Local Power of Fixed-T Panel Unit Root Tests with Serially Correlated Errors and Incidental Trends

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Motivation

• Panel data unit root tests are attractive because they are more powerful

• However, they have their own complications (see e.g. Moon et al (2007))

$$y_{it} = a_i + \beta_i t + \zeta_{it},$$

$$\zeta_{it} = \varphi \zeta_{it-1} + u_{it}$$

where

$$\varphi = 1 - rac{1}{T\sqrt{N}}$$



Research Question

• In this paper we examine the problem for fixed T tests

• Short term serial correlation is also a major factor

 Previous work in the area: Bond et al. (2005), Kruiniger (2008,2009), Madsen (2010), Westerlund(2014)

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Implementation

• Two tests are relative in this framework

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IV test (De Wachter et al. (2007) intercepts trends WG test (Kruiniger and Tzavalis (2002) intercepts trends
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 Derive local power functions of the IV test for serial correlation and incidental trends

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Contributions

- 1 The IV test is dominates WG test in terms of power
- The effects of serial correlation depend on the type of tests and on the deterministic specification
- The presence of serial correlation does not necessarily mean loss of power
- The incidental trends problem may not exist in the presence of serial correlation
- The IV test does not suffer from the incidental trends problem

Incidental Intercepts

• Consider the AR(1) model:

$$y_{it} = a_i + \zeta_{it}$$

 $\zeta_{it} = \varphi \zeta_{it-1} + u_{it}, \qquad i = 1, ..., N \text{ and } t = 1, ..., T$

Incidental Intercepts

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$$egin{array}{lcl} y_{it} &=& a_i + \zeta_{it} \ & \zeta_{it} &=& \varphi \zeta_{it-1} + u_{it}, \end{array} \qquad i=1,...,N ext{ and } t=1,...,T \end{array}$$

• Stacked over i

$$y_i = a_i e + \zeta_i,$$

 $\zeta_i = \varphi \zeta_{i-1} + u_i,$

$$y_i = (y_{i1},...,y_{iT})', \ e = (1,...,1)', \ \zeta_i = (\zeta_{i1},...,\zeta_{iT})', \ u_i = (u_{i1},...,u_{iT})'$$

 $\zeta_{i-1} = (\zeta_{i0},...,\zeta_{iT-1})'$ and also $y_{i-1} = (y_{i0},...,y_{iT-1})'$

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Incidental Intercepts

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Hypothesis of interest

$$arphi_{N}=1-rac{c}{\sqrt{N}}$$

$$H_0$$
 : $c = 0$
 H_1 : $c > 0$

Individual Intercepts - Assumption

- $\{u_{it}\}$ have $E(u_{it}) = 0$, and are independent and homogeneous across i.
- $\{u_{it}\}$ are serially correlated across time but $E(u_{i1}u_{iT})=0$.
- The u_{it} , are independent of a_i and y_{i0} and all variables have bounded $4 + \varepsilon$ moments.

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Individual Intercepts - The Estimators

• The WG estimator

$$\hat{\varphi}_{WG} = \frac{\sum_{i=1}^{N} y'_{i-1} Q y_i}{\sum_{i=1}^{N} y'_{i-1} Q y_{i-1}}$$

where $Q = I_T - e(e'e)^{-1}e'$.

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Individual Intercepts - The Estimators

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• The IV estimator

$$E\left[\sum_{t=1}^{T-
ho-1}y_{it}u_{i,t+
ho+1}(arphi)
ight]=0, \quad ext{or}$$
 $E(y_{i-1}'\Pi_
ho u_i)=0$

thus

$$\hat{arphi}_{IV} = rac{\sum\limits_{i=1}^{N} y_{i-1}' \Pi_{
ho} y_{i}}{\sum\limits_{i=1}^{N} y_{i-1}' \Pi_{
ho} y_{i-1}}$$



Individual Intercepts - Tests

• **Theorem 1:** Under Assumption A and as $N \to \infty$

$$\sqrt{N}\hat{V}_{WG}^{-\frac{1}{2}}\hat{\delta}(\hat{\varphi}_{WG}-1-\frac{\hat{b}}{\hat{\delta}}) \stackrel{d}{\longrightarrow} N(-ck_{WG},1)$$

where

$$\frac{\hat{b}}{\hat{\delta}} = \frac{tr(\Psi \hat{\Gamma})}{\frac{1}{N} \sum_{i=1}^{N} y'_{i,-1} Q y_{i,-1}}, \quad V_{WG} = 2tr((A_{WG} \Gamma)^2),$$

and
$$A_{WG} = rac{1}{2}(\Lambda'Q + Q\Lambda - \Psi - \Psi')$$
 $\hat{\Gamma} = rac{1}{N}\sum_{i=1}^N \Delta y_i \Delta y_i'$

