Ec240a, Fall 2015
Professor Bryan Graham

## Problem Set 4

Due: December 9th, 2015
Problem sets are due at 5PM in the GSIs mailbox (commented code and execution files should be e-mailed to the GSI prior to that time). You may work in groups, but each student should turn in their own write-up (including individually commented and executed code).

## 1 Average regression: identification

Let $Y$ be a scalar outcome interest, $X$ a $K \times 1$ vector of regressors with a constant as its first element (the other elements may be discretely- or continuously-valued) and $W \in\left\{w_{1}, \ldots, w_{L}\right\}$ a discretely-valued 'proxy variable' with $L$ points of support. For a random draw from the population $Y$ is generated according to

$$
\begin{equation*}
Y=X^{\prime} B \tag{1}
\end{equation*}
$$

where $B$ is a $K \times 1$ vector of random coefficients. Assume that

$$
\begin{equation*}
\mathbb{E}[B \mid X, W=w]=\mathbb{E}[B \mid W=w]=\beta(w) \tag{2}
\end{equation*}
$$

[a] Outline a concrete economic model which fits into the general set-up of (1) and (2). Assess the plausibility of condition (2) for your chosen example. One possibility is to discuss this set-up in light of the Card and Krueger (1996) schooling model, but you may choose another model if you like.
[b] Show that, for $l=1, \ldots, L$

$$
\beta\left(w_{l}\right)=\mathbb{E}\left[X X^{\prime} \mid W=w_{l}\right]^{-1} \times \mathbb{E}\left[X Y \mid W=w_{l}\right]
$$

You may assume all of the relevant matrices are well-defined.
[c] Consider the average linear regression

$$
m^{\mathrm{ar}}(x)=x^{\prime} \bar{\beta}
$$

for $\bar{\beta}=\mathbb{E}[\beta(W)]$. Interpret this function; outline a policy question for which knowledge of $m^{\text {ar }}(x)$ might be useful.
[d] Let $D$ be a $L \times 1$ vector with a 1 in the $l^{\text {th }}$ row if $W=w_{l}$ and zeros elsewhere. Let $R=(D \otimes X)$ and $\beta=\left(\beta\left(w_{1}\right)^{\prime}, \ldots, \beta\left(w_{L}\right)^{\prime}\right)^{\prime}$. Show that

$$
\beta=\mathbb{E}\left[R R^{\prime}\right]^{-1} \times \mathbb{E}[R Y]
$$

and also, for $S=\left(I_{K} \otimes D\right)$, that

$$
\bar{\beta}=\mathbb{E}\left[S^{\prime} \beta\right]
$$

[e] Assume that conditional on the event $W=w_{l}$ the distribution of $X$ is degenerate. What problems might
such a situation create? Comment in light of your empirical example of part [a] above.
[f] Let $\underline{0}$ be a $K \times 1$ vector of zeros and

$$
\mathbf{R}=\left(\begin{array}{cc}
R^{\prime} & \underline{0}_{1 \times K} \\
S^{\prime} & -I_{K}
\end{array}\right) \quad \mathbf{Z}=\left(\begin{array}{cc}
R^{\prime} & \underline{0}_{1 \times K} \\
\underline{0}_{K \times K L} & -I_{K}
\end{array}\right)
$$

and $\mathbf{Y}=\left(Y, \underline{0}^{\prime}\right)^{\prime}$. Show that

$$
\binom{\beta}{\bar{\beta}}=\mathbb{E}\left[\mathbf{Z}^{\prime} \mathbf{R}\right]^{-1} \times \mathbb{E}\left[\mathbf{Z}^{\prime} \mathbf{Y}\right]
$$

