

Kap. 9.  $\Leftrightarrow$  γχων Στάθμοι Απλών.

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[9.1] Οριστική και χρήστικη πρόδοση απλών.

$$c_t = \beta_1 + \beta_2 y_t + \varepsilon_t.$$

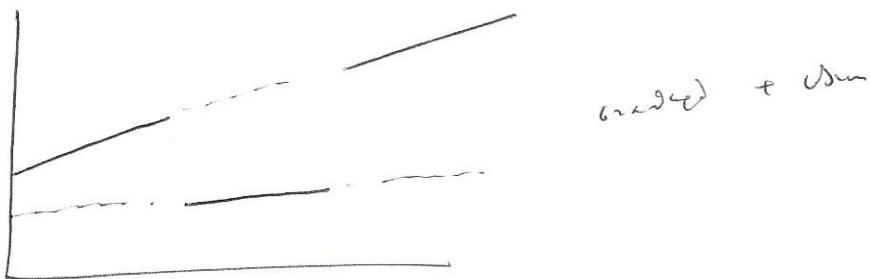
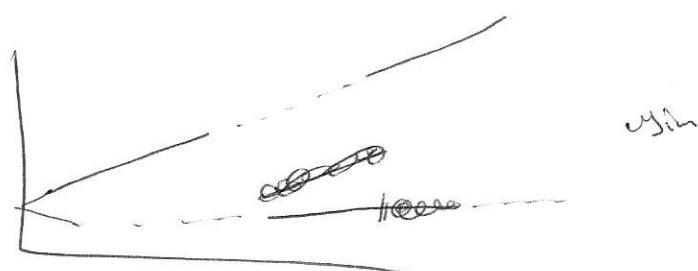
$$M_1: c_t = \beta_1 + \delta d_t + \beta_2 y_t + \varepsilon_t.$$

$$M_2: c_t = \beta_1 + \beta_2 y_t + \delta(d_t y_t) + \varepsilon_t.$$

$$M_3: c_t = \beta_1 + \delta_1 d_t + \beta_2 y_t + \delta_2 (d_t y_t) + \varepsilon_t$$

• προβλήματα σε χρήση μεταβλητών για πρόδοση απλών.

• ψευδογραφίες εποικιών επιδρίσεων.



9.2 Στατιστικής Ελάχησης (Α πρόγνωση).

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Ελάχηση Chow

$$y = X\beta + \varepsilon.$$

If: Σε διάφορους χρόνους έχουμε επεξεργαστές για χρονοδιάστημα  $T_1$  και  $T_2$  ανεπικρίτως.

1)  $T_1$  παραγράφεται:  $\hat{\beta}_{T_1} = (X'_{T_1} X_{T_1})^{-1} \cdot X'_{T_1} y_{T_1}$

2) Η βάση για  $\hat{\beta}_{T_1}$  βρίσκεται προβληματική για  $t = T_1+1, T_1+2, \dots, T$ .

$$\hat{y}_{T_2} = X_{T_2} \cdot \hat{\beta}_{T_1}$$

3) Διανούμε ερωτήσεων απόβλητων.

$$d = y_{T_2} - \hat{y}_{T_2} = y_{T_2} - X_{T_2} \cdot \hat{\beta}_{T_1} \quad \left. \right\} \Rightarrow$$

$$d \Leftrightarrow V_{T_2} = X_{T_2} \cdot B + \varepsilon_{T_2}$$

~~$$d = X_{T_2} \cdot B + \varepsilon_{T_2} - X_{T_2} \cdot \hat{\beta}_{T_1} \Rightarrow$$~~

$$\Rightarrow d = \varepsilon_{T_2} - X_{T_2} \cdot (\hat{\beta}_{T_1} - B).$$

$$E(d) = E \left[ \varepsilon_{T_2} - X_{T_2} (\hat{\beta}_{T_1} - B) \right] = \underbrace{E(\varepsilon_{T_2})}_{0} - \underbrace{E \left[ X_{T_2} (\hat{\beta}_{T_1} - B) \right]}_{0} = 0.$$

$$\begin{aligned} \text{Var}(d) &= E(d d') = E \left\{ \left[ \varepsilon_{T_2} - X_{T_2} (\hat{\beta}_{T_1} - B) \right] \left[ \varepsilon_{T_2} - X_{T_2} (\hat{\beta}_{T_1} - B) \right]' \right\} = \\ &= \sigma^2 \cdot I_{T_2} + X_{T_2} \cdot \text{Var}(\hat{\beta}_{T_1}) \cdot X'_{T_2} = \end{aligned}$$

$$= \sigma^2 \cdot \left[ I_{T_2} + X_{T_2} \cdot (X'_{T_1} X_{T_1})^{-1} \cdot X'_{T_2} \right].$$

