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Functional central limit theorems for nearly nonstationary processes and applications for testing epidemic change

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AR(1) process

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First order autoregressive AR(1) process is generated according to the scheme

$$y_k = \phi y_{k-1} + \varepsilon_k, \quad k \ge 1, \tag{1}$$

where  $(\varepsilon_k)$  are the innovations at time k, and  $\phi$  is an unknown parameter.

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where  $(\varepsilon_k)$  are the innovations at time k, and  $\phi$  is an unknown parameter. **(**)  $|\phi| < 1$ , then (1) is stationary process

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- **(**)  $|\phi| < 1$ , then (1) is stationary process;
- $|\phi| > 1$ , then (1) is explosive process

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AR(1) process

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where  $(\varepsilon_k)$  are the innovations at time k, and  $\phi$  is an unknown parameter.

- **(**)  $|\phi| < 1$ , then (1) is stationary process;
- 2  $|\phi| > 1$ , then (1) is explosive process;
- $\phi = 1$ , then (1) is nonstationary process.

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## $\phi$ is "close" to 1

• For  $|\phi| \leq 1$  the asymptotics :

$$\left(\sum_{k=1}^{n} y_{k-1}^{2}\right)^{1/2} \left(\widehat{\phi} - \phi\right) \xrightarrow[n \to \infty]{\mathbb{R}} \mathfrak{N}(0, 1).$$
(2)

**2** For  $\phi = 1$ 

$$\left(\sum_{k=1}^{n} y_{k-1}^{2}\right)^{1/2} \left(\widehat{\phi} - 1\right) \xrightarrow[n \to \infty]{\mathbb{R}} \frac{\frac{1}{2}(W^{2}(1) - 1)}{\left(\int_{0}^{1} W^{2}(t) \,\mathrm{d}t\right)^{1/2}}$$
(3)

Because of P(τ ≤ 0) = P(W<sup>2</sup>(1) ≤ 1) = 0.684, (2) may not be a satisfactory approximation when φ is "close" to 1 and the sample size is moderate. Also (3) could be used. However, neither (2) nor (3) seems to be intuitive approximations.

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## Nearly nonstationary process

Suppose we have first-order autoregressive process  $(y_{n,k})$  given by

$$y_{n,k} = \phi_n y_{n,k-1} + \varepsilon_k, \quad k \ge 1, \quad n \ge 1,$$
(4)

where

$$\begin{array}{l} \bullet & \phi_n \to 1, \text{ as } n \to \infty, \\ \bullet & (\varepsilon_k) \text{ is a sequence of i.i.d.random variables with } \mathbb{E}\varepsilon_k = 0 \text{ and } \\ \mathbb{E}\varepsilon_k^2 = 1, \end{array}$$

•  $y_{n,1} \dots, y_{n,n}$  are observations and *n* is a sample size,

• for simplicity 
$$y_{n,0} = y_0 = 0$$
.

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AR(1) process definition Definition of nearly nonstationary process

# Parametrisation of $\phi_n$

## Case 1

 $\phi_n = e^{\gamma/n}$  with constant  $\gamma < 0$ . This parametrisation was suggested by Phillips (1987 m.).

## Case 2

 $\phi_n = 1 - \frac{\gamma_n}{n}$ ,  $\gamma_n \to \infty$  slower than *n*. This parametrisation was suggested by Phillips and Giraitis (2006 m.)

We will use LSE estimate

$$\widehat{\phi}_n = \frac{\sum_{k=1}^n y_k y_{k-1}}{\sum_{k=1}^n y_{k-1}^2}$$

based on observations  $y_1, \ldots, y_n$  with  $y_0 = 0$  for convenience.

