

## Statistics 521, Problem Set 1, Solutions

Wellner; 10/3/12

- (a) Suppose that  $\{\mathcal{A}_n\}$  is an increasing sequence of algebras, i.e.  $\mathcal{A}_n \subset \mathcal{A}_{n+1}$  for all  $n \geq 1$ . Show that  $\cup_{n=1}^{\infty} \mathcal{A}_n$  is an algebra (i.e. a field).  
(b) Suppose that the  $\mathcal{A}_n$  of (a) are  $\sigma$ -algebras (i.e. a sigma-field). Show by constructing a counter-example that  $\cup_{n=1}^{\infty} \mathcal{A}_n$  need not be a  $\sigma$ -algebra (a sigma-field).

**Solution:** (a) If  $A \in \cup_{n=1}^{\infty} \mathcal{A}_n$ , then  $A \in \mathcal{A}_m$  for some  $m$ , and since  $\mathcal{A}_m$  is an algebra,  $A^c \in \mathcal{A}_m$ . Hence  $A^c \in \cup_{n=1}^{\infty} \mathcal{A}_n$ . If  $A, B \in \cup_{n=1}^{\infty} \mathcal{A}_n$ , then  $A \in \mathcal{A}_m$  for some  $m$  and  $B \in \mathcal{A}_n$  for some  $n$ . Without loss we can assume that  $m \leq n$ , and since  $\mathcal{A}_m \subset \mathcal{A}_n$  it follows that  $A, B \in \mathcal{A}_n$ . Since  $\mathcal{A}_n$  is an algebra, it follows that  $A \cup B \in \mathcal{A}_n$ , and hence that  $A \cup B \in \cup_{n=1}^{\infty} \mathcal{A}_n$ .

(b) Take  $\Omega = [0, 1]$ . Let  $\mathcal{A}_1 = \{\emptyset, \Omega\}$ ,  $\mathcal{A}_2 = \sigma[\mathcal{A}_0, [0, 1/2]]$ ,  $\dots$ ,  $\mathcal{A}_n = \sigma[\mathcal{A}_{n-1}, [0, 1 - 1/n]]$ ,  $\dots$ . Then  $\mathcal{A}_n \subset \mathcal{A}_{n+1}$  by construction, but  $\cup_{n=1}^{\infty} \mathcal{A}_n$  is not a sigma field: if we let  $A_k = [0, 1 - 1/k]$  for each  $k = 1, 2, \dots$ , then  $A_k \in \cup_{n=1}^{\infty} \mathcal{A}_n$  since  $A_k \in \mathcal{A}_k$  by construction, but  $[0, 1) = \cup_{k=1}^{\infty} A_k \notin \cup_{n=1}^{\infty} \mathcal{A}_n$ .

- Write out a proof of Proposition 1.1(j), PfS, page 3: There exists a minimal field,  $\sigma$ -field, or monotone class generated by (or containing) any specified class  $\mathcal{C}$  of subsets of  $\Omega$ .

**Solution:** By proposition 1.1.1(i), arbitrary intersections of fields,  $\sigma$ -fields, or monotone classes are again fields,  $\sigma$ -fields, or monotone classes. Hence

$$\phi[\mathcal{C}] \equiv \cap \{ \mathcal{A}_\alpha : \mathcal{A}_\alpha \text{ is a } \sigma\text{-field of subsets of } \Omega \text{ for which } \mathcal{C} \subset \mathcal{A}_\alpha \}$$

is again a field, and it is the smallest such field: if  $\mathcal{D}$  is the minimal field containing  $\mathcal{C}$  so that  $\mathcal{D} \subset \phi[\mathcal{C}]$ , then we also have  $\phi[\mathcal{C}] \subset \mathcal{D}$  by construction of  $\phi[\mathcal{C}]$ , and hence  $\phi[\mathcal{C}] = \mathcal{D}$ . The argument is the same for  $\sigma$ -fields and monotone classes with  $\phi[\mathcal{C}]$  replaced by  $\sigma[\mathcal{C}]$  and  $\text{mon}[\mathcal{C}]$  respectively.

3. PfS, Exercise 1.1.1, page 3. Let  $\mathcal{C}_1$  and  $\mathcal{C}_2$  denote two collections of subsets of the set  $\Omega$ . If  $\mathcal{C}_1 \subset \sigma[\mathcal{C}_2]$  and  $\mathcal{C}_2 \subset \sigma[\mathcal{C}_1]$ , then  $\sigma[\mathcal{C}_1] = \sigma[\mathcal{C}_2]$ .

