

## Statistics 581, Problem Set 3

Wellner; 10/11/2017

**Reading:** Course Notes: Chapter 2, pages 1 - 25;

Ferguson ACILST pages 26-60;

Van der Vaart Asymp. Statistics, sections 2.2 - 2.9 & 3.1 - 3.2, pages 12 - 31.

**Due:** Wednesday, October 18, 2017.

1. Ferguson, ACILST, page 34, problem 1(a) (modified slightly)

Suppose that  $X_1, X_2, \dots$  are i.i.d. in  $R^2$  with distribution giving probability  $\theta_1$  to  $(1, 0)'$ , probability  $\theta_2$  to  $(0, 1)'$ ,  $\theta_3$  to  $(0, 0)'$  and  $\theta_4$  to  $(-1, -1)'$  where  $\theta_j \geq 0$  for  $j = 1, 2, 3, 4$  and  $\theta_1 + \theta_2 + \theta_3 + \theta_4 = 1$ .

(a) Find  $\mu = E(X_1)$ .

(b) Compute  $E(X_1 X_1^T)$  and  $\Sigma = E(X_1 - \mu)(X_1 - \mu)^T$ .

(c) Find the limiting distribution of  $\sqrt{n}(\bar{X}_n - \mu)$  and describe the resulting approximation to the distribution of  $\bar{X}_n$ .

(d) Find values of  $(\theta_1, \dots, \theta_4)$  such that  $\Sigma$  has rank 1 and  $\det(\Sigma) = 0$ .

2. Suppose that  $X_1, X_2, \dots$  are i.i.d.  $(\mu, \sigma^2)$  with  $\mu_4 < \infty$ . Let  $\bar{X}_n = n^{-1} \sum_{i=1}^n X_i$  and  $S_n^2 = (n-1)^{-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2$  be the sample mean and sample variance respectively.

(a) Show that

$$\sqrt{n} \begin{pmatrix} \bar{X}_n - \mu \\ S_n^2 - \sigma^2 \end{pmatrix} \rightarrow_d \underline{Z} \sim N_2(0, \Sigma)$$

where

$$\begin{pmatrix} \sigma^2 & \mu_3 \\ \mu_3 & \mu_4 - \sigma^4 \end{pmatrix}.$$

(b) Suppose  $\mu \neq 0$ . Use (a) to find the limiting distribution of the sample *coefficient of variation*  $C_n \equiv S_n/\bar{X}_n$ ; i.e. show that  $\sqrt{n}(C_n - c) \rightarrow_d N(0, V^2)$  with  $c \equiv \sigma/\mu$  and find  $V^2$ .

3. Ferguson, ACILST, page 34, problem 1(b) (modified slightly)

Suppose that  $X_1, \dots, X_n$  is a sample from the Poisson distribution with parameter  $\lambda > 0$ :  $P(X_1 = k) = \exp(-\lambda)\lambda^k/k!$ ,  $k = 0, 1, \dots$ . Let  $Z_n = (1/n) \sum_{i=1}^n 1_{[X_i=0]}$ .

(a) What is the joint asymptotic distribution of

$$\sqrt{n}((\bar{X}_n, Z_n)' - (\lambda, e^{-\lambda})')?$$

(b) Let  $p_0(\lambda) \equiv P_\lambda(X_1 = 0)$ . What is the asymptotic distribution of  $\hat{p}_0 \equiv p_0(\hat{\lambda}_n)$  where  $\hat{\lambda}_n = \bar{X}_n$ ?

(c) What is the joint asymptotic distribution of  $(Z_n, \hat{p}_0)$  (after centering and rescaling)?

(d) Compute the ratio of the asymptotic variances of the two estimators  $Z_n$  and  $\hat{p}_0$  of  $p_0(\lambda)$ . Which estimator would you prefer if the Poisson model (assumption) holds? Which estimator would you prefer if the Poisson model (assumption) fails?

4. A sequence of random variables  $Y_n$  is *bounded in probability* and we write  $Y_n = O_p(1)$  if

$$\lim_{\lambda \rightarrow \infty} \limsup_{n \rightarrow \infty} P(|Y_n| > \lambda) = 0;$$

i.e. for each  $\epsilon > 0$  there exist  $\lambda_\epsilon$  and  $N_\epsilon$  such that  $P(|Y_n| > \lambda_\epsilon) < \epsilon$  for all  $n > N_\epsilon$ .

