



# On Reproducible Econometric Research

Achim Zeileis

<http://eeecon.uibk.ac.at/~zeileis/>

# Overview

Joint work with **Roger Koenker** (University of Urbana-Champaign).

Koenker R, Zeileis A (2009). “On Reproducible Econometric Research.” *Journal of Applied Econometrics*, **24**(5), 833–847. doi:10.1002/jae.1083

And with **Christian Kleiber** (Universität Basel).

Zeileis A, Kleiber C (2005). “Validating Multiple Structural Change Models – A Case Study.” *Journal of Applied Econometrics*, **20**(5), 685–690.  
doi:10.1002/jae.856

Kleiber C, Zeileis A (2008). *Applied Econometrics with R*. Springer-Verlag, New York. URL <http://CRAN.R-project.org/package=AER>

Kleiber C, Zeileis A (2011). “Reproducible Econometric Simulations.” *Working Paper 2011-02*, Working Papers in Economics and Statistics, Research Platform Empirical and Experimental Economics, Universität Innsbruck.  
URL <http://EconPapers.RePEc.org/RePEc:inn:wpaper:2011-02>

# Overview

- Forensic econometrics: Case studies
  - Cross-country growth regressions
  - Multiple structural change models
- Software tools
  - Version control
  - Data technologies and data archiving
  - Programming environments
  - Document preparation systems
  - Literate programming
- Challenges and conclusions

# Forensic econometrics: Growth regressions

**Investigation:** Cross-country growth behavior based on extended Solow model.

- Durlauf and Johnson (1995, *Journal of Applied Econometrics*) extend analysis by Mankiw, Romer, Weil (1992, *The Quarterly Journal of Economics*).
- Of interest: Output (GDP per capita) growth from 1960 to 1985 for 98 non-oil-producing countries.
- Variables: Real GDP per capita; fraction of real GDP devoted to investment; population growth; fraction of population in secondary schools; and adult literacy rate.
- Data taken from MRW. DJ added literacy rate. Available as `data.dj` in JAE data archive.

**Models:** OLS regressions for full sample and breaks based on initial output and literacy.

# Forensic econometrics: Growth regressions

Dependent variable:  $\log(Y/L)_{i,1985} - \log(Y/L)_{i,1960}$ .

	Full sample	$(Y/L)_{i,1960} < 1950$ $LR_{i,1960} < 54\%$	$(Y/L)_{i,1960} \geq 1950$ $LR_{i,1960} \geq 54\%$
Observations	98	42	42
Constant	3.040 (0.831)	1.400 (1.850)	0.450 (0.723)
$\log(Y/L)_{i,1960}$	-0.289 (0.062)	-0.444 (0.157)	-0.434 (0.085)
$\log(I/Y)_i$	0.524 (0.087)	0.310 (0.114)	0.689 (0.170)
$\log(n + 0.05)_i$	-0.505 (0.288)	-0.379 (0.468)	-0.545 (0.283)
$\log(SCHOOL)_i$	0.233 (0.060)	0.209 (0.094)	0.114 (0.164)

# Forensic econometrics: Growth regressions

**Replication:** Data is available from JAE archive, and OLS regression should be trivial ... right?

**Data:** Read, code missing values, and select non-oil countries.

```
R> dj <- read.table("data.dj", header = TRUE,  
+   na.strings = c("-999.0", "-999.00"))  
R> dj <- subset(dj, NONOIL == 1)
```

**Model:** R formula (converting percentages to fractions).

```
R> f1 <- I(log(GDP85) - log(GDP60)) ~ log(GDP60) +  
+   log(IONY/100) + log(POPGR0/100 + 0.05) + log(SCHOOL/100)
```

**Regression:** OLS fit for full sample and subsamples.

```
R> mrw <- lm(f1, data = dj)  
R> sub1 <- lm(f1, data = dj, subset = GDP60 < 1950 & LIT60 < 54)  
R> sub2 <- lm(f1, data = dj, subset = GDP60 >= 1950 & LIT60 >= 54)
```

# Forensic econometrics: Growth regressions

**Full sample results:** Success! Only minor deviations.

```
R> mrw <- lm(f1, data = dj)
R> coeftest(mrw)
```

	Durlauf & Johnson	Replication
Observations	98	98
Constant	3.040 (0.831)	3.022 (0.827)
$\log(Y/L)_{i,1960}$	-0.289 (0.062)	-0.288 (0.062)
$\log(I/Y)_i$	0.524 (0.087)	0.524 (0.087)
$\log(n + 0.05)_i$	-0.505 (0.288)	-0.506 (0.289)
$\log(SCHOOL)_i$	0.233 (0.060)	0.231 (0.059)

