## ECON 836 Midterm 2016

Each of the eight questions is worth 4 points. You have 2 hours. No calculators, I-devices, computers, and phones. No open books, and everything must be on the floor.

## Good luck!

- 1) Snappers:
  - a) [Study Question 6] Why are residuals in regressions mean-zero?
    - i) because the FOC of least squares is X'e=0, and X includes a constant.
  - b) [Study Question 8] How does use of the White hetero-robust covariance matrix estimator affect the estimated coefficients, compared to the OLS estimates?
    - i) it does not affect the estimated coefficients.
  - c) Why is the Maximum likelihood estimator for the linear model with normal errors equal to the OLS estimator for the linear model with exogenous regressors? You may answer with a proof if you like.
    - i) The optimisation for ML is to maximize the likelihood of observing the observed Y,X by choice of Beta. With normal errors and a linear model, this implies minimizing the sum of square errors. The first order condition of that is X'e=0, which is the same as the moment condition for the linear model with exogenous regressors.
  - d) In a panel model with regressors, time effects and unit effects, under what conditions does the OLS regression of the dependent variable on regressors and time effects (but not unit effects) deliver unbiased estimates of the parameters multiplying the regressors.
    - i) only if the unit effects are orthogonal to the regressors and time effects in expectation (aka: are uncorrelated with them).
- 2) [Study Question 1, modified] Pendakur and Pendakur (2011) control for personal characteristics *X*, but do not control for job characteristics *Z* in some of their regressions. If *X*=[*V W*] where *W* represents variables of interest (in this case, Aboriginal status) and *V* represents variables that are controlled for, but are not of direct interest, (in this case, age, education and so on) does it matter if:
  - a) *V* and *W* are correlated? Can you give an example of this kind of correlation? How does it affect your interpretation of the estimated coefficient on *W*?
    - i) V and W are both regressors, and they are correlated. E.g., education, V, is correlated with Aboriginal status, W.
    - ii) The estimated coefficient on W is unbiased, so its interpretation is unchanged. But it might have high variance due to the correlatedness of the regressors.
  - b) E[ZZ'] depends on W? Can you give an example of this kind of dependence? How does it affect your interpretation of the estimated coefficient on W?
    - i) the error term can be interpreted as a function of Z. So, ZZ' is like epsilon epsilon'. So this means we have heteroskedasticity of known form: V(epsilon) is a function of W, which is observed. E.g., literacy (Z) has higher variance for Aboriginal people (W) than for non-Aboriginal people. The estimated coefficient on W is unbiased, so its interpretation is unchanged. But it might have high variance due to not correcting for heteroskedasticity of known form.

