

Panel Data Methods using Stata

4A: Panels - Unit Root Tests

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Introduction

- Test whether y_{it} is nonstationary, specifically a unit root
 - ▶ Does $y_{it} = y_{i,t-1} + \varepsilon_{it}$?
- In pure time series case, if unit root then test statistics are not normal
 - ▶ instead functions of Wiener processes / Brownian Motion
 - ▶ and have low power: cannot easily distinguish between $y_{it} = 0.9y_{i,t-1} + \varepsilon_{it}$ and $y_{it} = y_{i,t-1} + \varepsilon_{it}$.
- Panel data may give more power
 - ▶ again nonstandard asymptotic theory + intercept(?) + trend(?)
 - ▶ in particular, may need to recenter and rescale test statistic
 - ▶ but then usually normally distributed
- But - there are many different tests of unit roots in panels
 - ▶ $N \rightarrow \infty$ or $T \rightarrow \infty$ or both; heterogeneity; cross-section correlation
 - ▶ for panel case there is no general clear best test
 - ▶ good idea to read the original paper before using a test
 - ▶ and rely on economic theory to give likely model for y_{it} .

Outline

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2. Time Series Unit Root Tests Consequences

- y_t has a unit root if need to first difference to get stationary process
 - ▶ examples: random walk with drift and random walk without drift.
- Random walk without drift
 - ▶ If $y_t = y_{t-1} + \varepsilon_t$ then $y_t = (y_{t-2} + \varepsilon_{t-1}) + \varepsilon_t = \dots$
 so $y_t = y_0 + \sum_{s=0}^t \varepsilon_s$
 - ★ shocks are permanent as $\Delta \varepsilon_0 = 1 \implies \Delta y_t = 1$
- Compared to AR(1)
 - ▶ Versus $y_t = \rho y_{t-1} + \varepsilon_t$, $\rho < 1$, then $y_t = \rho(\rho y_{t-2} + \varepsilon_{t-1}) + \varepsilon_t = \dots$
 so $y_t = \rho^t y_0 + \sum_{s=0}^t \rho^{t-s} \varepsilon_s$
 - ▶ shocks disappear as $\Delta \varepsilon_0 = 1 \implies \Delta y_t = \rho^t \rightarrow 0$.
- Fundamental result: shocks are permanent if unit root.
- Random walk with drift: $y_t = \alpha + y_{t-1} + \varepsilon_t$
 - ▶ implies $y_t = y_0 + \alpha t + \sum_{s=0}^t \varepsilon_s$
 - ▶ so induces a linear time trend as well as being nonstationary.

Dickey-Fuller Test

- Fuller (1976). Rewrite the model

$$\begin{aligned}y_t &= \rho y_{t-1} + \varepsilon_t \\ \Delta y_t &= (\rho - 1)y_{t-1} + \varepsilon_t \\ &= \phi y_{t-1} + \varepsilon_t.\end{aligned}$$

- ▶ unit root if $\phi = 0$ and stationary if $\phi < 0$.
- Dickey Fuller test: test $H_0 : \phi = 0$ against $H_a : \phi < 0$ in

$$\Delta y_t = \phi y_{t-1} + \varepsilon_t$$

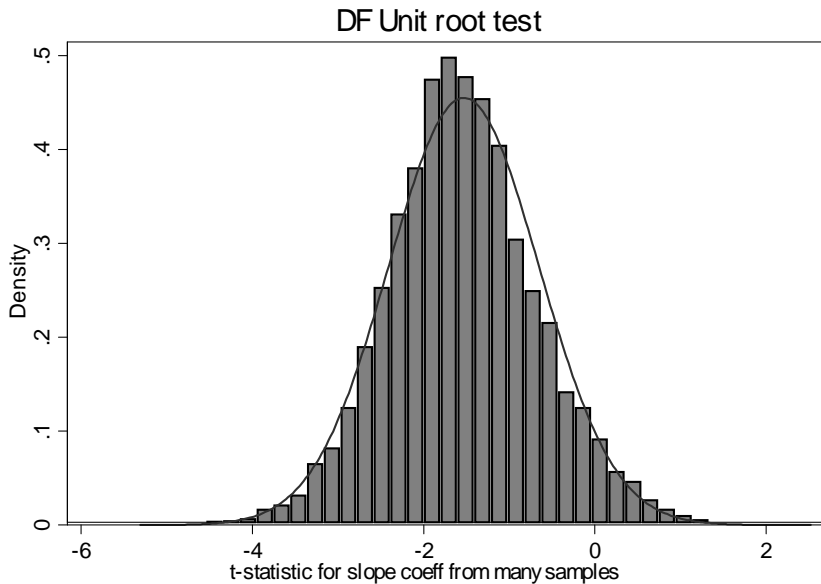
- Obvious approach is to test for $\phi = 0$ in regress y_t on y_{t-1}
 - ▶ But under $H_0 : \phi = 0$ test statistic has nonstandard distribution
 - ▶ use special tables not t tables
 - ▶ Stata command `dfuller`

Simulation

- DGP: $y_t = y_{t-1} + \varepsilon_t$ where ε_t iid $N[0, 1]$ and $T = 50$.
- Stata code

```
set seed 10101
program unitroot, rclass
drop _all
set obs 50
generate time = _n
tsset time
generate epsilon = rnormal(0,1)
generate yrwalk = 0
replace yrwalk = L.yrwalk + epsilon if time > 1
regress D.yrwalk L.yrwalk
return scalar b2 = _b[L.yrwalk]
return scalar se2 = _se[L.yrwalk]
return scalar t2 = (_b[L.yrwalk]-0)/_se[L.yrwalk]
end
```

- Do 10,000 simulations and summarize results
 quietly simulate b2=r(b2) se2=r(se2) t2=r(t2), ///
 reps(1000) saving(unitroot, replace) nolegend nodots: unitroot
 summarize b2 se2 t2
 mean b2 se2 t2
 histogram t2, normal title("DF Unit root test Case 2")
 centile t2, centile(1 2.5 5 10 90 95 97.5 99)
- Centiles for t-statistic: 1 2.5 5 10 90 95 97.5 99
 - ▶ $N[0,1]$: -2.57, -1.96, -1.64, -1.28, 1.28, 1.64, 1.96, 2.57
 - ▶ Simulation: -3.56, -3.22, -2.92, -2.60, -0.40, -0.02, 0.30, 0.63
 - ▶ DF Tables: -3.58, -3.22, -2.93, -2.60, -0.40, -0.03, 0.29, 0.66
- Definitely not normal or $T(48)$.



Dickey-Fuller Tests (continued)

- Complication 1: There are two test statistics

- ▶ The t-statistic $\hat{\phi}/se(\hat{\phi})$
- ▶ The z-statistic $T\hat{\phi}$

★ we can use the z-statistic in addition to the t statistic because when $\phi = 0$ its distribution does not depend on unknown parameters .

