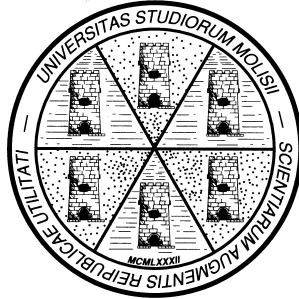


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**Change in persistence tests for panels:
an update and some new results**

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Change in persistence tests for panels: an update and some new results.*

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Abstract

In this paper we propose a set of new panel tests to detect changes in persistence. The test statistics are used to test the null hypothesis of stationarity against the alternative of a change in persistence from $I(0)$ to $I(1)$, from $I(1)$ to $I(0)$ and in an unknown direction. The limiting distributions of the panel tests are derived, and small sample properties are investigated by Monte Carlo experiments under the hypothesis that the individual series are independently cross-section distributed. These tests have a good size and power properties. Cross-sectional dependence is also considered. A procedure of de-factorizing, proposed by Stock and Watson (2002), is applied. The defactored panel tests have good size and power. The empirical results obtained from applying these tests to a panel covering 21 OECD countries observed between 1970 and 2007 suggest that inflation rate changes from $I(1)$ to $I(0)$ when cross-correlation is considered.

Keywords: Persistence, Stationarity, Panel data.

JEL Classification: C12, C23.

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1 Introduction

Recent time series literature has shown that economic and financial data are characterized by a change in persistence between separate $I(1)$ and $I(0)$ regimes rather than simply $I(1)$ or $I(0)$ behavior. For example, Cogley and Sargent (2001) and Emery (1994), using post World War II data, argued that persistence in U.S. inflation has decreased substantially since the early 1980s. Strikingly, Emery (1994) finds that U.S. inflation in the 1980s can best be described as a white noise. Further evidence of change in persistence from $I(1)$ to $I(0)$ behavior in U.S. inflation is also reported in Kim (2000), Busetti and Taylor (2004) and Leybourne et al. (2003). Other variables for which changes in persistence have been observed include real output (e.g Taylor, 2005) and short-term interest rates (e.g. Mankiw *et al.*, 1987)).

A number of testing procedures have been developed to test for changes in persistence. The most popular of these are the ratio-based change in persistence tests of Kim (2000), Kim *et al.* (2002), Busetti and Taylor (2004) and Harvey *et al.* (2006). These test the null hypothesis that a series is a constant $I(0)$ process against the alternative that it displays a change in persistence, either from $I(0)$ to $I(1)$, or viceversa. Kim (2000) and Kim *et al.* (2002) proposed a residual-based ratio test against changes in persistence in a time series, focusing on the case of a shift from $I(0)$ to $I(1)$ at some point in the sample. Kim (2000) also discussed the possibility of $I(1)$ to $I(0)$ shifts but did not provide tests against such an alternative. Busetti and Taylor (2004) proposed new ratio-based tests and breakpoint estimators which are consistent under $I(1)$ to $I(0)$ changes, and they demonstrated that the ratio-based tests which are consistent against changes from $I(1)$ to $I(0)$ are not consistent against changes from $I(0)$ to $I(1)$, and viceversa, with neither consistent against constant $I(1)$ processes. Harvey *et al.* (2006) developed a set of new tests which are based on a modified version of the ratio-base statistics of Kim (2000), Kim *et al.* (2002) and Busetti and Taylor (2004). These modifications use the variable addition approach of Vogelsang (1998), and a recent generalization by Sayginsoy (2003), and yielding tests which, by design, have the same critical values regardless of whether the process is $I(0)$ or (near) $I(1)$ throughout. This technique can only be used with the ratio based test of the null $I(0)$ because other tests of the $I(0)$ ($I(1)$) null are based on statistics which are divergent under constant $I(1)$ ($I(0)$) processes. Hence the null hypothesis is constant persistence (either a constant $I(0)$ process or a constant $I(1)$ process), and the alternative is a change in persistence. Using a panel framework, Costantini and Gutierrez (2007) proposed new recursive ADF unit root tests to detect changes in persistence based on the inverse normal Z test suggested by Choi (2001). The small sample properties of the recursive tests are investigated by Monte Carlo experiments. The panel tests have good size and power.

In this paper we propose a set of new panel tests to detect changes in persistence. The test statistics are used to test the null hypothesis of stationarity against the alternative of a change in persistence from $I(0)$ to $I(1)$, from $I(1)$ to $I(0)$ and an unknown direction. Two set of panel tests are proposed. The first set is based on the hypothesis that the individual series are independently

cross-section distributed. The second one uses the hypothesis of cross-sectional dependence.

The paper is organized as follows. In section 2 we present a set of new panel tests to detect a change in persistence under the hypothesis of cross-section independence. Section 3 describes the panel tests under the cross-section dependence hypothesis. Section 4 presents Monte Carlo simulations. In section 5 we apply the tests to analyze a panel of 21 OECD inflation rate series for the period 1970.1-2007.3 . Section 6 concludes. The main technical proofs and derivations are in the Appendix.

2 Persistence tests without cross-section correlation

2.1 The model

Consider the following Gaussian unobserved components model for a sample of N cross-sections observed over T time periods:

$$y_{i,t} = d_{i,t} + \mu_{i,t} + \varepsilon_{i,t}, \quad i = 1, \dots, N, t = 1, \dots, T, \quad (1)$$

we allow for the following three cases:

- Case 1: $I(0) \rightarrow I(1)$

$$\mu_{i,t} = \mu_{i,t-1} + 1(t > [\tau T])\eta_{i,t}, \quad i = 1, \dots, N, t = 1, \dots, T, \quad (2)$$

- Case 2: $I(1) \rightarrow I(0)$

$$\mu_{i,t} = \mu_{i,t-1} + 1(t \leq [\tau T])\eta_{i,t} \quad i = 1, \dots, N, t = 1, \dots, T. \quad (3)$$

- Case 3: unknown direction $I(0) \rightarrow I(1)$ or $I(1) \rightarrow I(0)$

where $1(\cdot)$ is the indicator function, $d_{i,t}$ is a deterministic component, $\varepsilon_{i,t}$ and $\eta_{i,t}$ are mutually independent mean zero iid gaussian processes with $\sigma_{\varepsilon_i}^2$ and $\sigma_{\eta_i}^2$ variance. The deterministic components are taken to be the unity vector.

From (2), it can be easily seen that for each cross section i , the data generating process yields a process which is stationary up to and including time $[\tau T]$, with the change-point proportion $\tau \in (0, 1)$, but is $I(1)$ after the break, if and only if, $\sigma_{\eta_i}^2 > 0$. Otherwise, from (3), we note that for each cross section i , the data generating process yields a process which is $I(1)$ up to and including time $[\tau T]$ but it is stationary after the break, if and only if, $\sigma_{\eta_i}^2 > 0$.

Therefore, the panel test of stationarity against a shift in persistence from stationarity to a unit root or viceversa can be framed in testing the null hypothesis as:

$$\mathbf{H}_0 = \sigma_{\eta_i}^2 = 0, \quad \forall i \quad (4)$$

against the alternative hypothesis

$$\mathbf{H}_1 = \sigma_{\eta_i}^2 > 0, \text{ at least for some } i. \quad (5)$$

When $I(0) \rightarrow I(1)$, we denote the alternative hypothesis as \mathbf{H}_{01} . If $I(1) \rightarrow I(0)$, then we use \mathbf{H}_{10} .

The following assumption plays a key role in the rest of the paper.

Assumption 1 *The process $\{\mu_{i,t}\}_{i,t=0}^{+\infty}$ is such that for each i*

1. $\mathbf{E}[\mu_i] = 0$;
2. $\mathbf{E}|\mu_i|^4 < +\infty$;
3. fixed i , then $\{\mu_{i,t}\}_{t=0}^{+\infty}$ is ϕ -mixing with mixing coefficients $\phi_{i,m}$ such that

$$\sum_{m=1}^{\infty} \phi_{i,m}^{\gamma_i} < +\infty,$$

for some $\gamma_i > 0$;

4. There exists the long-run variance

$$\sigma_{\mu i}^2 = \sum_{j=0}^{\infty} \mathbf{E}[\mu_{i,j+1} \mu'_{i,1}];$$

5. for each $s \in (0, 1)$, we have

$$\lim_{T \rightarrow \infty} \mathbf{V} \left[\frac{1}{\sqrt{T}} \sum_{t=1}^{[sT]} \mu_{i,t} \right] = s \sigma_{\mu i}^2$$

and

$$\lim_{T \rightarrow \infty} \mathbf{V} \left[\frac{1}{\sqrt{T}} \sum_{t=[sT]+1}^T \mu_{i,t} \right] = (1-s) \sigma_{\mu i}^2$$

The above conditions have been used by Phillips (1987), Phillips and Perron (1988) and Phillips and Solo (1992), among others, to prove results on the asymptotic distribution of a stochastic process. Finally, note that throughout the next sections we use sequential limits, where $T \rightarrow \infty$ is followed by $N \rightarrow \infty$.

2.2 Panel ratio-based tests: $I(0) \rightarrow I(1)$

In this section we present new panel tests to detect changes in persistence as in (2) and investigate their asymptotic behavior. We show that panel tests are standard normally distributed.

Consider the gaussian process (1)-(2). We want to test the null hypothesis \mathbf{H}_0 in (4) against \mathbf{H}_1 in (5). Let $\tilde{\varepsilon}_{i,t}$, $i = 1, \dots, N$ and $t = 1, \dots, T$, be the residuals from the regression of $y_{i,t}$ on intercept. If a structural change occurs at time $t = [\tau T]$ for $\tau \in (0, 1)$, the following partial sum process can be defined as:

$$\begin{cases} S_{i,t}^{(0)} = \sum_{j=1}^t \tilde{\varepsilon}_{i,j} & t = 1, \dots, [T\tau]; \quad i = 1, \dots, N, \\ S_{i,t}^{(1)} = \sum_{j=[T\tau]+1}^t \tilde{\varepsilon}_{i,j} & t = [T\tau] + 1, \dots, T; \quad i = 1, \dots, N, \end{cases} \quad (6)$$

Then, we consider the following test statistic:

$$\mathcal{K}_{T,N}(\tau) = \frac{\sqrt{N}}{\sigma} \cdot \left[\frac{(T - [T\tau])^{-2}}{[T\tau]^{-2}} \cdot \frac{1}{N} \cdot \sum_{i=1}^N \frac{\sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{\sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2} - \mu \right], \quad (7)$$

where

$$\mu = \frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T \mathbf{E}[S_{i,t}^{(1)}(\tau)^2]}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} \mathbf{E}[S_{i,t}^{(0)}(\tau)^2]} \quad (8)$$

and

$$\sigma = \sqrt{\frac{(T - [T\tau])^{-4} \sum_{t=[T\tau]+1}^T \mathbf{V}[S_{i,t}^{(1)}(\tau)^2]}{[T\tau]^{-4} \sum_{t=1}^{[T\tau]} \mathbf{V}[S_{i,t}^{(0)}(\tau)^2]}}. \quad (9)$$

Theorem 2.1 *Suppose that Assumption 1 is true under the null hypothesis \mathbf{H}_0 . Then*

$$\lim_{N \rightarrow +\infty} \lim_{T \rightarrow +\infty} \mathcal{K}_{T,N}(\tau) = \mathcal{K}(\tau) \sim N(0, 1). \quad (10)$$

In (7), the true value of τ is unknown. If the true change period is unknown three transformations of the tests $\mathcal{K}_{T,N}(\tau)$ defined in (7) for testing changes in persistence with unknown break point $[T\tau]$ can be considered:

- A maximum-Chow-type test, as used in Davies (1977), Hawkins (1987), Kim and Siegmund (1989), and Andrews (1993) is

$$\mathbf{H}_1(\mathcal{K}_{T,N}(\tau)) := \sup_{\tau \in (0,1)} \mathcal{K}_{T,N}(\tau). \quad (11)$$

- The mean score test proposed by Hansen (1991)

$$\mathbf{H}_2(\mathcal{K}_{T,N}(\tau)) := \int_{\tau \in (0,1)} \mathcal{K}_{T,N}(\tau) d\tau. \quad (12)$$

- The mean-exponential test introduced by Andrews and Ploberger (1994), i.e.

$$\mathbf{H}_3(\mathcal{K}_{T,N}(\tau)) := \log \left\{ \int_{\tau \in (0,1)} \exp[\mathcal{K}_{T,N}(\tau)] d\tau \right\}. \quad (13)$$

The asymptotic distribution of the tests defined in (11), (12) and (13) are given in the next result.

