

# 13-Identification of Breakpoints in Time Series

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## 1 Text needs to be written

The example is adapted from the help pages of R package “strucchange”, see Zeileis, A. et al. (2002), Journal of Statistical Software, 7(2), <https://www.jstatsoft.org/v07/i02/>

```
library(strucchange)
```

Lade nötiges Paket: zoo

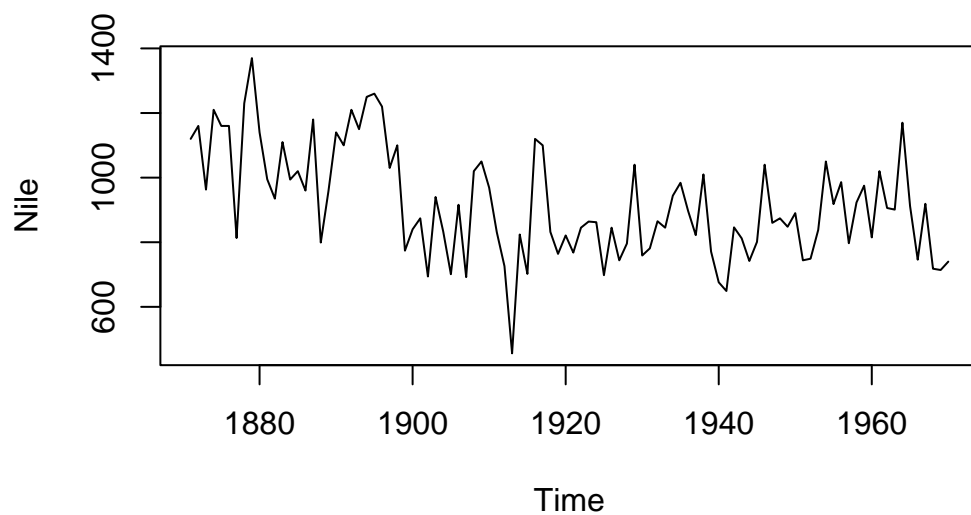
Attache Paket: 'zoo'

Die folgenden Objekte sind maskiert von 'package:base':

```
as.Date, as.Date.numeric
```

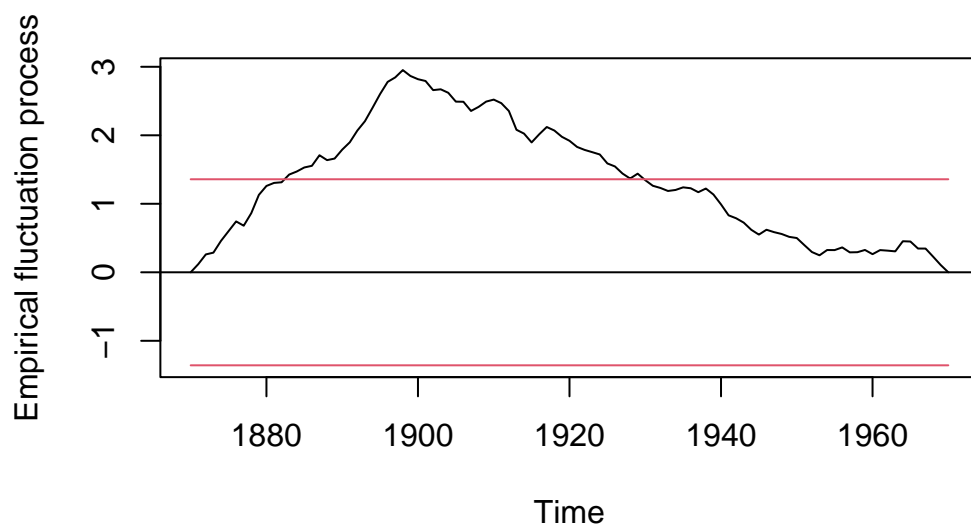
Lade nötiges Paket: sandwich

```
data("Nile")  
plot(Nile)
```



```
## OLS-CUSUM test for structural breaks in the time series
## are there periods with different discharge?
ocus <- efp(Nile ~ 1, type = "OLS-CUSUM")
plot(ocus)
```