Παρα:  $\text{Var}(\hat{\beta}_{T_1}) = E(\hat{\beta}_{T_1} - \beta_1)(\hat{\beta}_{T_1} - \beta_1)' = \sigma^2 \cdot (X'_{T_1} X_{T_1})^{-1}.$

$$\text{d} \sim N_{\text{ind}}(0, \sigma^2)$$

then  $d \sim N(0, \text{var}(a))$ .

$$d' [ \text{Var}(d) ]^{-1} \cdot d \sim \chi^2_{T_2}$$

$$\frac{\hat{\epsilon}'_{T_1} \hat{\epsilon}_{T_1}}{\sigma^2} \sim \chi^2_{T_2}$$

ii)  $T_1, T_2$  uncorrelated:

$$\begin{aligned} F_{\text{chow}} &= d' [ \text{Var}(d) ]^{-1} \cdot d / T_2 = d' \left[ \left[ I_{T_2} + X_{T_2} (X'_{T_1} X_{T_1})^{-1} X'_{T_2} \right] \cdot \hat{\epsilon}'_{T_1} \right]^{-1} d / T_2 \\ &= \frac{d' \left[ I_{T_2} + X_{T_2} (X'_{T_1} X_{T_1})^{-1} X'_{T_2} \right] d}{\hat{\epsilon}'_{T_1} / T_1 - k}. \end{aligned}$$

$$\text{and } \hat{\epsilon}'_{T_1} = \frac{\hat{\epsilon}'_{T_1} \hat{\epsilon}_{T_1}}{T_1 - k}.$$

$$F_{\text{chow}} \sim F_{T_2, T_1 - k}.$$

Accept. H<sub>0</sub>  $\Leftrightarrow F_{\text{chow}} > F_{T_2, T_1 - k, \alpha}$ .

• Advantages of F-tests.

$$1) Y_{T_1} = X_{T_1} \beta + \varepsilon_{T_1} \text{ known, } \rightarrow \text{partial R.S.S.}_{T_1} = \hat{\epsilon}'_{T_1} \hat{\epsilon}_{T_1}$$

$$2) Y = X \beta + \varepsilon \text{ all unknown data} \rightarrow \text{partial R.S.S.}_{T_1} = \hat{\epsilon}' \hat{\epsilon}.$$

$$3) F_{\text{chow}} = \frac{(R.S.S. - R.S.S._{T_1}) / T_2}{R.S.S._{T_1} / (T_1 - k)} = \frac{(\hat{\epsilon}' \hat{\epsilon} - \hat{\epsilon}'_{T_1} \hat{\epsilon}_{T_1}) / T_2}{\hat{\epsilon}'_{T_1} \hat{\epsilon}_{T_1} / (T_1 - k)} \sim F_{T_2, T_1 - k}.$$

Glejxar Allaxun Gross konfideras enos urodia fakos.

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• Eufijnen konfideras ja lajoptuna difara aukcenjum.

►  $y = XB + \varepsilon$ . 2 uaderorax

1:  $t=1, \dots, T_1$ .

2:  $t=T_1+1, T_1+2, \dots, T$ .

To uaderofha dekretax:

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_{T_1} \end{pmatrix} = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \\ \vdots & \vdots \\ 0 & X_{T_1} \end{bmatrix} \begin{bmatrix} B_1 \\ B_2 \\ \vdots \\ B_{T_1} \end{bmatrix} + \varepsilon. \quad (3)$$

$y_1 : T_1 \times 1$   
 $y_2 : T_2 \times 1 = [T - T_1] \times 1$ .

$X_1 : T_1 \times K$   
 $X_2 : T_2 \times K = (T - T_1) \times K$

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_{T_1} \\ y_{T_1+1} \\ y_{T_1+2} \\ \vdots \\ y_T \end{pmatrix} = \begin{bmatrix} 1 & X_{1,1} & \dots & X_{1,K} & 0 & 0 & \dots & 0 \\ 1 & X_{2,1} & \dots & X_{2,K} & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & X_{T_1,1} & \dots & X_{T_1,K} & 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & 1 & X_{T_1+1,1} & \dots & X_{T_1+1,K} \\ 0 & 0 & \dots & 0 & 1 & X_{T_1+2,1} & \dots & X_{T_1+2,K} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 & X_{T_1+K,1} & \dots & X_{T_1+K,K} \end{bmatrix} \begin{bmatrix} B_{1,1} \\ B_{1,2} \\ \vdots \\ B_{1,K} \\ B_{2,1} \\ B_{2,2} \\ \vdots \\ B_{2,K} \end{bmatrix}$$