Theorem 2.2 *The following conditions hold true.*

(i) *It results*

$$\lim_{T \rightarrow +\infty} \lim_{N \rightarrow +\infty} \mathbf{H}_j(\mathcal{K}_{T,N}(\tau)) = \mathbf{H}_j(\mathcal{K}(\tau)), \quad j = 1, 2, 3.$$

(ii) *For each $j = 1, 2, 3$, we have $\mathbf{H}_j(\mathcal{K}(\tau)) \sim N(0, 1)$.*

2.3 Panel reverse test: $\mathbf{I}(1) \rightarrow \mathbf{I}(0)$

Consider the gaussian process (1)-(3). In this case, the null hypothesis refers to the stationary process and the alternative to a shift from $\mathbf{I}(1)$ to $\mathbf{I}(0)$. The following reverse test statistic is proposed:

$$\mathcal{K}_{T,N}^*(\tau) = \frac{\sqrt{N}}{\sigma} \cdot \left[\frac{[T\tau]^{-2}}{(T - [T\tau])^{-2}} \cdot \frac{1}{N} \cdot \sum_{i=1}^N \frac{\sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2}{\sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2} - \mu \right], \quad (14)$$

where

$$\mu = \mathbf{E} \left[\frac{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2}{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2} \right]$$

and

$$\sigma = \sqrt{\mathbf{V} \left[\frac{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2}{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2} \right]}.$$

The asymptotic distribution of the statistics defined in (14) is shown by the following result.

Theorem 2.3 *Suppose that Assumption 1 is true under the null hypothesis \mathbf{H}_0 . Then*

$$\lim_{N \rightarrow +\infty} \lim_{T \rightarrow +\infty} \mathcal{K}_{T,N}^*(\tau) = \mathcal{K}^*(\tau) \sim N(0, 1). \quad (15)$$

In the next result we give the asymptotic distributions of the transformations \mathbf{H}_1 , \mathbf{H}_2 and \mathbf{H}_3 of the test \mathcal{K}^* .

Theorem 2.4 *The following propositions hold.*

(i) *It results*

$$\lim_{T \rightarrow +\infty} \lim_{N \rightarrow +\infty} \mathbf{H}_j(\mathcal{K}_{T,N}^*(\hat{\tau})) = \mathbf{H}_j(\mathcal{K}^*(\tau)), \quad j = 1, 2, 3.$$

(ii) *For each $j = 1, 2, 3$, we have $\mathbf{H}_j(\mathcal{K}^*(\tau)) \sim N(0, 1)$.*

2.4 Panel tests with unknown direction

We now discuss the case of unknown direction of changes in persistence. Three panel tests are developed and their asymptotic distributions are derived. The tests are:

$$\mathcal{M}_{T,N}^{j,*} = \frac{\sqrt{N}}{\sigma_j^*} \cdot \frac{1}{N} \cdot \sum_{i=1}^N [\max\{\mathbf{H}_j(\tilde{\mathcal{K}}_{T,i}), \mathbf{H}_j(\tilde{\mathcal{K}}_{T,i}^*)\} - \mu_j^*], \quad j = 1, 2, 3; \quad (16)$$

where

$$\tilde{\mathcal{K}}_{T,i} = \frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2},$$

$$\tilde{\mathcal{K}}_{T,i}^* = (\tilde{\mathcal{K}}_{T,i})^{-1},$$

$$\begin{aligned}\mu_j^* &= \mathbf{E} \left[\max\{\mathbf{H}_j(\tilde{\mathcal{K}}_{T,i}), \mathbf{H}_j(\tilde{\mathcal{K}}_{T,i}^*)\} \right] \quad j = 1, 2, 3; \quad i = 1, \dots, N, \\ \sigma_j^* &= \sqrt{\mathbf{V} \left[\max\{\mathbf{H}_j(\tilde{\mathcal{K}}_{T,i}), \mathbf{H}_j(\tilde{\mathcal{K}}_{T,i}^*)\} \right]} \quad j = 1, 2, 3; \quad i = 1, \dots, N.\end{aligned}$$

The asymptotic distributions of these tests are now derived.

Theorem 2.5 *It results*

$$\lim_{N \rightarrow +\infty} \lim_{T \rightarrow +\infty} \mathcal{M}_{T,N}^{j,*} \sim N(0, 1).$$

2.5 Modified panel tests

In this section we propose panel tests that are based on the modified version of the test statistics developed in subsections 2.2-2.4. These tests have the same critical value in the limit as the corresponding unmodified tests under the null hypothesis \mathbf{H}_0 , and the same limiting critical value is also appropriate under the alternative hypothesis \mathbf{H}_1 . The modification proposed has no asymptotic effect under the null \mathbf{H}_0 , so that the limiting distribution of the modified tests is the same of the corresponding unmodified tests. The modified panel tests developed are:

$$\text{MH}_j^d(\mathcal{K}_{T,N}(\tau)) := \exp(-bJ_{1,N,T}) \cdot \mathbf{H}_j(\mathcal{K}_{T,N}(\tau)), \quad j = 1, 2, 3; \quad (17)$$

$$\text{MH}_j^d(\mathcal{K}_{T,N}^*(\tau)) := \exp(-bJ_{1,N,T}) \cdot \mathbf{H}_j(\mathcal{K}_{T,N}^*(\tau)), \quad j = 1, 2, 3; \quad (18)$$

$$\text{MM}_{T,N}^{j,*} := \exp(-bJ_{1,N,T}) \cdot \mathcal{M}_{T,N}^{j,*} \quad j = 1, 2, 3; \quad (19)$$

where b is a finite constant and $J_{1,N,T}$ is the arithmetic mean on N of the truncated sequences of T^{-1} times the Wald statistic $J_{1,T}^{(i)}$ for testing the joint hypothesis $\varsigma_{i,k+1} = \dots = \varsigma_{i,9} = 0$ in panel regression

$$y_{i,t} = \varepsilon_{i,t} + \sum_{j=k+1}^9 \varsigma_{i,j} t^j + \text{error}, \quad t = 1, \dots, [\tau T]; \quad i = 1, \dots, N. \quad (20)$$

Under the null hypothesis, Harvey *et al.* (2006) show that

$$\lim_{T \rightarrow +\infty} J_{1,T}^{(i)} = 1, \quad \forall i = 1, \dots, N.$$

Therefore, since we assumed independence and identical distribution with respect to the cross-sectional dimension i , we have

$$\lim_{T,N \rightarrow +\infty} J_{1,N,T} = \lim_{T,N \rightarrow +\infty} \frac{1}{N} \sum_{i=1}^N J_{1,T}^{(i)} = 1.$$

Consequently, modified panel tests have the same limiting distribution under \mathbf{H}_0 as the unmodified tests. Under the alternative hypothesis, using Harvey *et al.*'s (2006) results and the fact that the asymptotic distributions of tests $\mathbf{H}_j(\mathcal{K}_{T,N}(\tau))$, $\mathbf{H}_j(\mathcal{K}_{T,N}^*(\tau))$, and $\mathcal{M}_{T,N}^{j,*}$, $j = 1, 2, 3$ are standard gaussian (see Theorem 2.2, Theorem 2.4 and Theorem 2.5), for each $j = 1, 2, 3$, we have

$$\lim_{N \rightarrow +\infty} T^{-2} (\text{MH}_j^d(\mathcal{K}_{T,N}(\tau)) - \mathbf{H}_j(\mathcal{K}_{T,N}(\tau))) =$$

$$\begin{aligned}
&= T^{-2} \lim_{N \rightarrow +\infty} (\text{MH}_j^d(\mathcal{K}_{T,N}(\tau)) - \text{H}_j(\mathcal{K}_{T,N}(\tau))) = \\
&= T^{-2} \lim_{N \rightarrow +\infty} \{\exp[-bJ_{1,N,T}] - 1\} \text{H}_j(\mathcal{K}_{T,N}(\tau)) = O_p(1)O_p(1) = O_p(1).
\end{aligned}$$

Analogously, it results

$$\lim_{N \rightarrow +\infty} T^{-2} (\text{MH}_j^d(\mathcal{K}_{T,N}^*(\tau)) - \text{H}_j(\mathcal{K}_{T,N}^*(\tau))) = O_p(1)$$

and

$$\lim_{N \rightarrow +\infty} T^{-2} (\text{MM}_{T,N}^{j*} - \mathcal{M}_{T,N}^{j*}) = O_p(1).$$

Thus, we obtain a test which rejects for large values of the modified tests and retains the same rate of consistency under the alternative \mathbf{H}_{01} as the original unmodified H_j tests. For the alternative \mathbf{H}_{10} , the modified tests are also $O_p(1)$. A more appropriate modification procedure is proposed for testing the null against the alternative \mathbf{H}_{01} . Following Harvey *et al.* (2006), we modify tests (17), (18) and (19) by introducing

$$J_{N,\min} := \min_{\tau \in (0,1)} J_{1,N,[\tau T]}.$$

We define

$$\text{MH}_{j,\min}(\mathcal{K}_{T,N}(\tau)) := \exp(-bJ_{N,\min}) \cdot \text{H}_j(\mathcal{K}_{T,N}(\tau)), \quad j = 1, 2, 3; \quad (21)$$

$$\text{MH}_{j,\min}(\mathcal{K}_{T,N}^*(\tau)) := \exp(-bJ_{N,\min}) \cdot \text{H}_j(\mathcal{K}_{T,N}^*(\tau)), \quad j = 1, 2, 3; \quad (22)$$

$$\text{MM}_{\min,T,N}^{j*} := \exp(-bJ_{N,\min}) \cdot \mathcal{M}_{T,N}^{j*} \quad j = 1, 2, 3; \quad (23)$$

and

$$J_{\min} := \lim_{N \rightarrow +\infty} J_{N,\min}.$$

in this case also we can derive the asymptotic analysis using the alternative hypothesis of Harvey *et al.* (2006). We have

$$\begin{aligned}
&\lim_{N \rightarrow +\infty} T^{-2} (\text{MH}_{j,\min}(\mathcal{K}_{T,N}(\tau)) - \text{H}_j(\mathcal{K}_{T,N}(\tau))) = \\
&= T^{-2} \lim_{N \rightarrow +\infty} (\text{MH}_{j,\min}(\mathcal{K}_{T,N}(\tau)) - \text{H}_j(\mathcal{K}_{T,N}(\tau))) = \\
&= T^{-2} \lim_{N \rightarrow +\infty} \{\exp[-bJ_{N,\min}] - 1\} \text{H}_j(\mathcal{K}_{T,N}(\tau)) = \\
&= T^{-2} \{\exp[-bJ_{\min}] - 1\} \lim_{N \rightarrow +\infty} \text{H}_j(\mathcal{K}_{T,N}(\tau)) = o_p(1)O_p(1) = o_p(1),
\end{aligned}$$

and, analogously,

$$\lim_{N \rightarrow +\infty} T^{-2} (\text{MH}_{j,\min}(\mathcal{K}_{T,N}^*(\tau)) - \text{H}_j(\mathcal{K}_{T,N}^*(\tau))) = o_p(1)$$

and

$$\lim_{N \rightarrow +\infty} T^{-2} (\text{MM}_{\min,T,N}^{j*} - \mathcal{M}_{T,N}^{j*}) = o_p(1).$$

Thus, the new modification has no asymptotic effect under the alternative \mathbf{H}_{01} , unlike to the original MH_j^d modifications. Under the alternative \mathbf{H}_{10} , it is easily shown that the modified min tests are $O_p(1)$.

2.6 Estimation of the break

In this section we present a procedure to estimate the unknown change point. Consider the following estimator:

$$\Pi_{N,T}(\tau) = \frac{1}{N} \sum_{i=1}^N \frac{\sum_{t=[T\tau]+1}^T \tilde{\mu}_{i,t}^2 / (T - [T\tau])^2}{\sum_{t=1}^{[T\tau]} \tilde{\mu}_{i,t}^2 / [\tau T]}. \quad (24)$$

In order to explore the asymptotic behavior of the estimated unknown change point, the following assumption is necessary.

Assumption 2 Let $\tilde{\mu}_{i,s+1}, \tilde{\mu}_{i,s+2}, \dots, \tilde{\mu}_{i,s+m}$, for $s \in 0, \dots, T-1$ and $m \leq T-s$ be a sequence of stationary variables. Assume that $m^{-1} \sum_{t=s+1}^{s+m} \tilde{\mu}_{i,t}^2 \rightarrow E[\mu_i^2]$ for $E[\mu_i^2] < \infty$, $\forall i = 1, \dots, N$.

Now, let $\hat{\tau}$ be:

$$\hat{\tau} = \left\{ \operatorname{argmax}_{\tau \in (0,1)} \Pi_{N,T}(\tau) \right\}. \quad (25)$$

The following theorem shows the asymptotic properties of $\hat{\tau}$:

Theorem 2.6 Suppose that Assumption 2 holds. Under the alternative hypothesis, it results

$$(\hat{\tau} - \tau) = o_p(1), \quad (26)$$

$$T(\hat{\tau} - \tau) = O_p(1), \quad (27)$$

3 Persistence test with cross-section correlation

Previous testing procedures are valid under the assumption that the units are cross-section independent. However, this requirement is rarely likely to be satisfied in empirical economic applications where the countries or regions depend on each other. Thus in this section we therefore generalize our earlier results to suit the case of dependence among the cross-sectional units, by filtering out common factor in the panel structure. Basically, the estimation procedure is based on the Stock and Watson (2002), Bai (2003) and Bai and Ng (2004) approaches. It consists of two steps. In the first step, the common factors are estimated using the Principal Component (PC) method. In order to estimate the true number of common factors accurately in this first step, we compute the number of factors using Bai and Ng's (2002) selection criteria. In the second step, defactored data is constructed.