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 $\hat{\Gamma} = rac{1}{N}\sum_{i=1}^{N}\Delta y_i \Delta y_i'$

• **Theorem 2:** Under Assumption A and as $N \to \infty$

$$\sqrt{N}(\hat{\varphi}_{IV}-1)\hat{V}_{IV}^{-\frac{1}{2}} \stackrel{d}{\longrightarrow} N(-ck_{IV},1)$$

where

$$V_{IV} = rac{2tr((A_{IV}\Gamma)^2)}{tr(\Lambda'\Pi_p\Lambda\Gamma)^2}, A_{IV} = rac{1}{2}(\Lambda'\Pi_p + \Pi'_p\Lambda)$$

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Individual Intercepts - General Local Power Functions

where

$$k_{WG} = \frac{tr(\Lambda'Q\Lambda\Gamma) + tr(F'Q\Gamma) - tr(\Psi\Lambda\Gamma) - tr(\Lambda'\Psi\Gamma)}{\sqrt{V_{WG}}}$$

and

$$k_{IV} = \frac{1}{\sqrt{V_{IV}}}$$

Then the asymptotic local power function is

$$\Phi(z_a + ck)$$

Individual Intercepts - Examples

$$\begin{array}{lll} \text{k} & \text{WG} & \text{IV} \\ p{=}0 & \frac{\sqrt{3}(T{-}1)}{\sqrt{T^2{-}2T - \frac{4}{T} + 5}} & \sqrt{\frac{1}{2}(T^2 - T)} \\ p{=}1 & \frac{\sqrt{3}(T^2{-}3T + 2)}{T\sqrt{T^2{-}6T - \frac{24}{T} + \frac{12}{T^2} + 17}} & \sqrt{\frac{T^2}{2} - \frac{3T}{2} + 1} \\ p{=}2 & \frac{\sqrt{3}(T^2{-}5T + 6)}{T\sqrt{T^2{-}10T - \frac{80}{T} + \frac{60}{T^2} + 41}} & \sqrt{\frac{T^2}{2} - \frac{5T}{2} + 3} \\ p{=}3 & \frac{\sqrt{3}(T^2{-}7T + 12)}{T\sqrt{T^2{-}14T - \frac{196}{T} + \frac{192}{T^2} + 77}} & \sqrt{\frac{T^2}{2} - \frac{7T}{2} + 6} \end{array}$$

and for MA(1) with parameter θ

$$k_{WG} = \frac{(T-2)(T\theta^2 - \theta^2 + 3T\theta - 7\theta + T - 1)}{2T\sqrt{R_{1,WG}\theta^4 + R_{2,WG}\theta^3 + R_{3,WG}\theta^2 + R_{2,WG}\theta + R_{1,WG}}}$$

$$k_{IV} = \frac{D_{1,IV}\theta^2 + D_{2,IV}\theta + D_{1,IV}}{\sqrt{R_{1,IV}\theta^4 + R_{2,IV}\theta^3 + R_{3,IV}\theta^2 + R_{2,IV}\theta + R_{1,IV}}}$$

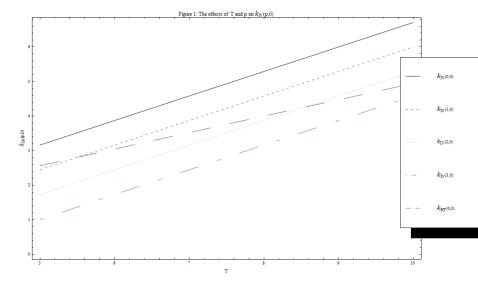
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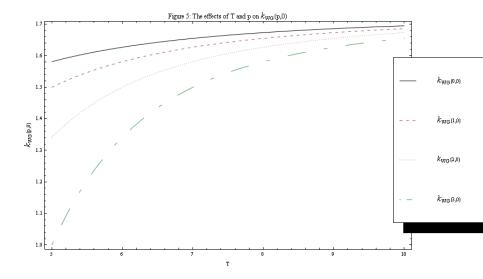
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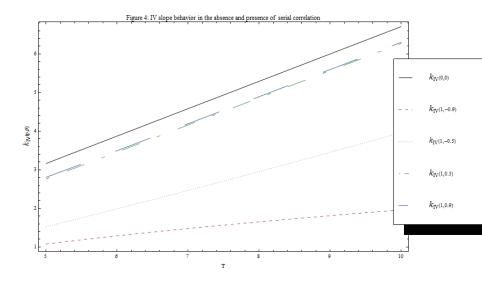
Individual Intercepts - Local Power Functions - IV