► To uaderofha (3) lajed = dekar uo t+ en xedon pefobrizaBunur:

$$y_t = B_{1,1} \cdot d_{t,1} + B_{1,2} \cdot X_{t,2} \cdot d_{t,1} + B_{1,3} \cdot X_{t,3} \cdot d_{t,1} + \dots + B_{1,K} \cdot X_{t,K} \cdot d_{t,1}$$

$$+ B_{2,1} \cdot d_{t,2} + B_{2,2} \cdot X_{t,2} \cdot d_{t,2} + \dots + B_{2,K} \cdot X_{t,K} \cdot d_{t,2} + \varepsilon_t. =$$

$$= (d_{t,1} \ X_{t,2} \ d_{t,1} \ \dots \ X_{t,K} \ d_{t,1}) \begin{pmatrix} B_{1,1} \\ B_{1,2} \\ \vdots \\ B_{1,K} \end{pmatrix} + (d_{t,2} \ X_{t,2} \ d_{t,2} \ \dots \ X_{t,K} \ d_{t,2}) \begin{pmatrix} B_{2,1} \\ B_{2,2} \\ \vdots \\ B_{2,K} \end{pmatrix} + \varepsilon_t$$

$$\text{dann } d_{t,1} = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \\ 0 \end{pmatrix} \leftarrow T_1 \times 1$$

$$\text{und } d_{t,2} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix} \leftarrow T_2 \times 1$$

► LS schätzen.

$$\begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} = \left\{ \begin{bmatrix} x_1 & 0 \\ 0 & x_2 \end{bmatrix}' \begin{bmatrix} x_1 & 0 \\ 0 & x_2 \end{bmatrix} \right\}^{-1} \cdot \begin{bmatrix} x_1' & 0 \\ 0 & x_2' \end{bmatrix}' \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} =$$

$$= \begin{bmatrix} x_1' x_1 & 0 \\ 0 & x_2' x_2 \end{bmatrix}^{-1} \cdot \begin{bmatrix} x_1' y_1 \\ x_2' y_2 \end{bmatrix} = \begin{bmatrix} (x_1' x_1)^{-1} \cdot x_1' y_1 \\ (x_2' x_2)^{-1} \cdot x_2' y_2 \end{bmatrix}.$$

$$\text{d.h. } \hat{\beta}_1 = (x_1' x_1)^{-1} x_1' y_1, \quad \hat{\beta}_2 = (x_2' x_2)^{-1} x_2' y_2.$$

$$V_{\text{var}} \begin{pmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix} = \sigma^2 \left\{ \begin{bmatrix} x_1 & 0 \\ 0 & x_2 \end{bmatrix}' \begin{bmatrix} x_1 & 0 \\ 0 & x_2 \end{bmatrix} \right\}^{-1} = \sigma^2 \cdot \begin{bmatrix} x_1' x_1 & 0 \\ 0 & x_2' x_2 \end{bmatrix}^{-1}.$$

$$\text{d.h. } \sigma^2 = \frac{\hat{\epsilon}' \hat{\epsilon}}{T-2k}.$$

$$\text{d.h. } \text{var}(\hat{\beta}_1) = \sigma^2 \cdot (x_1' x_1)^{-1}$$

$$\text{var}(\hat{\beta}_2) = \sigma^2 \cdot (x_2' x_2)^{-1}.$$

Da Voraussetzung der Schätzungen:  $\hat{\beta}_1 \rightarrow \text{aus } t=1 \dots T$ ,

$\hat{\beta}_2 \rightarrow \text{aus } t=T+1 \dots T$

$$\hat{\epsilon}' \hat{\epsilon} = \hat{\epsilon}_1' \hat{\epsilon}_1 + \hat{\epsilon}_2' \hat{\epsilon}_2.$$

Informationen direkt aus den Schätzungen:  $\begin{bmatrix} x_1' x_1 & 0 \\ 0 & x_2' x_2 \end{bmatrix}$  eine Blockdiagonale (nur diagonalen Blöcken).

► Hypothesen:  $H_0: \beta_1 = \beta_2$ .

