

# Multiple changes in coefficients of autoregressive models

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## Abstract

We deal with an  $F$  type test for detection of changes in parameters of an autoregressive model. We apply resampling methods to approximate the critical values of the test.

## 1. AR(p) model

$$y_i = \mathbf{x}'_i \boldsymbol{\beta}_j + e_i \quad t_{j-1} < i \leq t_j, \quad j = 1, \dots, k+1$$

unknown change points  $t_1, t_2, \dots, t_k$   
convention  $t_0 = p, t_{k+1} = n$

$$\mathbf{x}'_i = (1, y_{i-1}, y_{i-2}, \dots, y_{i-p}), \quad i = p+1, \dots, n$$

$$\boldsymbol{\beta}_j = (\beta_{j0}, \beta_{j1}, \dots, \beta_{jp})$$

$$\beta_{j0} = (1 - \sum_{l=1}^p \beta_{jl}) \mu_j, \quad \mu_j = E y_i, \quad t_{j-1} < i \leq t_j$$

## 2. Assumptions

- Stationary AR(p) in segments
- i.i.d. errors  $e_i$  with

$$E e_i = 0, \quad 0 < \text{Var}(e_i) = \sigma^2 < \infty, \quad E |e_i|^4 < \infty$$

- $t_j = \lfloor n \lambda_j \rfloor, \quad j = 1, \dots, k$   
 $0 = \lambda_0 < \lambda_1 < \dots < \lambda_{k+1} = 1.$

## 3. Estimation of change points for known $k$

- The minimal length of a segment is  $(n-p)\varepsilon$   
 $T_\varepsilon = \{(t_1, \dots, t_k) : t_{j+1} - t_j \geq n\varepsilon, \forall j = 0, \dots, k\}$
- SSR for a given partition  $(t_1, \dots, t_k)$

$$S_n(t_1, \dots, t_k) \equiv \sum_{j=1}^{k+1} \min_{\boldsymbol{\beta}_j} \sum_{i=t_{j-1}+1}^{t_j} (y_i - \mathbf{x}'_i \boldsymbol{\beta}_j)^2$$

- The change points are estimated as ([2, 7])

$$(\hat{t}_1, \dots, \hat{t}_k) = \arg \min_{t_1, \dots, t_k \in T_\varepsilon} S_n(t_1, \dots, t_k)$$

## 4. F type test [1]

No change versus  $k$  changes ( $k$  fixed)

$$F_n^\varepsilon(k, q) = \max_{t_1, \dots, t_k \in T_\varepsilon} F_n(t_1, \dots, t_k, q)$$

$$F_n(t_1, \dots, t_k, q) = \frac{1}{kq} \frac{SSR_0 - SSR_k}{\tilde{\sigma}_n^2},$$

where

$$\tilde{\sigma}_n^2 = \frac{SSR_k}{n-q} \rightarrow_p \sigma^2 \quad \text{under } H_0 \text{ and } H_A$$

$$SSR_0 = \min_{\boldsymbol{\beta}} \sum_{i=p+1}^n (y_i - \mathbf{x}'_i \boldsymbol{\beta})^2, \quad SSR_k = S_n(t_1, \dots, t_k)$$

$q$  - number of regressors

## 5. Equivalent expression of the test

$$\hat{e}_i = y_i - \mathbf{x}'_i \mathbf{C}_n^{-1} \sum_{i=p+1}^n \mathbf{x}_i y_i \stackrel{H_0}{=} e_i - \mathbf{x}'_i \mathbf{C}_n^{-1} \sum_{i=p+1}^n \mathbf{x}_i e_i$$

$$\mathbf{C}_{k,l} = \sum_{i=k+1}^l \mathbf{x}_i \mathbf{x}'_i, \quad \mathbf{C}_{p,l} = \mathbf{C}_l, \quad k+1, l = p+1, \dots, n$$

$$\begin{aligned} & SSR_0 - SSR_k \\ &= \sum_{j=1}^{k+1} \left( \sum_{i=t_{j-1}+1}^{t_j} \mathbf{x}_i \hat{e}_i \right)' \mathbf{C}_{t_{j-1}, t_j}^{-1} \left( \sum_{i=t_{j-1}+1}^{t_j} \mathbf{x}_i \hat{e}_i \right) \\ & \stackrel{H_0}{=} \left\{ - \left( \sum_{i=p+1}^n \mathbf{x}_i e_i \right)' \mathbf{C}_n^{-1} \left( \sum_{i=p+1}^n \mathbf{x}_i e_i \right) \right. \\ & \quad \left. + \sum_{j=1}^{k+1} \left( \sum_{i=t_{j-1}+1}^{t_j} \mathbf{x}_i e_i \right)' \mathbf{C}_{t_{j-1}, t_j}^{-1} \left( \sum_{i=t_{j-1}+1}^{t_j} \mathbf{x}_i e_i \right) \right\} \end{aligned} \quad (1)$$

