

OLS

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$$y = X\beta + u$$

$$E(u|X) = 0$$

$$E(uu^T) = \sigma^2 I_n$$

Recall,

$$\hat{\beta} = (X^T X)^{-1} X^T y \quad - \text{LS}$$

$$\hat{\beta} \sim N(\beta, (X^T X)^{-1} X^T \sigma^2 X (X^T X)^{-1})$$

but is not efficient.

Let Ψ be a matrix ($n \times n$, pos def)
such that $\Psi u \sim (0, \sigma^2 I_n)$

using Ψ^T to transform model

$$\Psi^T y = \Psi^T X \beta + \Psi^T u$$

$$y^* = X^* \beta + u^*$$

u^* obeys Assump
of GM. \therefore use OLS

$$\begin{aligned} \hat{\beta}_{OLS} &= (X^{*T} X^*)^{-1} X^{*T} y^* \\ &= (X^T \Psi^T \Psi X)^{-1} X^T \Psi^T \Psi y \end{aligned}$$

Note:

$$u \sim (0, \Sigma)$$

$$\Psi^T u \sim (0, \Psi^{\text{T}} \Sigma \Psi^{\text{T}})$$

$$\Psi^{\text{T}} \Sigma \Psi^{\text{T}} = I$$

$$\begin{aligned} &= \Psi^{\text{T}} \Sigma \Psi^{\text{T}} = \Psi^{\text{T}} (\Psi^{\text{T}})^{-1} = \Psi^{\text{T}} (\Psi^{\text{T}})^{-1} \\ &= (\Psi^{\text{T}} \Psi^{\text{T}})^{-1} \end{aligned}$$

$$\Rightarrow (\Psi^{\text{T}} \Psi^{\text{T}})^{-1} = \Sigma \quad \text{or}$$

$$\Psi^{\text{T}} \Psi^{\text{T}} = \Sigma^{-1}$$

$$\therefore \hat{\beta}_{OLS} = (X^T \Sigma^{-1} X)^{-1} X^T \Sigma^{-1} y$$

$$E(\hat{\beta}_{OLS}) =$$

$$\text{Cov}(\hat{\beta}_{OLS}) = E \left[(\hat{\beta}_{OLS} - E(\hat{\beta}_{OLS})) (\hat{\beta}_{OLS} - E(\hat{\beta}_{OLS}))^T \right]$$

$$= (X^T X)^{-1} \Sigma^{-1}$$

=

Actually, this is the estimate you get if minimizing the GLS objective function

$$= (y - X\beta_{OLS})^T \Omega^{-1} (y - X\beta_{OLS}) \quad (7.06)$$

$$= (y - X\beta_{OLS})^T \Psi \Psi^T (y - X\beta_{OLS})$$

$$= (\Psi^T y - \Psi^T X \beta_{OLS})^T (\Psi^T y - \Psi^T X \beta_{OLS})$$

$$= (y^* - X^* \beta_{OLS})^T (y^* - X^* \beta_{OLS})$$

The FOC is

$$X^{*T} (y^* - X^* \beta_{OLS}) = 0$$

$$X^T \Psi \Psi^T (y - X \beta_{OLS}) = 0$$

$$X^T \Psi \Psi^T (y - X \beta_{OLS}) = 0$$

$$X^T \Omega^{-1} (y - X \beta_{OLS}) = 0 \quad (7.07)$$

$$\begin{aligned}
 \hat{\beta} &= (X^T \Omega^{-1} X)^{-1} X^T \Omega^{-1} y \\
 &= (X^T \Omega^{-1} X)^{-1} X^T \Omega^{-1} (X\beta + u) \\
 &= \beta + (X^T \Omega^{-1} X)^{-1} X^T \Omega^{-1} u \\
 E(\hat{\beta}) &= \beta + E[(X^T \Omega^{-1} X)^{-1} X^T \Omega^{-1} u] \\
 &= \beta \quad \text{since } E(u|X) = 0
 \end{aligned}$$

$$E((\hat{\beta} - \beta)(\hat{\beta} - \beta)^T) =$$

$$\begin{aligned}
 &= E\left[(X^T \Omega^{-1} X)^{-1} X^T \Omega^{-1} u u^T \Omega^{-1} X (X^T \Omega^{-1} X)^{-1}\right] \\
 &= E\left[(X^T \Omega^{-1} X)^{-1} X^T \Omega^{-1} \Sigma \Omega^{-1} X (X^T \Omega^{-1} X)^{-1}\right] \\
 &= (X^T \Omega^{-1} X)^{-1}
 \end{aligned}$$

$$\therefore \hat{\beta}_{GLS} \sim (\beta, (X^T \Omega^{-1} X)^{-1})$$

which by GM is efficient.