Specifically, let, as before, $y_{i,t}$ be the observation on the i -th cross section unit at time t for $i = 1, \dots, N$, $t = 1, \dots, T$ and suppose that it is generated according to the following linear heterogeneous panel data model:

$$y_{i,t} = d_{i,t} + \mu_{i,t} + F_t \lambda_i + \varepsilon_{i,t} \quad (28)$$

As in the previous section, three cases are considered:

- Case 1: $I(0) \rightarrow I(1)$

$$\mu_{i,t} = \mu_{i,t-1} + 1(t > [\tau T]) \eta_{i,t} \quad (29)$$

- Case 2: $I(1) \rightarrow I(0)$

$$\mu_{i,t} = \mu_{i,t-1} + 1(t \leq [\tau T])\eta_{i,t} \quad (30)$$

- Case 3: unknown direction $I(0) \rightarrow I(1)$ or $I(1) \rightarrow I(0)$

where, as before, $\varepsilon_{i,t}$ is a stationary process, $\tau \in (0, 1)$ and the deterministic component $d_{i,t}$ is taken to be the unity vector. Further we assume that F_t in (28) is a stationary $(r \times 1)$ vector of common factors and λ_i is the corresponding vector of factor loadings.

Let us write the factors F_t and factor loadings λ_i in (28) in matrix notation. We write $F = (F_1, \dots, F_T)'$ and $\Lambda = (\lambda_1, \dots, \lambda_N)'$. The following assumptions are required:

Assumption 3 *the loadings λ_i is either deterministic such that $\|\lambda_i\|^4 \leq M$ or stochastic such that $E\|\lambda_i\|^4 \leq M$, in either case $\Lambda_i'\Lambda_i/N \rightarrow \Sigma_\Lambda$, as $N \rightarrow \infty$ for some $(r \times r)$ positive definite matrix Σ_Λ .*

Assumption 4 *$E\|F_t\|^4 \leq M$ and $\frac{1}{T} \sum_{t=1}^T F_t F_t' \rightarrow \Sigma_F$, for a $(r \times r)$ positive definite matrix Σ_F .*

Assumption 5 *The errors $\{\varepsilon_{i,t}\}$ and $\{\eta_{i,t}\}$, the factor $\{F_t\}$ and the loadings $\{\lambda_i\}$ are four mutually independent stochastic variables.*

Assumption 3 and 4 imply r common factors and assumption 5 is standard in factor analysis. The common factors are estimated as they are in Stock and Watson (2002), i.e. using principal components. Specifically, the principal component of F , denoted as \tilde{F} , is \sqrt{T} times the first r eigenvectors, corresponding to the first r largest eigenvalues of the $(T \times T)$ matrix of demeaned and standardized $\tilde{y}_i \tilde{y}_i'$. Under the normalization $\tilde{F} \tilde{F}'/T = I_r$, the estimated loading matrix is $\tilde{\Lambda} = \tilde{F}' \tilde{y}_i/T$. Thus the estimated residuals are defined as

$$\tilde{z}_{i,t} = \tilde{y}_{i,t} - \tilde{F}_t \tilde{\lambda}_i \quad (31)$$

From the data generating process (28)-(29) and (31), one can, for example, see that for each cross section i , the process $\tilde{z}_{i,t}$ is stationary up to and including time $[\tau T]$ but is $I(1)$ after the break if, and only if, $\sigma_{\eta_i}^2 > 0$. Naturally the converse it's true if we adopt (30) instead of (29).

Thus our strategy is to apply the panel test statistics presented in section 2 to the de-factored data $\tilde{z}_{i,t}$.

4 Monte Carlo simulation results

In this section we use Monte Carlo experiments to examine the finite sample properties of the panel persistence tests. We consider two sets of Monte Carlo experiments. The first set focuses on the model (1)-(3), i.e. where we assume cross-section independence, while the second set of experiments is based on the model (28)-(31) where we allow for the presence of dependence across the different units in the panel. We start the analysis considering the empirical

rejection frequencies of the tests when the data are generated according to the I(0)-I(1) switch data generating process embraced in (1)-(3) under the hypothesis of cross-section independence. We investigate the impact of varying the signal-noise-ratio among $\sigma_{\eta_i} = 0, 0.10, 0.50$ and $\sigma_{\varepsilon_i} \sim U [0.5, 1.5]$ and the breakpoint among $\tau = 0.3, 0.5, 0.7$. The simulation results were performed using 1000 Monte Carlo replications and the RNDN function of Gauss 6.0. As is often the case in the literature, we fix for all the tests $\Lambda = [0.2, 0.8]$ and $T = 50, 100$ and $N = 25, 50, 100$. In Table 1 we present the moments of Kim's (2000, 2002) and Busetti and Taylor's (2004) tests which were used to standardize the panel tests. Their values were computed using 50,000 replications.

Table 1 about here

The size results for the benchmark model (1)-(3) are reported in Table 2.

Table 2 about here

All the panel test statistics seem to have good size for both small and large T, N .

Looking at power of the tests for the case $I(0) \rightarrow I(1)$ (Tables 3-5), many interesting results emerge.

Table 3 about here

Table 4 about here

Table 5 about here

Comparing the power of panel tests derived using Kim's (2000), Busetti and Taylor's (2004) and Harvey *et al.*'s (2006) methods we find no significant differences. As expected, the power of tests grows, with the exception of reverse panel tests, as the signal to noise ratio rises and τ is smaller. This occurs because the higher σ_{η} is, the stronger is the random walk component. We have that the smaller τ is, the greater is the proportion of the sample containing a random walk component. Finally, the previous finding are reversed for the panel reverse tests. This because the fact that we are testing a change from I(0) to I(1). The opposite results are found for a shift from $I(1) \rightarrow I(0)$. Here, as expected, the power of tests grows largely for the reverse tests and it is striking that the results mimic those in Tables 3-5. Thus $H_1(\mathcal{K}^*)$, $H_2(\mathcal{K}^*)$ and $H_3(\mathcal{K}^*)$ show better properties than $H_1(\mathcal{K})$, $H_2(\mathcal{K})$, $H_3(\mathcal{K})$ and \mathcal{M}^{1*} , \mathcal{M}^{2*} and \mathcal{M}^{3*} (see Tables 6-8).

Table 6 about here

Table 7 about here

Table 8 about here

We now present the empirical size of tests when cross-section dependence is included in the model, as in equations (28)-(30). We consider two cases of cross section dependence, a "low cross section dependence", where $\lambda_i \sim iidU [0.0, 0.2]$, and "high cross section dependence", with $\lambda_i \sim iidU [-1, 3]$. In the simulation we include only one factor $F_t \sim N(0, 1)$. Results are reported in panels A (low case) and B (high case) of Table 9.

Table 9 about here

As expected, the extent of over-rejection of the tests very much depends on the degree of cross-section dependence. Both for low as well as for strong cross-section dependence the panel tests are distorted with over-rejection, which grows as the degree of cross-section dependence rise. Thus panel tests that do not allow for cross-section dependence can be seriously biased if the degree cross-section dependence is large. To take into account of cross-section dependence we use the method proposed in Stock and Watson (2002) and Bai and Ng (2004). The method basically consists of filtering out the individual-specific cross-sections y_{it} by the factor component computed using the principal component method. The number of factors are computed using the methodology proposed in Bai and Ng (2002). To be precise throughout the Monte Carlo simulation analyses the number of factors are computed using the $IC(3)$ criterion proposed in Bai and Ng (2002) with a maximum number of three factors. As before, we use only one factor $F_t \sim N(0, 1)$, and $\lambda \sim U[0, 1]$.

In Tables 10 and in 11-13, we present, respectively, the size and power of defactored panel tests using the Stock and Watson (2002) methodology. Looking at the results we note first that the tests have now generally good size. As expected, the power of tests grows for larger values of T and N .

Table 10 about here

Table 11 about here

Table 12 about here

Table 13 about here

Finally, results of the power of tests for the case of a change from $I(1)$ to $I(0)$ are reported in Tables 14-16. As expected, the power of tests grows large for the reverse tests.

Table 14 about here

Table 15 about here

Table 16 about here

5 Empirical applications

We apply the panel tests described in this paper to a panel of 21 OECD quarterly inflation rate series observed for the period 1970.1-2007.3.¹ The series are calculated as the first difference of the logarithm of the (seasonally adjusted) consumer price index. The data are taken from OECD Main Economic Indicators. In Table 17 the panel tests results are reported.

Table 17 about here

Looking at their values, we note first that for first set of test statistics, i.e. test statistics which are computed not taking into account possible cross-section

¹The countries included in the panel are Australia, Austria, Belgium, Canada, Denmark, Finland, France, Germany, Greece, Italy, Japan, Luxembourg, the Netherlands, New Zealand, Norway, Portugal, Spain, Sweden, Switzerland, the United Kingdom, the United States.

dependence, reverse tests strongly suggest a change of persistence from $I(1)$ to $I(0)$. Change in persistence are also shown by the \mathcal{M}^j tests as well as their modified version. Mixed responses are obtained from the $H_1(\mathcal{K})$ test, when this is compared with $H_2(\mathcal{K})$ and $H_3(\mathcal{K})$ test statistics. While the former test rejects the null hypothesis, the latter tests do not reject the null hypothesis for a process which is $I(0)$ throughout. In order to take into account possible cross-dependence across the countries, we first compute the number of factors. The $IC(3)$ criterion suggests three factors (we allow for a maximum number of five factors). We use the estimated factors and factor loadings to compute $\hat{\tilde{z}}_{i,t}$, as in equation (31). Looking at the results of panel test statistics we note that the previous results are now partially reversed. Here both $H_1(\mathcal{K})$, and $H_2(\mathcal{K})$, $H_3(\mathcal{K})$ test statistics do not reject the null hypothesis, allowing for a process which is $I(0)$ throughout the sample. Reversed test statistics indicate a change from $I(1)$ to $I(0)$ and \mathcal{M}^j strongly reject the null hypothesis of constant persistence (constant $I(0)$). Given these results, we conclude that the inflation process is characterized by a change in persistence from $I(1)$ to $I(0)$. Using the cross-sectional dependence adjusted series, the change in persistence, computed using (25), took place in 1980.4.

6 Conclusions

In this paper we present new panel test statistics for a change in persistence which are based on a modified time series version of the ratio-based statistics presented in Busetti and Taylor (2004) and Harvey *et al.* (2006). These statistics are used to test the null hypothesis of stationarity against the alternative of a change in persistence from $I(0)$ to $I(1)$, or viceversa. The alternative of an unknown direction is also considered. Asymptotic distributions of the new panel tests under the hypothesis of cross-section independence are derived and Monte Carlo analysis suggests that these tests perform very well. Cross-section dependence is also considered.

We show first that when testing for a change in persistence from $I(0)$ ($I(1)$) to $I(1)$ ($I(0)$) panel tests have good properties. Secondly, we report the importance of taking into account possible cross-sectional dependence when computing the panel test statistics, especially for highly dependent panels. Finally, we apply the panel tests to a panel of 21 OECD inflation rates observed during the period 1970.1 - 2007.3. The results were consistent with a change of persistence from $I(1)$ to $I(0)$ in April 1980.

Appendix A. Proof of theorems

Proof of Theorem 2.1. Fixed $i = 1, \dots, N$, $t_1 = 1, \dots, [T\tau]$ and $t_2 = [T\tau] + 1, \dots, T$, it results $S_{i,t_1}^{(1)}$ and $S_{i,t_2}^{(0)}$ mutually independent. Therefore, (8) and (9) can be rewritten as

$$\mu = \mathbf{E} \left[\frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2} \right]$$

and

$$\sigma = \sqrt{\mathbf{V} \left[\frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2} \right]}.$$

By Theorem 3.1 in Kim (2000), it results

$$\lim_{T \rightarrow +\infty} \frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2} = \frac{(1 - \tau)^{-2} \int_{\tau}^1 V_i(r - \tau)^2 dr}{\tau^{-2} \int_0^{\tau} V_i(r)^2 dr},$$

where $\{V_i\}_{i=1}^{+\infty}$ is a sequence of standard brownian bridges that are independent and identically distributed. Furthermore, by the hypotheses stated in Assumption 1, we have that

$$\begin{aligned} \lim_{T \rightarrow +\infty} \mathbf{E} \left[\frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2} \right] &= \bar{\mu}, \\ \lim_{T \rightarrow +\infty} \mathbf{V} \left[\frac{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2}{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2} \right] &= \bar{\sigma}^2, \end{aligned}$$

with

$$\bar{\mu} = \mathbf{E} \left[\frac{(1 - \tau)^{-2} \int_{\tau}^1 V_i(r - \tau)^2 dr}{\tau^{-2} \int_0^{\tau} V_i(r)^2 dr} \right] \quad (32)$$

and

$$\bar{\sigma}^2 = \mathbf{V} \left[\frac{(1 - \tau)^{-2} \int_{\tau}^1 V_i(r - \tau)^2 dr}{\tau^{-2} \int_0^{\tau} V_i(r)^2 dr} \right] \quad (33)$$

for each $i = 1, \dots, N$. Therefore

$$\lim_{N \rightarrow +\infty} \lim_{T \rightarrow +\infty} \mathcal{K}_{T,N}(\tau) = \lim_{N \rightarrow +\infty} \frac{\sqrt{N}}{\bar{\sigma}} \cdot \frac{1}{N} \cdot \sum_{i=1}^N \left[\frac{(1 - \tau)^{-2} \int_{\tau}^1 V_i(r - \tau)^2 dr}{\tau^{-2} \int_0^{\tau} V_i(r)^2 dr} - \bar{\mu} \right],$$

for $i = 1, \dots, N$. The Central Limit Theorem guarantees that

$$\lim_{N \rightarrow +\infty} \frac{\sqrt{N}}{\bar{\sigma}} \cdot \frac{1}{N} \cdot \sum_{i=1}^N \left[\frac{(1 - \tau)^{-2} \int_{\tau}^1 V_i(r - \tau)^2 dr}{N \tau^{-2} \int_0^{\tau} V_i(r)^2 dr} - \bar{\mu} \right] \sim N(0, 1),$$

and the Theorem is completely proved. ■

Proof of Theorem 2.2.