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# Process built on the $y_k$ 's

We focus on polygonal line processes built on the  $y_k$ 's :

$$S_n^{\mathrm{pl}}(t) := \sum_{k=1}^{[nt]} y_{k-1} + (nt - [nt]) y_{[nt]}, \quad t \in [0,1], \quad n \ge 1.$$

and on polygonal line process built on the  $\widehat{\varepsilon}_k{\,}'{\rm s}$  :

$$\widehat{W}_n^{\mathrm{pl}}(t) := \sum_{k=1}^{[nt]} \widehat{\varepsilon}_k + (nt - [nt])\widehat{\varepsilon}_{[nt]+1}, \quad t \in [0,1], \quad n \ge 1,$$

where  $\widehat{\varepsilon}_k$  are residuals of the process  $y_k$  defined by

$$\widehat{\varepsilon}_k = y_k - \widehat{\phi}_n y_{k-1}$$

# Function spaces

The polygonal line process  $S_n^{\rm pl}$  can be viewed as a random element in Hölder space  ${\rm H}^o_{\alpha}[0,1]$ . For  $\alpha \in (0,1)$ 

$$\mathrm{H}^{o}_{lpha}[0,1]:=\left\{f\in \mathcal{C}[0,1]:\lim_{\delta
ightarrow0}\omega_{lpha}(f,\delta)=0
ight\},$$

endowed with the norm  $\|f\|_{lpha}:=|f(0)|+\omega(f,1)$ , where

$$\omega_{lpha}(f,\delta):=\sup_{\substack{s,t\in[0,1]\0< t-s<\delta}}rac{|f(t)-f(s)|}{|t-s|^{lpha}}$$

is a separable Banach space.

FCLT for sums of process : Case 1 FCLT for sums of process : Case 2 FCLT for residuals of the process : Case 1 FCLT for residuals of the process : Case 2

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# Convergence in $H^o_{\alpha}[0, 1]$ spaces

## Theorem

Suppose that  $(y_k)$  is generated by (4),  $\phi_n = e^{\gamma/n}$  with  $\gamma < 0$  and that the sequence of polygonal lines  $n^{-1/2}W_n^{\rm pl}$  converges weakly to the standard Brownian motion W in  $\mathrm{H}^o_{\alpha}[0,1]$  for some  $0 < \alpha < 1/2$ . Then  $n^{-3/2}S_n^{\rm pl}$  converges weakly in the space under consideration to the integrated Ornstein-Uhlenbeck process J defined by :

$$J(t) := \int_0^t U_\gamma(s) \,\mathrm{d}s, \quad 0 \le t \le 1,$$
 (5)

where  $U_{\gamma}(s) = \int_0^s e^{\gamma(s-r)} dW(r)$ .

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# Convergence in $H^0_\beta[0,1]$ space

#### Theorem

Suppose  $(y_k)$  is generated by (4) and  $\phi_n = 1 - \gamma_n/n$ , where  $(\gamma_n)$  is a sequence of non negative numbers,  $\gamma_n \to \infty$  and  $\gamma_n/n \to 0$  as  $n \to \infty$ . Assume also that the innovations  $(\varepsilon_k)$  are i.i.d. and satisfy condition  $\lim_{t\to\infty} t^p P(|\varepsilon_0| > t) = 0$  for some p > 2. Put  $\alpha = \frac{1}{2} - \frac{1}{p}$ . Then for  $0 < \beta < \alpha$ ,

$$n^{-1/2}(1-\phi_n)S_n^{\mathrm{pl}} \xrightarrow[n \to \infty]{\mathrm{H}^0_\beta[0,1]} W,$$

provided that

$$\liminf_{n\to\infty}\gamma_n n^{-\frac{\beta}{\alpha}}>0.$$

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# Convergence in $H^o_{\alpha}[0, 1]$

## Theorem

Let  $\alpha \in (0, 1/2)$ . Suppose that  $(y_k)$  is generated by (4),  $\phi_n = e^{\gamma/n}$ . Also  $(\varepsilon_k)$  are independent identically distributed random variables with  $\mathbb{E}\varepsilon_0 = 0$ . Then

$$n^{-1/2}\widehat{W}_n^{\mathrm{pl}} \xrightarrow[n \to \infty]{} W - A^{-1}B'J, \qquad (6)$$

if and only if condition

$$\lim_{t\to\infty} t^{1/(1/2-\alpha)} P(|\varepsilon_1| \ge t) = 0.$$
(7)

holds. Here  $B' = \int_0^1 U_\gamma(r) dW(r)$ ,  $A = \int_0^1 U_\gamma(r)^2 dr$  and  $J(t) := \int_0^t U_\gamma(s) ds$ .

