## EC408 Topics in Applied Econometrics

B Fingleton, Dept of Economics, Strathclyde University

## Limitations of graphical methods



Figure 3 gives good visual evidence of a stochastic trend, but Figure 4 is less clear cut, although it also is an I(1) variable

For the Figure 4 variable a more accurate test is need because visual inspection may be misleading. A good test would show that the variable is also I(1), even though visually it appears to be a TSP.

# Limitations of graphical methods

Given an I(1) variable, say  $Y_t$ , then by definition  $\Delta Y = Y_t - Y_{t-1} = u_t \sim I(0)$ , in which case

$$Y_{t} = \rho Y_{t-1} + u_{t} \qquad (21)$$

$$\rho = 1$$
If  $\rho = 1$  then  $Y_{t} \sim I(1)$  if  $\rho < 1$ ,

if 
$$\rho < 1$$
, then  $Y_t \sim I(0)$ 

 $Y_t$  has a constant mean and variance, the correlation between pair of values  $Y_t$  and  $Y_{t+k}$ the same for any *t* and *k*.





Fig. 3 A Stochastic trend

## Set up for basic DF test

tests of stationarity are tests of the null hypothesis Ho:  $\rho = 1$ The alternative is Ha:  $\rho < 1$ 

 $Y_{t-1}$  subtracted from each side

$$Y_{t} - Y_{t-1} = \rho Y_{t-1} - Y_{t-1} + u_{t}$$

$$\Delta Y_{t} = (\rho - 1)Y_{t-1} + u_{t}$$

$$\Delta Y_{t} = \delta Y_{t-1} + u_{t}$$
(22)

Ho :  $\delta = 0$  in  $\Delta Y_t = \delta Y_{t-1} + u_t$  implying  $Y \sim I(1)$ 

Ha:  $\delta < 0$  in  $\Delta Y_t = \delta Y_{t-1} + u_t$  implying Y~I(0)

when *Y* is I(1), the theoretical t-distribution does not apply to the statistic  $\tau = \hat{\delta}/s.e.(\hat{\delta})$ .

the correct distribution for  $\tau = \hat{\delta}/s.e.(\hat{\delta})$ under Ho :  $\delta = 0$  is the Dickey-Fuller distribution.

there is a family of Dickey-Fuller distributions depending on kind of I(1) processes

 $\tau_{nc}, \tau_c, \tau_{ct}, \tau_{ctt}$  are appropriate respectively for models with no constant, models with constant, models with constant and trend, models with constant plus trend and trend-squared

## Dickey-Fuller significance levels



### Fig. 8 Asymptotic densities for Dickey-Fuller test statistics

#### 638 CHAPTER 20 + Time-Series Models

		Sample Size			
	25	50	100	$\infty$	
F ratio (D–F) <sup>a</sup>	7.24	6.73	6.49	6.25	
F ratio (standard)	3.42	3.20	3.10	3.00	
AR model <sup>b</sup> (random wa	lk)				
0.01	-2.66	-2.62	-2.60	-2.58	
0.025	-2.26	-2.25	-2.24	-2.23	
0.05	-1.95	-1.95	-1.95	-1.95	
0.10	-1.60	-1.61	-1.61	-1.62	
0.975	1.70	1.66	1.64	1.62	
AR model with constant	t (random walk with	n drift)			
0.01	-3.75	-3.59	-3.50	-3.42	
0.025	-3.33	-3.23	-3.17	-3.12	
0.05	-2.99	-2.93	-2.90	-2.86	
0.10	-2.64	-2.60	-2.58	-2.57	
0.975	0.34	0.29	0.26	0.23	
AR model with constan	t and time trend (tre	end stationary)			
0.01	-4.38	-4.15	-4.04	-3.96	
0.025	-3.95	-3.80	-3.69	-3.66	
0.05	-3.60	-3.50	-3.45	-3.41	
0.10	-3.24	-3.18	-3.15	-3.13	
0.975	-0.50	-0.58	-0.62	-0.66	

<sup>a</sup>From Dickey and Fuller (1981, p. 1063). Degrees of freedom are 2 and T - p - 3. <sup>b</sup>From Fuller (1976, p. 373 and 1996, Table 10.A.2).