- Complication 2: Under H_0 these statistics have nonstandard distributions - functionals of Wiener processes (Brownian motion)

- ▶ asymptotic results use a functional central limit theorem that does not require ε_t to be normally distributed
- ▶ letting $\Psi = \Psi(r)$ denote a detrended Wiener process

$$\star T\hat{\phi} \xrightarrow{d} \frac{\int \Psi(r)d\Psi(r)}{\int \Psi(r)^2 dr} \quad \text{and} \quad \frac{\hat{\phi}}{se(\hat{\phi})} \xrightarrow{d} \frac{\int \Psi(r)d\Psi(r)}{\sqrt{\int \Psi(r)^2 dr}}$$

- ▶ finite sample results assume ε_t is normally distributed

Dickey-Fuller Tests (continued)

- Complication 3: These distributions change when deterministic terms are added to the regression, so $y_t = \mathbf{d}'_t \boldsymbol{\gamma} + \rho y_{t-1} + \varepsilon_t$ where \mathbf{d}_t is constant plus possibly trend
 - ▶ Case 1: $y_t = \rho y_{t-1} + \varepsilon_t$ estimated and $H_0 : y_t = y_{t-1} + \varepsilon_t$
 - ▶ Case 2: $y_t = \alpha + \rho y_{t-1} + \varepsilon_t$ estimated and $H_0 : y_t = y_{t-1} + \varepsilon_t$
 - ▶ Case 3: $y_t = \alpha + \rho y_{t-1} + \varepsilon_t$ estimated and $H_0 : y_t = \delta + y_{t-1} + \varepsilon_t$
 - ▶ Case 4: $y_t = \alpha + \delta t + \rho y_{t-1} + \varepsilon_t$ estimated and $H_0 : y_t = \delta + y_{t-1} + \varepsilon_t$
- Most often use cases 2 and 4
 - ▶ case 3 unrealistic as H_0 model $y_t = y_0 + \delta t + (\varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_1)$ has time trend
 - ▶ lose power as move from case 1 to case 2 to case 4
- Stata `dfuller` uses the t-statistic $\hat{\phi} / se(\hat{\phi})$
 - ▶ cases 1-4 are, respectively, options (1) noconstant, (2) the default, (3) drift, (4) trend

Unit Root Tests for Time Series

- Complication 4: ε_t is serially correlated
- Dickey and Fuller (1979) - Augmented Dickey-Fuller (ADF) test
 - ▶ easiest to understand
 - ▶ add lags until ε_t is serially uncorrelated and use original DF tables
 - ▶ so estimate $\Delta y_t = \mathbf{d}'_t \gamma + \phi y_{t-1} + \sum_{k=1}^K \gamma_k \Delta y_{t-k} + \varepsilon_t$.
 - ▶ t-statistic is $\hat{\phi} / \text{se}(\hat{\phi})$ and z-statistic is $T\hat{\phi} / (1 - \hat{\gamma}_1 - \dots - \hat{\gamma}_T)$
 - ▶ determine # lags using AIC or BIC or by specifying K , doing regular t-test on $\hat{\gamma}_K$, then $\hat{\gamma}_{K-1}$, then, ...
- Phillips and Perron (1988) test
 - ▶ correct the original (unaugmented) DF test for ε_t serially correlated
 - ▶ then use the original DF tables
- KPSS (1992) test of $H_0 : \phi < 0$ against $H_a : \phi = 0$
 - ▶ OLS regress $y_t = \mathbf{d}'_t \gamma + u_t$, $\hat{u}_t = y_t - \mathbf{d}'_t \hat{\gamma}$, $S_t = \sum_{s=1}^t \hat{u}_s^2$
 - ▶ $LM = \sum_{t=1}^T S_t^2 / (T^2 f_0)$; f_0 is estimate of the long-run variance of \hat{u}_t
- Other tests include DF-GLS test coming next.

Unit Root Tests for Time Series (continued)

- Elliott, Rothenberg & Stock (1996) test - DF-GLS test
 - ▶ this is viewed as the best unit root test
 - ▶ most powerful test against local alternatives $H_0 : \phi = cT$ for fixed c
 - ▶ Stata command `dfgls`
- $y_t = \mathbf{d}'_t \boldsymbol{\gamma} + \rho y_{t-1} + \sum_{k=1}^K \gamma_k \Delta y_{t-k} + \varepsilon_t$; $d_t = 1$ or $\mathbf{d}_t = (1, t)$
 - ▶ DF test jointly estimates $\boldsymbol{\gamma}$ and ρ and $\gamma'_k s$
 - ▶ when $d_t = 1$ equivalently ρ (and $\gamma'_k s$) from OLS of $(y_t - \bar{y})$ on $\rho(y_{t-1} - \bar{y}_{t-1})$ and $\Delta y'_{t-k} s$.
- 1. Instead (GLS step) estimate $\boldsymbol{\gamma}$ by $\hat{\boldsymbol{\gamma}}$ from OLS in

$$(y_t - \bar{\rho} y_{t-1}) = (\mathbf{d}_t - \bar{\rho} \mathbf{d}_t)' \boldsymbol{\gamma} + v_t$$
 - ▶ where $\bar{\rho} = 1 - 7/T$ for $d_t = 1$ and $\bar{\rho} = 1 - 13.5/T$ for $\mathbf{d}_t = (1, t)$
- 2. ϕ (and $\gamma'_k s$) by OLS in $\Delta y_t^* = \phi y_{t-1}^* + \sum_{k=1}^K \gamma_k \Delta y_{t-k}^* + w_t$ where $y_t^* = y_t - \mathbf{d}'_t \hat{\boldsymbol{\gamma}}$
- 3. Use t-statistic for $\hat{\phi} = 0$
 - ▶ for $d_t = 1$ use DF tables (case 2)
 - ▶ for $\mathbf{d}_t = (1, t)$ use tables in Elliott et al.

3. Panel Example: Real Exchange Rates and PPP

- Real Exchange Rate annual data 1970-2003 on 151 countries.
 - ▶ Balanced with $T = 34$ and $N = 151$
 - ▶ Sometimes restrict to OECD $N = 27$ or G7 $N = 6$
 - ▶ USA is the reference country
 - ▶ Data from Stata Manual [XT] xtunitroot.
- Real exchange rate = nominal exchange rate \times (price in home country / price in foreign country)

$$\lambda = EP^* / P$$

$$\ln \lambda = \ln E + \ln P^* - \ln P$$

- Purchasing power parity says **no unit root (instead I(0))**
 - ▶ reason: real exchange rate is mean-reverting
 - ▶ essentially even if $\ln E$, $\ln P^*$ and $\ln P$ are I(1)
 - ★ they are cointegrated so that $\ln E + \ln P^* - \ln P$ is I(0).
- Also there is no reason to believe there is a trend
 - ▶ so unit root tests here do not include a trend.

