

Prelim Examination

Friday June 7, 2013, Time limit: 150 minutes

Instructions:

- (i) The exam consists of two parts. The total number of points for each part is 50. The number of points for each question is given below.
- (ii) The exam is closed book and closed notes.
- (iii) To receive full credit for your answers you have to explain your calculations.
You may state additional assumptions.

Part I

Question 1: Probability Density Functions and Regressions (13 Points)

Two random variables, X , and Y , have a joint distribution characterized by the following probability density function (pdf):

$$p(x, y) = \frac{1}{c}y(x + 1), \quad 0 \leq x \leq 1, 0 \leq y \leq 1.$$

- (i) (2 Points) Find the value of the normalization constant c that ensures that $p(x, y)$ is a proper (pdf).
- (ii) (2 Points) Derive the marginal densities $p(x)$ and $p(y)$.
- (iii) (2 Points) Derive the conditional pdf $p(y|x)$.
- (iv) (3 Points) Derive the cumulative density function (cdf) $F(x, y)$.

Now suppose you have a sample of n iid draws (Y_i, X_i) from the above distribution. Based on this sample, you estimate the (misspecified) regression model

$$Y_i = \beta X_i + U_i$$

using the OLS estimator

$$\hat{\beta}_n = \frac{\sum_{i=1}^n X_i Y_i}{\sum_{i=1}^n X_i^2}.$$

- (v) (4 Points) Derive the probability limit *plim* of $\hat{\beta}_n$ as $n \rightarrow \infty$. Note: you can calculate the *plim* numerically.

Question 2: Do Stopping Rules Matter? (26 Points)

Consider the following experiment:

$$Y_i \sim iidN(\theta, 1), \quad i = 1, \dots, n$$

The goal is to compare frequentist and Bayesian inference with respect to θ . Define

$$\hat{\theta}_n = \frac{1}{n} \sum_{i=1}^n Y_i.$$

Now consider the following two stopping rules:

1. Fixed sample size: terminate the experiment when the sample size has reached $n = 2$.
2. Random sample size: terminate the experiment when the t -statistic for the hypothesis $H_0 : \theta = 0$ exceeds the 5% critical value:

$$|\sqrt{n}\hat{\theta}_n| > 1.96.$$

Consider the Stopping Rule 1. The sample size is fixed at $n = 2$.

- (i) (3 Points) Derive the distribution of $\hat{\theta}|\theta$.
- (ii) (5 Points) Derive the posterior distribution $p(\theta|Y_{1:2})$ under the improper (the probability mass is infinite) prior $p(\theta) = 1$, i.e. the prior is uniform on the real line.

Now consider Stopping Rule 2. Before you start your calculations, think about this: we are terminating the experiment when we reject the null hypothesis $\theta = 0$. At this point the frequentist gag reflex should kick in!

Suppose that the experiment happens to be terminated at $n = 2$. The goal is to derive the pdf of $\hat{\theta}$ conditional on the sampling being terminated at $n = 2$. We also condition on the “true” θ being equal to zero.

You can use $\phi(x)$ and $\Phi(x)$ to denote the pdf and cdf of a standard $N(0, 1)$ random variable. We break the frequentist calculation into several steps:

- (iii) (1 Points) Derive $p(y_1, y_2 | \theta = 0, |y_1| \leq 1.96)$.
- (iv) (3 Points) Using a change of variables $(Y_1, Y_2) \mapsto (\hat{\theta}_1, \hat{\theta}_2)$ derive $p(\hat{\theta}_1, \hat{\theta}_2 | \theta = 0, |\hat{\theta}_1| \leq 1.96)$.
- (v) (4 Points) Finally, derive $p(\hat{\theta}_2 | \theta = 0, n = 2)$.
- (vi) (1 Points) Does the stopping rule affect the sampling distribution of $\hat{\theta}_n$ and subsequent inference such as hypothesis testing and confidence intervals? Explain.

Now we repeat the Bayesian analysis under Stopping Rule 2.

- (vii) (4 Points) Treating the sample size n as random variable that can take values $n = 1, 2, 3, \dots$, derive the density of $Y_{1:n}$ given θ , denoted by $p(y_1, y_2, \dots, y_n | \theta)$. Note: this density will take the form of a sum. Each summand corresponds to a size that the sample can take. You only need to write out the first couple of terms.
- (viii) (3 Points) In our observed sample $n = 2$. Using the general density derived in (xii) provide an expression for the likelihood function (given the observed sample).
- (ix) (1 Points) Combine the likelihood function in (xiii) with the improper prior $p(\theta) = 1$ and derive the posterior $p(\theta | y_1, y_2, n = 2)$.
- (x) (1 Points) For better or worse, is Bayesian inference affect by the stopping rule?

