# Package 'qfa'

November 26, 2024

```
Type Package
Title Quantile-Frequency Analysis (QFA) of Time Series
Version 3.0
Date 2024-11-26
Maintainer Ta-Hsin Li <th1024@outlook.com>
Description
     Quantile-frequency analysis (QFA) of univariate or multivariate time series based on trigonomet-
     ric quantile regression. See Li, T.-H. (2012) ``Quantile periodograms", Journal of the Ameri-
     can Statistical Association, 107, 765–776, <doi:10.1080/01621459.2012.682815>; Li, T.-
     H. (2014) Time Series with Mixed Spectra, CRC Press, <doi:10.1201/b15154>; Li, T.-
     H. (2022) ` Quantile Fourier transform, quantile series, and nonparametric estimation of quan-
     tile spectra", <doi:10.48550/arXiv.2211.05844>.
Depends R (>= 3.5)
Imports RhpcBLASctl,
     doParallel,
     fields,
     foreach,
     mgcv,
     nlme,
     parallel,
     quantreg,
     splines,
     stats,
     graphics,
     colorRamps,
     MASS
License GPL (>=2)
URL https://www.r-project.org, https://github.com/IBM/qfa
NeedsCompilation yes
Encoding UTF-8
RoxygenNote 7.3.2
Contents
```

ar2qspec

	2qspec Quantile Spectrum from AR M	Model of Quantile Series or Quantile	
dex			31
			45
	tsqr.fit		
	tqr.fit		
	sqr.fit		
	sqdft		
	sar.gc.test		27
	sar.gc.bootstrap		23 26
	•		25
	sar.eq.bootstrap		23
	qspec2qcoh		23
	qspec.sqrlw		21 22
	qspec.sar		20
	qspec.qslw		19
	qspec.lwqs		18
	qspec.lw		17
	qspec.ar		16
	qsmooth.qper		15
	qsmooth.qdft		14
	qser2sar		13
	qser2qacf		12
	qser2ar		11
	qser		11
	qper2		10
	qper		9
	qkl.divergence		9
	qfa.plot		8
	qdft2qser		7
	qdft2qper		6
	qdft2qacf		5
	qdft		5
	qcser		4
	qacf		3
	per		3

# Description

This function computes quantile spectrum (QSPEC) or quantile-crossing spectrum (QCSPEC) from an AR model of quantile series (QSER) or quantile-crossing series (QCSER).

Crossing Series

```
ar2qspec(fit, freq = NULL)
```

per 3

#### **Arguments**

fit object of AR model from qser2sar() or qser2ar()

freq sequence of frequencies in [0,1) (default = NULL: all Fourier frequencies)

### Value

a list with the following elements:

spec matrix or array of quantile spectrum or quantile-crossing spectrum

freq sequence of frequencies

per

Periodogram (PER)

# Description

This function computes the periodogram or periodogram matrix for univariate or multivariate time series.

# Usage

per(y)

# Arguments

y time series (if multivariate, nrow(y) = length of time series)

#### Value

A vector or array of periodogram

### **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y.per <- per(y)
plot(y.per)</pre>
```

qacf

Quantile Autocovariance Function (QACF)

### **Description**

This function computes quantile autocovariance function (QACF) from time series or quantile discrete Fourier transform (QDFT).

```
qacf(y, tau, y.qdft = NULL, n.cores = 1, cl = NULL)
```

4 qcser

# **Arguments**

У	time series (if multivariate, nrow(y) = length of time series)
tau	sequence of quantile levels in $(0,1)$
y.qdft	matrix or array of pre-calculated QDFT (default = NULL: compute from y and tau); if y.qdft is supplied, y and tau can be left unspecified
n.cores	number of cores for parallel computing of QDFT if y . $qdft = NULL (default = 1)$
cl	pre-existing cluster for repeated parallel computing of QDFT (default = NULL)

#### Value

matrix or array of quantile autocovariance function

## **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
# compute from time series
y.qacf <- qacf(y,tau)
# compute from QDFT
y.qdft <- qdft(y,tau)
y.qacf <- qacf(y.qdft=y.qdft)</pre>
```

qcser

Quantile-Crossing Series (QCSER)

# **Description**

This function creates the quantile-crossing series (QCSER) for univariate or multivariate time series.

#### Usage

```
qcser(y, tau, normalize = FALSE)
```

#### **Arguments**

```
y time series (if multivariate, nrow(y) = length of time series)
tau sequence of quantile levels in (0,1)
normalize TRUE or FALSE (default): normalize QCSER to have unit variance
```

# Value

A matrix or array of quantile-crossing series

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.qser <- qcser(y,tau)
dim(y.qser)</pre>
```

qdft 5

qdft

Quantile Discrete Fourier Transform (QDFT)

### **Description**

This function computes quantile discrete Fourier transform (QDFT) for univariate or multivariate time series.

### Usage

```
qdft(y, tau, n.cores = 1, cl = NULL)
```

### **Arguments**

```
y time series (if multivariate, nrow(y) = length of time series)
tau sequence of quantile levels in (0,1)
n. cores number of cores for parallel computing (default = 1)
cl pre-existing cluster for repeated parallel computing (default = NULL)
```

#### Value

matrix or array of quantile discrete Fourier transform of y

# **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
tau <- seq(0.1,0.9,0.05)
y.qdft <- qdft(y,tau)
# Make a cluster for repeated use
n.cores <- 2
cl <- parallel::makeCluster(n.cores)
parallel::clusterExport(cl, c("tqr.fit"))
doParallel::registerDoParallel(cl)
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y.qdft <- qdft(y1,tau,n.cores=n.cores,cl=cl)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y.qdft <- qdft(y2,tau,n.cores=n.cores,cl=cl)
parallel::stopCluster(cl)</pre>
```

qdft2qacf

Quantile Autocovariance Function (QACF)