## 6. Limit distribution under $H_0$

Using results from [3], one can show

$$F_n^\varepsilon(k, q) \rightarrow_d F^\varepsilon(k, q) = \sup_{\lambda_1, \dots, \lambda_k \in \Lambda_\varepsilon} F(\lambda_1, \dots, \lambda_k, q),$$

where

$$\begin{aligned} & F(\lambda_1, \dots, \lambda_k, q) \\ &= \frac{1}{kq} \sum_{j=1}^k \frac{\|\lambda_j \mathbf{W}(\lambda_{j+1}) - \lambda_{j+1} \mathbf{W}(\lambda_j)\|^2}{\lambda_j \lambda_{j+1} (\lambda_{j+1} - \lambda_j)}, \end{aligned}$$

$\mathbf{W}(t)$  is a vector of  $q$  independent standard Wiener processes and

$$\Lambda_\varepsilon = \{(\lambda_1, \dots, \lambda_k) : \lambda_{j+1} - \lambda_j \geq \varepsilon, \forall j = 0, \dots, k\}.$$

## 7. Bootstrapping

- $\tilde{e}_{p+1}^*, \dots, \tilde{e}_n^*$  bootstrap sample with replacement from  $\tilde{e}_{p+1}, \dots, \tilde{e}_n$

$$\tilde{e}_i = y_i - \mathbf{x}'_i \mathbf{C}_{\hat{t}_{j-1}, \hat{t}_j}^{-1} \sum_{i=\hat{t}_{j-1}+1}^{\hat{t}_j} \mathbf{x}_i y_i, \quad \hat{t}_{j-1} < i \leq \hat{t}_j$$

- Bootstrap version of  $F_n^\varepsilon(k, q)$

$$F_n^{*\varepsilon}(k, q) = \max_{t_1, \dots, t_k \in T_\varepsilon} F_n^*(t_1, \dots, t_k, q)$$

$$F_n^*(t_1, \dots, t_k, q) = \frac{1}{kq} \frac{SSR_0^* - SSR_k^*}{\tilde{\sigma}_n^2},$$

$SSR_0^* - SSR_k^*$  as in (1) with  $e_i$  replaced by  $\tilde{e}_i^*$ .

## 8. Theorem

Let the data follow  $H_0$  or alternatives. Then, under given assumptions,

$$\sup_x \{P(F_n^{*\varepsilon}(k, q) \leq x | \mathbf{y}) - P(F^\varepsilon(k, q) \leq x)\} \rightarrow_p 0$$

Proof - using key results from [4].

## 9. Simulation results [6, 7]

$$y_i = x_i \beta_j + e_i, \quad t_{j-1} < i \leq t_j, \quad j = 1, \dots, m+1$$

$$x_i = y_{i-1}$$

Graphs - inspiration from [5]