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FCLT for sums of process : Case 1 FCLT for sums of process : Case 2 FCLT for residuals of the process : Case 1 FCLT for residuals of the process : Case 2

# Convergence in $H^o_{\alpha}[0, 1]$

## Theorem

Suppose  $(y_k)$  is generated by (4) and  $\phi_n = 1 - \gamma_n/n$ , where  $\gamma_n$  is a sequence of non negative constants,  $\gamma_n \to \infty$  and  $\gamma_n/n \to 0$  as  $n \to \infty$ . Assume also that the innovations  $(\varepsilon_k)$  are i.i.d. and satisfy condition

$$\lim_{t \to \infty} t^{\rho} \mathbb{P}(|\varepsilon_0| > t) = 0$$
(8)

for some p > 2. Put  $\alpha = \frac{1}{2} - \frac{1}{p}$ . Then for  $0 < \beta \le \alpha$ ,

$$n^{-1/2}\widehat{W}_{n}^{\mathrm{pl}} \xrightarrow[n \to \infty]{\mathrm{H}_{\beta}^{0}[0,1]} W, \qquad (9)$$

where W is a standard Wiener process if

$$\liminf_{n \to \infty} \gamma_n n^{-\frac{2\beta}{1+2\alpha}} > 0. \tag{10}$$

Test statistics Testing with  $\gamma_k$ 's Testing with  $\varepsilon_k$ 's

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## Epidemic change in mean

## Hypothesis for $y_k$ 's

$$z_{n,k} = a_n \mathbf{1}_{\{k^* < k \le k^* + l^*\}} + y_{n,k}$$
  

$$H_0 : a_n = 0;$$
  

$$H_A : a_n \neq 0.$$

## Hypothesis for $\varepsilon_k$ 's

$$\begin{aligned} &H_0: \mathbb{E}\varepsilon_k = 0; \\ &H_A: \mathbb{E}\varepsilon_k = a \mathbf{1}_{\{k^* < k \le k^* + l^*\}}. \end{aligned}$$



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Test statistics Testing with  $\gamma_k$ 's Testing with  $\varepsilon_k$ 's

## Test statistics with $y_k$ 's

We construct uniform increment statistics :

$$UI_{\mathsf{S}}(n,\alpha) := \sup_{k^*,l^*} \frac{\left| S_n^{\mathrm{st}}(\mathbb{I}_n^*) - \frac{l^*}{n} S_n^{\mathrm{st}}(n) \right|}{\left| \frac{l^*}{n} \left( 1 - \frac{l^*}{n} \right) \right|^{\alpha}}$$

where

$$\mathbb{I}_{n}^{*} = \{k^{*} + 1, \dots, k^{*} + l^{*}\}$$
$$S_{n}^{\mathrm{st}}(\mathbb{I}_{n}^{*}) = \sum_{k \in \mathbb{I}_{n}^{*}} y_{n,k-1}$$
$$S_{n}^{\mathrm{st}}(n) = \sum_{k=1}^{n} y_{n,k-1}$$

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Test statistics Testing with  $\mathcal{F}_{k}$ 's Testing with  $\varepsilon_{k}$ 's

# Behaviour under null hypothesis in case 1

#### Theorem

Suppose that  $(y_k)$  is generated by (4),  $\phi_n = e^{\gamma/n}$  with  $\gamma < 0$  and that the sequence  $(\varepsilon_k)$  are i.i.d. random variables with mean 0 and satisfy condition  $\lim_{t\to\infty} t^{1/(1/2-\alpha)}P(|\varepsilon_1| \ge t) = 0$ . Then under hypothesis  $H_0$ 

$$n^{-3/2}UI_{\mathcal{S}}(n,\alpha) \xrightarrow[n \to \infty]{\mathbb{R}} UI_{\mathcal{S},1}(\alpha).$$
 (11)

Here

$$UI_{S,1}(\alpha) = \sup_{t,s} \frac{|J(t) - J(s) - (t-s)J(1)|}{|(t-s)(1-(t-s))|^{\alpha}}$$

where  $J(t) = \int_0^t U_{\gamma}(r) dr$  and  $U_{\gamma}(r) = \int_0^s e^{\gamma(r-\nu)} dW(\nu)$ .