### **Description**

This function computes quantile autocovariance function (QACF) from QDFT.

```
qdft2qacf(y.qdft, return.qser = FALSE)
```

6 qdft2qper

#### **Arguments**

y.qdft matrix or array of QDFT from qdft() or SQDFT from sqdft()
return.qser if TRUE, return quantile series (QSER) along with QACF

#### Value

matrix or array of quantile autocovariance function if return.sqer = FALSE (default), else a list with the following elements:

qacf matirx or array of quantile autocovariance function

qser matrix or array of quantile series

### **Examples**

```
# single time series
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.qdft <- qdft(y1,tau)
y.qacf <- qdft2qacf(y.qdft)
plot(c(0:9),y.qacf[c(1:10),1],type='h',xlab="LAG",ylab="QACF")
y.qser <- qdft2qacf(y.qdft,return.qser=TRUE)$qser
plot(y.qser[,1],type='l',xlab="TIME",ylab="QSER")
# multiple time series
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y.qdft <- qdft(cbind(y1,y2),tau)
y.qacf <- qdft2qacf(y.qdft)
plot(c(0:9),y.qacf[1,2,c(1:10),1],type='h',xlab="LAG",ylab="QACF")</pre>
```

qdft2qper

Quantile Periodogram and Cross-Periodogram (QPER)

# Description

This function computes quantile periodogram/cross-periodogram (QPER) from QDFT.

# Usage

```
qdft2qper(y.qdft)
```

# **Arguments**

y.qdft matrix or array of QDFT from qdft()

#### Value

matrix or array of quantile periodogram/cross-periodogram

qdft2qser 7

### **Examples**

```
# single time series
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.qdft <- qdft(y1,tau)
y.qper <- qdft2qper(y.qdft)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
qfa.plot(ff[sel.f],tau,Re(y.qper[sel.f,]))
# multiple time series
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y.qdft <- qdft(cbind(y1,y2),tau)
y.qper <- qdft2qper(y.qdft)
qfa.plot(ff[sel.f],tau,Re(y.qper[1,1,sel.f,]))
qfa.plot(ff[sel.f],tau,Re(y.qper[1,2,sel.f,]))</pre>
```

qdft2qser

Quantile Series (QSER)

# **Description**

This function computes quantile series (QSER) from QDFT.

### Usage

```
qdft2qser(y.qdft)
```

#### **Arguments**

y.qdft

matrix or array of QDFT from qdft() or SQDFT from sqdft()

### Value

matrix or array of quantile series

```
# single time series
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.qdft <- qdft(y1,tau)
y.qser <- qdft2qser(y.qdft)
plot(y.qser[,1],type='l',xlab="TIME",ylab="QSER")
# multiple time series
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y.qdft <- qdft(cbind(y1,y2),tau)
y.qser <- qdft2qser(y.qdft)
plot(y.qser[1,,1],type='l',xlab="TIME",ylab="QSER")</pre>
```

8 qfa.plot

qfa.plot

Quantile-Frequency Plot

# Description

This function creates an image plot of quantile spectrum.

# Usage

```
qfa.plot(
  freq,
  tau,
  rqper,
  rg.qper = range(rqper),
  rg.tau = range(tau),
  rg.freq = c(0, 0.5),
  color = colorRamps::matlab.like2(1024),
  ylab = "QUANTILE LEVEL",
  xlab = "FREQUENCY",
  tlab = NULL,
  set.par = TRUE,
  legend.plot = TRUE
)
```

# Arguments

freq	sequence of frequencies in $(0,0.5)$ at which quantile spectrum is evaluated
tau	sequence of quantile levels in $(0,1)$ at which quantile spectrum is evaluated
rqper	real-valued matrix of quantile spectrum evaluated on the freq x tau grid
rg.qper	<pre>zlim for qper (default = range(qper))</pre>
rg.tau	<pre>ylim for tau (default = range(tau))</pre>
rg.freq	xlim for freq (default = $c(0, 0.5)$ )
color	<pre>colors (default = colorRamps::matlab.like2(1024))</pre>
ylab	label of y-axis (default = "QUANTILE LEVEL")
xlab	label of x-axis (default = "FREQUENCY")
tlab	title of plot (default = NULL)
set.par	if TRUE, par() is set internally (single image)
legend.plot	if TRUE, legend plot is added

### Value

no return value

qkl.divergence 9

qkl.divergence Kullback-Leibler Divergence of Quantile Spectral Estimate
--

# Description

This function computes Kullback-Leibler divergence (KLD) of quantile spectral estimate.

# Usage

```
qkl.divergence(y.qper, qspec, sel.f = NULL, sel.tau = NULL)
```

### **Arguments**

y.qper	matrix or array of quantile spectral estimate from, e.g., qspec.lw()
qspec	matrix of array of true quantile spectrum/cross-spectrum (same dimension as $y.qper$ )
sel.f	index of selected frequencies for computation (default = NULL: all frequencies)
sel.tau	index of selected quantile levels for computation (default = $NULL$ : all quantile levels)

#### Value

real number of Kullback-Leibler divergence

qper	Quantile Periodogram and Cross-Periodogram (QPER)	

# Description

This function computes quantile periodogram/cross-periodogram (QPER) from time series or quantile discrete Fourier transform (QDFT).

