

Prelim 1 is on March 27th Thursday @ 10:10 AM

Office Hours: Tomorrow Wednesday 9AM-11AM

(Not exhaustive!)

Topics to Review for Prelim

- - a) best livear predictor (projection) vs. causal livear model
 - b) decomposition of the projection: FWL, OVB
 - c) finite sample theory of DLS:
 - assumptions
 - results: unbiasedness, finite sample variance, Gauss - Markov theorem
 - exact distribution and hypothesis testing under normality
 - d) asymptotic theory:
 - · assumptions

 - · results: consistency, asymptotic normality · inference: joint inference, multiple hypothesis testing

2 Instrumental Variables

- a) intuition
- b) assumptions: relevance + validity is whether and when they are testable / not testable
- c) results: consistent, but not unbiasedness, asymptotics of simple IV (vs. ols)
- d) IV vs. TSLS leads to overidentification
- e) Weak Instrument (Intuition)

(3) <u>GMM</u>

- a) definition
- b) relationship w/ TSLS
- c) assumptions
- d) routs: consistency, asymptotic normality, asymptotic efficiency is optimal very lets 4 efficient TSGMM

estimator

- e) overidentification test
- f) multiple equation GMM:
 - · Ex: SUR, common coefs.,...
 - pro/coms of joinet estimation

4 Panel Data

s go over fixed effects derivations

2016 RI

1. Poisson Regression. A researcher observes the random variables (y_i, \mathbf{x}_i) , where (conditionally on \mathbf{x}_i) y_i is distributed independently according to a Poisson distribution with parameter

$$\lambda_i = \mathbf{x}_i' \boldsymbol{\beta}.$$

Recall that the Poisson distribution has probability mass function

$$Pr(y_i = y) = \lambda^y \exp(-\lambda)/y!, y \ge 0,$$

and expectation and variance equal to λ .

- **1.1** A researcher estimates β by running an OLS regression of y_i on \mathbf{x}_i . Write down the estimator $\widehat{\beta}$. Is $\widehat{\beta}$ consistent?
- **1.2** Is $\widehat{\boldsymbol{\beta}}$ unbiased? Is it linear (in the sense required by Gauss-Markov)? Is it best linear unbiased?
- 1.3 Fitted values from the above regression can be negative. To avoid that, one could model y_i as being Poisson with parameter

$$\lambda_i = \exp\left(\mathbf{x}_i'\boldsymbol{\beta}\right).$$

Assuming this model is true, can β be estimated by OLS regression of $\log y_i$ on \mathbf{x}_i ?

1.4 Can you sketch out a method to estimate β in the model from 1.3?

2024 Q1

- **2** This question is about scalar random variables (Y, X, Z). Researchers observe i.i.d. realizations $(Y_i, X_i, Z_i)_{i=1}^n$ and are interested in the causal effect of X on Y. Assume homoskedasticity throughout.
 - **2.1** Researcher 1 uses the following two-step procedure:
 - 1. Regress X on Z.
 - 2. Regress Y on the fitted values \hat{X} of the first regression.

Name the estimator and state standard assumptions under which this estimator will be appropriate in some asymptotic sense.

- 2.2 Researcher 2 uses the following two-step procedure:
- 1. Regress X on Z.
- 2. Regress Y on the residuals $\hat{\eta} := X \hat{X}$ of the first regression.

Name the estimator and state standard assumptions under which this estimator will be appropriate in some asymptotic sense.

2.3 Suppose all assumptions that you stated hold. What estimator (of the above or other) would you propose?