$H_1: \beta_1 \neq \beta_2$ .

$$R \cdot \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = r$$

$$(k \times 2k) \cdot (2k \times 1) = (k \times 1).$$

$$H_0: \begin{cases} \beta_{11} = \beta_{21} \\ \beta_{12} = \beta_{22} \\ \vdots \\ \beta_{1k} = \beta_{2k} \end{cases}$$

$$\text{d.h. } R = \begin{bmatrix} 1 & 0 & \dots & 0 & -1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 & 0 & -1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & 0 & 0 & \dots & -1 \end{bmatrix}, \quad B = \begin{bmatrix} \beta_{11} \\ \beta_{12} \\ \vdots \\ \beta_{1k} \\ \hline \beta_{21} \\ \beta_{22} \\ \vdots \\ \beta_{2k} \end{bmatrix}, \quad r = \begin{pmatrix} 0 \\ 0 \\ \vdots \end{pmatrix}$$

$$F = \frac{\left( R \cdot \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} - r \right)' \cdot k}{R \cdot \text{Var} \left( \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} \right) \cdot R'}^{-1} \cdot \left( R \cdot \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} - r \right) \sim F_{k, T-2k}.$$

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$\Leftrightarrow F > F_{k, T-2k, \alpha}$  resp.  $H_0$

### • Ισοδυναμός τύπωσης ( $\beta$ )

Τυποληφτική ( $\beta$ ):  $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} x_1 & 0 \\ 0 & x_2 \end{pmatrix} \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} + \varepsilon$ . ευτίθεται  $\begin{pmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix}$  ως βίαιης  $\hat{\varepsilon}$ , ως αδροίδηα τετελεστή μεταβολής  $\hat{\varepsilon}'\hat{\varepsilon}$ .

Τυποληφτική ( $\beta$ ):  $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \beta + \varepsilon^*$ ;  $\gamma = X\beta + \varepsilon^*$ , ευτίθεται  $\hat{\beta}$ , υποδογή  $\hat{\varepsilon}^*$ ,  $\Rightarrow \hat{\varepsilon}^*\hat{\varepsilon}^*$ .

$$F = \frac{[(\hat{\varepsilon}^*\hat{\varepsilon}^*) - (\hat{\varepsilon}'\hat{\varepsilon})]/k}{\hat{\varepsilon}'\hat{\varepsilon} / (T-2k)} \sim F_{k, T-2k}.$$

$$\text{Βασικό} \quad f = \frac{(RSS_R - RSS_U)/J}{RSS_U/(N-k)} = \frac{(\hat{\varepsilon}^{*'}\hat{\varepsilon}^* - \hat{\varepsilon}'\hat{\varepsilon})/J}{\hat{\varepsilon}'\hat{\varepsilon}/(N-k)} \sim F_{J, N-k}$$

(ΚΕΠ. 4)

ΕΣΩΤΙΚΟ ΑΠΛΑΧΣ ΣΤΑ ΣΩΑΘΕΡΑ.

• Βασική  $\hat{Y}_t$ :  $\begin{pmatrix} \hat{y}_1 \\ \hat{y}_2 \end{pmatrix} = \begin{pmatrix} I_1 & 0 & \tilde{X}_1 \\ 0 & I_2 & \tilde{X}_2 \end{pmatrix} \begin{pmatrix} \hat{\beta}_{11} \\ \hat{\beta}_{21} \\ \tilde{B} \end{pmatrix} + \varepsilon_t$ .

$$\hat{X}_1 = \begin{pmatrix} X_{11} & \dots & X_{1K} \\ X_{21} & \dots & X_{2K} \\ \vdots & \ddots & \vdots \\ X_{T_1,1} & \dots & X_{T_1,K} \end{pmatrix} \quad \tilde{X}_2 = \begin{pmatrix} X_{T_1+1,2} & \dots & X_{T_1+1,K} \\ X_{T_1+2,2} & \dots & X_{T_1+2,K} \\ \vdots & & \vdots \\ X_{T_2,2} & \dots & X_{T_2,K} \end{pmatrix} \quad \hat{\beta} = \begin{pmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \\ \vdots \\ \hat{\beta}_K \end{pmatrix}$$

xwps σειμ διε σειρά

$$I_1 = \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix} \leftarrow T_1 x_1 \text{ vector.} \quad I_2 = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \leftarrow (T - T_1) x_1 \text{ vector.}$$

•  $\hat{y}_t = \hat{\beta}_{11} \cdot d_{t1} + \hat{\beta}_{21} \cdot d_{t2} + \hat{\beta}_2 X_{t+2} + \dots + \hat{\beta}_K X_{t+K} + \varepsilon_t$ .

•  $H_0: \hat{\beta}_{12} = \hat{\beta}_{21}$ .