# Forensic econometrics: Growth regressions

**Subsample results:** Failure! Not even sample size is correct.

```
R> sub2 <- lm(f1, data = dj, subset = GDP60 >= 1950 & LIT60 >= 54)
R> coeftest(sub2)
```

	Durlauf & Johnson	Replication
Observations	42	39
Constant	0.450 (0.723)	3.952 (1.337)
$\log(Y/L)_{i,1960}$	-0.434 (0.085)	-0.425 (0.104)
$\log(I/Y)_i$	0.689 (0.170)	0.653 (0.187)
$\log(n + 0.05)_i$	-0.545 (0.283)	-0.587 (0.361)
$\log(SCHOOL)_i$	0.114 (0.164)	0.137 (0.180)



# Forensic econometrics: Growth regressions

**Problem 1:** Grid search plus educated guessing leads to different breaks.

```
R> sub2b <- lm(f1, data = dj, subset = GDP60 >= 1800 & LIT60 >= 50)
R> coeftest(sub2b)
```

	Durlauf & Johnson	Replication
Observations	42	42
Constant	0.450 (0.723)	4.147 (1.230)
$\log(Y/L)_{i,1960}$	-0.434 (0.085)	-0.435 (0.096)
$\log(I/Y)_i$	0.689 (0.170)	0.689 (0.178)
$\log(n + 0.05)_i$	-0.545 (0.283)	-0.545 (0.345)
$\log(SCHOOL)_i$	0.114 (0.164)	0.114 (0.171)

# Forensic econometrics: Growth regressions

**Problem 2:** Population growth and schooling not fractions but percent.

```
R> sub2c <- update(sub2b, . ~ log(GDP60) +  
+   log(IONY) + log(POPGRO/100 + 0.05) + log(SCHOOL))
```

	Durlauf & Johnson	Replication
Observations	42	42
Constant	0.450 (0.723)	0.450 (0.899)
$\log(Y/L)_{i,1960}$	-0.434 (0.085)	-0.435 (0.096)
$\log(I/Y)_i$	0.689 (0.170)	0.689 (0.178)
$\log(n + 0.05)_i$	-0.545 (0.283)	-0.545 (0.345)
$\log(SCHOOL)_i$	0.114 (0.164)	0.114 (0.171)

# Forensic econometrics: Growth regressions

**Problem 3:** Robust sandwich standard errors.

```
R> coeftest(sub2c, vcov = sandwich)
```

	Durlauf & Johnson	Replication
Observations	42	42
Constant	0.450 (0.723)	0.450 (0.723)
$\log(Y/L)_{i,1960}$	-0.434 (0.085)	-0.435 (0.085)
$\log(I/Y)_i$	0.689 (0.170)	0.689 (0.170)
$\log(n + 0.05)_i$	-0.545 (0.283)	-0.545 (0.283)
$\log(SCHOOL)_i$	0.114 (0.164)	0.114 (0.164)

# Forensic econometrics: Growth regressions

## Summary:

- Cutoffs actually used did not match those indicated.
- Usage of standard errors inconsistent.
- Scaling of variables (and hence intercepts) inconsistent.
- Other models in DJ paper: Similar problems, and some inference not reproducible at all.

## Implications:

- Casts doubt results. (Even though – in this case, so far – qualitative results remain unchanged.)
- Very hard to track down without original code.
- Might have been impossible for less standard models.
- *Hence*: Provide replication code even for simple things and details.

# Forensic econometrics: Structural change models

**Investigation:** Multiple structural change model for level of US ex-post real interest rate (Jan 1961–Jul 1986).

**Source:** Bai and Perron (2003, *Journal of Applied Econometrics*).

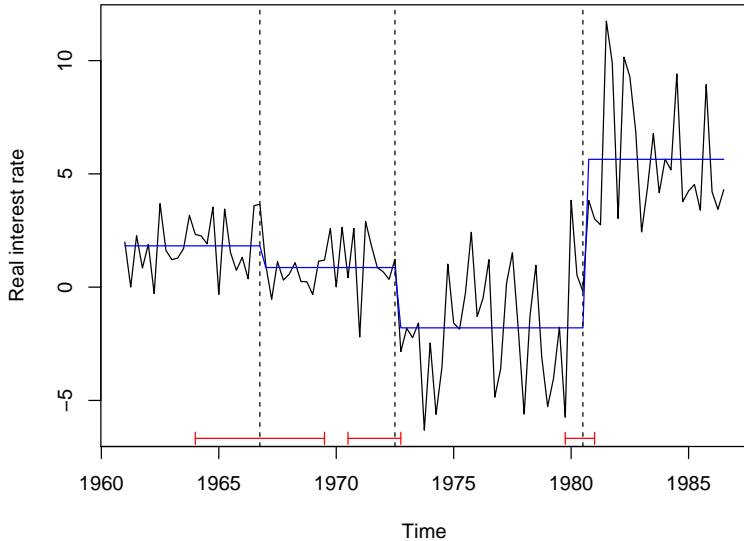
- Comprehensive discussion of computational aspects of multiple structural change models.
- Empirical examples, with data in JAE archive.
- GAUSS software and replication code!