- (i) The result follows from the continuous mapping theorem and the continuity of the functionals.
- (ii) Since $\mathcal{K}(\tau) \sim N(0, 1)$, for each τ , then $\mathcal{K}(\tau)$ is an iid continuous-time stochastic process. Therefore, we can define the random variable $\mathcal{K} \sim N(0, 1)$ such that $\mathcal{K}(\tau) \equiv \mathcal{K}$, for each $\tau \in (0, 1)$.

Then we have

$$\begin{aligned} \mathbf{H}_1(\mathcal{K}(\tau)) &= \sup_{\tau \in (0,1)} (\mathcal{K}) = \mathcal{K} \sim N(0, 1); \\ \mathbf{H}_2(\mathcal{K}(\tau)) &= \int_{\tau \in (0,1)} \mathcal{K} d\tau = \mathcal{K} \cdot \int_{\tau \in (0,1)} d\tau = \mathcal{K} \sim N(0, 1); \\ \mathbf{H}_3(\mathcal{K}(\tau)) &= \log \left\{ \int_{\tau \in (0,1)} \exp[\mathcal{K}] d\tau \right\} = \log \{ \exp[\mathcal{K}] \} \cdot \int_{\tau \in (0,1)} d\tau = \mathcal{K} \sim N(0, 1) \end{aligned}$$

The result is completely proved. ■

Proof of Theorem 2.3. By Theorem 3.1 in Busetti and Taylor (2004), it results

$$\lim_{T \rightarrow +\infty} \frac{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2}{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2} = \frac{\tau^{-2} \int_0^\tau [V_i^{***}(r)]^2 dr}{(1 - \tau)^{-2} \int_\tau^1 [V_i^{**}(r)]^2 dr},$$

where

$$\begin{aligned} V_i^{**}(r) &= V_i(r) - V_i(\tau) - (r - \tau)(1 - \tau)^{-1}(V_i(1) - V_i(\tau)) \\ V_i^{***}(r) &= V_i(r) - r\tau^{-1}V_i(\tau) \end{aligned}$$

and

$$V_i(r) = W_0(r) + c \left\{ \int_0^{\min\{r,\tau\}} W_c(s) ds + 1(r > \tau)[(r - \tau)W_c(\tau)] \right\},$$

where W is a standard Wiener process. Hypotheses stated in Assumption 1 imply

$$\begin{aligned} \lim_{T \rightarrow +\infty} \mathbf{E} \left[\frac{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2}{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2} \right] &= \bar{\mu}, \\ \lim_{T \rightarrow +\infty} \mathbf{V} \left[\frac{[T\tau]^{-2} \sum_{t=1}^{[T\tau]} S_{i,t}^{(0)}(\tau)^2}{(T - [T\tau])^{-2} \sum_{t=[T\tau]+1}^T S_{i,t}^{(1)}(\tau)^2} \right] &= \bar{\sigma}^2, \end{aligned}$$

with

$$\bar{\mu} = \mathbf{E} \left[\frac{\tau^{-2} \int_0^\tau [V_i^{***}(r)]^2 dr}{(1 - \tau)^{-2} \int_\tau^1 [V_i^{**}(r)]^2 dr} \right] \quad (34)$$

and

$$\bar{\sigma}^2 = \mathbf{V} \left[\frac{\tau^{-2} \int_0^\tau [V_i^{***}(r)]^2 dr}{(1 - \tau)^{-2} \int_\tau^1 [V_i^{**}(r)]^2 dr} \right] \quad (35)$$

for each $i = 1, \dots, N$. Hence

$$\lim_{N \rightarrow +\infty} \lim_{T \rightarrow +\infty} \mathcal{K}_{T,N}^*(\tau) = \lim_{N \rightarrow +\infty} \frac{\sqrt{N}}{\bar{\sigma}} \cdot \frac{1}{N} \cdot \sum_{i=1}^N \left[\frac{\tau^{-2} \int_0^\tau [V_i^{***}(r)]^2 dr}{(1 - \tau)^{-2} \int_\tau^1 [V_i^{**}(r)]^2 dr} - \bar{\mu} \right],$$

for $i = 1, \dots, N$. The Central Limit Theorem guarantees that

$$\lim_{N \rightarrow +\infty} \frac{\sqrt{N}}{\bar{\sigma}} \cdot \frac{1}{N} \cdot \sum_{i=1}^N \left[\frac{\tau^{-2} \int_0^\tau [V_i^{***}(r)]^2 dr}{(1-\tau)^{-2} \int_\tau^1 [V_i^{**}(r)]^2 dr} - \bar{\mu} \right] \sim N(0, 1),$$

and the Theorem is completely proved. ■

Proof of Theorem 2.4. Analogous to the proof of Theorem 2.2. ■

Proof of Theorem 2.5. It is a direct consequence of the Central Limit Theorem. ■

Proof of Theorem 2.6. Since

$$M_T^{(i)}(\tau) := \frac{\sum_{t=[T\tau]+1}^T \tilde{\mu}_{i,t}^2 / (T - [T\tau])^2}{\sum_{t=1}^{[T\tau]} \tilde{\mu}_{i,t}^2 / [\tau T]} \geq 0, \quad \forall i = 1, \dots, N,$$

then

$$\begin{aligned} \hat{\tau} &= \operatorname{argmax}_{\tau \in (0,1)} M_T^{(i)}(\tau) \Rightarrow \\ \Rightarrow \hat{\tau} &= \operatorname{argmax}_{\tau \in (0,1)} \left\{ \frac{1}{N} \cdot \sum_{i=1}^N M_T^{(i)}(\tau) \right\} = \operatorname{argmax}_{\tau \in (0,1)} \Pi_{N,T}(\tau). \end{aligned}$$

Therefore, Theorem 3.5 in Kim (2000) guarantees the thesis. ■

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Table 1: Simulated moments for individual Kim(2000, 2002) and Busetti and Taylor tests (2004)

T	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}
Mean - drift case									
50	1.817	1.608	6.173	1.836	1.624	6.204	2.783	2.632	9.214
100	1.803	1.566	6.386	1.803	1.564	6.391	2.749	2.545	9.436
500	1.800	1.537	6.788	1.779	1.506	6.660	2.741	2.478	9.868
Mean - linear trend case									
50	1.415	0.906	3.785	1.412	0.908	3.796	1.992	1.334	5.297
100	1.377	0.844	3.738	1.374	0.843	3.831	1.924	1.222	5.145
500	1.362	0.815	3.718	1.353	0.822	3.839	1.886	1.177	5.115
Std. deviation - drift case									
50	1.580	2.282	5.856	1.586	2.370	6.003	1.737	2.903	6.874
100	1.548	2.178	5.841	1.526	2.106	5.728	1.669	2.644	6.590
500	1.528	2.170	6.041	1.503	1.980	5.701	1.667	2.562	6.318
Std. deviation - linear trend case									
50	0.869	0.863	2.759	0.866	0.861	2.777	0.869	1.030	3.068
100	0.814	0.687	2.451	0.798	0.685	2.447	0.794	0.781	2.618
500	0.764	0.607	2.428	0.734	0.651	2.401	0.709	0.701	2.604

Table 2: Size of Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. Intercept case.

T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}
50	25	0.075	0.074	0.078	0.069	0.061	0.058	0.056	0.061	0.063
	50	0.078	0.074	0.068	0.064	0.060	0.047	0.074	0.069	0.067
	100	0.073	0.072	0.078	0.065	0.063	0.056	0.065	0.062	0.077
100	25	0.054	0.054	0.051	0.076	0.072	0.073	0.069	0.072	0.070
	50	0.041	0.040	0.046	0.056	0.069	0.069	0.058	0.059	0.064
	100	0.051	0.060	0.064	0.053	0.065	0.055	0.049	0.051	0.051

T	N	$MH_1^d(\mathcal{K})$	$MH_2^d(\mathcal{K})$	$MH_3^d(\mathcal{K})$	$MH_1^d(\mathcal{K}^*)$	$MH_2^d(\mathcal{K}^*)$	$MH_3^d(\mathcal{K}^*)$	MM^{1*}	MM^{2*}	MM^{3*}
50	25	0.075	0.075	0.078	0.069	0.063	0.062	0.062	0.065	0.064
	50	0.078	0.078	0.069	0.065	0.061	0.047	0.075	0.070	0.067
	100	0.073	0.072	0.070	0.065	0.066	0.056	0.066	0.064	0.073
100	25	0.055	0.054	0.052	0.078	0.078	0.076	0.070	0.075	0.072
	50	0.043	0.043	0.048	0.056	0.071	0.069	0.058	0.060	0.065
	100	0.051	0.060	0.066	0.054	0.065	0.056	0.050	0.052	0.052

T	N	$MH_{1m}(\mathcal{K})$	$MH_{2m}(\mathcal{K})$	$MH_{3m}(\mathcal{K})$	$MH_{1m}(\mathcal{K}^*)$	$MH_{2m}(\mathcal{K}^*)$	$MH_{3m}(\mathcal{K}^*)$	MM_m^{1*}	MM_m^{2*}	MM_m^{3*}
50	25	0.075	0.075	0.076	0.069	0.063	0.062	0.059	0.064	0.063
	50	0.068	0.068	0.069	0.065	0.061	0.047	0.075	0.070	0.067
	100	0.073	0.072	0.070	0.065	0.065	0.056	0.065	0.064	0.072
100	25	0.055	0.054	0.051	0.077	0.073	0.072	0.060	0.074	0.071
	50	0.043	0.044	0.046	0.056	0.068	0.068	0.058	0.059	0.065
	100	0.051	0.060	0.064	0.054	0.065	0.056	0.049	0.051	0.052

Notes : Empirical sizes corresponding to a 5% nominal size.

Table 3: Power of Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. $I(0) \rightarrow I(1)$. Intercept case.

τ	σ_η	T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}		
0.3	0.10	50	25	0.582	0.610	0.611	0.193	0.209	0.107	0.544	0.574	0.561		
			50	0.845	0.858	0.854	0.272	0.316	0.130	0.798	0.821	0.802		
			100	0.979	0.985	0.987	0.417	0.490	0.147	0.971	0.981	0.968		
			100	25	0.998	0.998	0.997	0.801	0.832	0.470	0.999	0.999	0.998	
				50	1.000	1.000	1.000	0.950	0.965	0.640	1.000	1.000	1.000	
				100	1.000	1.000	1.000	0.997	0.997	0.806	1.000	1.000	1.000	
		0.5	50	25	0.534	0.557	0.640	0.035	0.048	0.012	0.346	0.372	0.444	
				50	0.819	0.818	0.902	0.030	0.042	0.007	0.588	0.609	0.669	
				100	0.943	0.959	0.983	0.014	0.022	0.000	0.775	0.813	0.879	
			100	25	0.994	0.994	0.999	0.125	0.152	0.004	0.983	0.990	0.992	
				50	1.000	1.000	1.000	0.117	0.158	0.001	0.998	0.998	1.000	
				100	1.000	1.000	1.000	0.097	0.164	0.000	0.998	0.998	1.000	
0.7	50	25	0.280	0.305	0.406	0.025	0.032	0.020	0.150	0.165	0.226			
		50	0.438	0.456	0.628	0.016	0.023	0.005	0.199	0.210	0.361			
		100	0.639	0.662	0.866	0.008	0.011	0.003	0.317	0.349	0.554			
	100	25	0.909	0.920	0.973	0.003	0.009	0.002	0.701	0.749	0.885			
		50	0.989	0.990	0.999	0.001	0.003	0.000	0.888	0.921	0.984			
		100	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000			
0.3	0.50	50	25	1.000	1.000	1.000	0.994	0.995	0.930	1.000	1.000	1.000		
			50	1.000	1.000	1.000	1.000	1.000	0.996	1.000	1.000	1.000		
			100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
			100	10	1.000	1.000	1.000	0.974	0.985	0.861	1.000	1.000	1.000	
				25	1.000	1.000	1.000	1.000	1.000	0.995	1.000	1.000	1.000	
				50	1.000	1.000	1.000	1.000	0.999	1.000	1.000	1.000	1.000	
		0.5	50	100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	
				100	25	1.000	1.000	1.000	0.636	0.685	0.051	1.000	1.000	1.000
					50	1.000	1.000	1.000	0.834	0.872	0.034	1.000	1.000	1.000
			100		1.000	1.000	1.000	0.952	0.979	0.018	1.000	1.000	1.000	
			0.7	50	25	1.000	1.000	1.000	0.914	0.933	0.232	1.000	1.000	1.000
					50	1.000	1.000	1.000	0.982	0.987	0.263	1.000	1.000	1.000
100	1.000	1.000			1.000	0.999	0.999	0.244	1.000	1.000	1.000			
100	25	1.000		1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000			
	50	1.000		1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000			
	100	1.000		1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000			
0.7	100	25	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000			
		50	1.000	1.000	1.000	0.001	0.000	0.000	1.000	1.000	1.000			
		50	1.000	1.000	1.000	0.001	0.000	0.000	1.000	1.000	1.000			
		100	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000			

Table 4: Power of Modified Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. $I(0) \rightarrow I(1)$. Intercept case.