Data Description

```
. describe
```

```
Contains data from pennxrate.dta
```

```
obs:      5,134
vars:      10                               15 Oct 2012 22:17
size:      200,226 (99.7% of memory free)  (_dta has notes)
```

variable name	storage type	display format	value label	variable label
country	str3	%9s		
year	int	%8.0g		
xrate	float	%9.0g		Nominal exchange rate
ppp	float	%9.0g		PWT Purchasing Power Parity index
id	float	%9.0g		group(country)
capt	float	%9.0g		
realxrate	float	%9.0g		Real exchange rate
lnrxrate	float	%9.0g		Log real exchange rate
oecd	byte	%8.0g		
g7	byte	%8.0g		

```
Sorted by:  id  year
```

```
. * Summarize data
. sum
```

Variable	Obs	Mean	Std. Dev.	Min	Max
country	0				
year	5134	1986.5	9.811664	1970	2003
xrate	5134	1412.495	36112.62	1.67e-12	1507226
ppp	5134	664.0886	18089.86	1.54e-12	852999.1
id	5134	94.36424	52.87679	1	188
capt	5134	34	0	34	34
realxrate	5134	.632419	.5703317	.005522	11.02636
lnrxrate	5134	-.6375265	.5680524	-5.19902	2.400289
oecd	5134	.1788079	.3832287	0	1
g7	5134	.0397351	.1953552	0	1
lnrxrate	5134	1.856842	4.194947	-27.11847	14.22578
lnppp	5134	1.197976	4.097343	-27.19693	13.65651

Panel Summary

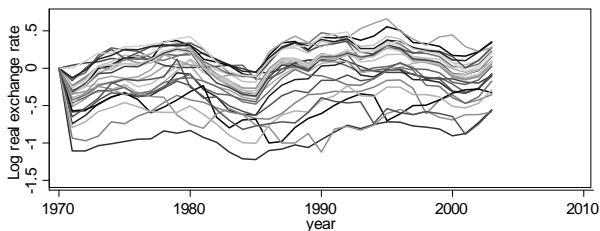
- Key variables have both within and between variation
 - ▶ Inrxrate is key variable
 - ▶ lnppp is ln price ratio and lnrxrate is ln nominal exchange rate

```
. xtsum lnrxrate lnrxrate lnppp
```

Variable		Mean	Std. Dev.	Min	Max	Observations
lnrxrate	overall	-.6375265	.5680524	-5.19902	2.400289	N = 5134
	between		.4317351	-1.560863	.7197967	n = 151
	within		.370792	-5.137827	1.797904	T = 34
lnrxrate	overall	1.856842	4.194947	-27.11847	14.22578	N = 5134
	between		3.451942	-16.00139	7.287732	n = 151
	within		2.399643	-12.46177	23.86295	T = 34
lnppp	overall	1.197976	4.097343	-27.19693	13.65651	N = 5134
	between		3.403155	-16.52128	6.778892	n = 151
	within		2.29809	-13.06217	22.3583	T = 34

Line Charts (for OECD)

- Seem to move together against the reference USA
 - ▶ `xtline lnrxrate if oecd == 1, overlay`
 - ▶ control for cross-country correlation by including $\bar{y}_t = \frac{1}{T} \sum_{i=1}^N y_{it}$.



— id = 9/id = 85	— id = 10/id = 88
— id = 13/id = 95	— id = 31/id = 105
— id = 32/id = 112	— id = 47/id = 126
— id = 53/id = 127	— id = 56/id = 129
— id = 58/id = 137	— id = 61/id = 140
— id = 63/id = 159	— id = 69/id = 171
— id = 77	— id = 80
— id = 83	

4. Time Series Unit Root Tests for GBR-USA

- Consider ADF test for a single time series
 - ▶ Inrxrate is GBR-USA log real exchange rate

```
. * ADF test with 5 lags
. dfuller lnrxrate if country == "GBR", lags(5)
```

Augmented Dickey-Fuller test for unit root Number of obs = 28

	Test Statistic	Interpolated Dickey-Fuller		
		1% Critical Value	5% Critical Value	10% Critical Value
z(t)	-1.632	-3.730	-2.992	-2.626

Mackinnon approximate p-value for z(t) = 0.4666

- Do not reject H_0 as $p > .05$
 - ▶ conclude there is a unit root!
- But this is very black box.

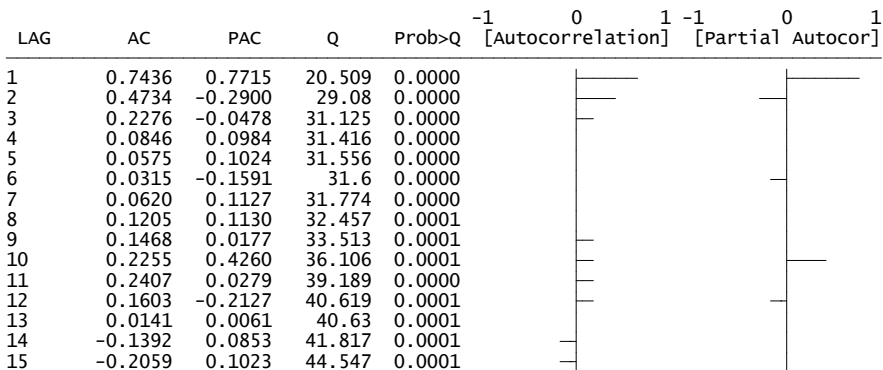
- The following also shows the associated ADF regression
`dfuller lnrxrate if country == "GBR", lags(5) regress`

D.lnrxrate	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnrxrate						
L1.	-.2397465	.1469375	-1.63	0.118	-.5453198	.0658268
LD.	.4322423	.2210537	1.96	0.064	-.0274641	.8919487
L2D.	.0879437	.2176313	0.40	0.690	-.3646454	.5405327
L3D.	-.2028981	.2105659	-0.96	0.346	-.6407938	.2349977
L4D.	-.1348889	.2074825	-0.65	0.523	-.5663723	.2965946
L5D.	.1590723	.1595226	1.00	0.330	-.1726731	.4908177
_cons	-.0083102	.0219679	-0.38	0.709	-.053995	.0373746

- The estimated coefficient $-.239$ is a long way from 0
 - Note, however, that $\hat{\phi}$ is considerably biased below zero if $\phi = 0$
 - Fail to reject H_0 in part because very noisily estimated
 - And note that H_0 is unit root, not H_0 is no unit root.
- Aside: `dfgls` with 5 lags also does not reject H_0

- The autocorrelations look a lot like AR(1) with $\rho \simeq 0.7$, not like unit root.

```
. * ACF
. corrgram lnrxrate if country == "GBR"
```



5. Panel Unit Roots: Overview

- Levin and Lin (1992) seminal paper that introduced many of the ideas
 - need to recenter and rescale DF but ultimately normal
 - results vary with heterogeneity, deterministic trends,

- A quite general model for heterogeneous panels is

$$\begin{aligned}\Delta y_{it} &= \phi_i y_{i,t-1} + \alpha_i + \delta_i t + \theta_t + u_{it} \\ &= \phi_i y_{i,t-1} + \mathbf{z}'_{it} \gamma_i + \theta_t + u_{it}\end{aligned}$$

- Three cases of deterministic trends:
 - no constant: $\alpha_i = 0$; $\delta_i = 0$ so $\mathbf{z}'_{it} \gamma_i = 0$
 - constant, no trend: $\delta_i = 0$ so $\mathbf{z}'_{it} \gamma_i = \alpha_i$
 - trend: both nonzero so $\mathbf{z}'_{it} \gamma_i = \alpha_i + \delta_i t$.
- Different tests make different restrictions on this.
 - initially u_{it} i.i.d. (and possibly normal for finite sample results)
 - relax to serially correlated ($\Rightarrow y_{i,t-1}$ endogenous)
 - relax to u_{it} correlated over i (more complicated).

