

# Panel Unit Root Tests with Structural Breaks

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# Motivation

- Structural breaks are shocks which are exogenous to the model but have a lasting effect.
- They mislead unit root tests to accept the null of unit root when in fact it is stationary.
- Ignorance of structural breaks can distort power of tests and lead to deceptive conclusions.
- In real world, structural breaks can be caused by many factors, including changes in policy regime or important worldwide events, e.g., the Great Depression, World War II, oil price shock, Covid19.

# Contribution

- **xtbunitroot** implements the econometric method suggested by Karavias and Tzavalis (2014).
- It performs panel unit root tests that allow for breaks in the intercepts of the individual series or in both intercepts and linear trends.
- **xtbunitroot** allows for one or two breaks at either known or unknown dates.
- It is the first Stata command which allows for panel unit root tests with structural breaks and can be viewed as a complement to the official **xtunitroot** command.

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## Underlying models

- M1 model tests against a structural break in the intercepts of the series:

$$y_{i,t} = \varphi y_{i,t-1} + (1 - \varphi)[\alpha_{1,i}I(t \leq b) + \alpha_{2,i}I(t > b)] + u_{i,t}$$

- $\alpha_{1,i}$  and  $\alpha_{2,i}$  are the fixed effects before and after the break.
- The break happens on date  $b$  and the notation  $I(\cdot)$  denotes indicator function.

## Underlying models

- M2 model tests against a structural break in both intercepts and linear trends:

$$y_{i,t} = \varphi y_{i,t-1} + \varphi[\beta_{1,i}I(t \leq b) + \beta_{2,i}I(t > b)] \\ + (1 - \varphi)[\alpha_{1,i}I(t \leq b) + \alpha_{2,i}I(t > b)] \\ + (1 - \varphi)[\beta_{1,i}tI(t \leq b) + \beta_{2,i}tI(t > b)] + u_{i,t}$$

- $\beta_{1,i}$  and  $\beta_{2,i}$  are coefficients of linear trends before and after the break.
- The break is allowed to be in  $I_1 = \{1, 2, \dots, T - 1\}$  for M1 and in  $I_2 = \{2, \dots, T - 2\}$  for M2.
- Both models can be extended to the case of two structural breaks.



# Hypothesis

- The null hypothesis  $H_0$  : All panel time series are unit root processes without breaks ( $\varphi = 1$ ).
- The alternative hypothesis  $H_1$  : Some or all of the panel time series are stationary processes with breaks ( $\varphi < 1$ ).
- Structural breaks can only occur under the alternative hypothesis (Zivot and Andrews, 1992).
- The alternative hypothesis is homogeneous across different individuals but it is also evidenced that the test has power against heterogeneous alternatives (Karavias and Tzavalis, 2016).

# Test Statistics

## Known breaks

- For a given break date  $b$ , autoregressive parameter  $\varphi$  can be estimated with the following pooled least squares estimator:

$$\hat{\varphi} = \left( \sum_{i=1}^N y'_{i,-1} Q_m^b y_{i,-1} \right)^{-1} \left( \sum_{i=1}^N y'_{i,-1} Q_m^b y_i \right), \quad m = \{M1, M2\}$$

- The orthogonal projection matrix  $Q_m^b$  is defined as  $Q_m^b = I_T - X_m^b (X_m^{b'} X_m^b)^{-1} X_m^{b'}$ , where  $X_{M1}^b = (e_1, e_2)$  and  $X_{M2}^b = (e_1, e_2, \tau_1, \tau_2)$ .
- Karavias and Tzavalis (2014) show that the estimator  $\hat{\varphi}$  is inconsistent and must be modified.

# Test Statistics

## Known breaks

- The inconsistency of  $\hat{\varphi}$  is given by:

$$B^b = \underset{N \rightarrow \infty}{plim} (\hat{\varphi} - 1) = \frac{tr[\Lambda' Q_m^b]}{tr(\Lambda' Q_m^b \Lambda)}, \text{ for } m = \{M1, M2\}$$

- $T \times T$  matrix  $\Lambda$  is defined as:  $[\Lambda]_{r,c} = 1$  if  $r > c$  and 0 otherwise, where  $r, c \in \{1, \dots, T\}$ .
- The consistent test statistics is:

$$Z(b) = \sqrt{N} [C^b(k_u, \sigma_u^2)]^{-\frac{1}{2}} (\hat{\varphi} - 1 - B^b) \xrightarrow{L} N(0, 1)$$

- Assume errors are not serially correlated, the estimated variance is:

$$C^b(k_u, \sigma_u^2) = \left\{ k \sum_{j=1}^T [A^b]_{j,j}^2 + 2\sigma_u^4 tr(A^{b2}) \right\} \left\{ \sigma_u^2 tr(\Lambda' Q_m^b \Lambda) \right\}^{-2}$$

# Test Statistics

## Known breaks

- The consistent estimators of parameters  $k_u$  and  $\sigma_u^2$  are given by Harris and Tzavalis (2004).
- For heteroskedastic errors,  $C^b(k_u, \sigma_u^2)$  is replaced with

$$\bar{C}^b(k_u, \sigma_u^2) = \frac{1}{N} \sum_{i=1}^N \left\{ \left[ k_i \sum_{j=1}^T [A^b]_{j,j}^2 + 2\sigma_{u,i}^4 \text{tr}(A^{b^2}) \right] [\sigma_{u,i}^2 \text{tr}(\Lambda' Q_m^b \Lambda)]^{-2} \right\}$$

- For independently, normally distributed errors:

$$C^b = 2\text{tr}[(A^b)^2] / \text{tr}(\Lambda' Q_m^b \Lambda)^2$$

# Test Statistics

## Unknown breaks

- If the break date is unknown, one break point  $b_{min}$  is chosen which minimize test statistic over all possible break dates (Zivot and Andrews, 1992)

$$\min \mathcal{Z} = \min_{b \in I_m} Z(b) \text{ for } m = \{M_1, M_2\}$$

- Following Karavias and Tzavalis (2019), a bootstrap algorithm is implemented to derive the critical values and p-values of  $\min \mathcal{Z}$ .