Question 3: Inference for Variance Parameters (11 Points)

Consider the model $Y_i \sim iidN(0, \theta)$, $i = 1, \dots, n$.

- (i) (2 Points) Derive the score $s(\theta) = \partial \ln p(Y_{1:n}|\theta)/\partial \theta$.
- (ii) (2 Points) Derive the maximum likelihood estimator $\hat{\theta}$ for θ .
- (iii) (4 Points) Assume that the “true” value is θ_0 and derive the limit distribution of the (properly normalized) score evaluated at $\theta = \theta_0$.
- (iv) (3 Points) Construct the Lagrange multiplier/score test for the hypothesis $H_0 : \theta = \theta_0$ and state 95% critical value as well as the acceptance and rejection region.

Part II

Question 4: Linear Instrumental Variable Regression (10 Points)

Take a linear instrumental variable equation

$$\begin{aligned}y_i &= x_i\beta_1 + z_i\beta_2 + e_i, \\E(e_i|z_i) &= 0\end{aligned}$$

where both x_i and z_i are scalars.

- (i) (2 Points) Can the coefficients (β_1, β_2) be estimated by two-stage-least-squares (2SLS) estimator using z_i as an instrument for x_i ?
- (ii) (2 Points) Can the coefficients (β_1, β_2) be estimated using z_i and z_i^2 as instruments?
- (iii) (2 Points) For the 2SLS estimator suggested in (ii), what is the exclusion restriction on instrument exogeneity?
- (iv) (4 points) For the 2SLS estimator suggested in (ii), what is the implicit assumption about instrument relevance? Write down the implicit reduced form equation for x_i to explain the instrument relevance.

Question 5: Inference for Linear Instrumental Variable Regression (10 Points)

Take a linear equation with endogeneity and a just-identified linear reduced form

$$\begin{aligned}y_i &= x_i\beta + e_i, \\x_i &= z_i\gamma + u_i,\end{aligned}$$

where both x_i and z_i are scalars. Assume that

$$\begin{aligned}E(z_i e_i) &= 0, \\E(z_i u_i) &= 0.\end{aligned}$$

- (i) (4 Points) Write down the standard 2SLS estimator $\widehat{\beta}_{2SLS}$ for β using z_i as an instrument for x_i .
- (ii) (6 points) Derive the asymptotic distribution for $\widehat{\beta}_{2SLS}$. Write the asymptotic variance as a function of $\Omega = E(z_i^2 e_i^2)$, $Q = E(z_i^2)$, and γ . Be clear when you use the weak law of large numbers and the central limit theorem.

Question 6: Hypothesis Testing (15 Points)

Suppose you have two independent i.i.d. samples $\{y_{1i}, x_{1i}, z_{1i} : i = 1, \dots, n\}$ and $\{y_{2i}, x_{2i}, z_{2i} : i = 1, \dots, n\}$. The dependent variables y_{1i} and y_{2i} are scalars and the regressors x_{1i} and x_{2i} and the instruments z_{1i} and z_{2i} are $k \times 1$ vectors. The model is standard just-identified linear instrumental regression

$$\begin{aligned}y_{1i} &= x'_{1i}\beta_1 + e_{1i}, \\E(z_{1i}e_{1i}) &= 0. \\y_{2i} &= x'_{2i}\beta_2 + e_{2i}, \\E(z_{2i}e_{2i}) &= 0.\end{aligned}$$

You want to test $H_0 : \beta_1 = \beta_2$, that the two samples have the same coefficients.

- (i) (5 Points) Develop a test statistic for H_0 .
- (ii) (6 Points) Derive the asymptotic distribution of the test.
- (iii) (4 Points) Describe the testing procedure.

Question 7: Regression with Binary Dependent Variable (15 Points)

Consider the following model

$$Y_i^* = X_i' \beta + e_i,$$

where the error has a conditional distribution $e_i|X_i \sim N(0, \sigma^2)$. Only the sign of Y_i^* is reported. That is, the dataset contains the variable

$$Y_i = \begin{cases} 1 & \text{if } Y_i^* \geq 0 \\ 0 & \text{if } Y_i^* < 0. \end{cases}$$

- (i) (2 Points) Explain why β and σ cannot be identified separately.
- (ii) (3 Points) Suppose σ is normalized to be 1 here and below. Write $E(Y_i|X_i)$ as a function of X_i and β .
- (iii) (5 Points) Write down the log-likelihood function for the maximum likelihood estimation of β .
- (iv) (5 Points). What is the asymptotic distribution of this maximum likelihood estimator? How to estimate its standard error?

END OF EXAM