τ	σ_η	T	N	MH ₁ ^d (κ)	MH ₂ ^d (κ)	MH ₃ ^d (κ)	MH ₁ ^d (κ^*)	MH ₂ ^d (κ^*)	MH ₃ ^d (κ^*)	MM ^{1*}	MM ^{2*}	MM ^{3*}	
0.3	0.10	50	25	0.588	0.612	0.615	0.197	0.214	0.108	0.557	0.578	0.574	
			50	0.851	0.863	0.854	0.275	0.317	0.132	0.802	0.827	0.808	
			100	0.980	0.985	0.988	0.420	0.491	0.148	0.972	0.981	0.969	
			100	25	0.998	0.998	0.997	0.803	0.836	0.471	0.999	0.999	0.998
				50	1.000	1.000	1.000	0.951	0.965	0.643	1.000	1.000	1.000
				100	1.000	1.000	1.000	0.997	0.997	0.809	1.000	1.000	1.000
		0.5	50	25	0.542	0.560	0.647	0.038	0.048	0.013	0.356	0.382	0.451
				50	0.823	0.822	0.903	0.033	0.042	0.007	0.591	0.619	0.676
				100	0.945	0.960	0.983	0.015	0.022	0.000	0.778	0.814	0.881
			100	25	0.994	0.994	0.999	0.126	0.155	0.004	0.984	0.990	0.992
				50	1.000	1.000	1.000	0.118	0.159	0.001	0.999	0.999	1.000
				100	1.000	1.000	1.000	0.099	0.164	0.000	0.982	0.987	0.999
0.7	50	25	0.289	0.309	0.410	0.025	0.033	0.021	0.159	0.170	0.230		
		50	0.442	0.458	0.634	0.017	0.023	0.005	0.201	0.221	0.364		
		100	0.640	0.664	0.866	0.008	0.012	0.003	0.321	0.353	0.560		
	100	25	0.912	0.920	0.973	0.003	0.009	0.002	0.708	0.751	0.889		
		50	0.989	0.990	0.999	0.001	0.003	0.000	0.889	0.924	0.984		
		100	1.000	1.000	1.000	0.000	0.000	0.000	0.982	0.987	0.999		
0.3	0.50	50	25	1.000	1.000	1.000	0.994	0.995	0.933	1.000	1.000	1.000	
			50	1.000	1.000	1.000	1.000	1.000	0.996	1.000	1.000	1.000	
			100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	
		100	25	1.000	1.000	1.000	1.000	1.000	1.000	0.995	1.000	1.000	1.000
			50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
			100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
		0.5	50	25	1.000	1.000	1.000	0.651	0.695	0.052	1.000	1.000	1.000
				50	1.000	1.000	1.000	0.835	0.876	0.034	1.000	1.000	1.000
				100	1.000	1.000	1.000	0.953	0.979	0.018	1.000	1.000	1.000
			100	25	1.000	1.000	1.000	0.916	0.935	0.237	1.000	1.000	1.000
				50	1.000	1.000	1.000	0.982	0.987	0.265	1.000	1.000	1.000
				100	1.000	1.000	1.000	0.999	0.999	0.247	1.000	1.000	1.000
0.7	50	10	0.999	1.000	1.000	0.002	0.005	0.001	0.990	0.995	1.000		
		25	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
	100	100	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
		25	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	0.001	0.001	0.000	1.000	1.000	1.000		
			100	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000	

Table 5: Power of Modified min Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. $I(0) \rightarrow I(1)$. Intercept case.

τ	σ_η	T	N	MH _{1m} (\mathcal{K})	MH _{2m} (\mathcal{K})	MH _{3m} (\mathcal{K})	MH _{1m} (\mathcal{K}^*)	MH _{2m} (\mathcal{K}^*)	MH _{3m} (\mathcal{K}^*)	$\mathcal{M}\mathcal{M}_m^{1*}$	$\mathcal{M}\mathcal{M}_m^{2*}$	$\mathcal{M}\mathcal{M}_m^{3*}$	
0.3	0.10	50	25	0.586	0.612	0.612	0.197	0.213	0.108	0.552	0.576	0.574	
			50	0.849	0.859	0.854	0.273	0.316	0.132	0.801	0.824	0.806	
			100	0.979	0.985	0.988	0.419	0.491	0.147	0.972	0.981	0.969	
			100	25	0.998	0.998	0.997	0.802	0.832	0.470	0.999	0.999	0.998
				50	1.000	1.000	1.000	0.951	0.965	0.642	1.000	1.000	1.000
				100	1.000	1.000	1.000	0.997	0.997	0.806	1.000	1.000	1.000
		0.5	50	25	0.539	0.558	0.647	0.037	0.048	0.013	0.354	0.379	0.450
				50	0.821	0.819	0.903	0.032	0.042	0.007	0.590	0.617	0.673
				100	0.998	0.998	0.997	0.802	0.832	0.470	0.999	0.999	0.998
			100	25	0.994	0.994	0.999	0.125	0.152	0.004	0.984	0.990	0.992
				50	1.000	1.000	1.000	0.117	0.158	0.001	0.998	0.998	1.000
				100	1.000	1.000	1.000	0.098	0.164	0.000	1.000	1.000	1.000
0.7	50	25	0.287	0.307	0.409	0.025	0.032	0.021	0.154	0.168	0.230		
		50	0.441	0.456	0.633	0.017	0.023	0.005	0.200	0.220	0.363		
		100	0.640	0.663	0.866	0.008	0.012	0.003	0.320	0.353	0.560		
	100	25	0.911	0.920	0.973	0.003	0.009	0.001	0.702	0.751	0.887		
		50	0.989	0.990	0.999	0.001	0.003	0.000	0.888	0.922	0.984		
		100	1.000	1.000	1.000	0.098	0.164	0.000	1.000	1.000	1.000		
0.3	0.50	50	25	1.000	1.000	1.000	0.994	0.995	0.930	1.000	1.000	1.000	
			50	1.000	1.000	1.000	1.000	1.000	0.996	1.000	1.000	1.000	
			100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	
			100	25	1.000	1.000	1.000	1.000	1.000	0.995	1.000	1.000	1.000
				50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
				100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
		0.5	50	25	1.000	1.000	1.000	0.642	0.687	0.051	1.000	1.000	1.000
				50	1.000	1.000	1.000	0.834	0.872	0.034	1.000	1.000	1.000
				100	1.000	1.000	1.000	0.952	0.979	0.018	0.932	0.949	0.933
			100	25	1.000	1.000	1.000	0.915	0.933	0.233	1.000	1.000	1.000
				50	1.000	1.000	1.000	0.982	0.987	0.263	1.000	1.000	1.000
				100	1.000	1.000	1.000	0.999	0.999	0.244	1.000	1.000	1.000
0.7	50	25	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
		100	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
	100	25	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	0.000	0.001	0.001	1.000	1.000	1.000		
		100	1.000	1.000	1.000	0.000	0.000	0.000	1.000	1.000	1.000		

Table 6: Power of Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. $I(1) \rightarrow I(0)$. Intercept case.

τ	σ_η	T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}	
0.3	0.10	50	25	0.030	0.037	0.021	0.266	0.284	0.391	0.160	0.182	0.249	
			50	0.021	0.036	0.012	0.427	0.428	0.593	0.206	0.218	0.355	
			100	0.008	0.011	0.005	0.639	0.674	0.839	0.319	0.358	0.549	
			100	25	0.002	0.003	0.001	0.891	0.913	0.968	0.682	0.736	0.875
				50	0.000	0.000	0.000	0.990	0.993	0.998	0.893	0.931	0.976
				100	0.000	0.000	0.000	1.000	1.000	1.000	0.983	0.990	1.000
		0.5	50	25	0.041	0.049	0.017	0.550	0.570	0.657	0.382	0.405	0.453
				50	0.032	0.047	0.007	0.792	0.810	0.890	0.563	0.598	0.703
				100	0.020	0.034	0.001	0.964	0.973	0.993	0.820	0.856	0.907
			100	25	0.095	0.127	0.004	0.995	0.996	0.998	0.975	0.981	0.985
				50	0.095	0.134	0.000	1.000	1.000	1.000	0.998	0.999	1.000
				100	0.089	0.141	0.000	1.000	1.000	1.000	1.000	1.000	1.000
0.7	50	25	0.220	0.247	0.120	0.623	0.639	0.636	0.593	0.619	0.599		
		50	0.326	0.370	0.134	0.858	0.860	0.860	0.814	0.841	0.819		
		100	0.437	0.512	0.174	0.984	0.985	0.981	0.967	0.978	0.975		
	100	25	0.789	0.817	0.429	0.996	0.999	0.996	0.995	0.997	0.996		
		50	0.941	0.950	0.595	1.000	1.000	1.000	1.000	1.000	1.000		
		100	0.999	0.998	0.758	1.000	1.000	1.000	1.000	1.000	1.000		
0.3	0.50	50	25	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			50	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			100	0.000	0.002	0.030	1.000	1.000	1.000	1.000	1.000	1.000	
			100	25	0.000	0.001	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000
		0.5	50	25	0.659	0.708	0.039	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.846	0.887	0.039	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.966	0.981	0.030	1.000	1.000	1.000	1.000	1.000	1.000
			100	25	0.910	0.921	0.195	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.095	0.133	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.999	0.999	0.238	1.000	1.000	1.000	1.000	1.000	1.000
0.7	50	25	0.999	0.999	0.937	1.000	1.000	1.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		100	0.966	0.981	0.030	1.000	1.000	1.000	1.000	1.000	1.000		
	100	25	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		

Table 7: Power of Modified Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. $I(1) \rightarrow I(0)$. Intercept case.

τ	σ_η	T	N	MH ₁ ^d (κ)	MH ₂ ^d (κ)	MH ₃ ^d (κ)	MH ₁ ^d (κ^*)	MH ₂ ^d (κ^*)	MH ₃ ^d (κ^*)	MM ^{1*}	MM ^{2*}	MM ^{3*}	
0.3	0.10	50	25	0.030	0.037	0.021	0.266	0.284	0.391	0.160	0.182	0.249	
			50	0.021	0.036	0.012	0.427	0.428	0.593	0.206	0.218	0.355	
			100	0.008	0.011	0.005	0.639	0.674	0.839	0.319	0.358	0.549	
			100	25	0.002	0.003	0.001	0.891	0.913	0.968	0.682	0.736	0.875
				50	0.000	0.000	0.000	0.990	0.993	0.998	0.893	0.931	0.976
				100	0.000	0.000	0.000	1.000	1.000	1.000	0.983	0.990	1.000
		0.5	50	25	0.041	0.049	0.017	0.550	0.570	0.657	0.382	0.405	0.453
				50	0.032	0.047	0.007	0.792	0.810	0.890	0.563	0.598	0.703
				100	0.020	0.034	0.001	0.964	0.973	0.993	0.820	0.856	0.907
			100	25	0.095	0.127	0.004	0.995	0.996	0.998	0.975	0.981	0.985
				50	0.095	0.134	0.000	1.000	1.000	1.000	0.998	0.999	1.000
				100	0.089	0.141	0.000	1.000	1.000	1.000	1.000	1.000	1.000
0.7	50	25	0.220	0.247	0.120	0.623	0.639	0.636	0.593	0.619	0.599		
		50	0.326	0.370	0.134	0.858	0.860	0.860	0.814	0.841	0.819		
		100	0.437	0.512	0.174	0.984	0.985	0.981	0.967	0.978	0.975		
	100	25	0.789	0.817	0.429	0.996	0.999	0.996	0.995	0.997	0.996		
		50	0.941	0.950	0.595	1.000	1.000	1.000	1.000	1.000	1.000		
		100	0.999	0.998	0.758	1.000	1.000	1.000	1.000	1.000	1.000		
0.3	0.50	50	25	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			50	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			100	0.000	0.002	0.030	1.000	1.000	1.000	1.000	1.000	1.000	
			100	25	0.000	0.001	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000
		0.5	50	25	0.659	0.708	0.039	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.846	0.887	0.039	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.966	0.981	0.030	1.000	1.000	1.000	1.000	1.000	1.000
			100	25	0.910	0.921	0.195	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.095	0.133	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.999	0.999	0.238	1.000	1.000	1.000	1.000	1.000	1.000
0.7	50	25	0.999	0.999	0.937	1.000	1.000	1.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		100	0.966	0.981	0.030	1.000	1.000	1.000	1.000	1.000	1.000		
	100	25	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		

Table 8: Power of Modified min Panel Tests When the Model Does Not Allow for Cross Sectional Dependence. $I(1) \rightarrow I(0)$. Intercept case.

