

**Stat 581 Homework 4: Outline solutions**

1. (a)  $\Pr(X_n \neq 0) = 1/n \rightarrow 0$ , so  $X_n \rightarrow_p 0$  for all real  $\alpha$ .  
 (b) From the Ferguson example,  $X_n = n^\alpha$  i.o. So if  $\alpha \geq 0$ ,  $X_n$  does not cge a.s. to 0. However, if  $\alpha < 0$ ,  $n^\alpha \rightarrow 0$ . So  $X_n \rightarrow_{a.s.} 0$ , iff  $-\infty < \alpha < 0$ .  
 (c)  $\mathbb{E}(|X_n - 0|^r) = n^{r\alpha-1} \rightarrow 0$  iff  $-\infty < \alpha < 1/r$ .
2. (a)  $\mathbb{E}(f_n(x)) = (2b_n)^{-1}(F(x + b_n) - F(x - b_n)) \rightarrow 0$  if  $b_n \rightarrow 0$ .  
 (b)

$$\begin{aligned} \text{var}(f_n(x)) &= (4b_n^2n)^{-1}\text{var}(I(x - b_n < X \leq x + b_n)) \\ &= (2nb_n)^{-1} \left( \frac{F(x + b_n) - F(x - b_n)}{2b_n} \right) (1 - (F(x + b_n) - F(x - b_n))) \\ &\rightarrow 0 \cdot f(x) \cdot 1 = 0 \text{ if } b_n \rightarrow 0, nb_n \rightarrow \infty. \end{aligned}$$

- (c)  $f_n(x) - \mathbb{E}(f_n(x)) = (2nb_n)^{-1} \sum_{i=1}^n I(x - b_n < X_i \leq x + b_n) - \mathbb{E}(f_n(x))$ ,  
 so have mean 0,  $\text{var}(I(x - b_n < X_i \leq x + b_n)/\sqrt{2b_n}) \rightarrow f(x)$  as above, and by CLT and Slutsky  $\sqrt{2nb_n}(f_n(x) - \mathbb{E}(f_n(x))) \rightarrow_d N(0, f(x))$  if  $b_n \rightarrow 0$ ,  $nb_n \rightarrow \infty$  as in (b).  
 (d) Need  $\sqrt{2nb_n}(\mathbb{E}(f_n(x) - f(x))) = \sqrt{2nb_n}((F(x + b_n) - F(x - b_n))/(2b_n) - f(x)) \rightarrow 0$  so  $b_n\sqrt{2nb_n} \rightarrow 0$  or  $nb_n^3 \rightarrow 0$  should do it.  
 (e) Suppose we have the conditions for (d) and apply  $g'$  theorem to (c), with  $g(y) = \sqrt{y}$ , so  $(g'(f(x)))^2 = 1/(4f(x))$  and

$$\sqrt{2nb_n}(\sqrt{f_n(x)} - \sqrt{f(x)}) \rightarrow_d N(0, (g'(f(x)))^2 f(x)) \equiv N(0, 1/4)$$

3. (a) With the change of variable suggested

$$I \equiv \int_1^\infty \frac{1}{x} \sin(2\pi x) dx = \int_0^1 \frac{1}{y} \sin(2\pi/y) dy$$

However, if  $Y_1, \dots, Y_n$  are i.i.d  $U(0, 1)$ ,

$$\begin{aligned} \mathbb{E} \left( \left| \frac{1}{Y_1} \sin(2\pi/Y_1) \right| \right) &= \int_0^1 \frac{1}{y} |\sin(2\pi/y)| dy = \int_1^\infty \frac{1}{x} |\sin(2\pi x)| dx \\ &= \sum_{k=3}^\infty \int_{(k-1)/2}^{k/2} \frac{1}{x} |\sin(2\pi x)| dx \geq \sum_{k=3}^\infty \frac{2}{k} \int_{(k-1)/2}^{k/2} |\sin(2\pi x)| dx \\ &= \sum_{k=3}^\infty \frac{2}{k} \frac{1}{\pi} = \infty \end{aligned}$$

So the SLLN fails, and this will probably not be a good way to estimate  $I$ .

- (b) Now

$$I_\alpha \equiv \int_1^\infty \frac{1}{x^\alpha} \sin(2\pi x) dx = \int_0^1 y^{\alpha-2} \sin(2\pi/y) dy$$

and in this case, if  $Y \sim U(0, 1)$

$$\mathbb{E}(|Y^{\alpha-2} \sin(2\pi/Y)|) \leq \mathbb{E}(Y^{\alpha-2}) = (\alpha-1)^{-1} [y^{\alpha-1}]_0^1 < \infty$$

if  $\alpha > 1$ . So by SLLN the estimator will converge *a.s.* to  $I_\alpha$  if  $\alpha > 1$ .