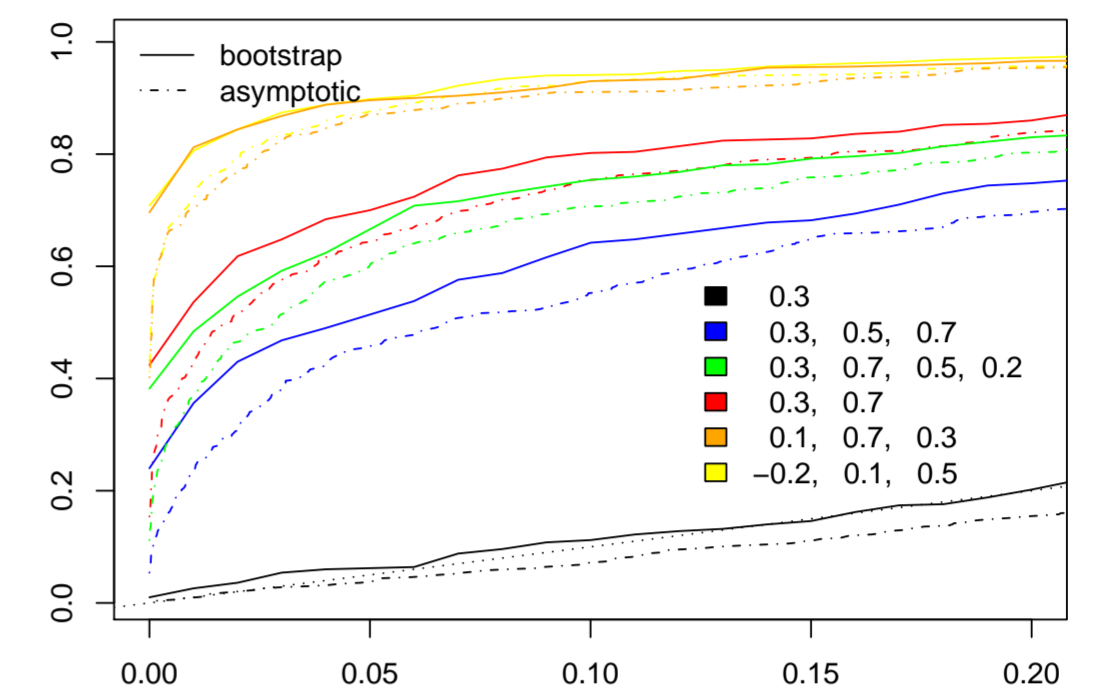


Figure 1: Size-power-curves plots (SPC-plots) showing empirical distribution function of  $p$ -values of the test  $F_n^\varepsilon(k, q)$  for the null hypothesis or some alternative with respect to the bootstrap distribution which was used to determine the critical values of the test. On the  $y$ -axis actual  $\alpha$ -errors or  $1 - \beta$ -errors for chosen quantiles on the  $x$ -axis. Here  $n = 200, k = 2, q = 1, \varepsilon = 0.15$ . For comparison  $p$ -values with respect to the asymptotic distribution  $F^\varepsilon(k, q)$  calculated (dotted lines).

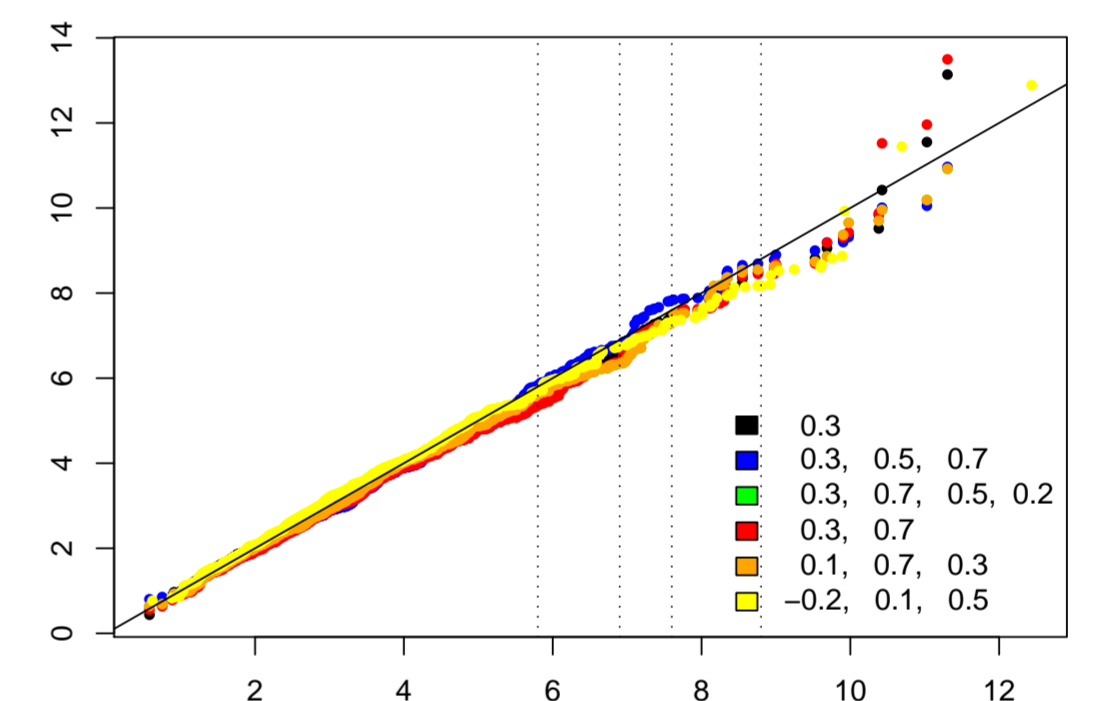


Figure 2: QQ-plot of empirical conditional quantiles of  $F_n^{*\varepsilon}(k, q)$  (1000 bootstrap samples), given one realisation  $\mathbf{y}$ , against empirical quantiles of  $F_n^\varepsilon(k, q)$  under  $H_0$  (1000 repetitions). Dotted lines are 90, 95, 97.5, 99% empirical quantiles of  $F_n^\varepsilon(k, q)$  under  $H_0$ . Here  $n = 200, k = 2, q = 1, \varepsilon = 0.15$ .

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