**Replication:**

- Re-implementation of methods in R (package *strucchange*).
- Successful replication of: Breakpoint estimates (OLS), coefficient estimates (OLS), coefficient standard errors (quadratic spectral kernel HAC with prewhitening).
- Problems: Confidence intervals of breakpoints.

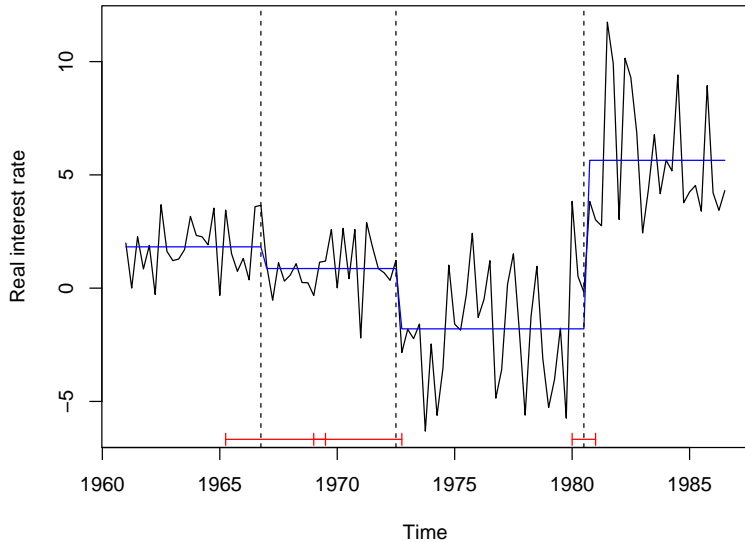
# Forensic econometrics: Structural change models

Bai & Perron



# Forensic econometrics: Structural change models

## Replication



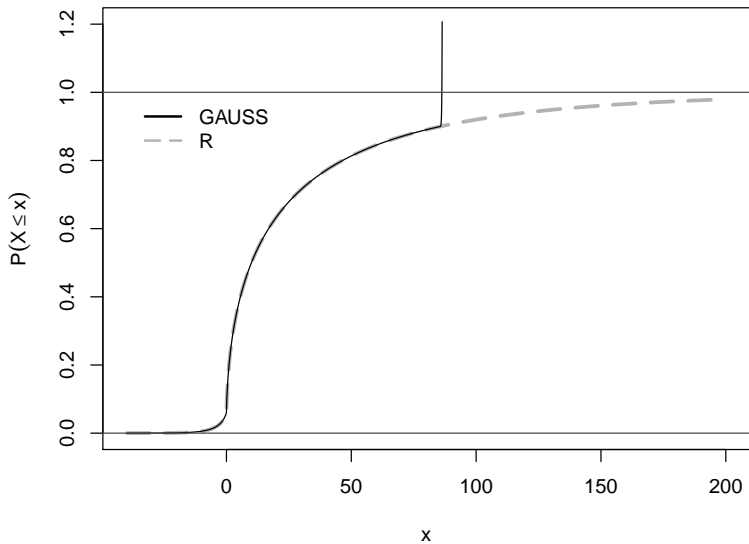
# Forensic econometrics: Structural change models

## What is going on?

- Computation of confidence intervals is based on asymptotic theory, leading to nonstandard distribution function.
- Quantiles need to be computed from functional of a two-sided Brownian motion with different scales and drifts.
- This also involves the term:  $\exp(ax) \cdot \Phi(-b\sqrt{x})$ .
- For second breakpoint:  $a \approx 8.31$ ,  $b \approx 4.08$ , and  $x \in [0, 300]$ .
- Product of a huge and a tiny number, numerically very instable.
- Better:  $\exp\{ax + \log \Phi(-b\sqrt{x})\}$  and compute  $\log \Phi$  directly.
- But still: GAUSS 3.2.38 (and even up to 6.0.8) chokes on this.