$H_1: \hat{\beta}_{11} \neq \hat{\beta}_{21}$ .

$$F = \frac{\left[ R \begin{pmatrix} \hat{\beta}_{11} \\ \hat{\beta}_{21} \\ \tilde{B} \end{pmatrix} - r \right]' \cdot \left[ R \cdot \text{Var} \begin{pmatrix} \hat{\beta}_{11} \\ \hat{\beta}_{21} \\ \tilde{B} \end{pmatrix} \cdot R' \right]^{-1} \cdot \left[ R \cdot \begin{pmatrix} \hat{\beta}_{11} \\ \hat{\beta}_{21} \\ \tilde{B} \end{pmatrix} - r \right]}{r}$$

$H_0: R \cdot \begin{pmatrix} \hat{\beta}_{11} \\ \hat{\beta}_{21} \\ \tilde{B} \end{pmatrix} = r$ .

$$\sim F_{L, T-K-L} \quad L, T-(K+1).$$

όπου  $R = \begin{pmatrix} 1 & -1 & 0 & \dots & 0 \\ & & \ddots & & \\ [1 \times (K+1)] & & & & \end{pmatrix}$

$$\begin{pmatrix} \hat{\beta}_{11} \\ \hat{\beta}_{21} \\ \tilde{B} \end{pmatrix}, \quad \begin{matrix} r \\ (K+1)x_1 \end{matrix}$$

Gegeben:  $y_{t_1}, \dots, y_{t_k}$  mit den entsprechenden Werten

$$\text{zu schätzen: } \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} l_1 & \tilde{x}_1 & 0 \\ l_2 & 0 & \tilde{x}_2 \end{pmatrix} \begin{pmatrix} B_1 \\ \tilde{B}_1 \\ \tilde{B}_2 \end{pmatrix} + \varepsilon.$$

$$\text{dann } \tilde{B}_1 = \begin{pmatrix} B_{11} \\ B_{12} \\ \vdots \\ B_{1k} \\ (k-1) \times 1 \end{pmatrix}, \quad \tilde{B}_2 = \begin{pmatrix} B_{21} \\ B_{22} \\ \vdots \\ B_{2k} \\ (k-1) \times 1 \end{pmatrix}, \quad \tilde{x}_1 = \begin{pmatrix} x_{11} & \dots & x_{1k} \\ x_{21} & \dots & x_{2k} \\ \vdots & \ddots & \vdots \\ x_{T_1,1} & \dots & x_{T_1,k} \end{pmatrix}, \quad \tilde{x}_2 = \begin{pmatrix} x_{T_1+1,1} & \dots & x_{T_1+1,k} \\ \vdots & & \vdots \\ x_{T_2,1} & \dots & x_{T_2,k} \end{pmatrix}$$

$$l_1 = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \leftarrow T_1 \times 1, \quad l_2 = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \leftarrow T_2 \times 1. = (T - T_1) \times 1.$$

$$\text{. } y_t = \beta_1 + \beta_{12} x_{t+2} d_{t_1} + \dots + \beta_{1k} x_{t+k} d_{t_1} + \beta_{21} x_{t+2} d_{t_2} + \dots + \beta_{2k} x_{t+k} d_{t_2} + \varepsilon_t.$$

$$\begin{array}{ll} \text{H}_0: \tilde{B}_1 = \tilde{B}_2 & \text{H}_0: R \cdot \begin{pmatrix} \tilde{B}_1 \\ \tilde{B}_1 \\ \tilde{B}_2 \end{pmatrix} = r. \\ \text{H}_1: \tilde{B}_1 \neq \tilde{B}_2. & \downarrow \\ & (k-1) \times [1 + 2 \cdot (k-1)]. \end{array} \quad \text{dann } R = \begin{pmatrix} 0 & 1 & 0 & \dots & 0 & -1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 & 0 & -1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \ddots & 0 & 0 & 0 & \ddots & -1 \end{pmatrix}_{k-1 \times k-1}.$$

$$F = \left[ R \cdot \begin{pmatrix} \tilde{B}_1 \\ \tilde{B}_1 \\ \tilde{B}_2 \end{pmatrix} - r \right]' \left[ R \cdot \text{Var} \left( \begin{pmatrix} \tilde{B}_1 \\ \tilde{B}_1 \\ \tilde{B}_2 \end{pmatrix} \right) \cdot R' \right]^{-1} \cdot \left[ R \cdot \begin{pmatrix} \tilde{B}_1 \\ \tilde{B}_1 \\ \tilde{B}_2 \end{pmatrix} - r \right] \xrightarrow{k-1} F_{k-1, T - 2(k-1)-1}.$$