| > \epsilon)$

answer The density function of U is simply the indicator function of $[0, 1]$. Therefore the expectation of D_n is given by

$$\mathbb{E} D_n = n^{-1} \sum_{i=1}^n \int_{u=0}^1 f(u)du - \int_{x=0}^1 f(x)dx = 0$$

Now by Chebyshev's inequality, we have

$$P(|D_n| > \epsilon) \leq \frac{1}{\epsilon^2} \text{var}(D_n)$$

writing $\int_0^1 f(x)dx = \mathbb{E} f(U)$, we have

$$\begin{aligned} \text{var}(D_n) &= \mathbb{E} \left(\frac{1}{n} \sum_{i=1}^n f(U_i) - \mathbb{E} f(U) \right)^2 && \text{since } \mathbb{E} D_n = 0 \\ &= \frac{1}{n^2} \mathbb{E} \left(\sum_{i=1}^n f(U_i) \right)^2 + (\mathbb{E} f(U))^2 - \frac{2}{n} \mathbb{E} f(U) \sum_{i=1}^n \mathbb{E} f(U_i) \\ &= \frac{1}{n^2} \left(\sum_{i=1}^n \mathbb{E} f^2(U_i) + 2 \sum_{1 \leq i < j \leq n} \mathbb{E}(f(U_i)f(U_j)) \right) + (\mathbb{E} f(U))^2 - 2(\mathbb{E} f(U))^2 \\ &= \frac{1}{n^2} \left(\sum_{i=1}^n \mathbb{E} f^2(U_i) + 2 \sum_{1 \leq i < j \leq n} \mathbb{E} f(U_i) \mathbb{E} f(U_j) \right) - (\mathbb{E} f(U))^2 && \text{by independence of } f(U_i)\text{'s} \\ &= \frac{1}{n} \mathbb{E} f^2(U) + \frac{2n(n-1)}{n^2} (\mathbb{E} f(U))^2 - (\mathbb{E} f(U))^2 \\ &= \frac{1}{n} \mathbb{E} f^2(U) + \frac{n-2}{n} (\mathbb{E} f(U))^2 \end{aligned}$$

which written in integral form, is

$$\text{var}(D_n) = \frac{1}{n} \int_{u=0}^1 f^2(u)du + \frac{n-2}{n} \left(\int_0^1 f(u)du \right)^2$$

2. Show this bound remains true if the U_i are only *pairwise* independent.

proof Pairwise independence is sufficient since in the above proof, it suffices that $E f(U_i)f(U_j) = E f(U_i) E f(U_j)$ for all i, j , which is true when U_i 's are pairwise independent.

(4.2) Let $X \geq 0$ and $Y \geq 0$ be independent r.v.'s with densities f_X and f_Y . Calculate the densities of XY and X/Y

answer

- Let f_{XY} be the density of XY . Then for all $c \in \mathbb{R}$,

$$P(XY \leq c) = \int_{u=0}^c f_{XY}(u) du$$

but we can also write

$$\begin{aligned} P(XY \leq c) &= \iint_{x \in \mathbb{R}_+, y \in \mathbb{R}_+} 1_{xy \leq c} f_X(x) f_Y(y) dx dy \\ &= \int_{x \in \mathbb{R}_+ \setminus \{0\}} \int_{y \in \mathbb{R}_+} 1_{xy \leq c} f_X(x) f_Y(y) dx dy && \text{since } \{0\} \text{ has measure } 0 \\ &= \int_{x \in \mathbb{R}_+ \setminus \{0\}} \left(\int_{y=0}^{c/x} f_Y(y) dy \right) f_X(x) dx \\ &= \int_{x \in \mathbb{R}_+ \setminus \{0\}} \left(\int_{u=c}^0 \frac{1}{x} f_Y(u/x) du \right) f_X(x) dx && \text{using the change of variable } u = xy \\ &= \int_{u=0}^c \left(\int_{x \in \mathbb{R}_+ \setminus \{0\}} \frac{1}{x} f_Y(u/x) f_X(x) dx \right) du && \text{by Fubini} \end{aligned}$$

therefore we have for all c ,

$$\int_{u=0}^c f_{XY}(u) du = \int_{u=0}^c \left(\int_{x \in \mathbb{R}_+ \setminus \{0\}} \frac{1}{x} f_Y(u/x) f_X(x) dx \right) du$$

and taking derivative with respect to c on both sides, we conclude that

$$f_{XY}(u) = \left(\int_{x \in \mathbb{R}_+ \setminus \{0\}} \frac{1}{x} f_Y(u/x) f_X(x) dx \right)$$