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Test statistics Testing with  $\gamma_k$ 's Testing with  $\varepsilon_k$ 's

## Consistency of the test in case 1

## Theorem

Let  $0 < \alpha < 1/2$ . Suppose  $(y_k)$  is generated by (4) and  $\phi_n = e^{\gamma/n}$ , where  $\gamma < 0$  is a constant. Under alternative  $H_A$ , if

$$\lim_{n \to \infty} |a_n| h_n^{1-\alpha} n^{-1/2} = \infty, \quad \text{where} \quad h_n := \frac{l^*}{n} \left( 1 - \frac{l^*}{n} \right)$$

then

$$n^{-3/2}UI_{\mathcal{S}}(n,\alpha) \xrightarrow[n \to \infty]{P} \infty.$$

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# Behaviour under null hypothesis in case 2

## Theorem

Suppose  $(y_k)$  is generated by (4) and  $\phi_n = 1 - \gamma_n/n$ , where  $(\gamma_n)$  is a sequence of non negative numbers,  $\gamma_n \to \infty$  and  $\gamma_n/n \to 0$ , as  $n \to \infty$ . Assume also that the innovations  $(\varepsilon_k)$  are i.i.d. and satisfy condition  $\lim_{t\to\infty} t^p P(|\varepsilon_0| > t) = 0$  for some p > 2. Put  $\alpha = \frac{1}{2} - \frac{1}{p}$ . Also for  $0 < \beta < \alpha$  condition  $\liminf_{n \to \infty} \gamma_n n^{-\frac{\beta}{\alpha}} > 0$  holds. Then under hypothesis H<sub>0</sub>

$$n^{-1/2}(1-\phi_n)UI_S(n,\alpha) \xrightarrow[n\to\infty]{\mathbb{R}} UI_2(\alpha),$$

Here

$$UI_{2}(\alpha) = \sup_{t,s} \frac{|B(t) - B(s)|}{|(t-s)(1-(t-s))|^{\alpha}}$$

where B(t) = W(t) - tW(1) is Brownian bridge.

Test statistics Testing with  $\mathscr{P}_{k}$ 's Testing with  $\varepsilon_{k}$ 's

## Consistency of the test in case 2

#### Theorem

Let  $0 < \alpha < 1/2$ . Suppose  $(y_k)$  is generated by (4) and  $\phi_n = 1 - \gamma_n/n$ , where  $(\gamma_n)$  is a sequence of non negative numbers,  $\gamma_n \to \infty$  and  $\gamma_n/n \to 0$ , as  $n \to \infty$ . Under alternative  $H_A$ , if

$$\lim_{n \to \infty} |a_n| h_n^{1-\alpha} n^{-1/2} \gamma_n = \infty, \quad \text{where} \quad h_n := \frac{l^*}{n} \left( 1 - \frac{l^*}{n} \right)$$

then

$$(1-\phi_n)n^{-1/2}UI_S(n,\alpha) \xrightarrow[n\to\infty]{\mathrm{P}} \infty.$$

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Testing with  $\varepsilon_k$ 's

# Test statistics with $\hat{\varepsilon}_k$ 's

We construct uniform increment statistics :

$$UI_{W}(n,\alpha) := \sup_{k^{*},l^{*}} \frac{\left|\widehat{W}_{n}^{\mathrm{st}}(\mathbb{I}_{n}^{*}) - \frac{l^{*}}{n}\widehat{W}_{n}^{\mathrm{st}}(n)\right|}{\left|\frac{l^{*}}{n}\left(1 - \frac{l^{*}}{n}\right)\right|^{\alpha}}$$

where

$$\mathbb{I}_n^* = \{k^* + 1, \dots, k^* + l^*\}$$
$$\widehat{W}_n^{\mathrm{st}}(\mathbb{I}_n^*) = \sum_{k \in \mathbb{I}_n^*} \widehat{\varepsilon}_k$$
$$\widehat{W}_n^{\mathrm{st}}(n) = \sum_{k=1}^n \widehat{\varepsilon}_k$$