## 2 Average linear regression: estimation and inference.

Let $\left\{\left(Y_{i}, X_{i}, W_{i}\right)\right\}_{i=1}^{N}$ be a random sample of size $N$ draw from a population in which (1) and (2) and additional 'regularity conditions' hold.
[a] Show that

$$
\hat{\theta}=\left[\frac{1}{N} \sum_{i=1}^{N} \mathbf{Z}_{i}^{\prime} \mathbf{R}_{i}\right]^{-1} \times\left[\frac{1}{N} \sum_{i=1}^{N} \mathbf{Z}_{i}^{\prime} \mathbf{Y}_{i}\right]
$$

consistently estimates $\theta=\left(\beta^{\prime}, \bar{\beta}^{\prime}\right)^{\prime}$. Briefly discuss any needed regularity conditions on $F_{Y, X, W}$.
[b] Let $\mathbf{U}_{i}=\mathbf{Y}_{i}-\mathbf{R}_{i}{ }^{\prime} \theta$ and

$$
\Gamma=\mathbb{E}\left[\mathbf{Z}^{\prime} \mathbf{R}\right] \quad \Omega=\mathbb{E}\left[\mathbf{Z}^{\prime} \mathbf{U} \mathbf{U}^{\prime} \mathbf{Z}\right]
$$

show that, for $\Lambda=\Gamma^{-1} \Omega \Gamma^{-1^{\prime}}$,

$$
\sqrt{N}(\hat{\theta}-\theta) \xrightarrow{D} \mathcal{N}(0, \Lambda) .
$$

Briefly discuss any needed regularity conditions on $F_{Y, X, W}$.
Bonus (5 points of aggregate homework score): Provide an 'elegant' expression for the lower-right-hand $K \times K$ block of $\Lambda$.

## 3 Average linear regression: computation/illustration

The file brazil_pnad96_ps4.out contains 65,801 comma delimited records drawn from the 1996 round of the Brazilian Pesquisas Nacional por Amostra de Domicilos (PNAD96). The population corresponds to employed males between the ages of 20 and 60. Respondents with incomplete data are dropped from the sample. Each record contains MONTHLY_EARNINGS, YRSSCH, AgeInDays, Dad_NoSchool_c, Dad_1stPrim_c, Dad_2ndPrim_c, Dad_Sec_c, Dad_DK_c, Mom_NoSchool_c, Mom_1stPrim_c, Mom_2ndPrim_c, Mom_Sec_c, Mom_DK_c and ParentsSchooling. The first three variables equal monthly earnings, years of completed schooling and age in years (but measured to the precision of a day). The next 5 variables are dummies for father's level of education (no school, first primary cycle completed, second primary cycle completed, secondary or more and 'don't know'). The next 5 variables are the corresponding dummies for mother's level of education. The final variable takes on 25 values corresponding to each possible combination of parent's schooling.
[a] Compute the least squares fit of Log(MONTHLY_EARNINGS) onto a constant YRSSCH, AgeInDays, and AgeInDays squared. Construct a 95 percent confidence interval for the coefficient on YrsSch.
[b] Compute the least squares fit of Log(MONTHLY_EARNINGS) onto a constant YRSSCH, AgeInDays, AgeInDays squared, Dad_NoSchool_c, Dad_1stPrim_c, Dad_2ndPrim_c, Dad_Sec_c, Mom_NoSchool_c, Mom_1stPrim_c, Mom_2ndPrim_c, and Mom_Sec_c. Compare the resulting coefficient on YRSSCH with that in part [a] above. Provide an explanation for any differences found.
[c] Let $X=\left(1, \text { YRSSCH, AgeInDays, } \text { AgeInDays }^{2}\right)^{\prime}$ and $W=$ ParentsSchooling, using the results derived above compute an estimate of $\bar{\beta}$ and as well as a set of estimated standard errors. Compare your results with those of parts [a] and [b]. Provide an explanation for any differences found.
[d] Using the Bayes Boostrap to approximate a posterior distribution for $\bar{\beta}$. How does this posterior distribution compare with the estimated asymptotic sampling distribution calculated in part [c].