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# Basic Syntax

```
xtbunitroot varname [if] [in] [, trend known(integer  
integer) unknown(numlist integer) normal csd het  
nobootstrap]
```

- Dataset must be xtset before using the command.
- If no option is specified, the default will be  $M_1$  model with a single break in the intercept, at an unknown date with 100 bootstrap replications.

# Main Options

- known(*break1 break2*)
  - the number and places of breaks. This option must be used when the dates of the breaks are known.
- unknown(*numbreaks numboot*)
  - the number of unknown breaks and the number of bootstrap replications.
- trend
  - the common breaks affect both intercepts and trends.
- normal
  - the errors are normally distributed.
- csd
  - demeaning procedure for cross-sectionally dependent errors.
- het
  - errors are cross-sectionally heteroskedastic.



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# Empirical examples

- To examine the stationarity of banking variable: returns on assets.
- The quarterly data is collected from the Federal Deposit Insurance Corporation (FDIC), composed of a random sample of 500 banks, from 2018q3 to 2020q4.
- This period includes the COVID19 pandemic which may have caused breaks in the intercepts and trends of the series (started in 2020q1).

# Results

- First assume that the date of the break is known to be 2020q1.
- Errors are assumed to be normally distributed but cross-sectionally dependent.
- The output below shows that the null hypothesis of non-stationarity is rejected at 1% significance level.

## Output

```
. use xtbunitroot_example.dta, clear
. xtset fed_rssd time
      panel variable:  fed_rssd (strongly balanced)
      time variable:  time, 1 to 10
                delta:  1 unit

. xtbunitroot roa, known(7) normal csd
Karavias and Tzavalis (2014) panel unit root test for roa
```

---

H0: All panel time series are unit root processes

H1: Some or all of the panel time series are stationary processes

---

Number of panels:	500	Number of periods:	10
Number of breaks:	1		
Cross-section dependence:	Yes	Linear time trend:	No
Cross-section heteroskedasticity:	No	Normal errors:	Yes
Result: the null is rejected		Known break date(s):	7

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	Statistic	5% Asymptotic critical-value	p-value
Z-statistic	-15.3360	-1.6450	0.0000

---

# Output

- Secondly the break date is assumed to be unknown and determined from the data.

```
. xtbunitroot roa, unknown(1) normal csd
Karavias and Tzavalis (2014) panel unit root test for roa
```

---

```
H0: All panel time series are unit root processes
H1: Some or all of the panel time series are stationary processes
```

---

Number of panels:	500	Number of periods:	10
Number of breaks:	1	Bootstrap replications:	100
Cross-section dependence:	Yes	Linear time trend:	No
Cross-section heteroskedasticity:	No	Normal errors:	Yes
Result: the null is rejected		Estimated break date(s):	6

---

	Statistic	5% Bootstrap critical-value	p-value
minZ-statistic	-25.1273	9.3046	0.0000

---

# Output

- Similarly the null hypothesis is rejected at 1% significance level.
- When the break date is unknown, the command reports the estimated break date: observation 6 in this case, which corresponds to 2019q4.
- Result: Returns on assets are found to be stationary with structural breaks in the intercepts.

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# Conclusion

- This presentation has introduced a new Stata command **xtbunitroot** suggested by Karavias and Tzavalis (2014).
- The test is more powerful when structural break exists in intercepts and trends.
- It allows for one or two breaks under the alternative which can be either known or unknown.
- The command can be downloaded in stata using: **ssc install xtbunitroot**



# References I

- Harris, Richard DF and Elias Tzavalis (2004). “Testing for unit roots in dynamic panels in the presence of a deterministic trend: re-examining the unit root hypothesis for real stock prices and dividends”. In: *Econometric Reviews* 23.2, pp. 149–166. ISSN: 0747-4938.
- Karavias, Yiannis and Elias Tzavalis (2014). “Testing for unit roots in short panels allowing for a structural break”. In: *Computational Statistics & Data Analysis* 76, pp. 391–407. ISSN: 0167-9473.
- (2016). “Local power of fixed T panel unit root tests with serially correlated errors and incidental trends”. In: *Journal of Time Series Analysis* 37.2, pp. 222–239. ISSN: 0143-9782.
  - (2019). “Generalized fixed T panel unit root tests”. In: *Scandinavian Journal of Statistics* 46.4, pp. 1227–1251. ISSN: 0303-6898.
- Zivot, Eric and Donald W K Andrews (1992). “Further evidence on the great crash, the oil-price shock, and the unit-root hypothesis”. In: *Journal of business & economic statistics* 20.1, pp. 25–44. ISSN: 0735-0015.

*Thanks!*