τ	σ_η	T	N	MH _{1m} (\mathcal{K})	MH _{2m} (\mathcal{K})	MH _{3m} (\mathcal{K})	MH _{1m} (\mathcal{K}^*)	MH _{2m} (\mathcal{K}^*)	MH _{3m} (\mathcal{K}^*)	$\mathcal{M}\mathcal{M}_m^{1*}$	$\mathcal{M}\mathcal{M}_m^{2*}$	$\mathcal{M}\mathcal{M}_m^{3*}$	
0.3	0.10	50	25	0.030	0.037	0.021	0.266	0.284	0.391	0.160	0.182	0.249	
			50	0.021	0.036	0.012	0.427	0.428	0.593	0.206	0.218	0.355	
			100	0.008	0.011	0.005	0.639	0.674	0.839	0.319	0.358	0.549	
			100	25	0.002	0.003	0.001	0.891	0.913	0.968	0.682	0.736	0.875
				50	0.000	0.000	0.000	0.990	0.993	0.998	0.893	0.931	0.976
				100	0.000	0.000	0.000	1.000	1.000	1.000	0.983	0.990	1.000
		0.5	50	25	0.041	0.049	0.017	0.550	0.570	0.657	0.382	0.405	0.453
				50	0.032	0.047	0.007	0.792	0.810	0.890	0.563	0.598	0.703
				100	0.020	0.034	0.001	0.964	0.973	0.993	0.820	0.856	0.907
			100	25	0.095	0.127	0.004	0.995	0.996	0.998	0.975	0.981	0.986
				50	0.095	0.134	0.000	1.000	1.000	1.000	0.998	0.999	1.000
				100	0.089	0.141	0.000	1.000	1.000	1.000	1.000	1.000	1.000
0.7	50	25	0.220	0.247	0.120	0.623	0.639	0.636	0.593	0.619	0.599		
		50	0.326	0.370	0.134	0.858	0.860	0.860	0.814	0.841	0.819		
		100	0.437	0.512	0.174	0.984	0.985	0.981	0.967	0.978	0.975		
	100	25	0.789	0.817	0.429	0.996	0.999	0.996	0.995	0.997	0.996		
		50	0.941	0.950	0.595	1.000	1.000	1.000	1.000	1.000	1.000		
		100	0.999	0.998	0.758	1.000	1.000	1.000	1.000	1.000	1.000		
0.3	0.50	50	25	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			50	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			100	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	
			100	25	0.000	0.001	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.000	0.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000
		0.5	50	25	0.659	0.708	0.039	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.846	0.887	0.039	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.966	0.981	0.030	1.000	1.000	1.000	1.000	1.000	1.000
			100	25	0.910	0.921	0.195	1.000	1.000	1.000	1.000	1.000	1.000
				50	0.984	0.991	0.238	1.000	1.000	1.000	1.000	1.000	1.000
				100	0.999	0.999	0.238	1.000	1.000	1.000	1.000	1.000	1.000
0.7	50	25	0.999	0.999	0.940	1.000	1.000	1.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	0.992	1.000	1.000	1.000	1.000	1.000		
		100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
	100	25	1.000	1.000	0.994	1.000	1.000	1.000	1.000	1.000	1.000		
		50	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		
		100	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000		

Table 9: Size of Panel Tests When the Model Allows for Low and Strong Cross Sectional Dependence. Intercept case.

A/Low										
T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}
50	25	0.181	0.183	0.185	0.081	0.062	0.076	0.165	0.156	0.165
	50	0.199	0.190	0.194	0.083	0.068	0.076	0.210	0.204	0.200
	100	0.230	0.223	0.230	0.105	0.093	0.104	0.236	0.229	0.228
100	10	0.179	0.176	0.174	0.070	0.075	0.070	0.176	0.173	0.177
	25	0.179	0.173	0.181	0.083	0.091	0.085	0.185	0.180	0.188
	50	0.229	0.216	0.227	0.104	0.090	0.103	0.231	0.221	0.235
T	N	$MH_1^d(\mathcal{K})$	$MH_2^d(\mathcal{K})$	$MH_3^d(\mathcal{K})$	$MH_1^d(\mathcal{K}^*)$	$MH_2^d(\mathcal{K}^*)$	$MH_3^d(\mathcal{K}^*)$	MM^{1*}	MM^{2*}	MM^{3*}
50	25	0.185	0.186	0.185	0.082	0.063	0.076	0.170	0.157	0.166
	50	0.200	0.191	0.197	0.085	0.069	0.077	0.210	0.207	0.202
	100	0.235	0.223	0.233	0.108	0.099	0.106	0.236	0.231	0.227
100	25	0.179	0.174	0.184	0.095	0.083	0.085	0.187	0.180	0.189
	50	0.230	0.217	0.228	0.104	0.090	0.106	0.231	0.223	0.238
	100	0.236	0.226	0.240	0.112	0.103	0.112	0.241	0.239	0.247
T	N	$MH_{1m}(\mathcal{K})$	$MH_{2m}^d(\mathcal{K})$	$MH_{3m}(\mathcal{K})$	$MH_{1m}(\mathcal{K}^*)$	$MH_{2m}(\mathcal{K}^*)$	$MH_{3m}(\mathcal{K}^*)$	MM_m^{1*}	MM_m^{2*}	MM_m^{3*}
50	25	0.182	0.185	0.185	0.082	0.063	0.076	0.169	0.157	0.166
	50	0.200	0.191	0.196	0.085	0.069	0.077	0.210	0.207	0.202
	100	0.235	0.223	0.232	0.108	0.098	0.104	0.236	0.231	0.227
100	25	0.179	0.174	0.183	0.095	0.083	0.085	0.185	0.180	0.188
	50	0.229	0.217	0.227	0.104	0.090	0.105	0.231	0.222	0.236
	100	0.245	0.233	0.241	0.118	0.101	0.104	0.236	0.231	0.227
B/High										
T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}
50	25	0.419	0.409	0.381	0.230	0.205	0.200	0.402	0.381	0.391
	50	0.477	0.458	0.433	0.255	0.246	0.215	0.425	0.406	0.425
	100	0.542	0.523	0.521	0.321	0.331	0.310	0.405	0.401	0.434
100	25	0.365	0.360	0.334	0.220	0.210	0.180	0.355	0.331	0.320
	50	0.431	0.423	0.401	0.255	0.246	0.215	0.429	0.416	0.409
	100	0.512	0.502	0.521	0.321	0.331	0.310	0.472	0.481	0.478
T	N	$MH_1^d(\mathcal{K})$	$MH_2^d(\mathcal{K})$	$MH_3^d(\mathcal{K})$	$MH_1^d(\mathcal{K}^*)$	$MH_2^d(\mathcal{K}^*)$	$MH_3^d(\mathcal{K}^*)$	MM^{1*}	MM^{2*}	MM^{3*}
50	25	0.420	0.410	0.383	0.230	0.208	0.203	0.409	0.402	0.384
	50	0.478	0.458	0.434	0.257	0.246	0.215	0.443	0.406	0.427
	100	0.541	0.523	0.519	0.323	0.331	0.310	0.432	0.401	0.436
100	25	0.367	0.362	0.334	0.221	0.212	0.180	0.357	0.333	0.321
	50	0.428	0.414	0.380	0.208	0.251	0.232	0.399	0.391	0.372
	100	0.539	0.510	0.512	0.302	0.298	0.310	0.412	0.342	0.412
T	N	$MH_{1m}(\mathcal{K})$	$MH_{2m}(\mathcal{K})$	$MH_{3m}(\mathcal{K})$	$MH_{1m}(\mathcal{K}^*)$	$MH_{2m}(\mathcal{K}^*)$	$MH_{3m}(\mathcal{K}^*)$	MM_m^{1*}	MM_m^{2*}	MM_m^{3*}
50	25	0.420	0.410	0.383	0.230	0.205	0.203	0.407	0.382	0.392
	50	0.477	0.459	0.434	0.256	0.215	0.246	0.442	0.427	0.406
	100	0.540	0.522	0.519	0.323	0.324	0.334	0.432	0.418	0.423
100	25	0.365	0.360	0.334	0.221	0.211	0.180	0.356	0.332	0.320
	50	0.428	0.414	0.379	0.251	0.230	0.207	0.399	0.391	0.371
	100	0.539	0.513	0.509	0.320	0.331	0.310	0.410	0.401	0.410

Notes : Empirical sizes corresponding to a 5% nominal size.

Table 10: Size of DeFactored Panel Tests. Intercept case

T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^{1*}	\mathcal{M}^{2*}	\mathcal{M}^{3*}
50	25	0.066	0.067	0.066	0.065	0.063	0.063	0.059	0.056	0.061
	50	0.082	0.076	0.077	0.061	0.063	0.050	0.065	0.070	0.070
	100	0.080	0.077	0.080	0.059	0.058	0.059	0.071	0.068	0.074
100	25	0.064	0.062	0.063	0.070	0.072	0.068	0.066	0.065	0.065
	50	0.055	0.057	0.048	0.069	0.066	0.068	0.062	0.065	0.059
	100	0.057	0.055	0.501	0.068	0.067	0.057	0.063	0.062	0.053

T	N	$MH_1^d(\mathcal{K})$	$MH_2^d(\mathcal{K})$	$MH_3^d(\mathcal{K})$	$MH_1^d(\mathcal{K}^*)$	$MH_2^d(\mathcal{K}^*)$	$MH_3^d(\mathcal{K}^*)$	MM^{1*}	MM^{2*}	MM^{3*}
50	25	0.069	0.067	0.070	0.067	0.066	0.064	0.062	0.058	0.064
	50	0.074	0.076	0.079	0.062	0.065	0.050	0.075	0.075	0.073
	100	0.080	0.077	0.081	0.060	0.058	0.059	0.073	0.069	0.074
100	25	0.064	0.062	0.063	0.069	0.072	0.068	0.065	0.065	0.064
	50	0.055	0.057	0.048	0.068	0.066	0.067	0.059	0.064	0.059
	100	0.057	0.054	0.051	0.067	0.065	0.057	0.061	0.061	0.053

T	N	$MH_{1m}(\mathcal{K})$	$MH_{2m}(\mathcal{K})$	$MH_{3m}(\mathcal{K})$	$MH_{1m}(\mathcal{K}^*)$	$MH_{2m}(\mathcal{K}^*)$	$MH_{3m}(\mathcal{K}^*)$	MM_m^{1*}	MM_m^{2*}	MM_m^{3*}
50	25	0.069	0.067	0.069	0.067	0.065	0.063	0.061	0.057	0.063
	50	0.072	0.076	0.079	0.061	0.064	0.050	0.075	0.075	0.073
	100	0.080	0.077	0.081	0.059	0.058	0.059	0.061	0.069	0.074
100	25	0.064	0.064	0.064	0.070	0.072	0.069	0.067	0.066	0.066
	50	0.056	0.058	0.049	0.069	0.066	0.068	0.063	0.065	0.059
	100	0.057	0.055	0.051	0.068	0.067	0.057	0.063	0.062	0.053

Notes :Empirical sizes corresponding to a 5% nominal size.

Table 11: Power of Defactored Panel Tests. $I(0) \rightarrow I(1)$. Intercept case.

