

Prelim Examination

Friday June 9, 2017. Time limit: 150 minutes

Instructions:

- (i) The total number of points is 44, the number of points for each problem 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.

Problem 1: Two-Step-Ahead Forecasting (13 Points)

Consider the following data generating process:

$$y_t = \rho y_{t-1} + \epsilon_t + \theta \epsilon_{t-1}, \quad 0 \leq \rho < 1, \quad \epsilon_t \sim iid(0, 1). \quad (1)$$

A forecaster considers the following (misspecified) AR(1) model:

$$y_t = \phi y_{t-1} + u_t, \quad u_t \sim iid(0, \sigma_u^2). \quad (2)$$

- (i) (2 Points) Suppose that the forecaster's model were correct and ϕ were known to the forecaster. Show that the optimal two-step ahead predictor under the quadratic forecast error loss function $L(e_{T+2|T}) = e_{T+2|T}^2$, where $e_{T+2|T} = y_{T+2} - \hat{y}_{T+2|T}$, is

$$\hat{y}_{T+2|T}^* = \phi^2 y_T. \quad (3)$$

- (ii) (4 Points) Now suppose that the forecaster tries to estimate $\psi = \phi^2$ in (3) by direct estimation as follows. Let

$$\hat{\psi}_T = \operatorname{argmin} \sum_{t=2}^T (y_t - \psi y_{t-2})^2. \quad (4)$$

Find the probability limit ψ_* of $\hat{\psi}_T$ under the assumption that y_1, \dots, y_T have been generated from model (1). Does this probability limit coincide with ρ^2 if $\theta = 0$?

- (iii) (4 Points) Provide a characterization of the asymptotic variance of

$$\sqrt{T}(\hat{\psi}_T - \psi_*).$$

- (iv) (3 Points) Approximate the frequentist prediction risk

$$\mathbb{E}_{\rho, \theta} [(y_{T+2} - \hat{\psi} y_T)^2]$$

by decomposing it into a part that is due to the conditional variance of y_{T+2} , a part that is due to the misspecification of the forecaster's model, and a part that is due to sampling variance of $\hat{\psi}_T$. Explain in what sense your calculation is approximate rather than exact.

Problem 2: Bayesian Estimation of AR(2) Model (11 Points)

Consider the process

$$y_t = \phi_1 y_{t-1} + \phi_2 y_{t-2} + u_t, \quad u_t \sim iidN(0, 1) \quad (5)$$

and the prior distribution

$$\phi = [\phi_1, \phi_2]' \sim N\left(\mu, \frac{1}{\lambda} I\right), \quad (6)$$

where I is the identity matrix.

- (i) (2 Points) For which values of ϕ is the process $\{y_t\}$ non-explosive?
- (ii) (5 Points) Using a conditional likelihood function, derive the posterior distribution of $\phi|(Y_{1:T}, y_{-1}, y_0, \lambda)$, which we denote by $p(\phi|Y, \lambda)$. Interpret the formula for the posterior mean. What happens as you vary λ ?
- (iii) (4 Points) Derive the marginal likelihood (or data density) $p(Y|\lambda)$. How can $p(Y|\lambda)$ be used to determine the scaling of the prior distribution in a data-driven way?

Problem 3: Missing Observations (13 Points)

Consider the AR(1) model

$$x_t = \phi x_{t-1} + u_t, u_t \sim N(0, 1). \quad (7)$$

Suppose that we observe $y_t = x_t$ if t is even ($t = 0, 2, 4, \dots, T$) and we don't have observations if t is odd ($t = 1, 3, 5, \dots, T-1$). This is a simplified version of a setting in which the period t is, say, a month, but we only observe the series, say, once a quarter.

- (i) (2 Points) Taking y_{t-2} as given, derive $p(y_t | y_{t-2}, \phi)$.
- (ii) (1 Points) Derive the (conditional) likelihood function

$$p(y_2, y_4, \dots, y_{T-2}, y_T | y_0, \phi).$$

- (iii) (2 Points) Combine the likelihood function with a prior density of the form $\phi \sim N(0, 1)$. Is the posterior distribution of ϕ normally distributed? Does this posterior appear to be more or less informative with respect to ϕ compared to a posterior based on observations for both even and odd time periods?
- (iv) (2 Points) Explain why

$$p(x_{t-1} | y_t, y_{t-2}, \phi) = p(x_{t-1} | y_0, y_2, y_4, \dots, y_T, \phi). \quad (8)$$

- (v) (4 Points) Derive $p(x_{t-1} | y_t, y_{t-2}, \phi)$ under the assumption that x_t is stationary and was initialized in the infinite past. Start by deriving the joint distribution $p(x_{t-1}, y_t, y_{t-2} | \phi)$ and then use the formula for the conditional mean $\mu_{x|y}$ and variance $\Sigma_{x|y}$ of a normal distribution.
- (vi) (2 Points) Explain how one can use the ideas of data augmentation and Gibbs sampling to implement Bayesian inference in this model.

Problem 4: Vector Autoregressions (7 Points)

Consider a bivariate structural vector autoregression written in first differences:

$$\Delta y_t = \Phi \Delta y_{t-1} + \Phi_\epsilon \epsilon_t, \quad \epsilon_t \sim iidN(0, I). \quad (9)$$

Suppose the y_t is composed of log GDP and the log consumer price index (CPI). Moreover, suppose that we would like to interpret $\epsilon_{1,t}$ as monetary policy shock and $\epsilon_{2,t}$ as innovation to technology.

- (i) (1 Points) Without further restrictions, are Φ and Φ_ϵ identifiable?
- (ii) (2 Points) According to model (9) what is the effect of a unit shock $\epsilon_{1,t} = 1$ on Δy_{t+h} ? Hint: you can introduce the vector $q = [1, 0]'$ to provide a formula and use the notation $\partial \Delta y_{t+h} / \partial \epsilon_{1,t}$.
- (iii) (2 Points) What is the effect of a unit shock $\epsilon_{1,t} = 1$ on the level of y_{t+h} , that is, $\partial y_{t+h} / \partial \epsilon_{1,t}$?
- (iv) (2 Points) What restriction is imposed on Φ_ϵ by the assumption that “monetary policy shocks do not have any effects on the level in the long-run?” Hint: you can interpret long-run as $h \rightarrow \infty$. Does this assumption allow you to identify all the coefficients of the bivariate structural VAR in (9)?