- 3) [Study Question 3, modified] Allen, Pendakur and Suen (2005) estimate a panel model with the standard deviation of the log of age at first marriage on the LHS and no-fault status and state and year dummies on the RHS.
  - a) They do not include any information about the population of the state in the model. Likewise, there is not information on education levels in the state. Under what conditions does it not matter? Are these plausible conditions?
    - i) If they are uncorrelated with the regressor no fault, then we get an unbiased estimate of the coefficient on no fault. A sufficient condition for this is that they are uncorrelated with all the regressors (no fault, state, time).
  - b) Suppose they wanted to implement the random effects FGLS estimator for this model. How would they do it?
    - i) They would need to assume that the state effects are uncorrelated with the regressors, including no fault.
    - ii) implement random effects:
      - (1) reg Y nofault i.year
      - (2) collect e
      - (3) get estimates of sigma\_theta and sigma\_epsilon, and form Omega
      - (4) create T=inverse matrix square root(Omega)
      - (5) create TY, Tnofault, Ti.year
      - (6) reg TY Tnofault Ti.year
  - c) How do you think the variance of this estimate would compare to that of the fixed effects estimator that they used?
    - i) it would be smaller, because FGLS is the asymptotically efficient estimator.
  - d) Why do you think they didn't use the random effects FGLS estimator?
    - i) we worried that the state effects were in fact correlated with the take up of no fault. E.g., if everyone marries early in Tennessee, maybe Tennessee people want to have fault.
- 4) [Study Question 8, modified] If you have heteroskedasticity of **unknown** form,
  - a) Why is it not possible to use FGLS?
    - i) heteroskedasticity of unknown form has  $E[\varepsilon\varepsilon'] = \Omega$  where  $\Omega = D$  is a diagonal matrix with elements  $d_i$ . D has N unknown elements, so it cannot be estimated consistently.
  - b) How does the White hetero-robust covariance matrix estimator get around this problem?
    - i) We form an estimate of X'DX, which is a KxK matrix, which can be estimated consistently.
  - c) How does use of the White covariance matrix affect the estimated coefficients, compared to the OLS estimates?
    - i) it does not affect the estimated coefficients.
  - d) Is the White covariance matrix valid at all sample sizes, or only asymptotically? Why?
    - i) because it uses a consistent pre-estimate of X'DX, which is only valid asymptotically, the White covariance matrix is only valid asymptotically.
- 5) Consider the following code and output from a log-wage regression using 2001 Census data on British Columbia residents.

```
generate insamp=pobp<11&agep<65&agep>24&cowp==1&hlosp~=.&
wagesp>0&provp==59;
generate logwage=log(wages);
generate alone=unitsp==1;
recode agep (25/29=1) (30/34=2) (35/39=3) (40/44=4) (45/49=5) (50/54=6) (55/59=7)
       (60/64=8) (else=0), gen(agegp);
generate vismin=visminp<5;</pre>
generate aborig=abethncp<3;</pre>
replace vismin=0 if aborig==1;
generate white=(vismin==0) & (aborig==0);
xi: regress logwage i.agegp i.hlosp i.marsthp i.cmap alone unitsp i.olnp vismin
aborig if(insamp==1&sexp==2);
the Stata output is
         Iagegp_0-8 (naturally coded; _Iagegp_0 omitted)
_Ihlosp_1-14 (naturally coded; _Ihlosp_1 omitted)
_Imarsthp_1-5 (naturally coded; _Imarsthp_1 omitted)
_Icmap_933-935 (naturally coded; _Icmap_933 omitted)
_Iolnp_1-4 (naturally coded; _Iolnp_1 omitted)
i.hlosp
i.marsthp
i.cmap
i.olnp
                                                Number of obs = 8758
F(32, 8725) = 39.91
Prob > F = 0.0000
R-squared = 0.1277
    Source | SS df MS
     Model | 1079.87668 32 33.7461464
   Residual | 7377.23906 8725 .845528832
                                                  Adj R-squared = 0.1245
Root MSE = .91953
     Total | 8457.11574 8757 .96575491
   logwage | Coef. Std. Err. t P>|t| [95% Conf. Interval]
  _Iagegp_1 | -.4752942 .0460058 -10.33 0.000 -.5654763 -.385112
  _Iagegp_6 |
              .007959 .0440889 0.18 0.857 -.0784657
                                                                 .0943837
  -.4212746 -.2038568
 .9456885
                                                                  .8775636
                                                                 1.061845
                                                                 1.104625
                                                                 1.053463
                                                                  1.061395
                                                                 1.155116
                                                                1.135422
                                                                  1.179861
                                                                  1.357049
                                                                 1.200997
                                                                 1.411732
                                                                 1.533136
                                                                 .4615268
 __Imarsthp_3 | .1048837 .0677711
                                     1.55 0.122 -.0279636
                                                                .2377311
-2.69 0.007 -.216326
-0.28 0.776 -.3353126
                                                      -.216326
                                                                -.0337634
                                                                  .2502113
  _Icmap_935 | -.1541335 .0268411 -5.74 0.000 -.2067484 -.1015185
   alone | .2102727 .0384597
unitsp | .0239406 .0092155
_Tolnp_2 | -1.468731 .651815

      5.47
      0.000
      .1348827
      .2856627

      2.60
      0.009
      .0058761
      .0420051

      -2.25
      0.024
      -2.746442
      -.1910201
```

## a Why is <code>\_lagegp\_7</code> dropped? You have to drop 1 element of a list of dummies. Stata chose this one for age.

- b Why is white not a regressor in the regression?
  - Here, the list of dummies would be white, vismin, aborig.
- c How is it that so many coefficients are significant, and yet R-squared is only 12%? Does this suggest a problem in the model?

R-squared measures the size of sigma-squared relative to the variance of *Y*. Just because sigma-squared is big doesn't have anything to do with how good *X* is at predicting the mean of *Y*. It just says that *Y* has a lot of conditional variance remaining after *X* is conditioned out.

- The constant is highly significant, with a t-value of 41. Is this surprising? Why or why not? The constant measures the expectation of the log of earnings for white man aged 55-69, with unreported level of schooling, etc. The t-test tests whether or not this is zero. Zero is the log of \$1. Thus, the t-stat is big because this kind of person has an expected earnings level very far from \$1. Not surprising.
- 6) Suppose that I have the nonlinear equation  $Y_i = \sum_{k=1}^K (X_i^k)^{\beta_k} + \varepsilon_i$ ,  $E[X'\varepsilon] = 0_K$ . Here, there are K parameters and X is an NxK matrix.
  - a) Show how to estimate the parameters of this model by least squares;
    - i)  $\min_{\beta_k} \sum_{k=1}^K (e_i)^2$ ,
    - ii) where

iii) 
$$e_i = Y_i - \sum_{k=1}^K \left(X_i^k\right)^{\beta_k}$$

- iv) get the computer to solve it. or derive the FOC and solve it yourself. but, it is not a one-step solution.
- b) Show how to estimate the parameters of this model by the method of moments