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Test statistics Testing with  $\hat{\mathcal{T}}_{k}$ 's Testing with  $\varepsilon_{k}$ 's

# Behaviour under null hypothesis in case 1

## Theorem

Suppose that  $(y_k)$  is generated by (4),  $\phi_n = e^{\gamma/n}$  with  $\gamma < 0$  and that the sequence  $(\varepsilon_k)$  are i.i.d. random variables with mean 0 and variance 1. Coefficient  $\hat{\phi}_n$  is estimated by LSE. Then under hypothesis  $H_0$ 

$$n^{-1/2}UI_W(n,\alpha) \xrightarrow[n \to \infty]{\mathbb{R}} UI_{W,1}(\alpha).$$
 (12)

Here

$$UI_{W,1}(\alpha) = \sup_{t,s} \frac{|B(t) - B(s) - A^{-1}B'(J(t) - J(s) - (t-s)J(1))|}{|(t-s)(1-(t-s))|^{\alpha}}$$

Here  $B' = \int_0^1 U_{\gamma}(r) dW(r)$ ,  $A = \int_0^1 U_{\gamma}(r)^2 dr$  and  $J(t) := \int_0^t U_{\gamma}(s) ds$ , B(t) is a Brownian bridge.

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Testing with  $\varepsilon_k$ 's

## Consistency of the test in case 1

## Theorem

Let  $0 < \alpha < 1/2$ . Suppose  $(y_k)$  is generated by (4) and  $\phi_n = e^{\gamma/n}$ , where  $\gamma < 0$  is a constant. Under alternative H<sub>A</sub>, if

$$\lim_{n \to \infty} |a| h_n^{1-\alpha} n^{1/2} = \infty, \quad \text{where} \quad h_n := \frac{l^*}{n} \left( 1 - \frac{l^*}{n} \right)$$

then

$$n^{-1/2}UI_W(n,\alpha) \xrightarrow[n \to \infty]{\mathbb{P}} \infty.$$

Test statistics Testing with  $\hat{\mathcal{T}}_{k}$ 's Testing with  $\varepsilon_{k}$ 's

# Behaviour under null hypothesis in case 2

## Theorem

Suppose  $(y_k)$  is generated by (4) and  $\phi_n = 1 - \gamma_n/n$ , where  $(\gamma_n)$  is a sequence of non negative numbers,  $\gamma_n \to \infty$  and  $\gamma_n/n \to 0$ , as  $n \to \infty$ . Assume also that the innovations  $(\varepsilon_k)$  are i.i.d. and satisfy condition  $\lim_{t\to\infty} t^p P(|\varepsilon_0| > t) = 0$  for some p > 2. Put  $\alpha = \frac{1}{2} - \frac{1}{p}$ . Also for  $0 < \beta \le \alpha$  condition  $\lim_{n\to\infty} \gamma_n n^{-\frac{2\beta}{1+2\alpha}} > 0$  holds. Then under hypothesis  $H_0$ 

$$n^{-1/2}UI_W(n,\alpha) \xrightarrow[n \to \infty]{\mathbb{R}} UI_2(\alpha),$$

Here

$$UI_{2}(\alpha) = \sup_{t,s} \frac{|B(t) - B(s)|}{|(t-s)(1-(t-s))|^{\alpha}}$$

where B(t) = W(t) - tW(1) is Brownian bridge.

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## Consistency of the test in case 2

#### Theorem

Let  $0 < \alpha < 1/2$ . Suppose  $(y_k)$  is generated by (4) and  $\phi_n = 1 - \gamma_n/n$ , where  $(\gamma_n)$  is a sequence of non negative numbers,  $\gamma_n \to \infty$  and  $\gamma_n/n \to 0$ , as  $n \to \infty$ . Under alternative  $H_A$ , if

$$\lim_{n \to \infty} |a| h_n^{1-\alpha} n^{1/2} = \infty, \quad \text{where} \quad h_n := \frac{l^*}{n} \left( 1 - \frac{l^*}{n} \right)$$

then

$$n^{-1/2}UI_W(n,\alpha) \xrightarrow[n \to \infty]{P} \infty.$$

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