Limit Theory

- Asymptotics may potentially be
 - ▶ N fixed $T \rightarrow \infty$
 - ▶ T fixed $N \rightarrow \infty$
 - ▶ $T \rightarrow \infty$ and $N \rightarrow \infty$.
- For $T \rightarrow \infty$ and $N \rightarrow \infty$ Phillips and Moon (1999 Ecta) consider
 - ▶ sequential limits: notation $(T, N \rightarrow \infty)_{\text{seq}}$
 - ★ e.g. fix N , let $T \rightarrow \infty$, then let $N \rightarrow \infty$.
 - ▶ diagonal path: notation $T = T(N)$
 - ★ $(T, N) \rightarrow \infty$ at rate $T = T(N)$
 - ▶ joint limits: notation $N, T \rightarrow \infty$
 - ★ simultaneously without any restrictions.

Pedagogical Example

- Levin and Lin (1992)
 - common ρ , heterogeneous intercept, homoskedasticity
 - $y_{it} = \rho y_{i,t-1} + \alpha_i + u_{it}$; $u_{it} \sim \text{iid}[0, \sigma^2]$
 - note: Levin, Lin, Chu (2002) consider more general models
- Demeaning takes care of α_i (Frisch-Waugh Theorem)
 - then $\hat{\phi} = \hat{\rho} - 1$ comes from OLS of Δy_{it-1} on $y_{it} - \bar{y}_i$
 - $$\sqrt{NT}\hat{\phi} = \sqrt{N} \frac{\sum_i \frac{1}{T} \sum_t \{(y_{i,t-1} - \bar{y}_i)/\sigma\} \times \Delta\{y_{i,t-1}/\sigma\}}{\sum_{i=1}^N \frac{1}{T^2} \sum_{t=1}^T \{(y_{i,t-1} - \bar{y}_i)/\sigma^2\}}$$
 - $$\sqrt{NT}\hat{\phi} = \sqrt{N} \frac{\sum_i a_{iT}}{\sum_i b_{iT}} = \sqrt{N} \frac{a_{NT}}{b_{NT}}$$
 defining a_{iT} , b_{iT} , a_{NT} , b_{NT} .
- 1. $T \rightarrow \infty$, find a_{iT} , b_{iT} properties using functional CLT's
 - a_{iT} is function of Wiener process and has mean 1/6 and variance 1/45
 - b_{iT} is function of Wiener process and has mean -1/2 and variance 1/12
- 2. $N \rightarrow \infty$ and apply CLT's and LLN's to get limit distribution of
 - $$\sqrt{N}(T\hat{\phi} - \frac{E[a_{NT}]}{E[b_{NT}]}) = \frac{\sqrt{N}(a_{NT} - E[a_{NT}])}{b_{NT}} + E[a_{NT}]\sqrt{N} \left(\frac{1}{b_{NT}} - \frac{1}{E[b_{NT}]} \right)$$

- Then for $T \rightarrow \infty$ followed by $N \rightarrow \infty$

$$T\sqrt{N}\hat{\phi} + 3\sqrt{N} \xrightarrow{d} N[0, 10.2]$$

- ▶ need N small relative to T so $\sqrt{N}/T \rightarrow 0$
- ▶ convergence rate is $T\sqrt{N}$ rather than usual \sqrt{TN} .
 - ★ Like single time series where T not \sqrt{T} .
- ▶ centering is wrong: $\hat{\phi} + 3/T$ is centered on zero ($\hat{\alpha}_i$; Nickell bias).
 - ★ e.g. $\hat{\phi} = -0.2$ looks like not unit root, but with $T = 15$,
 $\hat{\phi} + 3/T = -0.2 + 3/15 = 0.0$.
- ▶ asymptotic variance of $\hat{\phi}$ equals $10.2/NT^2$ is not the usual variance
- ▶ lesson: need to recenter and rescale $\hat{\phi}$.

- If we use the t-statistic $t_{\phi=0} = \hat{\phi}/se_{\hat{\phi}}$ (rather than $\hat{\phi}$) we get

$$\frac{\sqrt{1.25} \times t_{\phi=0} + \sqrt{1.875N}}{\sqrt{645/112}} \xrightarrow{d} N[0, 1].$$

- ▶ again need to recenter and rescale.

Summary of Approaches

- 1. Common ρ - one pooled augmented DF regression
 - ▶ Detrend data, rescale to common variance for each i , and do ADF
 - ▶ Levin and Lin (1992) and Levin, Lin and Chu (2002)
 - ▶ Breiting (2000) proposes first pre-whitening data - better power
 - ▶ Harris-Tzavalis (1999) have variation that works for T fixed
 - ▶ Breiting and Das (2005) extend to weak (spatial) cross-section dependence
- 2. Common ρ - generalize the KPSS test
 - ▶ Hadri (2000) LM test
- 3. Heterogeneous ρ_i - N separate ADF t-statistics are averaged
 - ▶ detrend data, get ADF t-statistic for each, use the average of these
 - ▶ Im, Pesaran and Shin (2003)
 - ▶ Pesaran (2006) extends to strong (common shocks) cross-section dependence
- 4. Heterogeneous ρ_i - separate tests and combine p-values
 - ▶ Fisher-type tests - Choi (2001) proposes four ways to combine.

Stata Command xtunitroot

Test	Options	Asymptotics	ρ under H_a	Panels
LLC	noconstant	$\sqrt{N}/T \rightarrow 0$	common	balanced
LLC		$N/T \rightarrow 0$	common	balanced
LLC	trend	$N/T \rightarrow 0$	common	balanced
HT	noconstant	$N \rightarrow \infty, T$ fixed	common	balanced
HT		$N \rightarrow \infty, T$ fixed	common	balanced
HT	trend	$N \rightarrow \infty, T$ fixed	common	balanced
Breitung	noconstant	$(T, N) \rightarrow_{\text{seq}} \infty$	common	balanced
Breitung		$(T, N) \rightarrow_{\text{seq}} \infty$	common	balanced
Breitung	trend	$(T, N) \rightarrow_{\text{seq}} \infty$	common	balanced
IPS		$N \rightarrow \infty, T$ fixed or N and T fixed	panel-specific	unbalanced
IPS	trend	$N \rightarrow \infty, T$ fixed or N and T fixed	panel-specific	unbalanced
IPS	lags()	$(T, N) \rightarrow_{\text{seq}} \infty$	panel-specific	unbalanced
IPS	trend lags()	$(T, N) \rightarrow_{\text{seq}} \infty$	panel-specific	unbalanced
Fisher-type		$T \rightarrow \infty, N$ finite or infinite	panel-specific	unbalanced
Hadri LM		$(T, N) \rightarrow_{\text{seq}} \infty$	(not applicable)	balanced
Hadri LM	trend	$(T, N) \rightarrow_{\text{seq}} \infty$	(not applicable)	balanced

6. Application: Tests follow Stata Manual [XT]

- Eviews 8 has the same tests except for Harris-Tzavalis.
- Generally default option of constant (α_i) no trend ($\delta_i = 0$) (for `lnxrate`).
 - ▶ 1. LLC: Levin-Lin-Chu (1993, 2002)
 - ▶ 2. HT: Harris-Tzavalis (1999)
 - ▶ 3. Breitung: Breitung (2000)
 - ▶ 4. IPS: Im-Pesaran-Shin (2003)
 - ▶ 5. IPS: Im-Pesaran-Shin (2003) with serially correlated errors
 - ▶ 6. Fisher: Fisher-type Maddala and Wu (1999), Choi (2001)
 - ▶ 7. Hadri: LM test (Hadri (2000))
- Test 1 is for small N relative to T so use only G7 $N = 7$
- Test 2 is for large N so all countries $N = 151$ (also use for test 6).
- Tests 3-5, 7 use intermediate N so use OECD $N = 27$
- All tests reject hypothesis that all unit roots. Some are stationary.