### OLS-based CUSUM test



```
sctest(ocus)
```

OLS-based CUSUM test

```
data:  ocus
S0 = 2.9518, p-value = 5.409e-08
```

```
## identify time of structural break (with respect to mean value)
bp.nile <- breakpoints(Nile ~ 1)
summary(bp.nile)
```

Optimal (m+1)-segment partition:

Call:

```
breakpoints.formula(formula = Nile ~ 1)
```

Breakpoints at observation number:

m = 1	28				
m = 2	28	83			
m = 3	28	68	83		
m = 4	28	45	68	83	
m = 5	15	30	45	68	83

Corresponding to breakdates:

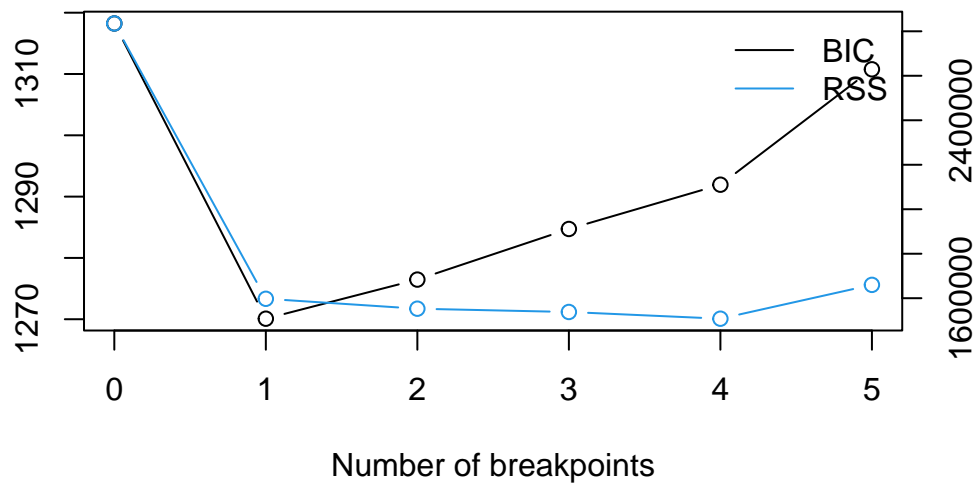
m = 1	1898				
m = 2	1898	1953			
m = 3	1898	1938	1953		
m = 4	1898	1915	1938	1953	
m = 5	1885	1900	1915	1938	1953

Fit:

m	0	1	2	3	4	5
RSS	2835157	1597457	1552924	1538097	1507888	1659994
BIC	1318	1270	1276	1285	1292	1311

```
## the BIC also chooses one breakpoint
plot(bp.nile)
```

## BIC and Residual Sum of Squares



```
## fit null hypothesis model and model with 1 breakpoint
fm0 <- lm(Nile ~ 1)
fm1 <- lm(Nile ~ breakfactor(bp.nile, breaks = 1))
plot(Nile)
lines(ts(fitted(fm0), start = 1871), col = 3)
lines(ts(fitted(fm1), start = 1871), col = 4)
lines(bp.nile)

## confidence interval
ci.nile <- confint(bp.nile)
ci.nile
```

Confidence intervals for breakpoints  
of optimal 2-segment partition:

Call:

```
confint.breakpointsfull(object = bp.nile)
```

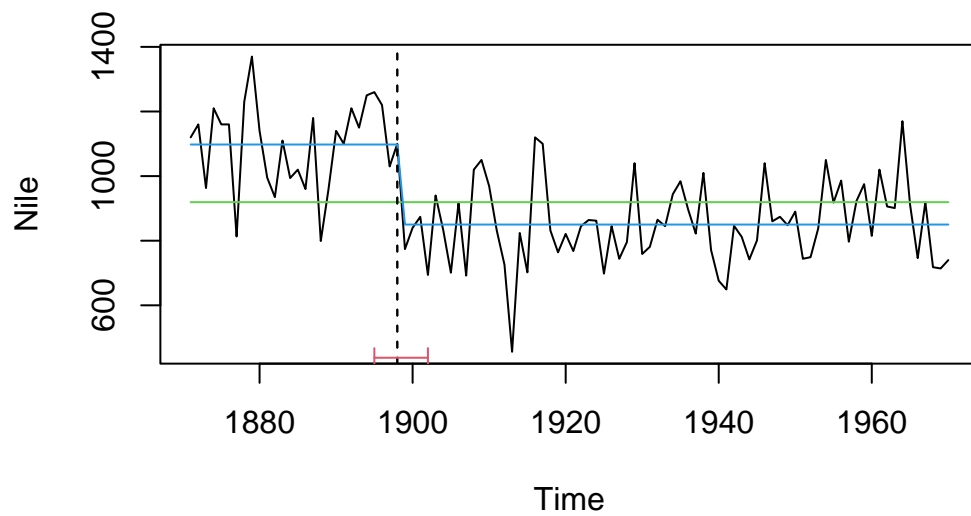
Breakpoints at observation number:

	2.5 % breakpoints	97.5 %
1	25	28 32

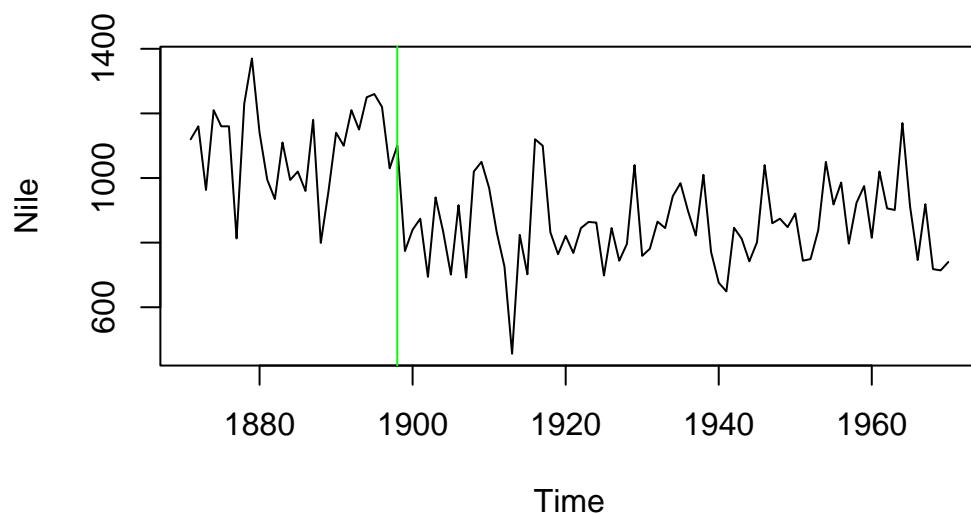
Corresponding to breakdates:

	2.5 % breakpoints	97.5 %
1	1895	1898 1902

```
lines(ci.nile)
```



```
## mark breakpoint using a simpler (and less fancy) method
plot(Nile)
dat <- data.frame(time = time(Nile), Q = as.vector(Nile))
abline(v=dat$time[bp.nile$breakpoints], col="green")
```



```
## ANOVA test whether the two models are significantly different
anova(fm0, fm1)
```

Analysis of Variance Table

Model 1: Nile ~ 1

Model 2: Nile ~ breakfactor(bp.nile, breaks = 1)

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	99	2835157				
2	98	1597457	1	1237700	75.93	7.439e-14 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
## alternative: AIC-based model comparison.
```

```
## The model with lower AIC is better
```

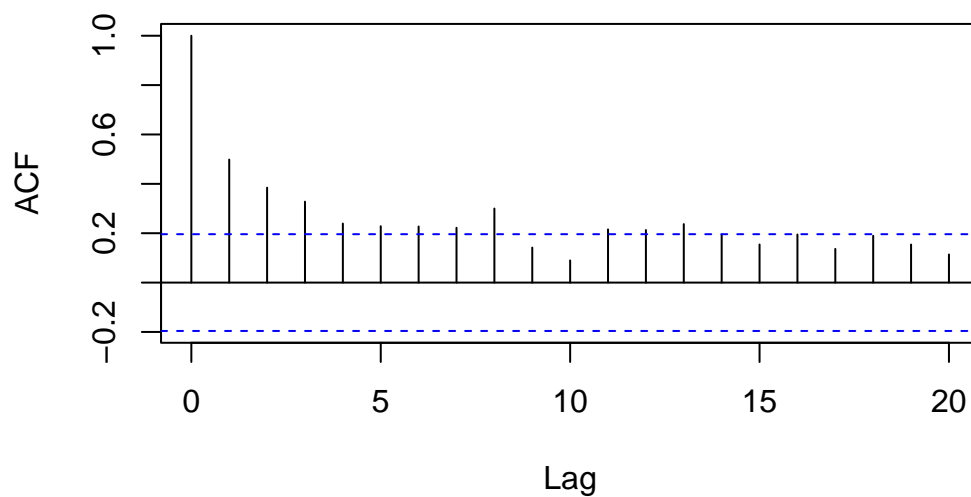
```
AIC(fm0,fm1)
```

	df	AIC
fm0	2	1313.031
fm1	3	1257.663