(a) Show that if  $Y_n \rightarrow_d Y$  for some random variable  $Y$ , then  $Y_n$  is bounded in probability. (This is Lehmann and Casella, problem 8.24, page 77.)

(b) Give an example of a sequence of random variables  $Y_n$  that is bounded in probability, but does not converge in distribution.

(c) Let  $X_n \sim t_n$ , the  $t$ -distribution with  $n$  degrees of freedom; thus  $X_n \stackrel{d}{=} Z/\sqrt{Y_n/n}$  where  $Y_n \sim \chi_n^2$  is independent of  $Z$ . Fix  $r > 0$  (large). Compute  $E|X_n|^r$  exactly as a function of  $n$  and  $r$  in terms of the Gamma function  $\Gamma(r) = \int_0^\infty t^{r-1}e^{-t}dt$ . Use Stirling's formula,  $\Gamma(r+1) \sim \sqrt{2\pi r}(r/e)^r$  to show that  $\limsup_{n \rightarrow \infty} E|X_n|^r < \infty$ . Use this to show that  $\{|X_n|^p : n \geq 1\}$  is uniformly integrable for any integer  $p \geq 1$ , and hence that  $E(X_n^p) \rightarrow E(Z^p)$ .

5. Suppose that  $X$  is a random variable with finite fourth moment;  $E|X|^4 < \infty$ . Then  $\mu_4 = E(X - \mu)^4$  is the fourth central moment of  $X$ . The ratio  $\mu_4/\sigma^4 \equiv \kappa$  is the *kurtosis* of  $X$  (or of the distribution function  $F$  of  $X$ ), and  $\gamma_2 \equiv \mu_4/\sigma^4 - 3$  is called the *excess of kurtosis*; note that for any  $N(\mu, \sigma^2)$  random variable,  $\gamma_2 = 0$ . Investigate the value of  $\gamma_2$  for various classical distributions ( $t_r$ , uniform, bernoulli, Poission( $\lambda$ ), ... ). How big can  $\gamma_2$  be? How small can  $\gamma_2$  be?

6. **Optional bonus problem 1.** Suppose that  $X_1, \dots, X_n$  are independent  $N(0, 1)$  random variables, and let  $Y_i = X_i^2$ , for  $i = 1, \dots, n$ . Thus  $\sum_1^n Y_i \sim \chi_n^2$ .

(a) Show that  $\sqrt{n}(\bar{Y}_n - 1) \rightarrow_d N(0, \text{"something"})$ , and find "something".

(b) Show that for each  $r > 0$ ,  $\sqrt{n}(\bar{Y}_n^r - 1) \rightarrow_d N(0, V^2(r))$  and find  $V^2(r)$  as a function of  $r$ .

(c) Show that  $V^2(1/3) = 2/9$  and that

$$\frac{\sqrt{n}(\bar{Y}_n^{1/3} - (1 - 2/(9n)))}{\sqrt{2/9}} \rightarrow_d N(0, 1).$$

(d) Make normal probability plots to compare the approximations in (a) and (c). [The transformation in (c) is called the "Wilson-Hilferty" transformation of a  $\chi^2$  random variable.]

7. **Optional bonus problem 2.** (Van der Vaart, page 24)

(a) Suppose that  $X_n$  and  $Y_n$  are independent random vectors with  $X_n \rightarrow_d X$  and  $Y_n \rightarrow_d Y$ . Show that  $(X_n, Y_n) \rightarrow (X, Y)$  where  $X$  and  $Y$  are independent.

(b) Suppose that  $P(X_n = i/n) = 1/n$  for  $i = 1, 2, \dots, n$ . From HW #1 we know that  $X_n \rightarrow_d X \sim \text{Uniform}(0, 1)$ . Show that there exists a Borel set  $B$  such that  $P(X_n \in B) = 1$  but  $P(X \in B) = 0$ . In particular, with  $P_n = \mathcal{L}(X_n)$  and  $P = \mathcal{L}(X)$ ,  $d_{TV}(P_n, P) = 1$  for each  $n$ .