Individual intercepts - Local Power Functions - WG

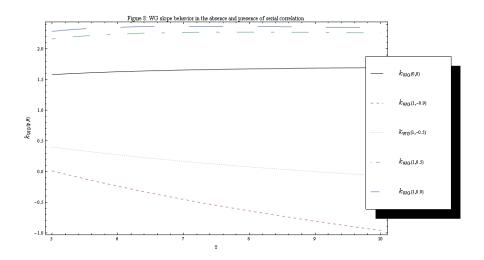


Individual Intercepts - Local Power Functions - IV



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Individual Intercepts - Local Power Functions - WG



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Incidental Trends - Estimators

• The incidental trends model is

$$y_i = a_i e + \beta_i \tau + \zeta_i,$$

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The WG estimator

$$\hat{\varphi}_{WG} = \left(\sum_{i=1}^{N} y'_{i-1} Q y_{i-1}\right)^{-1} \left(\sum_{i=1}^{N} y'_{i-1} Q y_{i}\right)$$

where $Q = I_T - X(X'X)^{-1}X'$ with $X = [e, \tau]$.

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where $Q = I_T - X(X'X)^{-1}X'$ with $X = [e, \tau]$.

• Taking first differences

$$E(y_{i-1}^{*\prime}\Pi_{p}^{*}u_{i}^{*})=0$$

where $u_i^* = \Delta u_i = (\Delta u_{i2},...,\Delta u_{iT})$ and $y_i^* = \Delta y_i = (\Delta y_{i2},...,\Delta y_{iT})$ which leads to

$$\hat{\varphi}_{IV} = \left(\sum_{i=1}^{N} y_{i-1}^{*'} \Pi_{\rho}^{*} y_{i-1}^{*}\right)^{-1} \left(\sum_{i=1}^{N} y_{i-1}^{*'} \Pi_{\rho}^{*} y_{i}^{*}\right)$$

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Incidental Trends - Tests

• **Theorem 3:** Under Assumption A and as $N \to \infty$

$$\sqrt{N} \hat{V}_{WG}^{-\frac{1}{2}} \hat{\delta}(\hat{\varphi}_{WG} - 1 - \frac{\hat{b}}{\hat{\delta}}) \rightarrow N(-ck_{WG}, 1)$$

where

$$\frac{\hat{b}}{\hat{\delta}} = \frac{tr(\Phi\hat{\Gamma})}{\frac{1}{N}\sum_{i=1}^{N}y'_{i,-1}Qy_{i,-1}} \text{ and } V_{WG} = vec(\Lambda'Q - \Phi')'\Theta vec(\Lambda'Q - \Phi')$$

and
$$\hat{\Theta} = \frac{1}{N} \sum_{i=1}^{N} \left(vec(\Delta y_i \Delta y_i') vec(\Delta y_i \Delta y_i')' \right)$$

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Incidental Trends - Tests

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• **Theorem 4:** Under Assumption A and as $N \to \infty$

$$\sqrt{N}(\hat{\varphi}_{IV}-1)\hat{V}_{IV}^{-\frac{1}{2}} \rightarrow N(-ck_{IV},1)$$

where
$$V_{IV} = \frac{2tr((A_{FDIV}\Theta)^2)}{tr(\Lambda^*\Pi_p^*\Lambda^*\Theta)^2}$$
, $A_{IV} = \frac{1}{2}(\Lambda^{*\prime}\Pi_p^* + \Pi_p^{*\prime}\Lambda^*)$ and $\Theta = 2\Gamma_1 - \Gamma_2 - \Gamma_2'$ where $\Gamma_1 = E(u_iu_i')$ and $\Gamma_2 = E(u_iu_{i-1}')$.

Individual Trends - Local Power Functions

where for the WG test

$$k_{WG} = \frac{tr(\Lambda'Q\Gamma) + tr(F'Q\Gamma) - tr(\Phi\Lambda\Gamma) - tr(\Lambda'\Phi\Gamma)}{2tr((A_{WGT}\Gamma)^2)}$$

and for the IV test

$$k_{IV} = \frac{1}{\sqrt{V_{IV}}}$$

Then

Table: Values of slope parameter *p*.

