

Likelihood Theory and Extensions (BIOS:7110)  
Breheny

Assignment 5

Due: Monday, October 5

1. *Consistency of the OLS estimator.* Suppose that for a sequence of constant vectors  $\mathbf{x}_1, \mathbf{x}_2, \dots$ , the random variables  $Y_1, Y_2, \dots$  are independent with mean  $\mathbf{x}_i^\top \boldsymbol{\beta}$  and variance  $\sigma^2$ . Prove that  $\widehat{\boldsymbol{\beta}} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top \mathbf{y}$  is a consistent estimator of  $\boldsymbol{\beta}$ . Do you need additional conditions on  $\mathbf{X}$  in order to guarantee that  $\widehat{\boldsymbol{\beta}}$  is consistent? If so, make any additional assumptions you wish, but clearly describe and label these additional conditions.
2. *A triangular array for binomial data.* Suppose  $X_n \sim \text{Binom}(n, \pi_n)$ ; note in particular that the binomial probability  $\pi$  here is changing with  $n$ . Use the Lindeberg-Feller central limit theorem to show that

$$\frac{X_n - n\pi_n}{\sqrt{n\pi_n(1 - \pi_n)}} \xrightarrow{d} N(0, 1).$$

As above, are there any additional conditions you need to require on the sequence  $\pi_n$  in order for this result to be true? If so, tell me what they are. In your proof, you will need to set up a triangular array; clearly describe what random variables you are using for this array and prove that it satisfies the requirements to be a triangular array as defined in the notes.

3. *Slutsky constant.* In Slutsky's theorem, we require that one of the two random variables is converging in probability to a constant. What happens if both  $\mathbf{x}_n$  and  $\mathbf{y}_n$  are converging in distribution? In other words, suppose  $\mathbf{x}_n \xrightarrow{d} \mathbf{x}$  and  $\mathbf{y}_n \xrightarrow{d} \mathbf{y}$ ; is it true that

$$\begin{bmatrix} \mathbf{x}_n \\ \mathbf{y}_n \end{bmatrix} \xrightarrow{d} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix}?$$

If this is not true, provide a counterexample.