(c) Now

$$\mathbb{E}(|Y^{\alpha-2} \sin(2\pi/Y)|^2) \leq \mathbb{E}(Y^{2\alpha-4}) = (2\alpha-3)^{-1} [y^{2\alpha-3}]_0^1 < \infty$$

if  $\alpha > 3/2$ . So then, by the CLT, if  $\alpha > 3/2$ ,

$$n^{\frac{1}{2}}(\widehat{I}_{n,\alpha} - I_\alpha) \rightarrow_d N(0, \sigma^2(\alpha))$$

where  $\sigma^2(\alpha) = \mathbb{E}((Y^{2\alpha-4}(\sin(2\pi/Y))^2) I_\alpha^2)$ .

4. (a)  $R = \overline{XY}/(\overline{X^2Y^2})^{1/2}$ .

By SLLN  $\overline{X^2} \rightarrow_{a.s.} \mathbb{E}(X_i^2) = 1$ ,  $\overline{Y^2} \rightarrow_{a.s.} \mathbb{E}(Y_i^2) = 1$ , and  $\overline{XY} \rightarrow_{a.s.} \mathbb{E}(X_i Y_i) = \rho$ .

By CLT,  $\sqrt{n}(\overline{XY} - \rho, \overline{X^2} - 1, \overline{Y^2} - 1)' \rightarrow_d N_3(0, V)$  where  $V_{11} = \text{var}(X_i Y_i) = \mathbb{E}(X^2 Y^2) - \rho^2$ ,  $V_{12} = \text{Cov}(X_i^2, X_i Y_i) = \mathbb{E}(X^3 Y) - \rho$ , etc.

(b) Consider  $g(z_1, z_2, z_3) = z_1 z_2^{-1/2} z_3^{-1/2}$ , so  $R = g(\overline{XY}, \overline{X^2}, \overline{Y^2})$ . Then

$$\left( \frac{\partial g}{\partial z_i} \right) |_{(\rho, 1, 1)} = ((z_2 z_3)^{-\frac{1}{2}}, -\frac{1}{2} z_1 z_2^{-\frac{3}{2}} z_3^{-\frac{1}{2}}, -\frac{1}{2} z_1 z_2^{-\frac{1}{2}} z_3^{-\frac{3}{2}}) = (1, -\rho/2, -\rho/2)$$

So, by  $g'$  theorem  $\sqrt{n}(R - \rho) \rightarrow_d (Z_1 - \frac{1}{2}\rho Z_2 - \frac{1}{2}\rho Z_3)$  where  $(Z_1, Z_2, Z_3)$  has the  $N_3(0, V)$  dsn of (a).

(c)  $X, Y$  indep  $\Rightarrow \rho = 0$ , and  $\text{var}(Z_1) = \mathbb{E}(X^2 Y^2) - \rho^2 = 1$ , so  $\sqrt{n}R \rightarrow_d Z_1 \sim N(0, 1)$ .

(d) If  $(X_1, Y_1)$  Normal

$$\begin{aligned} \text{var}(XY) &= \mathbb{E}(X^2 Y^2) - \rho^2 = \mathbb{E}(X^2 \mathbb{E}(Y^2|X)) - \rho^2 \\ &= \mathbb{E}(\rho^2 X^4 + (1 - \rho^2)X^2 - \rho^2) = 3\rho^2 + (1 - \rho^2) - \rho^2 = 1 + \rho^2 \\ \mathbb{E}(X^3 Y) - \rho &= \mathbb{E}(\rho X^4) - \rho = 2\rho = \mathbb{E}(XY^3) - \rho \\ \mathbb{E}(X^2 Y^2) - 1 &= 2\rho^2, \quad \text{var}(X^2) = \text{var}(Y^2) = 2 \end{aligned}$$

then direct calculation of  $(1, \rho/2, \rho/2)V(1, \rho/2, \rho/2)'$  gives  $\sqrt{n}(R - \rho) \rightarrow_d N(0, (1 - \rho^2)^2)$

(e)  $g(x) = (1/2)(\log(1+x) - \log(1-x))$ ,  $g'(x) = (1/2)((1+x)^{-1} + (1-x)^{-1}) = (2(1-x^2))^{-1}$ .