# Usage

```
qper(y, tau, y.qdft = NULL, n.cores = 1, cl = NULL)
```

# Arguments

У	time series (if multivariate, nrow(y) = length of time series)
tau	sequence of quantile levels in $(0,1)$
y.qdft	matrix or array of pre-calculated QDFT (default = NULL: compute from y and tau); if y.qdft is supplied, y and tau can be left unspecified
n.cores	number of cores for parallel computing of QDFT if y . $qdft = NULL (default = 1)$
cl	pre-existing cluster for repeated parallel computing of QDFT (default = NULL)

# Value

matrix or array of quantile periodogram/cross-periodogram

10 qper2

### **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
# compute from time series
y.qper <- qper(y,tau)
# compute from QDFT
y.qdft <- qdft(y,tau)
y.qper <- qper(y.qdft=y.qdft)</pre>
```

qper2

Quantile Periodogram Type II (QPER2)

# Description

This function computes type-II quantile periodogram for univariate time series.

# Usage

```
qper2(y, freq, tau, weights = NULL, n.cores = 1, cl = NULL)
```

# **Arguments**

У	univariate time series
freq	sequence of frequencies in [0,1)
tau	sequence of quantile levels in (0,1)
weights	sequence of weights in quantile regression (default = NULL: weights equal to 1)
n.cores	number of cores for parallel computing (default = 1)
cl	pre-existing cluster for repeated parallel computing (default = NULL)

### Value

matrix of quantile periodogram evaluated on freq \* tau grid

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qper2 <- qper2(y,ff,tau)
qfa.plot(ff[sel.f],tau,Re(y.qper2[sel.f,]))</pre>
```

qser 11

qser	Quantile Series (QSER)	

### **Description**

This function computes quantile series (QSER) from time series or quantile discrete Fourier transform (QDFT).

### Usage

```
qser(y, tau, y.qdft = NULL, n.cores = 1, cl = NULL)
```

#### **Arguments**

У	time series (if multivariate, $nrow(y) = length$ of time series)
tau	sequence of quantile levels in $(0,1)$
y.qdft	matrix or array of pre-calculated QDFT (default = NULL: compute from y and tau); if y.qdft is supplied, y and tau can be left unspecified
n.cores	number of cores for parallel computing of QDFT if y . $qdft = NULL (default = 1)$
cl	pre-existing cluster for repeated parallel computing of QDFT (default = NULL)

### Value

matrix or array of quantile series

# Examples

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
# compute from time series
y.qser <- qser(y,tau)
# compute from QDFT
y.qdft <- qdft(y,tau)
y.qser <- qser(y.qdft=y.qdft)</pre>
```

qser2ar

Autoregression (AR) Model of Quantile Series or Quantile-Crossing Series with or without Post-Smoothing

# **Description**

This function fits an autoregression (AR) model to quantile series (QSER) or quantile-crossing series (QCSER) for each quantile level using stats::ar().

```
qser2ar(y.qser, p = NULL, order.max = NULL, method = c("NA", "gamm", "sp"))
```

12 qser2qacf

#### **Arguments**

y.qser matrix or array of pre-calculated QSER or QCSER using qser() or qcser()

p order of AR model (default = NULL: selected by AIC)

order.max maximum order for AIC if p = NULL (default = NULL: determined by stats::ar())

method post-smoothing method: "gamm", "sp", or "NA" (default)

#### Value

a list with the following elements:

A matrix or array of AR coefficients

V vector or matrix of residual covariance

p order of AR modeln length of time seriesresiduals matrix or array of residuals

### **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.qser <- qcser(y,tau)
fit.ar <- qser2ar(y.qser,p=1,method="NA")
plot(tau,fit.ar$A[1,])
fit.ars <- qser2ar(y.qser,p=1,method="sp")
lines(tau,fit.ar$A[1,])</pre>
```

qser2qacf

ACF of Quantile Series or Quantile-Crossing Series

#### **Description**

This function creates the ACF of quantile series (QSER) or quantile-crossing series (QCSER)

# Usage

```
qser2qacf(y.qser)
```

# **Arguments**

```
y.qser QSER or QCSER computed from qser() or qcser()
```

### Value

A matrix or array of ACF

```
y \leftarrow stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)

tau \leftarrow seq(0.1,0.9,0.05)

y.qser \leftarrow qcser(y,tau)

y.qacf \leftarrow qser2qacf(y.qser)

dim(y.qacf)
```

qser2sar 13

qser2sar Spline Autoregression (SAR) Model of Quantile Series or q Crossing Series	Quantile-
---	-----------

# Description

This function fits spline autoregression (SAR) model to quantile series (QSER) or quantile-crossing series (QCSER).