Levin-Lin-Chu Panel Unit Root Test

- Levin, Lin and Chu (2002).
- For $N \rightarrow \infty$ slower than $T \rightarrow \infty$ so $N/T \rightarrow 0$
 - ▶ though in noconstant case $\sqrt{N}/T \rightarrow 0$.
- ADF model

$$\Delta y_{it} = \phi y_{i,t-1} + \mathbf{z}'_{it} \gamma_i + \sum_{j=1}^{p_i} \theta_{ij} \Delta y_{i,t-j} + u_{it}$$

$$u_{it} \sim \text{iid}[0, \sigma_i^2]$$

- Test $H_0 : \phi_i = 0$ for all i (all have unit roots)
against $H_a : \phi_i = \phi < 0$ for all i (all have common non-unit root).
- Implementation
 - ▶ detrend: form panel-by-panel (standardized) regression residuals
 - ★ $\tilde{\epsilon}_{it} = \hat{\epsilon}_{it} / \hat{\sigma}_{\epsilon i}$ where $\hat{\epsilon}_{it} = \Delta y_{it} - \sum_{j=1}^{p_i} \hat{\theta}_{ij} \Delta y_{i,t-j} - \mathbf{z}'_{it} \hat{\gamma}_i$
 - ★ $\tilde{v}_{i,t-1} = \hat{v}_{i,t-1} / \hat{\sigma}_{\epsilon i}$ where $\hat{v}_{i,t-1} = y_{i,t-1} - \sum_{j=1}^p \hat{\theta}_{ij} \Delta y_{i,t-j} - \mathbf{z}'_{it} \tilde{\gamma}_i$
 - ▶ base test on $\hat{\delta} / se(\hat{\delta})$ from $\tilde{\epsilon}_{it} = \hat{v}_{i,t-1} \delta + \text{error}$
 - ▶ recenter, rescale and use ratio of short-run to long-run variance.

- Test with lag length determined by AIC rejects H_0 as $p < .05$
 - ▶ conclude that stationary
 - ▶ if add option demean $p = 0.0187$ so again stationary

```
. * Levin-Lin-Chu unit root test
. xtunitroot llc lnrxrate if g7, lags(aic 10)
```

Levin-Lin-Chu unit-root test for lnrxrate

Ho: Panels contain unit roots	Number of panels =	6
Ha: Panels are stationary	Number of periods =	34

AR parameter: Common	Asymptotics: N/T -> 0
Panel means: Included	
Time trend: Not included	

ADF regressions: 1.00 lags average (chosen by AIC)
 LR variance: Bartlett kernel, 10.00 lags average (chosen by LLC)

	Statistic	p-value
Unadjusted t	-6.7538	
Adjusted t*	-4.0277	0.0000

- Previous chose only one lag ($p_i = 1$ on average) - seems too short
 - ▶ setting four lags of Δy_{it} we do not reject H_0
 - ▶ conclude that unit root present.

```
. * So do unit root tests with lag lengths set to 4
. xtunitroot llc lnrxrate if g7, lags(4)
```

Levin-Lin-Chu unit-root test for lnrxrate

Ho: Panels contain unit roots
Ha: Panels are stationary

Number of panels = 6
Number of periods = 34

AR parameter: Common
Panel means: Included
Time trend: Not included

Asymptotics: N/T -> 0

ADF regressions: 4 lags
LR variance: Bartlett kernel, 10.00 lags average (chosen by LLC)

	Statistic	p-value
Unadjusted t	-4.9507	
Adjusted t*	-1.1601	0.1230

Harris-Tsavalis Panel Unit Root Test

- Harris and Tsavalis (1999).
- For T fixed and $N \rightarrow \infty$

$$y_{it} = \rho y_{i,t-1} + \mathbf{z}'_{it} \boldsymbol{\gamma}_i + u_{it}$$

$$u_{it} \sim N[0, \sigma^2]$$

- ▶ **very strong assumptions** - i.i.d. error ! (DF not ADF)
- ▶ can relax normality (but then need to estimate kurtosis of u_{it})
- ▶ and can include time dummies
- Test $H_0 : \phi_i = 0$ for all i (all have unit roots)
against $H_a : \phi_i = \phi < 0$ for all i (all have common non-unit root).
- $\hat{\rho}$ biased for finite T via $\boldsymbol{\gamma}_i$
 - ▶ HT find the bias correction term when $\rho = 1$
 - ▶ e.g. if $\mathbf{z}'_{it} \boldsymbol{\gamma}_i = \alpha_i$ then

$$\sqrt{N}(\hat{\rho} - (1 - \frac{3}{T+1})) \xrightarrow{d} N\left[0, \frac{3(17T^2 - 20T + 17)}{5(T-1)(T+1)^3}\right].$$

- Reject H_0 as $p < .05$. Conclude that stationary.

```
. * Harris-Tsavalis unit root test
. xtunitroot ht lnrxrate, demean
```

Harris-Tzavalis unit-root test for lnrxrate

Ho: Panels contain unit roots	Number of panels =	151
Ha: Panels are stationary	Number of periods =	34
AR parameter: Common	Asymptotics: N -> Infinity	
Panel means: Included	T Fixed	
Time trend: Not included	Cross-sectional means removed	

	Statistic	z	p-value
rho	0.8184	-13.1239	0.0000

Breitung Panel Unit Root Test

- Breitung (2000) and Breitung and Das (2005)
- For $T \rightarrow \infty$ followed by $N \rightarrow \infty$
- Have heterogeneous intercept, common trend

$$y_{it} = \rho y_{i,t-1} + \gamma_i + \beta t + u_{it}; \quad u_{it} \sim \text{serially correlated and not iid}$$

- LLC controls for heterogeneous γ_i by mean differencing
 - ▶ instead use the long difference $y_{it} - y_{i,p+1}$
 - ★ with p lags lost due to serial correlation in u_{it}
 - ▶ then $y_{it} = \gamma_i + \beta t$ and $y_{i,p+1} = \gamma_i$ so $y_{it} - y_{i,p+1} = \beta(t - p + 1)$.
- To control for serial correlation of order p in u_{it} pre-whiten data
 - ▶ let $\Delta \hat{\varepsilon}_{it}$ be residual from OLS of Δy_{it} on $\Delta y_{i,t-1}, \dots, \Delta y_{i,t-p}$
 - ▶ let $\hat{\varepsilon}_{it}^*$ be residual from OLS of $y_{it}^* = y_{it} - y_{i,p+1}$ on $\Delta y_{i,t-1}, \dots, \Delta y_{i,t-p}$
 - ▶ $\hat{\lambda} = \frac{\sum_{i=1}^N \sum_{t=p+2}^T \hat{\varepsilon}_{it}^* \Delta \hat{\varepsilon}_{it} / \hat{\sigma}_i^2}{\sqrt{\sum_{i=1}^N \sum_{t=p+2}^T \Delta \hat{\varepsilon}_{it}^2 / \hat{\sigma}_i^2}} \xrightarrow{d} N[0, 1]$ where $\sigma_i^2 = \frac{1}{T-p-2} \sum_{t=p+2}^T \Delta \hat{\varepsilon}_{it}^2$.
- A variation (robust) permits weak cross-sectional correlation
 - ▶ though then need $T \gg N$.