### Some alternative DGPs

Pure random walk (y)

$$Y_{t} = \rho Y_{t-1} + u_{t}$$
$$\rho = 1$$
$$u \sim N(0, \sigma^{2})$$

 $Y_{t} = Y_{t-1} + \mu + u_{t}$ random walk with drift (ymu) Y  $\mu$  is a constant positive (or negative) contribution at each point in time, the Y series drifts consistently upwards (or downwards) TSP (yt)

Stationary autoregressive Process (ys)

$$Y_{t} = \rho Y_{t-1} + \mu + \gamma_{1} t + u_{t}$$
$$\rho < 1$$

$$Y_t = \mu + \rho Y_{t-1} + u_t$$
$$\rho < 1$$

## Alternative DGPs



Fig 7 Deterministic and stochastic trends

We test each data generating process against an alternative, To see if the DF test picks out the stationary from the nonstationary ones

the errors in these regressions may not be independent,'so-called 'white noise', but may in fact be autocorrelated.We therefore can whiten them by including lagged differences in the equation, so as to eliminate serial correlation from the errors.

linear time trend plus constant, augmented by lagged differences

$$\Delta Y_t = \mu + \delta Y_{t-1} + \gamma t + \gamma_1 \Delta Y_{t-1} + \gamma_2 \Delta Y_{t-2} + \dots + u_t$$
(30)

# ADF for random walk with drift (ymu)

#### Results from UNITROOT\_tests.fl

Augmented Dickey-Fuller test for ymu; regression of Dymu on:

	Coefficient	Std.Error	t-value
ymu_1	-0.0047175	0.0035605	-1.3250
Constant	0.46391	0.066725	6.9526
Trend	0.0024519	0.0017815	1.3763
Dymu_1	0.0063304	0.031837	0.19884
Dymu_2	-0.015645	0.031837	-0.49142

sigma = 0.988925 DW = 1.997 DW-ymu = 5.974e-005 ADF-ymu = -1.325 Critical values used in ADF test: 5%=-3.417, 1%=-3.972 RSS = 970.1479611 for 5 variables and 997 observations

# ADF

- Previous slide was output from PcGive
- It shows critical values that are same as Dickey-Fuller table (see earlier)
- Next slide is from gretl, it gives the p-value from the same reference distribution
- This can be obtained by running adf\_gretl.inp script
- Note that the p-value is NOT from the t distribution, which can be checked in gretl using gretl>tools>p-value finder

# ADF for random walk with drift (ymu)

```
Augmented Dickey-Fuller test for ymu
including 2 lags of (1-L)ymu
sample size 997
unit-root null hypothesis: a = 1
```

```
with constant and trend

model: (1-L)y = b0 + b1*t + (a-1)*y(-1) + \ldots + e

lst-order autocorrelation coeff. for e: -0.000

lagged differences: F(2, 992) = 0.141 [0.8687]

estimated value of (a - 1): -0.00471751

test statistic: tau_ct(1) = -1.32496

asymptotic p-value 0.8816
```

Augmented Dickey-Fuller regression OLS, using observations 4-1000 (T = 997) Dependent variable: d\_ymu

	coefficient	std. error	t-ratio	p-value	
const	0.463909	0.0667246	6.953	6.50e-012	***
ymu_1	-0.00471751	0.00356050	-1.325	0.8816	
d_ymu_1	0.00633042	0.0318374	0.1988	0.8424	
d_ymu_2	-0.0156454	0.0318374	-0.4914	0.6232	
time	0.00245187	0.00178151	1.376	0.1690	