**Solution:** Since  $\mathcal{C}_1 \subset \sigma[\mathcal{C}_2]$ , it follows immediately that  $\sigma[\mathcal{C}_1] \subset \sigma[\sigma[\mathcal{C}_2]] = \sigma[\mathcal{C}_2]$ . By a symmetric argument  $\sigma[\mathcal{C}_2] \subset \sigma[\mathcal{C}_1]$ . Hence  $\sigma[\mathcal{C}_2] = \sigma[\mathcal{C}_1]$ .

4. PfS, Exercise 1.1.2, page 7: We always have  $\mu(\liminf A_n) \leq \liminf \mu(A_n)$ , while  $\limsup \mu(A_n) \leq \mu(\limsup A_n)$  if  $\mu(\omega) < \infty$ .

**Solution:** First note that  $\liminf A_n = \bigcup_{n=1}^{\infty} \bigcap_{k=n}^{\infty} A_k = \bigcup_{n=1}^{\infty} B_n$  where  $B_n = \bigcap_{k=n}^{\infty} A_k$  is  $\uparrow$  since  $B_n = \bigcap_{k=n}^{\infty} A_k \subset \bigcap_{k=n+1}^{\infty} A_k = B_{n+1}$  for all  $n$ . Hence by Proposition 1.2(i),

$$\begin{aligned} \mu(\liminf A_n) &= \mu(\bigcup_{n=1}^{\infty} B_n) \\ &= \lim_{n \rightarrow \infty} \mu(B_n) \\ &= \lim_{n \rightarrow \infty} \mu(\bigcap_{k=n}^{\infty} A_k) \\ &\leq \lim_{n \rightarrow \infty} \inf_{m \geq n} \mu(A_m) \\ &= \liminf \mu(A_n) \end{aligned}$$

since  $\bigcap_{k=n}^{\infty} A_k \subset A_m$  for each  $m \geq n$  so that

$$\mu(\bigcap_{k=n}^{\infty} A_k) \leq \mu(A_m)$$

for each  $m \geq n$  and also  $\mu(\bigcap_{k=n}^{\infty} A_k) \leq \inf_{m \geq n} \mu(A_m)$ .

Similarly,  $\limsup A_n = \bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k = \bigcap_{n=1}^{\infty} B_n$  where  $B_n = \bigcup_{k=n}^{\infty} A_k$  is  $\downarrow$  since  $B_n = \bigcup_{k=n}^{\infty} A_k \supset \bigcup_{k=n+1}^{\infty} A_k = B_{n+1}$ . Thus by Proposition 1.1.2(j), if  $\mu(\Omega) < \infty$ ,

$$\begin{aligned} \mu(\limsup A_n) &= \mu(\bigcap_{n=1}^{\infty} B_n) \\ &= \lim_{n \rightarrow \infty} \mu(B_n) \\ &= \lim_{n \rightarrow \infty} \mu(\bigcup_{k=n}^{\infty} A_k) \\ &\geq \lim_{n \rightarrow \infty} \sup_{m \geq n} \mu(A_m) \\ &= \limsup \mu(A_n) \end{aligned}$$

since  $\bigcup_{k=n}^{\infty} A_k \supset A_m$  for each  $m \geq n$  so that

$$\mu(\bigcup_{k=n}^{\infty} A_k) \geq \mu(A_m)$$

for each  $m \geq n$  and also  $\mu(\bigcup_{k=n}^{\infty} A_k) \geq \sup_{m \geq n} \mu(A_m)$ .

5. PfS, Exercise 9.1.4, page 158: if  $np_n \rightarrow \lambda > 0$ , then

$$P(T_n = k) = \binom{n}{k} p_n^k (1 - p_n)^{n-k} \rightarrow \frac{\lambda^k}{k!} \exp(-\lambda) = P(Y = k)$$

where  $Y \sim \text{Poisson}(\lambda)$ .

**Solution:**

$$\begin{aligned} P(T_n = k) &= \binom{n}{k} p_n^k (1 - p_n)^{n-k} \\ &= \frac{n(n-1) \cdots (n-k+1)}{k! n^k} (np_n)^k \left(1 - \frac{np_n}{n}\right)^{n-k} \\ &\rightarrow \frac{1}{k!} \lambda^k e^{-\lambda} \end{aligned}$$

since  $(1 + x_n/n)^n \rightarrow e^x$  if  $x_n \rightarrow x$ .