- This version has no lags (should add lags) and gets cross-section correlation robust version
 - ▶ marginally reject H_0 at 5%.

```
. * Breitung unit root test
. xtunitroot breitung lnrxrate if oecd, robust
```

Breitung unit-root test for lnrxrate

Ho: Panels contain unit roots
Ha: Panels are stationary

Number of panels = 27
Number of periods = 34

AR parameter: Common
Panel means: Included
Time trend: Not included

Asymptotics: T,N -> Infinity
sequentially
Prewhitening: Not performed

	Statistic	p-value
lambda*	-1.6794	0.0465

* Lambda robust to cross-sectional correlation

Im-Pesaran-Shin Panel Unit Root Test

- Im, Pesaran and Shin (2003) allow different ϕ_i for each panel
- For T fixed (need normal errors) or $T \rightarrow \infty$ followed by $N \rightarrow \infty$.
- Augmented DF model

$$\begin{aligned}\Delta y_{it} &= \phi_i y_{i,t-1} + \mathbf{z}'_{it} \gamma_i + \sum_{j=1}^{p_i} \theta_{ij} \Delta y_{i,t-j} + u_{it} \\ u_{it} &\sim N[0, \sigma_i^2] \text{ in finite } T \text{ case}\end{aligned}$$

- Test $H_0 : \phi_i = 0$ for all i (all have unit roots)
against $H_a : \phi_i = \phi < 0$ for some i (some have non-unit root).

- Obtain t statistics for each panel

- ▶ these are usual augmented Dickey-Fuller t-test denote τ_i
- ▶ then τ_i under $H_0 : \phi_i = 0$ has mean $E[\tau_i]$ and variance $V[\tau_i]$ that are given in tables
- ▶ the average of these recentered and rescaled $\xrightarrow{d} N[0, 1]$ by CLT as $N \rightarrow \infty$

- ★ unbalanced panel

$$Z = \sqrt{N} \left[\frac{1}{N} \sum_{i=1}^N \tau_i - \frac{1}{N} \sum_{i=1}^N E[\tau_i] \right] / \sqrt{\frac{1}{N} \sum_{i=1}^N V[\tau_i]} \xrightarrow{d} N[0, 1]$$

- ★ balanced panel $Z = \sqrt{N} \left[\frac{1}{N} \sum_{i=1}^N \tau_i - E[\tau] \right] / \sqrt{V[\tau_i]} \xrightarrow{d} N[0, 1]$

- With serially uncorrelated errors (no lags)
 - ▶ reject as $p < 0$ and conclude some panels are stationary

```
. * Im-Pesaran-Shin unit root test
. xtunitroot ips lnrxrate if oecd, demean
```

Im-Pesaran-Shin unit-root test for lnrxrate

Ho: All panels contain unit roots
Ha: Some panels are stationary

Number of panels = 27
Number of periods = 34

AR parameter: Panel-specific
Panel means: Included
Time trend: Not included

Asymptotics: T,N -> Infinity
sequentially
Cross-sectional means removed

ADF regressions: No lags included

	Statistic	p-value	Fixed-N exact critical values		
			1%	5%	10%
t-bar	-3.1327		-1.810	-1.730	-1.680
t-tilde-bar	-2.5771				
Z-t-tilde-bar	-7.3911	0.0000			

- With serially correlated errors (AIC has average 1.5 lags)
 - ▶ reject as $p < 0$ and conclude some panels are stationary

```
. * Im-Pesaran-Shin unit root test with serially correlated errors
. xtunitroot ips lnrxrate if oecd, lags(aic 8) demean
```

Im-Pesaran-Shin unit-root test for lnrxrate

Ho: All panels contain unit roots
Ha: Some panels are stationary

Number of panels = 27
Number of periods = 34

AR parameter: Panel-specific
Panel means: Included
Time trend: Not included

Asymptotics: T,N -> Infinity
sequentially
Cross-sectional means removed

ADF regressions: 1.48 lags average (chosen by AIC)

	Statistic	p-value
w-t-bar	-7.3075	0.0000

Fisher-Type Panel Unit Root Test

- Maddala and Wu (1999) and Choi (2001).
- For $T \rightarrow \infty$ and N fixed
 - ▶ Model is completely heterogeneous
- Do separate time series unit root tests in each panel
 - ▶ Stata `fisher` does Dickey-Fuller and Phillips-Perron
- Then need to combine the N tests.
- This is like meta-analysis
 - ▶ combine the p values using methods of Fisher
 - ▶ e.g. $P = -2 \sum_{i=1}^N \ln p_i \sim \chi^2(2N)$ (reject H_0 if large)
 - ▶ there are other ways of combining - see the output.

- Here do Dickey-Fuller with two lags in each panel
 - ▶ all four ways have $p < .05$ so conclude at least one panel is stationary.

```
. * Fisher unit root test
. xtunitroot fisher lnrxrate, dfuller drift lags(2) demean
```

Fisher-type unit-root test for lnrxrate
Based on augmented Dickey-Fuller tests

Ho: All panels contain unit roots
Ha: At least one panel is stationary

Number of panels = 151
Number of periods = 34

AR parameter: Panel-specific
Panel means: Included
Time trend: Not included
Drift term: Included

Asymptotics: T -> Infinity
Cross-sectional means removed
ADF regressions: 2 lags

		Statistic	p-value
Inverse chi-squared(302)	P	975.9130	0.0000
Inverse normal	Z	-19.6183	0.0000
Inverse logit t(759)	L*	-20.9768	0.0000
Modified inv. chi-squared	Pm	27.4211	0.0000

P statistic requires number of panels to be finite.
Other statistics are suitable for finite or infinite number of panels.

Hadri LM Panel Unit Root Test

- Hadri (2000) generalizes the KPSS test.
- $T \rightarrow \infty$ and then $N \rightarrow \infty$
- Suppose

$$y_{it} = r_{it} + \beta_i t + \varepsilon_{it}, \quad \varepsilon_{it} \sim N[0, \sigma_\varepsilon^2]$$

$$r_{it} = r_{i,t-1} + u_{it}, \quad u_{it} \sim N[0, \sigma_u^2].$$

- In general y_{it} will be $I(1)$ due to the random walk r_{it} .
- But if $\sigma_u^2 = 0$ then $u_{it} = 0$, $r_{it} = r_{i,t-1} = \text{constant}$
 - ▶ so $y_{it} = \alpha + \beta_i t + \varepsilon_{it}$ is trend stationary.
- So test $H_0 : \sigma_u^2 / \sigma_\varepsilon^2 = 0$ against $H_a : \sigma_u^2 / \sigma_\varepsilon^2 = 0$
 - ▶ Note that now H_0 is no unit root against H_q unit root
 - ▶ simulations show this tends to over-reject.
- The test is an LM test that $\xrightarrow{d} N[0, 1]$ after recentering and rescaling
 - ▶ OLS regress $y_{it} = \mathbf{d}'_{it} \gamma + u_{it}$, $\hat{u}_{it} = y_{it} - \mathbf{d}'_{it} \hat{\gamma}$, $S_{it} = \sum_{s=1}^t \hat{u}_{is}^2$
 - ▶ $LM = \sum_{i=1}^N \left(\sum_{t=1}^T S_{it}^2 / (T^2 f_{i0}) \right)$; $f_{i0} = \hat{V}_{long-run}[\hat{u}_{it}]$.