# Usage

```
qser2sar(
  y.qser,
  tau,
  d = 1,
  p = NULL,
  order.max = NULL,
  spar = NULL,
  method = c("GCV", "AIC", "BIC"),
  weighted = FALSE
)
```

# Arguments

y.qser	matrix or array of pre-calculated QSER or QCSER using qser() or qcser()
tau	sequence of quantile levels where y. qser is calculated
d	subsampling rate of quantile levels (default = 1)
р	order of SAR model (default = NULL: automatically selected by AIC)
order.max	$maximum\ order\ for\ AIC\ if\ p=NULL\ (default=NULL:\ determined\ by\ stats::ar())$
spar	penalty parameter alla smooth.spline (default = NULL: automatically selected)
method	criterion for penalty parameter selection: "AIC" (default), "BIC", or "GCV"
weighted	if TRUE, penalty function is weighted (default = FALSE)

# Value

a list with the following elements:

A	matrix or array of SAR coefficients
V	vector or matrix of SAR residual covariance
р	order of SAR model
spar	penalty parameter
tau	sequence of quantile levels
n	length of time series
d	subsampling rate of quantile levels
weighted	option for weighted penalty function
fit	object containing details of SAR fit

14 gsmooth.qdft

qsmooth.qdft

Quantile Smoothing of Quantile Discrete Fourier Transform

# Description

This function computes quantile-smoothed version of quantile discrete Fourier transform (QDFT).

### Usage

```
qsmooth.qdft(
  y.qdft,
  method = c("gamm", "sp"),
  spar = "GCV",
  n.cores = 1,
  cl = NULL
)
```

### **Arguments**

```
y.qdft matrix or array of QDFT from qdft()
method smoothing method: "gamm" for mgcv::gamm() (default), "sp" for stats::smooth.spline()
spar smoothing parameter in smooth.spline() if method = "sp" (default = "GCV")
n.cores number of cores for parallel computing (default = 1)
cl pre-existing cluster for repeated parallel computing (default = NULL)
```

### Value

matrix or array of quantile-smoothed QDFT

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qdft <- qdft(cbind(y1,y2),tau)
y.qdft <- qsmooth.qdft(y.qdft,method="sp",spar=0.9)
y.qacf <- qdft2qacf(y.qdft)
y.qper.qslw <- qspec.lw(y.qacf,M=5)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.qslw[1,1,sel.f,]))</pre>
```

qsmooth.qper 15

qsmooth.qper

Quantile Smoothing of Quantile Periodogram or Spectral Estimate

# **Description**

This function computes quantile-smoothed version of quantile periodogram/cross-periodogram (QPER) or other quantile spectral estimate.

### Usage

```
qsmooth.qper(
  y.qper,
  method = c("gamm", "sp"),
  spar = "GCV",
  n.cores = 1,
  cl = NULL
)
```

# **Arguments**

```
y.qper matrix or array of quantile periodogram/cross-periodogram or spectral estimate
method smoothing method: "gamm" for mgcv::gamm() (default), "sp" for stats::smooth.spline()
spar smoothing parameter in smooth.spline() if method = "sp" (default = "GCV")
n.cores number of cores for parallel computing (default = 1)
cl pre-existing cluster for repeated parallel computing (default = NULL)
```

#### Value

matrix or array of quantile-smoothed quantile spectral estimate

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qdft <- qdft(cbind(y1,y2),tau)
y.qacf <- qdft2qacf(y.qdft)
y.qper.lw <- qspec.lw(y.qacf,M=5)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.lw[1,1,sel.f,]))
y.qper.lwqs <- qsmooth.qper(y.qper.lw,method="sp",spar=0.9)
qfa.plot(ff[sel.f],tau,Re(y.qper.lwqs[1,1,sel.f,]))</pre>
```

16 qspec.ar

qspec.ar	Autoregression (AR) Estimator of Quantile Spectrum or Quantile- Crossing Spectrum

# Description

This function computes autoregression (AR) estimate of quantile spectrum (QSPEC) or quantile-crossing spectrum (QCSPEC) from a time series (for QSPEC only) or from pre-calculated quantile series (QSER) or quantile-crossing series (QCSER).

# Usage

```
qspec.ar(
   y,
   tau,
   y.qser = NULL,
   p = NULL,
   order.max = NULL,
   freq = NULL,
   method = c("NA", "gamm", "sp"),
   n.cores = 1,
   c1 = NULL
)
```

# Arguments

У	time series (if multivariate, $nrow(y)$ = length of time series) for QSPEC only
tau	sequence of quantile levels in (0,1)
y.qser	pre-calculated QSER or QCSER from qdft2qser(), qser(), or qcser() (default = NULL: compute from y and tau for QSPEC only); if y.qser is supplied, y and tau can be left unspecified
р	order of AR model (default = NULL: automatically selected by AIC)
order.max	$maximum\ order\ for\ AIC\ if\ p\ =\ NULL\ (default\ =\ NULL:\ determined\ by\ stats::ar())$
freq	sequence of frequencies in [0,1) (default = NULL: all Fourier frequencies)
method	post-smoothing method: "gamm", "sp", or "NA" (default)
n.cores	number of cores for parallel computing of QDFT if y.qser = NULL (default = 1)
cl	pre-existing cluster for repeated parallel computing of QDFT (default = NULL)

# Value

a list with the following elements:

spec	matrix or array of AR quantile spectrum/cross-spectrum
freq	sequence of frequencies
fit	object of AR model
gser	matrix or array of quantile series if y.gser = NULL

qspec.lw 17

#### **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y <- cbind(y1,y2)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qspec.ar <- qspec.ar(y,tau,p=1)$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.ar[1,1,sel.f,]))
y.qspec.ar <- qspec.ar(y,qspec.ar[sel.f,]))
y.qspec.ar <- qspec.ar(y,qspec.ar[sel.f,]))</pre>
```

qspec.lw

Lag-Window (LW) Estimator of Quantile Spectrum or Quantile-Crossing Spectrum

### **Description**

This function computes lag-window (LW) estimate of quantile spectrum (QSPEC) or quantile-crossing spectrum (QCSPEC) from the ACF of quantile series (QSER) or quantile-crossing series (QCSER).

### Usage