τ	σ_η	T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^1	\mathcal{M}^{2*}	\mathcal{M}^{3*}		
0.3	0.10	50	25	0.400	0.408	0.405	0.161	0.171	0.088	0.350	0.364	0.335		
			50	0.500	0.521	0.503	0.191	0.193	0.093	0.456	0.442	0.418		
			100	0.684	0.672	0.657	0.255	0.264	0.103	0.624	0.625	0.595		
		100	25	25	0.834	0.821	0.798	0.454	0.485	0.201	0.821	0.811	0.737	
				50	0.877	0.829	0.763	0.325	0.287	0.051	0.810	0.843	0.814	
				100	0.912	0.893	0.865	0.574	0.291	0.034	0.900	0.893	0.861	
			50	25	25	0.467	0.480	0.545	0.049	0.056	0.029	0.310	0.321	0.369
					50	0.688	0.661	0.740	0.053	0.057	0.041	0.134	0.140	0.154
					100	0.886	0.901	0.947	0.021	0.027	0.008	0.674	0.721	0.796
100	25	25	0.899	0.900	0.908	0.079	0.105	0.006	0.850	0.859	0.860			
		50	0.555	0.506	0.494	0.082	0.078	0.013	0.437	0.404	0.319			
		100	0.625	0.513	0.399	0.112	0.081	0.019	0.436	0.344	0.130			
	50	25	25	0.297	0.315	0.419	0.027	0.026	0.019	0.153	0.166	0.249		
			50	0.438	0.436	0.434	0.025	0.030	0.031	0.199	0.210	0.298		
			100	0.593	0.633	0.817	0.007	0.008	0.004	0.275	0.317	0.484		
100		25	25	0.849	0.864	0.943	0.006	0.008	0.002	0.635	0.688	0.822		
			50	0.966	0.966	0.969	0.006	0.006	0.003	0.834	0.873	0.946		
			100	0.998	0.978	0.999	0.053	0.052	0.072	0.919	0.956	0.997		
0.3	0.50	50	25	0.678	0.625	0.583	0.064	0.068	0.039	0.394	0.372	0.283		
			50	0.761	0.656	0.632	0.027	0.031	0.017	0.436	0.412	0.391		
			100	0.876	0.777	0.748	0.007	0.007	0.004	0.611	0.575	0.560		
		100	25	25	0.997	0.992	0.984	0.076	0.082	0.042	0.940	0.925	0.854	
				50	0.999	0.997	0.997	0.026	0.030	0.014	0.991	0.988	0.966	
				100	1.000	1.000	1.000	0.001	0.001	0.002	0.999	0.999	0.997	
			50	25	25	0.609	0.594	0.455	0.025	0.021	0.014	0.698	0.679	0.644
					50	0.996	0.997	0.996	0.030	0.042	0.033	0.993	0.993	0.993
					100	0.999	0.999	0.999	0.018	0.016	0.027	0.998	0.999	0.999
100	25	25	0.947	0.920	0.949	0.009	0.014	0.000	0.692	0.665	0.608			
		50	0.995	0.985	0.994	0.001	0.001	0.000	0.837	0.805	0.747			
		100	1.000	1.000	1.000	0.000	0.000	0.000	0.944	0.919	0.897			
	50	25	25	0.174	0.144	0.112	0.059	0.047	0.055	0.312	0.295	0.263		
			50	0.242	0.218	0.198	0.247	0.174	0.319	0.335	0.329	0.311		
			100	0.347	0.284	0.260	0.429	0.372	0.481	0.503	0.559	0.570		
100		25	25	0.528	0.502	0.464	0.010	0.009	0.009	0.439	0.426	0.374		
			50	0.717	0.677	0.533	0.033	0.030	0.033	0.665	0.649	0.614		
			100	0.947	0.928	0.882	0.063	0.053	0.099	0.853	0.834	0.806		

Table 12: Power of Defactored Modified Panel Tests. $I(0) \rightarrow I(1)$. Intercept case.

τ	σ_η	T	N	MH ₁ ^d (κ)	MH ₂ ^d (κ)	MH ₃ ^d (κ)	MH ₁ ^d (κ^*)	MH ₂ ^d (κ^*)	MH ₃ ^d (κ^*)	MM ^{1*}	MM ^{2*}	MM ^{3*}		
0.3	0.10	50	25	0.409	0.412	0.410	0.165	0.174	0.089	0.357	0.372	0.343		
			50	0.551	0.568	0.559	0.189	0.200	0.092	0.473	0.498	0.465		
			100	0.684	0.673	0.661	0.258	0.264	0.104	0.626	0.625	0.596		
		100	25	0.834	0.821	0.798	0.453	0.485	0.199	0.821	0.810	0.737		
				50	0.877	0.829	0.814	0.324	0.287	0.305	0.882	0.856	0.834	
			50	0.969	0.927	0.901	0.290	0.246	0.034	0.979	0.958	0.921		
				100	0.477	0.486	0.546	0.051	0.057	0.029	0.317	0.327	0.377	
			0.5	50	50	0.688	0.735	0.810	0.054	0.058	0.043	0.498	0.510	0.623
					100	0.888	0.902	0.947	0.021	0.027	0.008	0.682	0.712	0.796
100	25	0.898			0.900	0.907	0.079	0.105	0.005	0.848	0.858	0.859		
100	50	0.939		0.943	0.951	0.082	0.078	0.013	0.900	0.910	0.916			
	100	0.993		0.999	0.999	0.112	0.081	0.019	0.967	0.987	0.978			
	100	25		0.300	0.320	0.421	0.027	0.027	0.019	0.159	0.173	0.262		
0.7	50	50	0.453	0.487	0.578	0.025	0.030	0.032	0.212	0.245	0.324			
		100	0.597	0.637	0.819	0.008	0.008	0.004	0.284	0.319	0.485			
		100	25	0.849	0.864	0.943	0.006	0.008	0.002	0.634	0.687	0.820		
	100	50	0.966	0.966	0.969	0.006	0.006	0.003	0.834	0.873	0.946			
		100	0.999	0.999	0.999	0.053	0.052	0.072	0.910	0.923	0.991			
		100	25	0.689	0.626	0.584	0.064	0.070	0.039	0.407	0.383	0.289		
0.3	0.50	50	50	0.762	0.659	0.635	0.028	0.032	0.018	0.589	0.579	0.465		
			100	0.877	0.781	0.750	0.007	0.007	0.004	0.724	0.710	0.634		
			100	25	0.997	0.992	0.984	0.076	0.082	0.042	0.940	0.924	0.853	
		100	50	0.999	0.997	0.997	0.026	0.030	0.014	0.991	0.987	0.963		
			100	1.000	1.000	1.000	0.001	0.002	0.001	0.999	0.997	0.997		
			50	25	0.375	0.308	0.294	0.025	0.023	0.012	0.187	0.157	0.124	
			50	50	0.510	0.489	0.467	0.027	0.021	0.015	0.298	0.246	0.210	
				100	0.789	0.720	0.698	0.018	0.016	0.028	0.410	0.378	0.324	
				100	25	0.947	0.919	0.948	0.009	0.014	0.000	0.691	0.664	0.605
0.5	50	50	0.995	0.985	0.994	0.001	0.001	0.000	0.837	0.802	0.746			
		100	0.999	0.999	1.000	0.000	0.000	0.000	0.942	0.999	0.997			
		100	25	0.176	0.145	0.114	0.063	0.051	0.057	0.116	0.095	0.067		
	50	50	0.245	0.212	0.189	0.250	0.179	0.322	0.176	0.149	0.123			
		100	0.349	0.286	0.246	0.430	0.375	0.481	0.207	0.196	0.178			
		100	25	0.528	0.501	0.464	0.010	0.009	0.009	0.439	0.426	0.374		
0.7	50	50	0.712	0.700	0.678	0.033	0.030	0.031	0.598	0.587	0.534			
		100	0.887	0.854	0.819	0.063	0.053	0.098	0.810	0.778	0.730			

Table 13: Power of Defactored Modified Min Panel Tests. $I(0) \rightarrow I(1)$. Intercept case.

τ	σ_η	T	N	MH _{1m} (\mathcal{K})	MH _{2m} (\mathcal{K})	MH _{3m} (\mathcal{K})	MH _{1m} (\mathcal{K}^*)	MH _{2m} (\mathcal{K}^*)	MH _{3m} (\mathcal{K}^*)	MM _m ^{1*}	MM _m ^{2*}	MM _m ^{3*}	
0.3	0.10	50	25	0.403	0.410	0.407	0.162	0.173	0.089	0.354	0.368	0.340	
			50	0.548	0.563	0.552	0.187	0.200	0.092	0.471	0.491	0.462	
			100	0.684	0.672	0.661	0.258	0.264	0.103	0.626	0.625	0.595	
		100	25	0.836	0.821	0.799	0.458	0.488	0.201	0.824	0.811	0.737	
			50	0.877	0.830	0.764	0.325	0.287	0.051	0.884	0.858	0.833	
			100	0.968	0.929	0.902	0.295	0.246	0.034	0.971	0.967	0.921	
		0.5	50	25	0.474	0.482	0.546	0.050	0.057	0.029	0.315	0.325	0.375
				50	0.683	0.731	0.798	0.053	0.057	0.042	0.496	0.508	0.621
				100	0.887	0.901	0.947	0.021	0.027	0.008	0.678	0.712	0.796
100	25		0.900	0.900	0.909	0.080	0.107	0.006	0.850	0.859	0.860		
	50		0.941	0.943	0.953	0.082	0.079	0.013	0.900	0.910	0.916		
	100		0.996	0.999	0.999	0.113	0.082	0.019	0.967	0.987	0.978		
0.7	50	25	0.297	0.317	0.421	0.027	0.027	0.019	0.156	0.168	0.259		
		50	0.450	0.483	0.578	0.025	0.030	0.032	0.210	0.234	0.321		
		100	0.597	0.636	0.819	0.007	0.008	0.004	0.282	0.318	0.484		
	100	25	0.849	0.865	0.943	0.006	0.008	0.002	0.640	0.693	0.823		
		50	0.966	0.966	0.969	0.006	0.006	0.003	0.834	0.873	0.947		
		100	0.998	0.999	0.999	0.054	0.052	0.073	0.912	0.922	0.990		
0.3	0.50	50	25	0.682	0.625	0.583	0.064	0.069	0.039	0.404	0.378	0.287	
			50	0.762	0.657	0.635	0.028	0.032	0.017	0.585	0.572	0.463	
			100	0.876	0.779	0.748	0.007	0.007	0.004	0.785	0.708	0.635	
		100	25	0.997	0.993	0.984	0.076	0.082	0.042	0.941	0.928	0.859	
			50	0.999	0.997	0.997	0.027	0.030	0.015	0.991	0.988	0.968	
			100	1.000	1.000	1.000	0.001	0.002	0.001	0.999	0.999	0.997	
		0.5	50	25	0.369	0.305	0.293	0.025	0.023	0.012	0.186	0.156	0.123
				50	0.500	0.482	0.466	0.026	0.021	0.014	0.297	0.246	0.209
				100	0.778	0.712	0.697	0.018	0.016	0.027	0.410	0.298	0.323
100	25		0.997	0.993	0.984	0.076	0.082	0.042	0.941	0.928	0.856		
	50		0.995	0.985	0.994	0.001	0.001	0.000	0.841	0.807	0.750		
	100		0.999	0.999	1.000	0.000	0.000	0.000	0.945	0.919	0.898		
0.7	50	25	0.175	0.144	0.114	0.062	0.050	0.057	0.116	0.095	0.067		
		50	0.244	0.211	0.200	0.248	0.178	0.322	0.036	0.176	0.123		
		100	0.347	0.285	0.265	0.429	0.374	0.481	0.207	0.210	0.179		
	100	25	0.529	0.503	0.464	0.010	0.010	0.009	0.441	0.426	0.375		
		50	0.711	0.702	0.678	0.033	0.031	0.034	0.601	0.587	0.374		
		100	0.875	0.857	0.818	0.064	0.053	0.101	0.810	0.778	0.730		

Table 14: Power of Defactored Panel Tests. $I(1) \rightarrow I(0)$. Intercept case.

τ	σ_η	T	N	$H_1(\mathcal{K})$	$H_2(\mathcal{K})$	$H_3(\mathcal{K})$	$H_1(\mathcal{K}^*)$	$H_2(\mathcal{K}^*)$	$H_3(\mathcal{K}^*)$	\mathcal{M}^1	\mathcal{M}^{2*}	\mathcal{M}^{3*}	
0.3	0.10	50	25	0.039	0.044	0.030	0.252	0.260	0.366	0.154	0.165	0.202	
			50	0.039	0.037	0.031	0.398	0.403	0.550	0.184	0.199	0.265	
			100	0.009	0.014	0.009	0.577	0.596	0.757	0.244	0.284	0.457	
		100	25	25	0.003	0.009	0.001	0.851	0.875	0.954	0.616	0.681	0.817
				50	0.000	0.004	0.002	0.953	0.955	0.964	0.839	0.883	0.939
				100	0.070	0.055	0.079	0.993	0.995	0.999	0.934	0.971	0.989
			50	25	0.045	0.057	0.026	0.422	0.431	0.540	0.300	0.315	0.368
				50	0.062	0.071	0.049	0.632	0.657	0.687	0.523	0.543	0.589
				100	0.030	0.043	0.006	0.878	0.888	0.938	0.682	0.725	0.812
100	25	25	0.058	0.087	0.009	0.921	0.923	0.924	0.868	0.884	0.888		
		50	0.067	0.062	0.010	0.965	0.967	0.970	0.921	0.943	0.951		
		100	0.101	0.081	0.014	0.978	0.999	0.998	0.987	0.999	0.999		
	50	25	0.138	0.155	0.095	0.386	0.382	0.384	0.346	0.365	0.357		
		50	0.120	0.117	0.087	0.510	0.508	0.509	0.472	0.483	0.476		
		100	0.264	0.304	0.114	0.667	0.647	0.634	0.638	0.633	0.600		
100	25	25	0.444	0.452	0.203	0.815	0.808	0.768	0.793	0.773	0.704		
		50	0.297	0.274	0.053	0.869	0.827	0.743	0.910	0.890	0.838		
		100	0.280	0.222	0.022	0.930	0.910	0.887	0.987	0.975	0.929		
	50	25	0.057	0.052	0.073	0.151	0.126	0.105	0.089	0.083	0.050		
		50	0.205	0.205	0.339	0.226	0.212	0.146	0.146	0.139	0.112		
		100	0.409	0.341	0.448	0.375	0.301	0.170	0.254	0.180	0.171		
100	25	25	0.010	0.012	0.010	0.523	0.500	0.467	0.427	0.415	0.366		
		50	0.023	0.021	0.024	0.718	0.693	0.640	0.578	0.561	0.470		
		100	0.055	0.036	0.087	0.836	0.825	0.798	0.741	0.721	0.653		
	50	25	0.034	0.035	0.015	0.357	0.313	0.294	0.215	0.195	0.129		
		50	0.021	0.021	0.015	0.756	0.678	0.633	0.374	0.328	0.223		
		100	0.027	0.024	0.037	0.881	0.849	0.819	0.532	0.471	0.378		
100	25	25	0.024	0.028	0.003	0.955	0.920	0.948	0.706	0.685	0.642		
		50	0.002	0.003	0.000	0.995	0.991	0.997	0.836	0.805	0.745		
		100	0.000	0.000	0.000	1.000	1.000	1.000	0.953	0.936	0.915		
	50	25	0.064	0.070	0.037	0.698	0.614	0.583	0.394	0.363	0.281		
		50	0.022	0.018	0.007	0.747	0.774	0.719	0.525	0.491	0.361		
		100	0.007	0.007	0.006	0.848	0.737	0.750	0.674	0.561	0.444		
100	25	0.071	0.082	0.034	0.997	0.993	0.989	0.940	0.926	0.850			
	50	0.117	0.017	0.021	0.998	0.998	0.997	0.992	0.988	0.969			
	100	0.003	0.003	0.003	1.000	1.000	1.000	0.998	0.997	0.994			

Table 15: Power of Defactored Modified Panel Tests. $I(1) \rightarrow I(0)$. Intercept case.