- Let $f_{X/Y}$ be the density of X/Y . Then for all $c \in \mathbb{R}$,

$$P(X/Y \leq c) = \int_{u=0}^c f_{X/Y}(u) du$$

but we can also write

$$\begin{aligned} P(X/Y \leq c) &= \iint_{x \in \mathbb{R}_+, y \in \mathbb{R}_+ \setminus \{0\}} 1_{x/y \leq c} f_X(x) f_Y(y) dx dy && \text{since } X \text{ and } Y \text{ are independent} \\ &= \int_{y \in \mathbb{R}_+ \setminus \{0\}} \left(\int_{x=0}^{yc} f_X(x) dx \right) f_Y(y) dy && \text{B Fubini} \\ &= \int_{y \in \mathbb{R}_+ \setminus \{0\}} \left(\int_{u=0}^c y f_X(uy) du \right) f_Y(y) dy && \text{using the change of variable } u = x/y \\ &= \int_{u=0}^c \left(\int_{y \in \mathbb{R}_+ \setminus \{0\}} y f_X(uy) f_Y(y) dy \right) du && \text{by Fubini} \end{aligned}$$

therefore we have for all c ,

$$\int_{u=0}^c f_{X/Y}(u) du = \int_{u=0}^c \left(\int_{y \in \mathbb{R}_+ \setminus \{0\}} y f_X(uy) f_Y(y) dy \right) du$$

and taking derivative with respect to c on both sides, we conclude that

$$f_{X/Y}(u) = \left(\int_{y \in \mathbb{R}_+ \setminus \{0\}} y f_X(uy) f_Y(y) dy \right)$$

(4.3) Let (X_i) be r.v.'s with $E X_i = 0$ and $E X_i X_j \leq r(j-i)$ for $1 \leq i \leq j < \infty$, where $r(n)$ is a deterministic sequence with $r(n) \rightarrow 0$ as $n \rightarrow \infty$. Prove that $n^{-1} \sum_{i=1}^n X_i \rightarrow 0$ in Probability.

proof Let $S_n = \sum_{i=1}^n X_i$. By Chebyshev's inequality, we have

$$\begin{aligned} P\left(\frac{|S_n|}{n} > \epsilon\right) &\leq \frac{1}{n^2 \epsilon^2} E[S_n^2] \\ &\leq \frac{1}{n^2 \epsilon^2} \sum_{i=1}^n \sum_{j=1}^n E[X_i X_j] && \text{by linearity of the expectation} \\ &\leq \frac{1}{n^2 \epsilon^2} \sum_{i=1}^n \sum_{j=1}^n r(|j-i|) \end{aligned}$$

Now for any fixed i ,

$$\sum_{j=1}^n r(|j-i|) = \sum_{j=1}^i r(i-j) + \sum_{j=i+1}^n r(j-i) = \sum_{k=0}^{i-1} r(k) + \sum_{l=1}^{n-i} r(l) \leq 2 \sum_{k=1}^n |r(k)|$$

Therefore

$$\begin{aligned} P\left(\frac{|S_n|}{n} > \epsilon\right) &\leq \frac{1}{n^2 \epsilon^2} 2n \sum_{k=1}^n |r(k)| \\ &= \frac{2}{\epsilon^2} \frac{\sum_{k=1}^n |r(k)|}{n} \end{aligned}$$

and since $(r(n))_n$ converges to 0, so does $(|r(n)|)_n$, and so does the sequence of Cesàro means $\left(\frac{\sum_{k=1}^n |r(k)|}{n}\right)_n$ (for any $\delta > 0$, there exists N such that for all $n > N$, $|r_n| < \delta$. Then we can write $\frac{1}{n} \sum_{k=1}^n |r_k| = \frac{1}{n} \sum_{k=1}^N |r_k| + \frac{n-N}{n} \delta \leq \frac{\sum_{k=1}^N |r_k|}{n} + \delta \leq 2\delta$ for n large enough).

Therefore $P\left(\frac{|S_n|}{n} > \epsilon\right)$ converges to 0 as $n \rightarrow \infty$

(4.4) Suppose events A_n satisfy $P(A_n) \rightarrow 0$ and

$$\sum_{n=1}^{\infty} P(A_n^c \cap A_{n+1}) < \infty$$

Prove that

$$P(A_n \text{ infinitely often}) = 0$$

proof Let us write $B_n = A_{n+1} \setminus A_n = A_{n+1} \cap A_n^c$.

By the first Borel-Cantelli Lemma, applied to the events B_n , we have $P(B_n \text{ infinitely often}) = 0$, i.e.

$$P(\cap_{m=1}^{\infty} \cup_{n \geq m} B_n) = 0$$

and since the sequence $(\cup_{n \geq m} B_n)_m$ is nested, we have $\lim_{m \rightarrow \infty} P(\cup_{n \geq m} B_n) = 0$. Now consider the union $\cup_{n \geq m} A_n$. We have

$$\cup_{n \geq m} A_n \subseteq A_m \cup \cup_{n \geq m} B_n \tag{1}$$

To prove this, we can show by induction on M that $\cup_{n=m}^M A_n \subseteq A_m \cup \cup_{n \geq m} B_n$:

- for $M = m$, we have $A_m \subseteq A_m \cup \cup_{n \geq m} B_n$
- suppose the statement is true for M , and consider $\cup_{n=m}^{M+1} A_n = A_{M+1} \cup \cup_{n=m}^M A_n$. It suffices to show the inclusion for A_{M+1} . Any element of A_{M+1} is either in A_M , which is a subset of $A_m \cup \cup_{n \geq m} B_n$ by induction hypothesis, or in $A_{M+1} \setminus A_M = B_M$ which is also a subset of $A_m \cup \cup_{n \geq m} B_n$. This concludes the induction.