```
qspec.lw(y.qacf, M = NULL)
```

#### **Arguments**

y.qacf pre-calculated ACF of QSER or QCSER from qdft2qacf() or qser2qacf()

M bandwidth parameter of lag window (default = NULL: quantile periodogram)

# Value

A list with the following elements:

spec matrix or array of quantile spectrum/cross-spectrum

lag-window sequence

### **Examples**

1w

```
# single time series
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.qdft <- qdft(y1,tau)
y.qacf <- qdft2qacf(y.qdft)
y.qspec.lw <- qspec.lw(y.qacf,M=5)$spec
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
qfa.plot(ff[sel.f],tau,Re(y.qspec.lw[sel.f,]))
# multiple time series</pre>
```

18 qspec.lwqs

```
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y.qdft <- qdft(cbind(y1,y2),tau)
y.qacf <- qdft2qacf(y.qdft)
y.qspec.lw <- qspec.lw(y.qacf,M=5)$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.lw[1,2,sel.f,]))</pre>
```

qspec.lwqs

Lag-Window-Quantile-Smoothing (LWQS) Estimator of Quantile Spectrum

# Description

This function computes lag-window-quantile-smoothing (LWQS) estimate of quantile spectrum/cross-spectrum from time series or quantile autocovariance function (QACF).

# Usage

```
qspec.lwqs(
   y,
   tau,
   y.qacf = NULL,
   M = NULL,
   method = c("gamm", "sp"),
   spar = "GCV",
   n.cores = 1,
   cl = NULL
)
```

# **Arguments**

у	time series (if multivariate, $nrow(y) = length$ of time series)
tau	sequence of quantile levels in $(0,1)$
y.qacf	matrix or array of pre-calculated QACF (default = NULL: compute from y and tau); if y.qacf is supplied, y and tau can be left unspecified
М	bandwidth parameter of lag window (default = NULL: quantile periodogram)
method	<pre>smoothing method: "gamm" for mgcv::gamm() (default), "sp" for stats::smooth.spline()</pre>
spar	<pre>smoothing parameter in smooth.spline() if method = "sp" (default = "GCV")</pre>
n.cores	number of cores for parallel computing (default = 1)
cl	pre-existing cluster for repeated parallel computing (default = NULL)

### Value

A list with the following elements:

spec matrix or array of quantile spectrum/cross-spectrum

spec.lw matrix or array of quantile spectrum/cross-spectrum without quantile smoothing

lw lag-window sequence

qacf matrix or array of quantile autocovariance function if y.qacf = NULL

qspec.qslw 19

### **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qper.lwqs <- qspec.lwqs(cbind(y1,y2),tau,M=5,method="sp",spar=0.9)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.lwqs[1,1,sel.f,]))</pre>
```

qspec.qslw

Quantile-Smoothing-Lag-Window (QSLW) Estimator of Quantile Spectrum

# Description

This function computes quantie-smoothing-lag-window (QSLW estimate of quantile spectrum/cross-spectrum from time series or quantile discrete Fourier transform (QDFT).

# Usage

```
qspec.qslw(
   y,
   tau,
   y.qdft = NULL,
   M = NULL,
   method = c("gamm", "sp"),
   spar = "GCV",
   n.cores = 1,
   cl = NULL
)
```

# **Arguments**

У	time series (if multivariate, $nrow(y)$ = length of time series)
tau	sequence of quantile levels in $(0,1)$
y.qdft	matrix or array of pre-calculated QDFT (default = NULL: compute from y and tau); if y.qdft is supplied, y and tau can be left unspecified
М	bandwidth parameter of lag window (default = NULL: quantile periodogram)
method	<pre>smoothing method: "gamm" for mgcv::gamm() (default), "sp" for stats::smooth.spline()</pre>
spar	<pre>smoothing parameter in smooth.spline() if method = 'sp' (default = "GCV")</pre>
n.cores	number of cores for parallel computing (default = 1)
cl	pre-existing cluster for repeated parallel computing (default = NULL)

### Value

A list with the following elements:

spec matrix or array of quantile spectrum/cross-spectrum

lw lag-window sequence

qdft matrix or array of quantile discrete Fourier transform if y.qdft = NULL

20 qspec.sar

#### **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qper.qslw <- qspec.qslw(cbind(y1,y2),tau,M=5,method="sp",spar=0.9)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.qslw[1,1,sel.f,]))</pre>
```

qspec.sar

Spline Autoregression (SAR) Estimator of Quantile Spectrum or Quatile-Crossing Spectrum

# Description

This function computes spline autoregression (SAR) estimate of quantile spectrum (QSPEC) or quantile-crossing spectrum (QCSPEC) from a time series (for QSPEC only) or from pre-calculated quantile series (QSER) or quantile-crossing series (QCSER).

# Usage

```
qspec.sar(
   y,
   y.qser = NULL,
   tau,
   d = 1,
   p = NULL,
   order.max = NULL,
   spar = NULL,
   method = c("GCV", "AIC", "BIC"),
   weighted = FALSE,
   freq = NULL,
   n.cores = 1,
   cl = NULL
)
```

### Arguments