```
## some tests for quality and assumptions of the fitted model
```

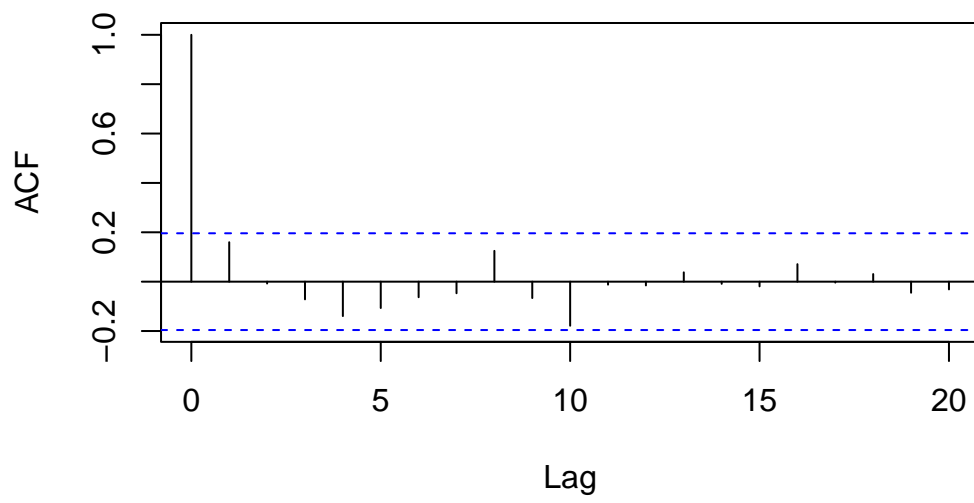
```
acf(residuals(fm0))
```

### Series residuals(fm0)



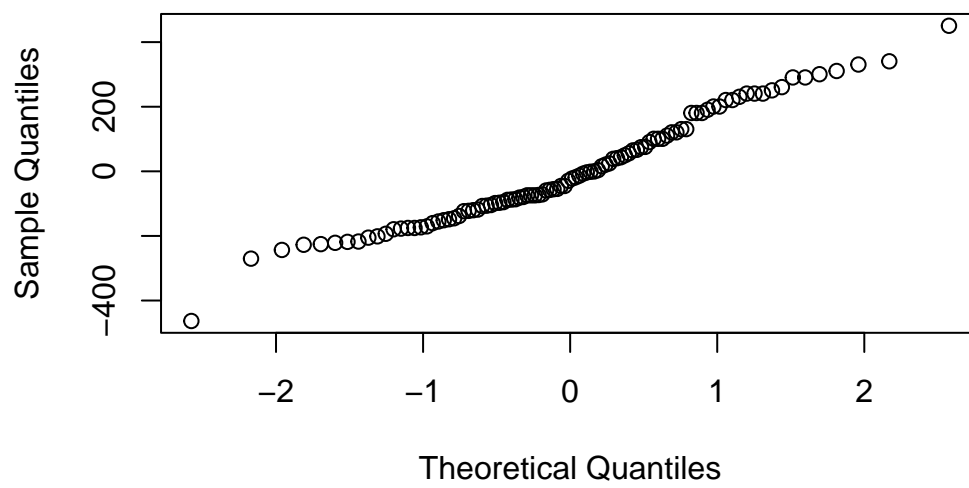
```
acf(residuals(fm1))
```

**Series residuals(fm1)**



```
qqnorm(residuals(fm0))
```

**Normal Q-Q Plot**



```
qqnorm(residuals(fm1))
```

**Normal Q-Q Plot**

