Statistics 581, Practice Midterm Exam Wellner; 10/29/2018

This exam is to be taken without any books or notes.

1. (24 points) Define any three of the following five terms.

- (a) A uniformly integrable sequence of random variables.
- (b) Convergence in *r*th mean of a sequence of random variables.
- (c) A normal random vector $Y = (Y_1, \ldots, Y_n)$.
- (d) The inverse or quantile function F^{-1} of a distribution function F.
- (e) The total variation distance between two probability measures P and Q on a measurable space $(\mathcal{X}, \mathcal{A})$.

Do either problem 2 or problem 3.

- 2. (40 points).
 - (a) State the ordinary (univariate) central limit theorem.
 - (b) State the Cramér-Wold device.
 - (c) State the multivariate central limit theorem.
 - (d) Use the ordinary (univariate) central limit theorem (a) and the Cramér-Wold

device (b) to prove the multivariate central limit theorem (c).

- 3. (40 points). State and prove the Glivenko-Cantelli theorem.
- 4. (36 points). Let X_1, \ldots, X_n be i.i.d. with exponential density $p_{\theta}(x) = \theta \exp(-\theta x) \mathbb{1}_{[0,\infty)}(x)$.
 - (a) Find a constant c so that $c\mathbb{F}_n^{-1}(p) \to_p \theta^{-1}$.
 - (b) For the $c = c_p$ you found in (a), show that $\sqrt{n}(c\mathbb{F}_n^{-1}(p) \theta^{-1}) \to_d N(0, \sigma^2)$ and find $\sigma^2 = \sigma^2(p)$.
 - (c) Show that the asymptotic variance $\sigma^2(p)$ is minimized by p satisfying 2p = $-\log(1-p).$
- 5. (36 points)

Suppose that X, X_1, \ldots, X_n are i.i.d. Exponential(θ) random variables so that $P_{\theta}(X > x) = \exp(-\theta x) = 1 - F_{\theta}(x) \text{ for } x > 0.$ (a) Fix $x_0 > 0$ and let $\mathbb{F}_n(x) = n^{-1} \sum_{i=1}^n \mathbb{1}_{[X_i \le x]} = n^{-1} \sum_{i=1}^n \mathbb{1}_{(-\infty,x]}(X_i)$ denote

the empirical distribution function. Show that

$$\sqrt{n} \left(\begin{array}{c} \overline{X}_n - 1/\theta \\ \mathbb{F}_n(x_0) - F_\theta(x_0) \end{array} \right) \to_d Y \sim N_2(0, \Sigma)$$

and find Σ .

(b) Let $g(\theta) \equiv F_{\theta}(x_0) = 1 - \exp(-\theta x_0)$, and consider the two estimators of $F = F_{\theta}$ given by $T_{n,1} \equiv g(\hat{\theta}_n)$ and $T_{n,2} \equiv \mathbb{F}_n(x_0)$ where $\hat{\theta}_n \equiv 1/\overline{X}_n$. Show that

$$\sqrt{n} \left(\begin{array}{c} T_{n,1} - F_{\theta}(x_0) \\ T_{n,2} - F_{\theta}(x_0) \end{array} \right) \to_d \tilde{Y}$$

and find the distribution of \tilde{Y} .

(c) What is the advantage of $T_{n,2} = \mathbb{F}_n(x_0)$ as an estimator even though it is inefficient when the exponential model holds?

6. (36 points).

Suppose that X, X_1, \ldots, X_n are i.i.d. with distribution function F given by $P(X > x) = 1 - F(x) = 1/x^5, x \ge 1, F(x) = 0, x \le 1.$

(a) For what values of r > 0 is $E|X|^r < \infty$? If they are finite compute $\mu = E(X)$ and $\sigma^2 = Var(X)$.

- (b) Compute $F^{-1}(t) = Q(t)$, the quantile function corresponding to F.
- (c) Which of the following are true? (Briefly indicate why or why not.)

(i)
$$\sum_{i=1}^{n} X_i = O_p(n^{1/2}).$$

(ii)
$$n^{1/3}(\overline{X}_n - \mu) = o_p(1).$$

- (iii) $n^{3/4}(\overline{X}_n \mu) = O_p(1).$
- (iv) $g(n^{1/3}(\overline{X}_n \mu)) \to_p 1/2$ where $g(x) = 1/(1 + e^{-x})$.
- (v) $h(n^{1/2}(\overline{X}_n \mu)) = O_p(1)$ with $h(x) = \log |x|$.
- (vi) $\sqrt{n}(\mathbb{F}_n^{-1}(1/2) F^{-1}(1/2)) \to_d N(0, (1/4)/[5(1/2)^{6/5}]^2).$