$$\begin{aligned}
 \text{Cov}(\hat{\beta}) - \text{Cov}(\hat{\beta}_{GLS}) &= \Delta \\
 (X^T X)^{-1} X^T \Sigma X (X^T X)^{-1} - (X^T \Omega^{-1} X)^{-1} &= \Delta \text{ p.s.d.}
 \end{aligned}$$

So, if Σ^{-1} is known

$$\hat{\beta}_{OLS} = (X^T \Sigma^{-1} X)^{-1} X^T \Sigma^{-1} y \quad \text{is efficient.}$$

Furthermore,

under general conditions $E(u|X) = 0$ or $E(u_i | T_i) = 0$

$$\sqrt{n}(\hat{\beta}_{OLS} - \beta) \xrightarrow{d} N(0, \sigma^2 \text{Cov}(X \frac{\Sigma^{-1} X}{n})^{-1})$$

Of course, if Σ^{-1} known, you could transform $y = X\beta + u$ using Ψ^T where $\Psi\Psi^T = \Sigma^{-1}$

$$\Psi^T y = \Psi^T X \beta + \Psi^T u$$

$$y^* = X^* \beta + u^*$$

$$\hat{\beta}_{OLS} \stackrel{a}{\sim} N(\beta, \sigma^2 (X^T \Sigma^{-1} X)^{-1})$$

EST. of Cov.

$$\hat{\sigma}_y^2 = (y - X\hat{\beta}_{OLS})^T \Sigma^{-1} (y - X\hat{\beta}_{OLS}) / (n-k)$$

\xrightarrow{a} and

$$\widehat{\text{Cov}}(\hat{\beta}) = \hat{\sigma}_y^2 (X^T X)^{-1} = \hat{\sigma}_y^2 (X^T \Sigma^{-1} X)^{-1}$$

Hypothesis Tests

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$$H_0: R\beta = r$$

$$H_1: R\beta \neq r$$

$$R \text{ } J \times K, \text{ Rank}(R) = J \leq K$$

$$r \text{ } J \times 1$$

$$\lambda_1 = \frac{(R\beta_{OLS} - r)' [R(X'X)^{-1}R']^{-1} (R\beta_{OLS} - r)}{J \hat{\sigma}_e^2}$$

$$\leadsto F_5 / 5 \text{ if } H_0 \text{ True.}$$

n approximate with $F_{5, n-4}$.

NOTE: on untransformed model

replace $X'X^{-1}X = X'Z^{-1}X$.

FEASIBLE GLS

unfortunately Σ and ψ are usually unknown.

$\frac{n(n+1)}{2}$ elements in Σ

So, the idea is to estimate Σ or ψ as a function of a small # of parameters, θ

$$\begin{aligned} \Sigma(\theta) &\text{ estimated by } \hat{\Sigma}(\hat{\theta}) \equiv \hat{\Sigma} \\ \psi(\theta) &\text{ " " } \hat{\psi}(\hat{\theta}) \equiv \hat{\psi} \end{aligned}$$

and FGCS

$$\hat{\beta}_{GLS} = (X^T \hat{\Sigma}^{-1} X)^{-1} X^T \hat{\Sigma}^{-1} y \quad \text{or} \quad (X^*{}^T X^*)^{-1} X^*{}^T y^*$$

$$\text{with } \begin{aligned} X^* &= \psi^T X \\ y^* &= \psi^T y \end{aligned}$$

$$\sqrt{n} (\hat{\beta}_{GLS} - \beta) \xrightarrow{L} N(0, \text{plim}(X^T \hat{\Sigma}^{-1} X)^{-1}) \quad \text{Some are GLS!}$$