## ADF for Pure random walk (y)

```
Augmented Dickey-Fuller test for y
including 2 lags of (1-L)y
sample size 997
unit-root null hypothesis: a = 1
```

with constant and trend model: (1-L)y = b0 + b1\*t + (a-1)\*y(-1) + ... + e1st-order autocorrelation coeff. for e: -0.001 lagged differences: F(2, 992) = 2.476 [0.0846]estimated value of (a - 1): -0.0207408 test statistic: tau\_ct(1) = -3.2372 asymptotic p-value 0.07718

Augmented Dickey-Fuller regression OLS, using observations 4-1000 (T = 997) Dependent variable: d\_y

	coefficient	std. error	t-ratio	p-value	
const	-0.278490	0.0859898	-3.239	0.0012 '	* * *
y_1	-0.0207408	0.00640702	-3.237	0.0772	*
d_y_1	-0.0357528	0.0317352	-1.127	0.2602	
d_y_2	0.0586645	0.0317069	1.850	0.0646	*
time	-0.000367871	0.000185549	-1.983	0.0477 **	

# ADF for Trend Stationary Process (yt)

```
Augmented Dickey-Fuller test for yt
including 2 lags of (1-L)yt
sample size 997
unit-root null hypothesis: a = 1
```

```
with constant and trend
model: (1-L)y = b0 + b1*t + (a-1)*y(-1) + ... + e
1st-order autocorrelation coeff. for e: 0.001
lagged differences: F(2, 992) = 1.477 [0.2289]
estimated value of (a - 1): -0.931959
test statistic: tau_ct(1) = -17.6711
asymptotic p-value 5.533e-057
```

Augmented Dickey-Fuller regression OLS, using observations 4-1000 (T = 997) Dependent variable: d\_yt

	coefficient	std. error	t-ratio	p-value	
const	1.01543	0.0675249	15.04	3.40e-046	* * *
yt_1	-0.931959	0.0527391	-17.67	5.53e-057	* * *
d_yt_1	-0.0298141	0.0441098	-0.6759	0.4993	
d_yt_2	0.0211984	0.0317835	0.6670	0.5050	
time	0.465897	0.0263646	17.67	5.75e-061	* * *

# Stationary Autoregressive series (ys)

```
Augmented Dickey-Fuller test for ys
including 2 lags of (1-L)ys
sample size 997
unit-root null hypothesis: a = 1
with constant and trend
model: (1-L)y = b0 + b1*t + (a-1)*y(-1) + ... + e
lst-order autocorrelation coeff. for e: -0.000
lagged differences: F(2, 992) = 0.552 [0.5759]
estimated value of (a - 1): -0.108455
test statistic: tau_ct(1) = -7.63655
asymptotic p-value 2.928e-011
```

Augmented Dickey-Fuller regression OLS, using observations 4-1000 (T = 997) Dependent variable: d\_ys

	coefficient	std. error	t-ratio	p-value	
const	1.10259	0.152919	7.210	1.11e-012	***
ys_1	-0.108455	0.0142021	-7.637	2.93e-011	* * *
d_ys_1	0.0283473	0.0315447	0.8986	0.3691	
d_ys_2	0.0189067	0.0314805	0.6006	0.5483	
time	-4.61587e-05	0.000111105	-0.4155	0.6779	

the ADF test successfully discriminates between

the two stationary series

Trend Stationary Process (yt) Stationary autoregressive series (ys),

and the two I(1) series

Pure random walk (y) random walk with drift (ymu).

if the null was stationarity rather than nonstationarity- simpler?

contradict the findings of the conventional DF tests, but also tend to contradict each other!

The majority of the (large number of) unit root tests have specified Ho as a unit root rather than stationarity.

# Type I and II error

state of world	decision accept Ho	decision Ho rejected
Ho true	correct Prob=1-α	Type I error
H, true	Type II	$\frac{Prob=\alpha}{correct}$
IIA UUC	error	$Prob = 1-\beta$
	Prob=β	= power

problems

size distortion

Type I error rate is not equal to what is tabulated <u>low power</u> Type II error rate is high

Even when the null of a unit root is not true, the ADF often fails to reject the null, so we think we have a unit root when we do not.