Since the statement is true for all M , the inclusion remains true for the union, which proves (1). Now from (1), we have

$$P(\cup_{n \geq m} A_n) \leq P(A_m) + P(\cup_{n \geq m} B_n)$$

and both terms converge to 0. Therefore $\lim_{m \rightarrow \infty} P(\cup_{n \geq m} A_n) = 0$, and since the sequence is nested, this is equivalent to

$$P(\cap_{m=1}^{\infty} \cup_{n \geq m} A_n) = 0$$

i.e.

$$P(A_n \text{ infinitely often}) = 0$$

(4.5)

1. Let Z have standard Normal distribution. Show

$$P(Z > z) \sim z^{-1}(2\pi)^{-1/2} \exp(-z^2/2) \text{ as } z \rightarrow \infty$$

proof Let $g(z) = z^{-1}(2\pi)^{-1/2} \exp(-z^2/2)$ and

$$F_Z(z) = P(Z > z) = (2\pi)^{-1/2} \int_{u>z} \exp(-u^2/2) du$$

We have $\lim_{z \rightarrow \infty} g(z) = 0$, and $\lim_{z \rightarrow \infty} F_Z(z) = 0$ by the *DCT* (consider the sequence of functions $f_n(u) = (2\pi)^{-1/2} \int_{u>z} \exp(-u^2/2) 1_{u \geq n}$, which are dominated by F_Z and converge to 0 pointwise). Therefore By L'Hopital's rule, we have

$$\lim_{z \rightarrow \infty} \frac{F_Z(z)}{g(z)} = \lim_{z \rightarrow \infty} \frac{F'_Z(z)}{g'(z)}$$

where

$$\begin{aligned} F'_Z(z) &= -(2\pi)^{-1/2} \exp(-z^2/2) \\ g'(z) &= (2\pi)^{-1/2} \exp(-z^2/2) \left(-z \frac{1}{z} - \frac{1}{z^2}\right) \end{aligned}$$

thus

$$\lim_{z \rightarrow \infty} \frac{F_Z(z)}{g(z)} = \lim_{z \rightarrow \infty} \frac{F'_Z(z)}{g'(z)} = \lim_{z \rightarrow \infty} \frac{1}{1 + 1/z^2} = 1$$

which proves the result.

2. Let (Z_1, Z_2, \dots) be independent with standard Normal distribution. Find constants $c_n \rightarrow \infty$ such that

$$\limsup_n Z_n / c_n = 1 \text{ a.s.}$$

answer Let us find sufficient conditions on c_n for the lim sup to be equal to 1.

- Consider the events $(Z_n/c_n > 1 + \epsilon)$. By the first Borel-Cantelli Lemma, if $\sum_n P(Z_n/c_n > 1 + \epsilon) < \infty$, then $P(Z_n/c_n > 1 + \epsilon \text{ infinitely often}) = 0$, i.e. with probability 1, $Z_n/c_n \leq 1 + \epsilon$ ultimately, therefore with probability 1, $\limsup_n Z_n/c_n \leq 1 + \epsilon$.
- Consider the events $(Z_n/c_n > 1 - \epsilon)$. By the second Borel-Cantelli Lemma, if $\sum_n P(Z_n/c_n > 1 - \epsilon) = \infty$, then since Z_n/c_n are independent, we have $P(Z_n/c_n > 1 - \epsilon \text{ infinitely often}) = 1$, thus with probability 1, $\limsup_n Z_n/c_n \geq 1 - \epsilon$.

Therefore if c_n satisfies

- (a) $\sum_n P(Z_n/c_n > 1 + \epsilon) < \infty$
- (b) $\sum_n P(Z_n/c_n > 1 - \epsilon) = \infty$

then $1 - \epsilon \leq \limsup_n Z_n/c_n \leq 1 + \epsilon$ for all $\epsilon > 0$, and it would follow that $\limsup_n Z_n/c_n = 1$.

By the first part, since $P(Z_n < z) \sim z^{-1}(2\pi)^{-1/2} \exp(-z^2/2)$, it suffices that

- (a) $\sum_n \frac{1}{c_n} \exp(-\frac{1}{2}c_n^2(1 + \epsilon)^2) < \infty$
- (b) $\sum_n \frac{1}{c_n} \exp(-\frac{1}{2}c_n^2(1 - \epsilon)^2) = \infty$

An example of sequence that satisfies these conditions is $c_n = (2 \ln n)^{1/2}$, since

- (a) $\frac{1}{c_n} \exp(-\frac{1}{2}c_n^2(1 + \epsilon)^2) = \frac{1}{(2 \ln n)^{1/2}} \frac{1}{n^{(1+\epsilon)^2}}$ and its series converges since $(1 + \epsilon)^2 < 1$
- (b) $\frac{1}{c_n} \exp(-\frac{1}{2}c_n^2(1 - \epsilon)^2) = \frac{1}{(2 \ln n)^{1/2}} \frac{1}{n^{(1-\epsilon)^2}}$ and its series diverges since $(1 - \epsilon)^2 > 1$