6. Let  $I = P(Z \geq 2) = .02275\dots$  where  $Z \sim N(0, 1)$ . Thus  $I = \int h(x)f(x)dx$  there  $f(x) = (2\pi)^{-1/2} \exp(-x^2/2)$  is the standard normal density and  $h(x) = 1_{[x \geq 2]}$ . The basic Monte Carlo estimator is  $\hat{I}_1 = n^{-1} \sum_{i=1}^n h(X_i)$  where  $X_1, \dots, X_n$  are i.i.d.  $N(0, 1)$ . When I carried this out I found  $\hat{I}_1 = .04$  and  $\widehat{Var}(\hat{I}_1) = .0239/100$ . Note that most observations are “wasted” in that they are not near the right tail. Now we try again with “importance sampling”: let  $g$  be the  $N(3, 1)$  density. When we sample from  $g$ , the corresponding estimator of  $I$  is given by  $\hat{I}_2 = n^{-1} \sum_{i=1}^n f(X_i)h(X_i)/g(X_i)$ . When I do this I find  $\hat{I}_2 = .0239\dots$  and  $\widehat{Var}(\hat{I}_2) = .00218/100$ . Note that  $\widehat{Var}(\hat{I}_1)/\widehat{Var}(\hat{I}_2) = .0239/.00218 \approx 10.9$ .

(a) Show that

$$Var(\hat{I}_1) = p(1-p)/n = .02275(1 - .02275)/100 = 0.0222\dots/100$$

where  $p = I = P(Z \geq 2) = .02275\dots$

(b) Show that

$$Var(\hat{I}_2) = n^{-1} Var_g(f(X)h(X)/g(X)) = .001805\dots/100,$$

and hence the variance has been reduced by a factor of 12.3.

Hint: You may compute

$$\int_{-\infty}^{\infty} \{h^2(x)f^2(x)/g(x)\} dx$$

numerically in (b).

**Solution:** (a)  $\hat{I}_1$  is the average of the i.i.d. Bernoulli random variables  $h(X_i) = 1_{[X_i \geq 2]}$  with  $Eh(X_1) = P(X_1 \geq 2) = .2275 \equiv p$ . Since the variance of a Bernoulli( $p$ ) random variable is  $p(1 - p)$  we find that

$$Var(\hat{I}_1) = n^{-2} \sum_1^n Var(1_{[X_i \geq 2]}) = n^{-1}p(1 - p) = .0222.../100.$$

(b)  $\hat{I}_2$  is the average of the i.i.d. random variables  $w(Y_i) = f(Y_i)h(Y_i)/g(Y_i)$  where  $Y_1, \dots, Y_n$  are i.i.d.  $N(3, 1)$ . Thus

$$Var(\hat{I}_1) = n^{-2} \sum_1^n Var(w(Y_i)) = n^{-1}Var(w(Y_1))$$

where

$$\begin{aligned} Var_g(w(Y_1)) &= Ew^2(Y_1) - \{Ew(Y_1)\}^2 = Ew^2(Y_1) - I^2 \\ &= \int_{-\infty}^{\infty} \frac{h^2(y)f^2(y)}{g(y)} dy - I^2. \end{aligned}$$

Now since

$$\frac{g(y)}{f(y)} = \frac{\exp(-(y - 3)^2/2)}{\exp(-y^2/2)} = \exp(3y - 3^2/2),$$

we can write

$$\begin{aligned} Ew^2(Y_1) &= \int_{-\infty}^{\infty} \frac{h^2(y)f^2(y)}{g(y)} dy \\ &= \int_2^{\infty} \frac{f(y)}{g(y)} f(y) dy \\ &= \int_2^{\infty} \exp(-3y + 3^2/2) \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy \\ &= e^{3^2} \int_2^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y + 3)^2\right) dy \\ &= e^9 \int_5^{\infty} \frac{1}{\sqrt{2\pi}} e^{-z^2/2} dz \\ &= e^9(1 - \Phi(5)) = 0.00232276. \end{aligned}$$

Thus  $var(w(Y_1)) = 0.00232276 - 0.02275^2 = 0.0018052$ , which agrees with a direct numerical integration.