i) model: 
$$Y_i = \sum_{k=1}^K (X_i^k)^{\beta_k} + \varepsilon_i$$

- ii) moment condition:  $E[X'\varepsilon] = 0_K$
- iii) substitute and solve:  $X'e = 0_K$  where  $e_i = Y_i \sum_{k=1}^K (X_i^k)^{\beta_k}$
- iv) get the computer to solve it. but it is not a one-step solution.
- 7) Consider the classical linear model with exogenous spherical errors.
  - a) Derive the variance of the estimator, using the formula for variance, the formula for the OLS estimator, and the model.
    - i) model:  $Y = X\beta + \varepsilon$
    - ii) exogeneity:  $E[X'\varepsilon] = 0_K$
    - iii) spherical errors:  $E[\varepsilon \varepsilon'] = \sigma^2 I_N$
    - iv)  $E[(X'X)^{-1}X'\varepsilon\varepsilon'X(X'X)^{-1}] = (X'X)^{-1}X'E[\varepsilon\varepsilon']X(X'X)^{-1} = (X'X)^{-1}X'\sigma^2I_NX(X'X)^{-1}$
    - $v) = \sigma^2 (X'X)^{-1}$

- b) Why is the variance of each estimated coefficient higher when the regressors are highly correlated?
  - i) When, regressors are highly correlated, the off-diagonals of X'X are big. This means the diagonals of X'X inverse big, because the determinant gets big when the off diagonal terms get big.
  - ii) Or, using Ballentine diagrams, we know that only the covariance that is exclusively between a regressor and Y is used to get precision on that coefficient. As the covariation between regressors rises, the covariance exclusively between each regressor and Y declines, and the joint variation between Y and all the regressors rises.
- c) Why doesn't the OLS estimate change if we leave out an uncorrelated regressor?
  - i) the exclusive covariation between a regressor and Y is used to identify that parameter. that covariation does not change when an uncorrelated regressor is introduced.
  - ii) or, use FWL to show that the partial estimator is unchanged when you partial out an uncorrelated regressor.
- d) Derive the variance of the OLS estimator for the case where errors are not spherical, but instead have a non-diagonal covariance matrix  $\Omega$ .
  - i) model:  $Y = X\beta + \varepsilon$
  - ii) exogeneity:  $E[X'\varepsilon] = 0_K$
  - iii) spherical errors:  $E \left[ \varepsilon \varepsilon' \right] = \Omega$
  - iv)  $E[(X'X)^{-1}X'\varepsilon\varepsilon'X(X'X)^{-1}] = (X'X)^{-1}X'E[\varepsilon\varepsilon']X(X'X)^{-1} = (X'X)^{-1}X'\Omega X(X'X)^{-1}$
- 8) Consider the problem of no-fault divorce and the data of Allen, Pendakur and Suen. Let *state* give the number from 1 to 50 indicating which state the data refer to, and *year* give the year from 1970 to 1989 indicating which year the data refer to. Let *nofault* be a dummy variable indicating whether or not that state-year is no-fault. Let *y* be the dependent variable.
  - a) Write Stata code to implement the fixed effects panel regression.
    - i) panel data program in stata:
      - (1) xtset state
      - (2) xtreg y nofault i.year, fe
    - ii) do it yourself
      - (1) reg y nofault i.year i.state
  - b) Suppose you wanted to run weighted least squares without using the weight suboption in Stata. Let *number* give the number of observations used to compute the variable *y* in each state-year. Write Stata code to do weighted least squares estimation of their fixed effects model without using the weight option in the regress command.
    - i) g y\_weight=y\*weight
    - ii) g nofault\_weight=nofault\*weight
    - iii) g state1\_weight=state1\*weight
    - iv) ...
    - v)  $g state 50_weight = state 50*weight$
    - vi) g year1\_weight=year1\*weight
    - vii)...
    - viii) g year25\_weight=year25\*weight

- ix) reg y\_weight nofault\_weight state2\_weight-state50\_weight year2\_weight year25\_weight
- c) Suppose nf1, nf2, and nf3 were indicator variables giving successively stronger definitions of no-fault divorce. These are not mutually exclusive, because if nf3=1 then both nf2 and nf1 equal 1, and if nf2=1, then nf1=1. Write Stata code that creates a set of mutually exclusive dummy variables, and describe what these dummy variables indicate.
  - i) g nofault3=nf3
  - ii) g nofault2=nf2-nofault3
  - iii) g nofault1=nf1-nofault2
- d) Write Stata code to implement the fixed effects panel regression, but with *fault* as the regressor, where *fault* is an indicator variable equal to 1 when the state-year is fault. How would the estimated coefficient from this regression relate to that from part a) above.
  - i) the coefficient on fault would be the negative of the coefficient on no fault in part a).