```
time series (if multivarite, nrow(y) = length of time series) for QSPEC only
У
                  pre-calculated QSER or QCSER from qdft2qser(), qser(), or qcser() (de-
y.qser
                  fault = NULL: compute from y and tau) for QSPEC); if y. qser is supplied, y can
                  be left unspecified
                  sequence of quantile levels in (0,1)
tau
d
                  subsampling rate of quantile levels (default = 1)
                  order of SAR model (default = NULL: automatically selected by AIC)
                  maximum order for AIC if p = NULL (default = NULL: determined by stats::ar())
order.max
                  penalty parameter alla smooth. spline (default = NULL: automatically selected)
spar
                  criterion for penalty parameter selection: "GCV", "AIC" (default), or "BIC"
method
```

qspec.sqrlw 21

```
weighted if TRUE, penalty function is weighted (default = FALSE)

freq sequence of frequencies in [0,1) (default = NULL: all Fourier frequencies)

n. cores number of cores for parallel computing of QDFT if y. qser = NULL (default = 1)

pre-existing cluster for repeated parallel computing of QDFT (default = NULL)
```

#### Value

a list with the following elements:

spec matrix or array of SAR quantile spectrum or quantile-crossing spectrum
freq sequence of frequencies
fit object of SAR model
qser matrix or array of quantile series if y.qser = NULL

#### **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
# compute from time series
y.sar <- qspec.sar(cbind(y1,y2),tau=tau,p=1)
qfa.plot(ff[sel.f],tau,Re(y.sar$spec[1,1,sel.f,]))
# compute from quantile series
y.qser <- qser(cbind(y1,y2),tau)
y.qspec.sar <- qspec.sar(y.qser=y.qser,tau=tau,p=1)$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.sar[1,1,sel.f,]))</pre>
```

qspec.sqrlw Spline-Quantile-Regression-Lag-Window (SQRLW) Estimator of Quantile Spectrum

### **Description**

This function computes spline-quantile-regression-lag-window (SQRLW) estimate of quantile spectrum/cross-spectrum from time series or spline quantile discrete Fourier transform (SQDFT).

```
qspec.sqrlw(
  y,
  tau,
  y.sqdft = NULL,
  M = NULL,
  c0 = 0.02,
  d = 4,
  weighted = FALSE,
  n.cores = 1,
  c1 = NULL
)
```

22 qspec2qcoh

# **Arguments**

У	time series (if multivariate, nrow(y) = length of time series)
tau	sequence of quantile levels in $(0,1)$
y.sqdft	matrix or array of pre-calculated SQDFT (default = $NULL$ : compute from y and tau); if y. sqdft is supplied, y and tau can be left unspecified
М	bandwidth parameter of lag window (default = NULL: quantile periodogram)
c0	penalty parameter for SQDFT
d	subsampling rate of quantile levels for SQDFT (default = 1)
weighted	if TRUE, SQR penalty function is weighted (default = FALSE)
n.cores	number of cores for parallel computing of SQDFT (default = 1)
cl	pre-existing cluster for repeated parallel computing of SQDFT (default = NULL)

#### Value

A list with the following elements:

spec matrix or array of quantile spectrum/cross-spectrum

lw lag-window sequence

sqdft matrix or array of spline quantile discrete Fourier transform if y. sqdft = NULL

# **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
n <- length(y1)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qper.sqrlw <- qspec.sqrlw(cbind(y1,y2),tau,M=5,c0=0.02,d=4)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.sqrlw[1,1,sel.f,]))</pre>
```

qspec2qcoh

Quantile Coherence Spectrum

#### **Description**

This function computes quantile coherence spectrum (QCOH) from quantile spectrum and cross-spectrum of multiple time series.

# Usage

```
qspec2qcoh(qspec, k = 1, kk = 2)
```

# **Arguments**

qspec array of quantile spectrum/cross-spectrum

k index of first series (default = 1)kk index of second series (default = 2)

sar.eq.bootstrap 23

#### Value

matrix of quantile coherence evaluated at Fourier frequencies in (0,0.5)

#### **Examples**

sar.eq.bootstrap

Bootstrap Simulation of SAR Coefficients for Testing Equality of Granger-Causality in Two Samples

#### **Description**

This function simulates bootstrap samples of selected spline autoregression (SAR) coefficients for testing equality of Granger-causality in two samples based on their SAR models under H0: effect in each sample equals the average effect.

### Usage

```
sar.eq.bootstrap(
   y.qser,
   fit,
   fit2,
   index = c(1, 2),
   nsim = 1000,
   method = c("ar", "sar"),
   n.cores = 1,
   mthreads = TRUE,
   seed = 1234567
)
```

### **Arguments**

```
matrix or array of QSER from qser() or qspec.sar()$qser
y.qser
fit
                  object of SAR model from qser2sar() or qspec.sar()$fit
fit2
                  object of SAR model for the other sample
index
                  a pair of component indices for multiple time series or a sequence of lags for
                  single time series (default = c(1,2))
nsim
                  number of bootstrap samples (default = 1000)
                  method of residual calculation: "ar" (default) or "sar"
method
                  number of cores for parallel computing (default = 1)
n.cores
                  if TRUE (default), multithread BLAS is enabled
mthreads
seed
                  seed for random sampling (default = 1234567)
```

24 sar.eq.test

#### Value

array of simulated bootstrap samples of selected SAR coefficients

### **Examples**

