

Equation 1
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Method of estimation = Ordinary Least Squares

Dependent variable: RCONS
Current sample: 1956 to 1997
Number of observations: 42

Mean of dependent variable = 149038.
Std. dev. of dependent var. = 78147.9
Sum of squared residuals = .127951E+10
Variance of residuals = .319878E+08
Std. error of regression = 5655.77
R-squared = .994890

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Adjusted R-squared = .994762
Durbin-Watson statistic = .116873
F-statistic (zero slopes) = 7787.70
Schwarz Bayes. Info. Crit. = 17.4101
Log of likelihood function = -421.469

Variable	Estimated Coefficient	Standard Error	t-statistic
C	-3317.80	1934.49	-1.71508
RYD	.854577	.968382E-02	88.2480

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Equation 2
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Method of estimation = Ordinary Least Squares

Dependent variable: RCONS
Current sample: 1956 to 1997
Number of observations: 42

Mean of dependent variable = 149038.
Std. dev. of dependent var. = 78147.9
Sum of squared residuals = .244501E+09
Variance of residuals = .643423E+07
Std. error of regression = 2536.58
R-squared = .999024

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Adjusted R-squared = .998946
Durbin-Watson statistic = .420979
F-statistic (zero slopes) = 12959.1
Schwarz Bayes. Info. Crit. = 15.9330
Log of likelihood function = -386.714

Variable	Estimated Coefficient	Standard Error	t-statistic
C	4204.11	1440.45	2.91861
D1	-39915.3	3154.24	-12.6545
RYD	.786609	.015024	52.3561
D1RYD	.194495	.018731	10.3839

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1. Equation 1

Significance test:

Equation 1 is:

$$RCONS = \beta_1 + \beta_2 RYD$$

$$H_0 : \beta_2 = 0$$

(No.1) *t* Test \Rightarrow Compare 10.3839 and $t(42 - 2)$.

(No.2) *F* Test \Rightarrow Compare $\frac{R^2/G}{(1 - R^2)/(n - k)} = \frac{.994890/1}{(1 - .994890)/(42 - 2)} = 7787.8$ and $F(1, 40)$.

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1% point of $F(1, 40) = 7.31$

$H_0 : \beta_2 = 0$ is rejected.

2. Equation 1 vs. Equation 2

Test the structural change between 1973 and 1974.

Equation 2 is:

$$RCONS = \beta_1 + \beta_2 D1 + \beta_3 RYD + \beta_4 RYD \times D1$$

$$H_0 : \beta_2 = \beta_4 = 0$$

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Restricted OLS \implies Equation 1

Unrestricted OLS \implies Equation 2

$$\frac{(\tilde{u}'\tilde{u} - e'e)/G}{e'e/(n-k)} = \frac{(.127951E+10 - .244501E+09)/2}{.244501E+09/(42-4)} = 80.43$$

which should be compared with $F(2, 38)$.

1% point of $F(2, 38) = 5.211 < 80.43$

$H_0 : \beta_2 = \beta_4 = 0$ is rejected.

\implies The structure was changed in 1974.

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First-Order Autocorrelation (一階の自己相関, 系列相関)

In the case of time series data, the subscript is conventionally given by t , not i .

$$u_t = \rho u_{t-1} + \epsilon_t, \quad \epsilon_t \sim \text{iid } N(0, \sigma_\epsilon^2)$$

$$\sigma^2 \Omega = \frac{\sigma_\epsilon^2}{1-\rho^2} \begin{pmatrix} 1 & \rho & \rho^2 & \dots & \rho^{n-1} \\ \rho & 1 & \rho & \dots & \rho^{n-2} \\ \rho^2 & \rho & 1 & \dots & \rho^{n-3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \rho^{n-1} & \rho^{n-2} & \rho^{n-3} & \dots & 1 \end{pmatrix}$$

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4. In general, when Ω is symmetric, Ω is decomposed as follows.

$$\Omega = A' \Lambda A$$

Λ is a diagonal matrix, where the diagonal elements of Λ are given by the eigen values.

A is a matrix consisting of egeenvectors.

When Ω is a positive definite matrix, all the diagonal elements of Λ are positive.

5. There exists P such that $\Omega = PP'$ (i.e., take $P = A' \Lambda^{1/2}$).

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8 Generalized Least Squares Method (GLS, 一般化最小自乗法)

1. Regression model: $y = X\beta + u, \quad u \sim (0, \sigma^2 \Omega)$

2. **Heteroscedasticity** (不等分散, 不均一分散)

$$\sigma^2 \Omega = \begin{pmatrix} \sigma_1^2 & 0 & \dots & 0 \\ 0 & \sigma_2^2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \sigma_n^2 \end{pmatrix}$$

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$$V(u_t) = \sigma^2 = \frac{\sigma_\epsilon^2}{1-\rho^2}$$

3. The generalized least squares (GLS) estimator of β , denoted by b , solves the following minimization problem:

$$\min_{\beta} (y - X\beta)' \Omega^{-1} (y - X\beta)$$

The GLSE of β is:

$$b = (X' \Omega^{-1} X)^{-1} X' \Omega^{-1} y$$

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Multiply P^{-1} on both sides of $y = X\beta + u$.

We have:

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

where $y^* = P^{-1}y, \quad X^* = P^{-1}X, \quad \text{and} \quad u^* = P^{-1}u$.

Note that

$$V(u^*) = V(P^{-1}u) = P^{-1}V(u)P^{-1} = \sigma^2 P^{-1} \Omega P^{-1} = \sigma^2 I_n,$$

because $\Omega = PP'$, i.e., $P^{-1} \Omega P^{-1} = I_n$.

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Accordingly, the regression model is rewritten as:

$$y^* = X^*\beta + u^*, \quad u^* \sim (0, \sigma^2 I_n)$$

Apply OLS to the above model.

That is,

$$\min_{\beta} (y^* - X^*\beta)'(y^* - X^*\beta)$$

is equivalent to:

$$\min_{\beta} (y - X\beta)'\Omega^{-1}(y - X\beta)$$

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$$b = (X^{*\prime}X^*)^{-1}X^{*\prime}y^* = (X'\Omega^{-1}X)^{-1}X'\Omega^{-1}y$$

$$b = \beta + (X^{*\prime}X^*)^{-1}X^{*\prime}u^* = \beta + (X'\Omega^{-1}X)^{-1}X'\Omega^{-1}u$$

$$E(b) = \beta$$

$$V(b) = \sigma^2(X^{*\prime}X^*)^{-1} = \sigma^2(X'\Omega^{-1}X)^{-1}$$

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