τ	σ_η	T	N	MH ₁ ^d (κ)	MH ₂ ^d (κ)	MH ₃ ^d (κ)	MH ₁ ^d (κ^*)	MH ₂ ^d (κ^*)	MH ₃ ^d (κ^*)	MM ^{1*}	MM ^{2*}	MM ^{3*}	
0.3	0.10	50	25	0.040	0.045	0.032	0.260	0.269	0.373	0.159	0.168	0.210	
			50	0.039	0.039	0.032	0.401	0.414	0.567	0.192	0.134	0.179	
			100	0.010	0.015	0.009	0.583	0.599	0.760	0.250	0.286	0.460	
		100	25	0.003	0.010	0.001	0.852	0.875	0.954	0.622	0.686	0.823	
			50	0.000	0.004	0.002	0.953	0.955	0.964	0.839	0.884	0.939	
			100	0.071	0.056	0.080	0.992	0.995	0.999	0.935	0.972	0.989	
		0.5	50	25	0.048	0.058	0.026	0.428	0.438	0.547	0.311	0.324	0.375
				50	0.065	0.071	0.051	0.639	0.663	0.701	0.542	0.561	0.610
				100	0.030	0.043	0.006	0.880	0.889	0.939	0.688	0.729	0.815
100	25		0.059	0.088	0.009	0.923	0.923	0.915	0.885	0.886	0.890		
	50		0.068	0.063	0.010	0.967	0.967	0.971	0.991	0.959	0.953		
	100		0.102	0.082	0.014	0.610	0.518	0.389	0.443	0.999	0.999		
0.7	50	25	0.142	0.156	0.097	0.391	0.388	0.390	0.355	0.373	0.362		
		50	0.120	0.118	0.087	0.515	0.510	0.511	0.478	0.490	0.478		
		100	0.269	0.307	0.114	0.669	0.647	0.636	0.640	0.633	0.603		
	100	25	0.449	0.457	0.207	0.819	0.811	0.769	0.795	0.777	0.705		
		50	0.298	0.277	0.054	0.869	0.828	0.799	0.797	0.790	0.780		
		100	0.280	0.222	0.022	0.930	0.911	0.876	0.878	0.901	0.864		
0.3	0.50	50	25	0.059	0.056	0.074	0.155	0.129	0.105	0.091	0.086	0.055	
			50	0.264	0.208	0.339	0.229	0.214	0.146	0.151	0.142	0.103	
			100	0.409	0.343	0.449	0.378	0.302	0.171	0.256	0.184	0.173	
		100	25	0.010	0.013	0.010	0.527	0.502	0.468	0.428	0.416	0.367	
			50	0.023	0.022	0.024	0.722	0.695	0.641	0.579	0.562	0.470	
			100	0.057	0.037	0.088	0.037	0.025	0.002	0.742	0.722	0.652	
		0.5	50	25	0.039	0.038	0.016	0.359	0.317	0.296	0.224	0.199	0.132
				50	0.028	0.024	0.019	0.759	0.695	0.631	0.379	0.325	0.228
				100	0.029	0.024	0.038	0.891	0.854	0.819	0.532	0.473	0.378
100	25		0.024	0.028	0.003	0.957	0.922	0.951	0.716	0.687	0.646		
	50		0.002	0.003	0.000	0.995	0.991	0.997	0.842	0.810	0.746		
	100		0.000	0.000	0.000	1.000	1.000	1.000	0.999	0.999	0.999		
0.7	50	25	0.064	0.072	0.037	0.702	0.625	0.590	0.407	0.373	0.287		
		50	0.021	0.021	0.015	0.758	0.682	0.637	0.545	0.513	0.365		
		100	0.007	0.008	0.006	0.850	0.737	0.723	0.678	0.643	0.445		
	100	25	0.072	0.082	0.035	0.997	0.993	0.989	0.944	0.927	0.855		
		50	0.017	0.021	0.009	0.998	0.998	0.997	0.992	0.989	0.971		
		100	0.003	0.003	0.003	1.000	1.000	1.000	0.998	0.997	0.995		

Table 16: Power of Defactored Modified Min Panel Tests. $I(1) \rightarrow I(0)$. Intercept case.

τ	σ_η	T	N	MH _{1m} (\mathcal{K})	MH _{2m} (\mathcal{K})	MH _{3m} (\mathcal{K})	MH _{1m} (\mathcal{K}^*)	MH _{2m} (\mathcal{K}^*)	MH _{3m} (\mathcal{K}^*)	MM _m ^{1*}	MM _m ^{2*}	MM _m ^{3*}	
0.3	0.10	50	25	0.040	0.045	0.032	0.256	0.263	0.371	0.157	0.167	0.209	
			50	0.039	0.038	0.032	0.401	0.407	0.558	0.189	0.201	0.271	
			100	0.010	0.015	0.009	0.581	0.597	0.760	0.249	0.284	0.460	
		100	25	0.003	0.009	0.001	0.852	0.875	0.954	0.617	0.682	0.819	
			50	0.000	0.004	0.002	0.953	0.955	0.964	0.839	0.883	0.939	
			100	0.070	0.055	0.080	0.994	0.996	0.999	0.934	0.971	0.989	
		0.5	50	25	0.047	0.057	0.026	0.424	0.432	0.545	0.307	0.321	0.373
				50	0.063	0.071	0.051	0.634	0.659	0.681	0.529	0.551	0.591
				100	0.030	0.043	0.006	0.879	0.889	0.938	0.685	0.727	0.814
100	25		0.058	0.087	0.009	0.923	0.923	0.924	0.870	0.884	0.889		
	50		0.068	0.063	0.010	0.972	0.972	0.970	0.920	0.942	0.949		
	100		0.101	0.082	0.014	0.991	0.991	0.970	0.986	0.999	0.999		
0.7	50	25	0.140	0.156	0.097	0.389	0.386	0.388	0.354	0.369	0.361		
		50	0.120	0.118	0.087	0.515	0.519	0.512	0.479	0.489	0.481		
		100	0.269	0.305	0.114	0.669	0.647	0.635	0.640	0.633	0.602		
	100	25	0.444	0.452	0.203	0.815	0.811	0.769	0.795	0.776	0.704		
		50	0.297	0.274	0.053	0.869	0.827	0.789	0.910	0.890	0.838		
		100	0.280	0.222	0.022	0.930	0.910	0.876	0.987	0.975	0.929		
0.3	0.50	50	25	0.059	0.054	0.074	0.154	0.129	0.105	0.090	0.086	0.055	
			50	0.263	0.208	0.339	0.229	0.218	0.146	0.146	0.137	0.116	
			100	0.409	0.342	0.449	0.378	0.302	0.171	0.255	0.184	0.172	
		100	25	0.010	0.012	0.010	0.523	0.501	0.467	0.428	0.415	0.366	
			50	0.023	0.022	0.024	0.718	0.695	0.640	0.579	0.562	0.480	
			100	0.055	0.036	0.088	0.836	0.821	0.799	0.744	0.722	0.663	
		0.5	50	25	0.037	0.037	0.016	0.359	0.316	0.295	0.220	0.196	0.131
				50	0.028	0.024	0.019	0.756	0.679	0.631	0.381	0.329	0.225
				100	0.029	0.024	0.037	0.882	0.850	0.829	0.551	0.471	0.379
100	25		0.024	0.028	0.003	0.956	0.920	0.949	0.709	0.686	0.642		
	50		0.002	0.003	0.000	0.995	0.991	0.997	0.839	0.806	0.745		
	100		0.000	0.000	0.000	1.000	1.000	1.000	0.953	0.936	0.915		
0.7	50	25	0.064	0.071	0.037	0.699	0.621	0.584	0.402	0.368	0.283		
		50	0.021	0.021	0.015	0.758	0.679	0.633	0.535	0.499	0.374		
		100	0.007	0.008	0.006	0.850	0.737	0.722	0.689	0.573	0.459		
	100	25	0.072	0.082	0.035	0.997	0.993	0.989	0.940	0.926	0.851		
		50	0.017	0.021	0.009	0.998	0.998	0.997	0.992	0.989	0.969		
		100	0.003	0.003	0.003	1.000	1.000	1.000	0.998	0.997	0.995		

Table 17: Inflation rate 1970.1-2007.3

No cross-sectional dependence adjusted tests					
Test Statistics	test – value (p-values)	Test Statistics	test – value (p-values)	Test Statistics	test – value (p-values)
$H_1(\mathcal{K})$	1.664 (0.048)	$H_1(\mathcal{K}^*)$	408.373 (0.000)	\mathcal{M}^{1*}	352.873 (0.000)
$H_2(\mathcal{K})$	-0.876 (0.809)	$H_2(\mathcal{K}^*)$	246.559 (0.000)	\mathcal{M}^{2*}	222.889 (0.000)
$H_3(\mathcal{K})$	1.415 (0.079)	$H_3(\mathcal{K}^*)$	549.627 (0.000)	\mathcal{M}^{3*}	436.048 (0.000)
$MH_1^d(\mathcal{K})$	1.731 (0.023)	$MH_1^d(\mathcal{K}^*)$	413.061 (0.000)	MM^{1*}	357.636 (0.000)
$MH_2^d(\mathcal{K})$	-0.844 (0.801)	$MH_2^d(\mathcal{K}^*)$	248.708 (0.000)	MM^{2*}	225.168 (0.000)
$MH_3^d(\mathcal{K})$	1.474 (0.070)	$MH_3^d(\mathcal{K}^*)$	557.550 (0.000)	MM^{3*}	443.112 (0.000)
$MH_{1_m}(\mathcal{K})$	1.667 (0.048)	$MH_{1_m}(\mathcal{K}^*)$	408.617 (0.000)	MM_m^{1*}	353.187 (0.000)
$MH_{2_m}(\mathcal{K})$	-0.874 (0.809)	$MH_{2_m}(\mathcal{K}^*)$	246.677 (0.000)	MM_m^{2*}	223.052 (0.000)
$MH_{3_m}(\mathcal{K})$	1.418 (0.078)	$MH_{3_m}(\mathcal{K}^*)$	550.015 (0.000)	MM_m^{3*}	436.187 (0.000)
Break Date : 1977.3					
Cross-correlation dependence adjusted tests					
Test Statistics	test – value (p-values)	Test Statistics	test – value (p-values)	Test Statistics	test – value (p-values)
$H_1(\mathcal{K})$	-0.846 (0.801)	$H_1(\mathcal{K}^*)$	55.396 (0.000)	\mathcal{M}^{1*}	46.288 (0.000)
$H_2(\mathcal{K})$	-3.103 (0.999)	$H_2(\mathcal{K}^*)$	39.024 (0.000)	\mathcal{M}^{2*}	33.092 (0.000)
$H_3(\mathcal{K})$	-0.946 (0.846)	$H_3(\mathcal{K}^*)$	70.435 (0.000)	\mathcal{M}^{3*}	54.639 (0.000)
$MH_1^d(\mathcal{K})$	-0.829 (0.797)	$MH_1^d(\mathcal{K}^*)$	55.649 (0.000)	MM^{1*}	46.547 (0.000)
$MH_2^d(\mathcal{K})$	-3.097 (0.999)	$MH_2^d(\mathcal{K}^*)$	39.161 (0.000)	MM^{2*}	33.236 (0.000)
$MH_3^d(\mathcal{K})$	-0.933 (0.824)	$MH_3^d(\mathcal{K}^*)$	70.830 (0.000)	MM^{3*}	54.994 (0.000)
$MH_{1_m}(\mathcal{K})$	-0.844 (0.801)	$MH_{1_m}(\mathcal{K}^*)$	55.419 (0.000)	MM_m^{1*}	46.321 (0.000)
$MH_{2_m}(\mathcal{K})$	-3.102 (0.999)	$MH_{2_m}(\mathcal{K}^*)$	33.113 (0.000)	MM_m^{2*}	33.113 (0.000)
$MH_{3_m}(\mathcal{K})$	-0.945 (0.828)	$MH_{3_m}(\mathcal{K}^*)$	70.468 (0.000)	MM_m^{3*}	54.682 (0.000)
Break Date : 1980.4					