```
y11 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y21 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
y12 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y22 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
tau <- seq(0.1,0.9,0.05)
y1.sar <- qspec.sar(cbind(y11,y21),tau=tau,p=1)
y2.sar <- qspec.sar(cbind(y12,y22),tau=tau,p=1)
A1.sim <- sar.eq.bootstrap(y1.sar$qser,y1.sar$fit,y2.sar$fit,index=c(1,2),nsim=5)
A2.sim <- sar.eq.bootstrap(y2.sar$qser,y2.sar$fit,y1.sar$fit,index=c(1,2),nsim=5)
dim(A1.sim)</pre>
```

sar.eq.test Wald Test and Confidence Band for Equality of SAR-Based Granger-Causality in Two Samples

# **Description**

This function computes Wald test and confidence band for equality of Granger-causality in two samples using bootstrap samples generated by sar.eq.bootstrap() based on the spline autoregression (SAR) models of quantile series (QSER).

### Usage

```
sar.eq.test(A1, A1.sim, A2, A2.sim, sel.lag = NULL, sel.tau = NULL)
```

#### **Arguments**

A1	matrix of selected SAR coefficients for sample 1
A1.sim	simulated bootstrap samples from $sar.eq.bootstrap()$ for sample 1
A2	matrix of selected SAR coefficients for sample 2
A2.sim	simulated bootstrap samples from sar.eq.bootstrap() for sample 2
sel.lag	indices of time lags for Wald test (default = NULL: all lags)
sel.tau	indices of quantile levels for Wald test (default = NULL: all quantiles)

#### Value

a list with the following elements:

test	list of Wald test result containing wald and p.value
D.u	matrix of upper limits of 95% confidence band for A1 – A2 $$
D.1	matrix of lower limits of 95% confidence band for A1 - A2

sar.gc.bootstrap 25

#### **Examples**

```
y11 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y21 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
y12 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
y22 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
tau <- seq(0.1,0.9,0.05)
y1.sar <- qspec.sar(cbind(y11,y21),tau=tau,p=1)
y2.sar <- qspec.sar(cbind(y12,y22),tau=tau,p=1)
A1.sim <- sar.eq.bootstrap(y1.sar$qser,y1.sar$fit,y2.sar$fit,index=c(1,2),nsim=5)
A2.sim <- sar.eq.bootstrap(y2.sar$qser,y2.sar$fit,y1.sar$fit,index=c(1,2),nsim=5)
A1 <- sar.gc.coef(y1.sar$fit,index=c(1,2))
A2 <- sar.gc.coef(y2.sar$fit,index=c(1,2))
sar.test <- sar.eq.test(A1,A1.sim,A2,A2.sim,sel.lag=NULL,sel.tau=NULL)</pre>
```

sar.gc.bootstrap

Bootstrap Simulation of SAR Coefficients for Granger-Causality Analvsis

#### **Description**

This function simulates bootstrap samples of selected spline autoregression (SAR) coefficients for Granger-causality analysis based on the SAR model of quantile series (QSER) under H0: (a) for multiple time series, the second series specified in index is not causal for the first series specified in index; (b) for single time series, the series is not causal at the lags specified in index.

### Usage

```
sar.gc.bootstrap(
   y.qser,
   fit,
   index = c(1, 2),
   nsim = 1000,
   method = c("ar", "sar"),
   n.cores = 1,
   mthreads = TRUE,
   seed = 1234567
)
```

#### **Arguments**

```
matrix or array of QSER from qser() or qspec.sar()$qser
y.qser
fit
                  object of SAR model from qser2sar() or qspec.sar()$fit
                  a pair of component indices for multiple time series or a sequence of lags for
index
                  single time series (default = c(1,2))
                  number of bootstrap samples (default = 1000)
nsim
                  method of residual calculation: "ar" (default) or "sar"
method
n.cores
                  number of cores for parallel computing (default = 1)
mthreads
                  if TRUE (default), multithread BLAS is enabled
                  seed for random sampling (default = 1234567)
seed
```

26 sar.gc.coef

#### Value

array of simulated bootstrap samples of selected SAR coefficients

#### **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
tau <- seq(0.1,0.9,0.05)
y.sar <- qspec.sar(cbind(y1,y2),tau=tau,p=1)
A.sim <- sar.gc.bootstrap(y.sar$qser,y.sar$fit,index=c(1,2),nsim=5)</pre>
```

sar.gc.coef

Extraction of SAR Coefficients for Granger-Causality Analysis

# Description

This function extracts the spline autoregression (SAR) coefficients from an SAR model for Granger-causality analysis. See sar.gc.bootstrap for more details regarding the use of index.

### Usage

```
sar.gc.coef(fit, index = c(1, 2))
```

### **Arguments**

fit object of SAR model from qser2sar() or qspec.sar()\$fit

index a pair of component indices for multiple time series or a sequence of lags for single time series (default = c(1,2))

### Value

matrix of selected SAR coefficients (number of lags by number of quantiles)

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.sar <- qspec.sar(cbind(y1,y2),tau=tau,p=1)
A <- sar.gc.coef(y.sar$fit,index=c(1,2))</pre>
```

sar.gc.test 27

sar.gc.test Wald Test and Confidence Band for SAR-Based Granger-Causality Analysis	lity
---	------

# **Description**

This function computes Wald test and confidence band for Granger-causality using bootstrap samples generated by sar.gc.bootstrap() based the spline autoregression (SAR) model of quantile series (QSER).

### Usage

```
sar.gc.test(A, A.sim, sel.lag = NULL, sel.tau = NULL)
```

### **Arguments**

Α	matrix of selected SAR coefficients
A.sim	simulated bootstrap samples from sar.gc.bootstrap()
sel.lag	indices of time lags for Wald test (default = NULL: all lags)
sel.tau	indices of quantile levels for Wald test (default = NULL: all quantiles)

#### Value

a list with the following elements:

test	list of Wald test result containing wald and p.value
A.u	matrix of upper limits of 95% confidence band of A
A.1	matrix of lower limits of 95% confidence band of A

### **Examples**