	•	•	•				
T=7							
-0.9	-0.5	0.0	0.5	0.9			
0.862	0.896	1.264 1.186		1.179			
0.694	0.466	0.00	-0.212	-0.248			
T=10							
-0.9 -0.5		0.0	0.5	0.9			
1.160 1.229		1.750	1.989	2.008			
1 040	0.645	0.00	0.216	-0.248			
	0.862 0.694 -0.9 1.160	0.862 0.896 0.694 0.466 -0.9 -0.5 1.160 1.229	-0.9 -0.5 0.0 0.862 0.896 1.264 0.694 0.466 0.00 T=10 -0.9 -0.5 0.0 1.160 1.229 1.750	-0.9 -0.5 0.0 0.5 0.862 0.896 1.264 1.186 0.694 0.466 0.00 -0.212 T=10 -0.9 -0.5 0.0 0.5			

Large T

If we assume T to be asymptotic the hypothesis of interest becomes

$$\varphi_{NT} = 1 - \frac{c}{T\sqrt{N}}$$

Theorem 5: If $T, N \to \infty$ jointly and the following condition holds: $\sqrt{N}/T \to 0$.

$$T\sqrt{N}(\sqrt{2})^{-1}(\hat{\varphi}_{IV}-1) \stackrel{d}{\longrightarrow} N(-c\frac{1}{\sqrt{2}},1),$$
 and

$$T\sqrt{N}(\sqrt{3})^{-1}\hat{\delta}_{WG}\left(\hat{\varphi}_{WG}-1-\frac{\hat{b}}{\hat{\delta}}\right)\stackrel{d}{\longrightarrow}N(-c0,1),$$

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Table 2: Slopes of large-T tests.

IV	MPP	LLC/HT	SGLS	IPS	WG
$1/\sqrt{2}$	$1/\sqrt{2}$	$(3/2)\sqrt{(5/51)}$	$1/\sqrt{3}$	0.282	0

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$$T\sqrt{N}(\sqrt{3})^{-1}\hat{\delta}_{WG}\left(\hat{\varphi}_{WG}-1-\frac{\hat{b}}{\hat{\delta}}\right)\stackrel{d}{\longrightarrow} N(-c0,1),$$

Table 2: Slopes of large-T tests.

IV	MPP	/	SGLS		WG
$1/\sqrt{2}$	$1/\sqrt{2}$	$(3/2)\sqrt{(5/51)}$	$1/\sqrt{3}$	0.282	0

• In the presence of incidental trends

$$k_{IV} = 0$$

$$k_{WG} = 0$$



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Heterogeneous Alternatives

For alternatives of the form

$$arphi_{Ni} = 1 - rac{c_i}{\sqrt{N}}$$

the hypothesis of interest is

$$H_0$$
: $c_i = 0$, for all i

$$H_1$$
: $c_i > 0$, for some i

where

• c_i are i.i.d. with support in a subset of a bounded interval $[0, M_c]$, for some $M_c \ge 0$.

Then, only the mean of c_i affects power, i.e.

$$\sqrt{N}(\hat{\varphi}_{IV}-1)\hat{V}_{IV}^{-\frac{1}{2}} \to N(-E(c_i)k_{IV},1)$$

Monte Carlo Simulations

Intercepts				Trend	ds				
N	50	200	1000	Theory	N	50	200	1000	Theory
$\theta = -0.5$				$\theta = -0.5$					
IV	0.285	0.444	0.567	0.793	IV	0.039	0.050	0.050	0.498
WG	0.057	0.068	0.087	0.069	WG	0.093	0.108	0.107	0.158
$\theta = 0$				heta=0					
IV	0.997	0.997	0.998	0.998	IV	0.050	0.051	0.048	0.760
WG	0.220	0.321	0.414	0.500	WG	0.169	0.122	0.089	0.050
heta=0.5				$\theta = 0$).5				
IV	0.979	0.988	0.993	0.994	IV	0.031	0.038	0.045	0.900
WG	0.388	0.57	0.678	0.730	WG	0.217	0.135	0.070	0.033

Conclusions

- The IV test is always better than the WG test and better than the HT test
- The effect of serial correlation is case specific
- Local power is possible in a fixed T therefore theoretically solving the incidental trends problem
- Power comes either from the presence of serial correlation or the use of double differences
- Non-trivial power **vanishes** when T is asymptotic

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