So  $\sqrt{n}(g(R) - \rho) \rightarrow_d N(0, 1/4)$  if  $(X_i, Y_i)$  is Normal.

5. (a)  $\mathbb{E}(X_i - \mu)^k = \mu_k$  so, by SLLN,  $B_k \rightarrow_{a.s.} \mu_k$ .

(b)  $\text{var}((X_i - \mu)^k) \exists$  finite, since  $\mu_{2k} < \infty$ . Also  $\mu_1 = \mathbb{E}(X_i - \mu) = 0$ .

So by CLT  $\sqrt{n}((B_1 - 1, \dots, B_k) - (0, \mu_2, \dots, \mu_k)') \rightarrow_d N(0, V)$  where  $V_{ij} = \text{Cov}((X_l - \mu)^i, (X_l - \mu)^j) = \mu_{i+j} - \mu_i \mu_j$

(c)  $nM_k = \sum_i (X_i - \overline{X_n})^k$ , and expanding  $(X_i - \overline{X_n})^k = ((X_i - \mu) - (\overline{X_n} - \mu))^k$  gives  $M_k = \sum_0^k (-1)^{k-j} C(k:j) B_j B_1^{k-j}$ , noting  $B_1 = (\overline{X_n} - \mu)$ . (Here  $C(k:j)$  denotes the combinatorial coeff “ $k$ -choose- $j$ ”.)

Now  $B_1 \rightarrow_{a.s.} 0$ , so  $M_k \rightarrow_{a.s.} (-1)^{k-k} C(k:k) \lim(B_k) \cdot 1 = 1$

(d)

$$\begin{aligned}
M_i &= \sum_0^i (-1)^{i-j} C(i:j) B_j B_1^{i-j} = B_i - i B_{i-1} B_1 + \frac{1}{2} i(i-1) B_{i-2} B_1^2 + \dots \\
(M_i - \mu_i) - (B_i - \mu_i - i \mu_{i-1} B_1) &= i \mu_{i-1} B_1 - i B_{i-1} B_1 + O_p(B_1^2) \\
&= -i B_1 (B_{i-1} - \mu_{i-1}) + O_p(B_1^2) \\
\sqrt{n}((M_i - \mu_i) - (B_i - \mu_i - i \mu_{i-1} B_1)) &= (-i(B_1 \sqrt{n}))(B_{i-1} - \mu_{i-1}) + (B_1 \sqrt{n}) O_p(B_1) \rightarrow_p 0
\end{aligned}$$

since  $B_1 \rightarrow_p 0$ ,  $\sqrt{n} B_1 \rightarrow_d N(0, 1)$ ,  $(B_{i-1} - \mu_{i-1}) \rightarrow_p 0$ .

(e) The limiting dsn of  $\sqrt{n}(M_i - \mu_i)$  is the same as that of  $\sqrt{n}(B_i - \mu_i - i \mu_{i-1} B_1)$  or  $N_{k-1}(0, V^*)$  where now  $i = (2, \dots, k)$  and

$$\begin{aligned}
V_{ij}^* &= \text{Cov}((X - \mu)^{i+1} - (i+1)\mu_i(X - \mu), (X - \mu)^{j+1} - (j+1)\mu_j(X - \mu)) \\
&= \mu_{i+j+2} - (j+1)\mu_{i+2}\mu_j - (i+1)\mu_i\mu_{j+2} + (i+1)(j+1)\mu_i\mu_j\mu_2 - \mu_{i+1}\mu_{j+1}
\end{aligned}$$

(f)  $G_2 \rightarrow_{a.s.} \mu_4/\mu_2^2 - 3 = 0$  and  $\mu_{2i+1} = 0$ . Then  $\sqrt{n}(M_2 - \mu_2, M_4 - \mu_4)$  cges to bivariate Normal, mean 0, variance  $(\mu_4 - \mu_2^2, \mu_6 - \mu_2\mu_4, \mu_8 - \mu_4^2)$ . Now let  $g(x, y) = y/x^2$ , so  $g' = (-2\mu_4/\mu_2^3, 1/\mu_2^2)$ , so limiting variance of  $\sqrt{n}G_2$  is  $(4\mu_4^2(\mu_4 - \mu_2^2) - 4\mu_4(\mu_6 - \mu_2\mu_4)\mu_2^2 + (\mu_8 - \mu_4^2)\mu_2^2)/\mu_2^6$ . (Arithmetic not guaranteed.)