- Reject H_0 as $p < .05$ here means some panels contain unit roots.

```
. * Hadri unit root test
. xtunitroot hadri lnrxrate if oecd, kernel(bartlett 5) demean
```

Hadri LM test for lnrxrate

Ho: All panels are stationary		Number of panels =	27
Ha: Some panels contain unit roots		Number of periods =	34
Time trend:	Not included	Asymptotics: T, N ->	Infinity
Heteroskedasticity:	Robust		sequentially
LR variance:	Bartlett kernel, 5 lags	Cross-sectional means removed	

	Statistic	p-value
z	9.6473	0.0000

7. Cross-section Dependence

- Levin, Lin and Chu (2003, p.13) state that cross-sectional dependence can be partially controlled for on their tests by first subtracting the “cross-sectional average” $\bar{y}_t = \frac{1}{N} \sum_{i=1}^N y_{it}$
 - ▶ equivalent to including a full set of time dummies in the original model for y_{it}
 - ▶ this is a single common shock that has an identical effect on all individuals in the panel.
 - ▶ Stata gives this as an option for all its unit tests, citing LLC
 - ▶ but LLC only say that this is okay for their test.
- Breiting and Pesaran (2008) call tests that allow for cross-section correlation of errors “second generation tests”.
 - ▶ ignoring cross-section dependence leads to size bias
 - ▶ especially if the panel units are cross-cointegrated e.g. PPP

Weak and Strong Cross-Section Dependence

- Weak dependence

- ▶ eigenvalues of covariance matrix of y_{it} are bounded as $N \rightarrow \infty$
- ▶ essentially this means that correlations across individuals need to dampen as they get “further” apart
- ▶ this is the case for spatial dependence

- Strong dependence

- ▶ at least one eigenvalue of covariance matrix of y_{it} diverges as $N \rightarrow \infty$
- ▶ this is the case for common factors $\mathbf{F}'_i \delta_t$.

Unit Root Tests

- Breitung and Pesaran (2008) section 9.4 have summary
 - ▶ see this for references
- Moon and Perron (2004) and Pesaran (2007)
 - ▶ model $y_{it} = (1 - \rho_i)\mu_i + \rho_i y_{i,t-1} + u_{it}$, $u_{it} = \gamma_i f_t + \varepsilon_{it}$
 - ▶ test of $\pi_i = 1$ is joint test that (1) all time series $I(1)$; and (2) they are not cointegrated.
- Bai and Ng (2004) propose Panel Analysis of Nonstationarity in Idiosyncratic and Common components (PANIC)
 - ▶ analyzes common factors and idiosyncratic components separately
 - ▶ nonstationarity can be pervasive, variable-specific or both
 - ▶ can determine the number of independent stochastic trends driving the common factors.

Pesaran (2007)

- Cross-sectionally augmented Dickey Fuller (CADF) test.
- One-factor model $u_{it} = \gamma_i f_t + \varepsilon_{it}$ for the cross-panel correlation.
- Begin with no serial correlation

$$y_{it} = (1 - \rho_i)\mu_i + \rho_i y_{i,t-1} + u_{it}; \quad u_{it} = \gamma_i f_t + \varepsilon_{it}$$

$$\implies \Delta y_{it} = \alpha_i + \phi_i y_{i,t-1} + \gamma_i f_t + \varepsilon_{it}; \quad \alpha_i = \mu_i(1 - \rho_i); \quad \phi_i = (\rho_i - 1)$$

- ▶ ε_{it} iid $[0, \sigma_\varepsilon^2]$, f_t iid $[0, \sigma_f^2 = 1]$, ε_{it} , f_t and γ_i independent for all i
- Test $H_0 : \phi_i = 0$ for all i against $H_a : \phi_i < 0$ for some i .
- Pesaran (2006) shows f_t can be proxied by cross-section mean $\bar{y}_t = \frac{1}{N} \sum_{i=1}^N y_{it}$ and lags if $N \rightarrow \infty$
 - ▶ So OLS estimate

$$\Delta y_{it} = a_i + b_i y_{i,t-1} + c_i \bar{y}_{t-1} + d_i \Delta \bar{y}_t + e_{it}$$

- ▶ If T is fixed replace y_{it} with $y_{it} - \bar{y}_0$; $\bar{y}_0 = \frac{1}{N} \sum_{i=1}^N y_{i0}$.
- CADF_{*i*} Test for panel i : $t_i = \hat{b}_i / se[\hat{b}_i]$.
- $\overline{\text{CADF}}$ Joint panel test: $\frac{1}{N} \sum_{i=1}^N t_i$ where truncate as in his equation (34)

- Pesaran (2007) with serially correlated error (AR(p)).
- OLS estimate (his equation (54))

$$\Delta y_{it} = a_i + b_i y_{i,t-1} + c_i \bar{y}_{t-1} + \sum_{j=0}^p d_{ij} \Delta \bar{y}_{t-j} + \sum_{j=1}^p \delta_{ij} \Delta \bar{y}_{i,t-j} + e_{it}$$

- $CADF_i$ Test for panel i : $t_i = \hat{b}_i / se[\hat{b}_i]$
 - ▶ has critical values in Tables 1.
- \overline{CADF} Joint panel test: $\frac{1}{T} \sum_{i=1}^N t_i$ where truncate as in his equation (34)
 - ▶ has critical values in Tables 2
 - ▶ or can combine $CADF_i$ test p-values e.g. $-2 \sum_{i=1}^N \ln(p_i) \sim \chi^2(2N)$.
- The critical values are obtained by simulation (no $N[0, 1]$ tests here)
 - ▶ asymptotic theory generally the same under sequential or joint limits
 - ▶ $N = 10, 20, 30, 40, 50, 100$ and $T = 10, 20, 30, 40, 50, 100$
 - ▶ generally works well even for small N and T .
- The paper also has good summary of other methods in the introduction and implements some of these in simulation and application.

Application: Pesaran (2007)

```

use pennxrate.dta, clear
rename lnrxrate yy
keep if oecd==1
bysort year: egen ytbar = mean(yy)
sort id year
statsby phi=_b[l.yy] se=_se[l.yy], by(id) clear: regress d.yy l.yy l.ytbar L(0/2)d.ytbar
L(1/2)d.yy
generate t = phi/se
* t is CADF_i test: Critical values are in Table 1 (individual panel)
format phi t %9.3f
list phi t, clean
* Panel ACDF test: Critical values are in Table 2 (average of panels)
mean t
* Fisher test combines the individual test p values
generate ln p = ln(p)
quietly summarize ln p
display "Fisher test = " -2*r(sum) " with p-value = " chi2tail(r(N),-2*r(sum))

```


Application (continued)

- Individual CADF_i t-tests
 - ▶ critical value -3.34 for $N = 30$ and $T = 30$ from Table 1(b)
 - ▶ here most negative t's are $-4.18, -4.09, -3.04$
 - ▶ so 2 of 27 are rejected - borderline reject H_0
- Overall $\overline{\text{CADF}}$ test
 - ▶ critical value -2.15 for $N = 30$ and $T = 30$ from Table 2(b)
 - ▶ here $\text{mean } t = -1.78$
 - ▶ do not reject $H_0 : \text{all } \phi_i = 0$ against $H_a : \text{some } \phi_i \neq 0$.
- Fisher test combining individual CADF_i t-tests
 - ▶ Fisher test = 8.6578493 with p-value = $.00325655$
 - ▶ reject $H_0 : \text{all } \phi_i = 0$ against $H_a : \text{some } \phi_i \neq 0$.

