

**Problem 1 (25 Points)**

Consider the model:

$$y = X\beta + e, \quad (1)$$

where the  $(T \times K)$ -regressor matrix is non-stochastic. Assume that the error term  $e = (e_1, \dots, e_T)'$  has the following structure:

$$e \sim (0, \Phi), \quad \text{with } \Phi \text{ positive definite.}$$

**1. (7 points)**

- a) Show that the OLS estimator  $b = (X'X)^{-1}X'y$  is still unbiased.
- b) Show that the covariance matrix of the OLS estimator  $b = (X'X)^{-1}X'y$  (under the assumption of a general error covariance matrix  $\Phi$ ) is given by

$$\text{Cov}(b) = (X'X)^{-1}(X'\Phi X)(X'X)^{-1}.$$

- c) Discuss briefly why in this framework the GLS estimator  $\hat{\beta}_G = (X'\Phi^{-1}X)^{-1}X'\Phi^{-1}y$  should be preferred to the OLS estimator  $b$ .

**2. (10 points)** Assume now that the error term  $e_t$  has the following structure:

$$E[e_t] = 0, \quad E[e_t^2] = \sigma_t^2, \quad E[e_t e_s] = 0 \quad \forall t \neq s.$$

- a) Write down the expectation vector and the covariance matrix for the error vector  $e$ .
- b) Find a suitable matrix  $P$ , such that  $e^* = Pe \sim (0, \sigma^2 I_T)$  holds for some  $\sigma^2 > 0$ .
- c) Show that the OLS estimator of the model transformed by  $P$  corresponds to the GLS estimator of model (1).

**3. (8 points)** Assume now that  $e_t \sim N(0, \sigma^2)i.i.d.$  Give a  $(1 - \alpha)$  confidence interval for  $x'_{(1)}\beta$ .

**Problem 2 (25 Points)**

Consider the following model:

$$\begin{aligned} y_1 &= X_1\beta_1 + e_1, \\ y_2 &= X_2\beta_2 + e_2, \end{aligned}$$

where  $X_1$  and  $X_2$  are  $(T \times K)$ –matrices of non–stochastic regressors. The error term is assumed to have the following structure:

$$e = \begin{pmatrix} e_1 \\ e_2 \end{pmatrix} \sim (0, \Sigma \otimes I_T) \quad \text{and} \quad \Sigma = \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix}.$$

- a) **(2 points)** Show that separate OLS estimations of  $\beta_1, \beta_2$  are unbiased, i.e.  $E[b_1] = \beta_1$  and  $E[b_2] = \beta_2$ .
- b) **(3 points)** Show that the GLS estimator  $\hat{\beta}_G$  of  $\beta = \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix}$  in the system is also unbiased, i.e.

$$E[\hat{\beta}_G] = \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix}.$$

- c) **(9 points)** Show that the GLS estimator equals the OLS estimator when  $X_1 = X_2 = \bar{X}$ . Under what other condition do GLS and OLS coincide?
- d) **(4 points)** Explain briefly one difference and one commonality between seemingly unrelated regression models and simultaneous equation models?
- e) **(7 points)** Show that

$$\frac{(y_1 - X_1 b_1)'(y_1 - X_1 b_1)}{T - K}$$

is an unbiased estimator for  $\sigma_{11}$ . Hint: You may use e.g. that  $E[z'Az] = \mu'A\mu + \text{tr}(A\Omega)$  if  $z \sim (\mu, \Omega)$  and  $A$  is some symmetric matrix.

### Problem 3 (25 Points)

Consider the following nonlinear model

$$y_t = f(x_t, \beta) + e_t, \quad (2)$$

where  $y_t$ ,  $x_t$  and  $\beta$  are scalars.

- a) (7 points) Derive the *linear pseudomodel* for equation (2) by using a first order Taylor approximation for  $f(x_t, \beta)$  around  $\beta_n$ . Explain how to derive the Gauss-Newton recursion formula from the *linear pseudomodel*.
- b) (7 points) Suppose a specific form of (2) represents a nonlinear consumption function

$$y_t = x_t^\beta + e_t, \quad (3)$$

where  $y_t$  denotes a measure of consumption and  $x_t$  is a measure of income. Derive the Newton-Raphson recursion formula for estimating equation (3) by nonlinear least squares (NLS).

- c) (7 points) A more general form of a consumption function may be written as

$$y_t = \beta_1 x_t^{\beta_2} + e_t. \quad (4)$$

With the economically plausible restriction  $\beta_2 = 1$ , this model becomes linear. How would you obtain reasonable starting values for NLS estimation of  $\beta_1$  and  $\beta_2$  in this case? How would you obtain starting values without any information about  $\beta_2$ ? Discuss the role of using different starting values in NLS estimation.

- d) (4 points) Suppose you want to test  $H : \beta_2 = 1$  in model (4). Would you prefer a Wald, a Lagrange Multiplier or a Likelihood Ratio test for this hypothesis? Explain your decision. (You don't have to compute the chosen test!)

### Problem 4 (25 Points)

Consider the following simultaneous equation model

$$\begin{aligned} y_{t1} &= \gamma_{21}y_{t2} + \gamma_{31}y_{t3} + \beta_{11}x_{t1} + e_{t1} \\ y_{t2} &= \gamma_{32}y_{t3} + \beta_{12}x_{t1} + e_{t2} \\ y_{t3} &= \beta_{23}x_{t2} + e_{t3} \end{aligned}$$

where  $y_i$  are the endogenous and  $x_i$  are exogenous variables. For the errors it is assumed that  $e \sim (0, \Sigma \otimes I_T)$ .

- a) (4 points) Write the model in the form  $Y\Gamma + XB + E = 0$ .
- b) (9 points) Check the identification of the 3 structural equations using the order and the rank criteria. Suggest a suitable estimation procedure for each structural equation.
- c) (6 points) The reduced form estimates are given by

$$\hat{\Pi} = \begin{pmatrix} \hat{\pi}_{11} & \hat{\pi}_{12} & \hat{\pi}_{13} \\ \hat{\pi}_{21} & \hat{\pi}_{22} & \hat{\pi}_{23} \end{pmatrix}.$$

Give efficient estimates for the structural parameters  $\Gamma$  and  $B$  using indirect least squares (ILS) for those equations where it is possible.

- d) (6 points) Check whether the first structural equation is identified when  $\beta_{11} = 0$ . What happens when  $\beta_{12} = 0$  in addition? Explain your result intuitively.