```
y1 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=32)
y2 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=32)
tau <- seq(0.1,0.9,0.05)
y.sar <- qspec.sar(cbind(y1,y2),tau=tau,p=1)
A <- sar.gc.coef(y.sar$fit,index=c(1,2))
A.sim <- sar.gc.bootstrap(y.sar$qser,y.sar$fit,index=c(1,2),nsim=5)
y.gc <- sar.gc.test(A,A.sim)</pre>
```

sqdft

Spline Quantile Discrete Fourier Transform (SQDFT)

# Description

This function computes spline quantile discrete Fourier transform (SQDFT) for univariate or multivariate time series through trigonometric spline quantile regression.

```
sqdft(y, tau, c0 = 0.02, d = 4, weighted = FALSE, n.cores = 1, cl = NULL)
```

28 sqr.fit

# **Arguments**

У		time series (if multivariate, nrow(y) = length of time series)
ta	u	sequence of quantile levels in $(0,1)$
c0		penalty parameter
d		subsampling rate of quantile levels (default = 1)
we	ighted	if TRUE, penalty function is weighted (default = FALSE)
n.	cores	number of cores for parallel computing (default = 1)
cl		pre-existing cluster for repeated parallel computing (default = NULL)

#### Value

matrix or array of the spline quantile discrete Fourier transform of y

# **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
y.sqdft <- sqdft(y,tau,c0=0.02,d=4)
n <- length(y)
ff <- c(0:(n-1))/n
sel.f <- which(ff > 0 & ff < 0.5)
y.qacf <- qdft2qacf(y.sqdft)
y.qper.sqrlw <- qspec.lw(y.qacf,M=5)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.sqrlw[sel.f,]))</pre>
```

sqr.fit

Spline Quantile Regression (SQR)

### **Description**

This function computes spline quantile regression (SQR) solution from response vector and design matrix. It uses the FORTRAN code rqfnb.f in the "quantreg" package with the kind permission of Dr. R. Koenker.

### Usage

```
sqr.fit(y, X, tau, c0, d = 1, weighted = FALSE, mthreads = TRUE)
```

# Arguments

у	response vector
X	<pre>design matrix (nrow(X) = length(y))</pre>
tau	sequence of quantile levels in (0,1)
c0	penalty parameter
d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)
mthreads	if TRUE (default), multithread BLAS is enabled

tqr.fit 29

#### Value

A list with the following elements:

coefficients matrix of regression coefficients number of iterations

tqr.fit

Trigonometric Quantile Regression (TQR)

# Description

This function computes trigonometric quantile regression (TQR) for univariate time series at a single frequency.

### Usage

```
tqr.fit(y, f0, tau, prepared = TRUE)
```

### **Arguments**

y time series

f0 frequency in [0,1)

tau sequence of quantile levels in (0,1)

prepared if TRUE, intercept is removed and coef of cosine is doubled when f0 = 0.5

# Value

```
object of rq() (coefficients in $coef)
```

### **Examples**

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
fit <- tqr.fit(y,f0=0.1,tau=tau)
plot(tau,fit$coef[1,],type='o',pch=0.75,xlab='QUANTILE LEVEL',ylab='TQR COEF')</pre>
```

tsqr.fit

Trigonometric Spline Quantile Regression (TSQR)

#### **Description**

This function computes trigonometric spline quantile regression (TSQR) for univariate time series at a single frequency.

```
tsqr.fit(y, f0, tau, c0, d = 1, weighted = FALSE, prepared = TRUE)
```

30 tsqr.fit

# **Arguments**

У	vector of time series
f0	frequency in [0,1)
tau	sequence of quantile levels in (0,1)
с0	penalty parameter
d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)

prepared if TRUE, intercept is removed and coef of cosine is doubled when f0 = 0.5

# Value

```
object of sqr.fit() (coefficients in $coef)
```

```
y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
tau <- seq(0.1,0.9,0.05)
fit <- tqr.fit(y,f0=0.1,tau=tau)
fit.sqr <- tsqr.fit(y,f0=0.1,tau=tau,c0=0.02,d=4)
plot(tau,fit$coef[1,],type='p',xlab='QUANTILE LEVEL',ylab='TQR COEF')
lines(tau,fit.sqr$coef[1,],type='l')</pre>
```

# **Index**

```
ar2qspec, 2
per, 3
qacf, 3
qcser, 4
qdft, 5
qdft2qacf, 5
qdft2qper, 6
qdft2qser, 7
qfa.plot, 8
{\tt qkl.divergence}, {\color{red} 9}
qper, 9
qper2, 10
qser, 11
qser2ar, 11
qser2qacf, 12
qser2sar, 13
{\tt qsmooth.qdft,}\ 14
qsmooth.qper, 15
qspec.ar, 16
qspec.lw, 17
qspec.lwqs, 18
qspec.qslw, 19
qspec.sar, 20
qspec.sqrlw, 21
qspec2qcoh, 22
sar.eq.bootstrap, 23
sar.eq.test, 24
sar.gc.bootstrap, 25
sar.gc.coef, 26
sar.gc.test, 27
sqdft, 27
sqr.fit, 28
tqr.fit, 29
\mathsf{tsqr}.\mathsf{fit}, 29
```