8. Random Walk is Asymptotically a Wiener Processes

- Let $y_t = y_{t-1} + u_t = \sum_{s=1}^t u_s$ where $u_s \sim \text{iid } N[0, 1]$
 - ▶ then $y_t \sim N[0, t]$
- Change the index from $0 \leq t \leq T$ to $0 \leq r \leq 1$
 - ▶ $t = [rT]$ where $[rT]$ is the integer part of rT
 - ▶ $y_t = y_{[rT]} \sim N[0, t]$
 - ▶ $\frac{1}{\sqrt{T}}y_{[rT]} \sim N[0, [rT]/T] \implies \frac{1}{\sqrt{T}}y_{[rT]} \sim N[0, r]$
 - ▶ $\frac{1}{\sqrt{T}}y_{[rT]} - \frac{1}{\sqrt{T}}y_{[r'T]} \sim N[0, r - r']; r' < r$ similarly
- But this is just a discrete time version of a continuous time Wiener process (Brownian motion) $W(r)$, defined as
 - ▶ $W(r) = 0$
 - ▶ $W(r)$ is continuous in r almost surely, $0 \leq r \leq 1$
 - ▶ $W(r) - W(r') \sim N[0, r - r']$ independently for any $0 \leq r' < r \leq 1$
- Key Result: $\frac{1}{\sqrt{T}}y_{[rT]} \stackrel{a}{=} W(r)$ or $\frac{1}{\sqrt{T}}y_t \stackrel{a}{=} W(\frac{t}{T})$

DF Test involves Sums of Functions of Random Walks

- Dickey-Fuller test from regress Δy_t on y_{t-1} yields
 - ▶ $\hat{\phi} = \hat{\rho} - 1 = \sum_{t=1}^T y_{t-1}(\Delta y_t) / \sum_{t=1}^T y_{t-1}^2$
- For nonrandom function $f(r)$ defined on $[0, 1]$
 - ▶ $\int_0^1 f(r) dr = \lim_{T \rightarrow \infty} \sum_{t=1}^T f(\frac{t}{T}) \times \frac{1}{T} = \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T f(\frac{t}{T})$
 - ▶ this extends to random functions such as $W(r)$
- So for sums of y_t and its products, given $\frac{1}{\sqrt{T}} y_t \stackrel{a}{=} W(\frac{t}{T})$
 - ▶ $\frac{1}{T^{3/2}} \sum_{t=1}^T y_t = \frac{1}{T} \sum_{t=1}^T \frac{1}{\sqrt{T}} y_t \stackrel{a}{=} \frac{1}{T} \sum_{t=1}^T W(\frac{t}{T}) \xrightarrow{d} \int_0^1 W(r) dr$
 - ▶ $\frac{1}{T^2} \sum_{t=1}^T y_t^2 = \frac{1}{T} \sum_{t=1}^T (\frac{1}{\sqrt{T}} y_t)^2 = \frac{1}{T} \sum_{t=1}^T W(\frac{t}{T})^2 \xrightarrow{d} \int_0^1 W(r)^2 dr$
 - ▶ $\frac{1}{T} \sum_{t=1}^T y_{t-1} \Delta y_t \xrightarrow{d} \frac{1}{2} [W(1)^2 - 1]$ after some algebra
- So Dickey-Fuller z-test statistic
 - ▶ $T\hat{\phi} = \frac{\frac{1}{T} \sum_{t=1}^T y_{t-1}(\Delta y_t)}{\frac{1}{T^2} \sum_{t=1}^T y_{t-1}^2} \xrightarrow{d} \frac{\frac{1}{2} [W(1)^2 - 1]}{\int_0^1 W(r)^2 dr}$ (no nuisance parameters)

Generalizations

- u_t in $y_t = y_{t-1} + u_t$ can be nonnormal, serially correlated, heterogeneously distributed if $T \rightarrow \infty$
 - ▶ we need $\frac{1}{\sqrt{T}}y_{[rT]} = \frac{1}{\sqrt{T}}\sum_{s=1}^t u_s \xrightarrow{d} \omega \times N[0, r]$ for some ω
 - ▶ this can be established using a (functional) central limit theorem
 - ▶ ω is the long-run variance of $y_{[rT]}$
- We often include an intercept and possibly trend in the DF test
 - ▶ $\Delta y_t = \alpha + \delta t + \phi y_{t-1} + u_t$
 - ▶ then need results for $\sum_{t=1}^T \Delta y_t$, $\sum_{t=1}^T t \Delta y_t$, $\sum_{t=1}^T y_{t-1}$, $\sum_{t=1}^T t y_{t-1}$, ... see Hamilton (1996, p.506)
 - ▶ or can use results for demeaned processes (Frisch-Waugh) ... see Hayashi (2000, p.570)

- Consider $\Delta y_t = \alpha + \phi y_{t-1} + u_t$
 - ▶ use demeaned random walk $y_t^\mu = y_t - \bar{y}$; where $\bar{y} = \frac{1}{T} \sum_{t=1}^T y_t$
 - ▶ continuous time analog is detrended standard Wiener

$$W^\mu(r) = W(r) - \int_0^1 W(s) ds$$
 - ▶ $\frac{1}{T^2} \sum_{t=1}^T (y_t^\mu)^2 \xrightarrow{d} \int_0^1 W^\mu(r)^2 dr$
 - ▶ $\frac{1}{T} \sum_{t=1}^T y_{t-1}^\mu \Delta y_t^\mu \xrightarrow{d} \frac{1}{2} [W^\mu(1)^2 - W^\mu(0)^2 - 1]$
- Consider $\Delta y_t = \alpha + \delta t + \phi y_{t-1} + u_t$ detrended random walk $y_t^\mu = y_t - \hat{\alpha} - \hat{\delta}t$ where $\hat{\alpha}$ and $\hat{\delta}$ are from OLS of y_t on intercept and linear time trend
 - ▶ then different functions of Wiener processes.

9. Summary

- Stata unit root test

Single Time Series `dfgls, dfuller`

Panel: cross-country correlation `xtunitroot`

Panel: cross-country correlation `—`

- For panel unit root tests with cross-country correlation can easily code up Pesaran (2007) CADF test and refer to tables in his paper.

10. Selected References

• Surveys

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- ▶ Kirchgässner. G. (2013), *Introduction to Modern Time Series Analysis*, chapter 7, Springer.
- ▶ Stata Manual [XT] Longitudinal Data / Panel Data entry xtunitroot Method and Formulas summarizes several panel unit root tests.
- ▶ Eviews 8 Users Guide II Chapter 16 summarizes several panel unit root tests.

- Unit Root Tests without cross-section correlation

- ▶ Im, K.S., M.H. Pesaran, and Y. Shin (2003), “Testing for Unit Roots in Heterogenous Panels,” *Journal of Econometrics*, 115, 53–74.
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- Unit Root Tests with cross-section correlation

- ▶ Breitung, J. and S. Das (2005), “Panel Unit Root Tests Under Cross Sectional Dependence,” *Statistica Neerlandica*, 59, 414–433.
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