# Forensic econometrics: Structural change models



# Forensic econometrics: Structural change models

Computation of  $\Phi(y)$  and  $\log(\Phi(y))$  with  $y = -4.08\sqrt{x}$  in GAUSS 3.2.38 and R (all versions at least since 2.1.1).

x	GAUSS cdfn(y)	R pnorm(y)	GAUSS lncdfn(y)	R pnorm(y, log.p=TRUE)
82	$4.23e - 299$	$4.23e - 299$	-687.03	-687.03
84	$2.46e - 306$	$2.46e - 306$	-703.69	-703.69
86	$2.23e - 308$	0	-720.35	-720.35
88	$2.23e - 308$	0	-737.01	-737.01
90	$2.23e - 308$	0	$-\infty$	-753.66

# Forensic econometrics: Structural change models

## Conclusions:

- Be careful about numerical precision of your own code . . .
- . . . and also the functions of your programming environment.
- Replication would not have been successful without access to GAUSS code of Bai and Perron.

## Epilogue:

- Aptech fixed `lncdfn()` in recent versions of GAUSS (after initial private e-mails to us claiming that our computations were wrong).
- Stata published a very good and openly available C implementation for normal log-probabilities.

# Software tools

**Typically:** An econometric analysis encompasses the following.

- Data handling.
- Data analysis in some programming environment.
- Document preparation with results of the analysis.

**Question:** Which software can assist the researcher in making such an analysis reproducible?

# Software tools: Version control

**Often:** Research is carried out

- over an extended period,
- by several authors,
- on several computers,
- and hence difficult to reconstruct *exactly*.

**Problem:** Files proliferate with inconsistent naming conventions, get overwritten or deleted or are ultimately archived upon paper acceptance . . . or next disk crash.

**Idea:** Employ version control tools.

- Only one current version of each file.
- But full history of all changes in database.
- Annotate changes in log files.
- Enable moving back and forth through revisions.

# Software tools: Version control

## Work flow:

- Initially, check out a repository of files.
- Subsequently, easily check out updates by other authors.
- Work on files and commit own changes.
- All changes, additions, removals stored in repository.

## Software:

- Popularized through internet and open software development.
- Various packages available: CVS, SVN, Git, Mercurial, . . .
- Probably most popular for small to medium sized projects:  
*Subversion (SVN)*.
- Only “diffs” stored in each revision.
- On Windows: TortoiseSVN integrates with Explorer.

# Software tools: Data technologies and data archiving

## Typically:

- Data is *not* extremely large or complex.
- Flat plain text file ideal for reproducibility: Portable, easy to store and access.

## Furthermore:

- Relational database management systems for complex data, e.g., open-source systems PostgreSQL or MySQL.
- New standards for web-based sharing of data, e.g., XML or PHP.

# Software tools: Programming environments

**In general:** Many aspects drive choice of programming environment.

**Here:** Focus on aspects directly relevant to reproducibility.

## **Desirable:**

- Command line interface (CLI) or at least script from some graphical user interface (GUI).
- Modular code that is easy to read, encapsulates conceptual tasks, and is reusable in other settings.
- Open sources to enable gradual refinement.



# Software tools: Document preparation systems

## Two approaches:

- WYSIWYG (what you see is what you get) text processors (e.g., Microsoft Word, OpenOffice.org, LibreOffice, ...).
- Markup languages (like  $\text{\LaTeX}$ , HTML, ...).

## Again: Focus on reproducibility.

- Stable open standards preferred.
- Proprietary binary formats (such as Microsoft Word) problematic.
- Flat text files ideal for combination with version control.

# Software tools: Literate programming

**Idea:** Merge text, documentation, and computer code to facilitate keeping everything in sync.

## Literate programming:

- Single file contains documentation and computer code.
- *Tangling*: Extract computer code.
- *Weaving*: Produce documentation that optionally shows or hides the code.

## Literate data analysis:

- Extend weaving step: Execute code to produce all numeric output, tables, figures, etc.
- Sweave in R: Combines R code with  $\text{\LaTeX}$  (or HTML, ODF, ...).
- Results in “dynamic” or “revivable” documents.

# Challenges and conclusions

**Real challenge:** Better incentives from journals and funding agencies for archiving and distribution of details underlying empirical and/or computational work.

**Goal:** Convince authors that providing such details will enhance the chances for publication and citation of their work.

**However:** Advances in (open) software make it relatively easy to enhance reproducibility without too much extra effort.

**At eeecon:** New server and support by the admins provide web-based services, in particular web space, working paper series, SVN, . . .