

# Package ‘qfa’

January 8, 2025

**Type** Package

**Title** Quantile-Frequency Analysis (QFA) of Time Series

**Version** 4.0

**Date** 2025-01-06

**Maintainer** Ta-Hsin Li <th1024@outlook.com>

## Description

Quantile-frequency analysis (QFA) of time series based on trigonometric quantile regression. Spline quantile regression (SQR) for regression coefficient estimation.

References:

- [1] Li, T.-H. (2012) ``Quantile periodograms'', Journal of the American Statistical Association, 107, 765–776, <doi:10.1080/01621459.2012.682815>.
- [2] Li, T.-H. (2014) Time Series with Mixed Spectra, CRC Press, <doi:10.1201/b15154>
- [3] Li, T.-H. (2022) ``Quantile Fourier transform, quantile series, and nonparametric estimation of quantile spectra'', <doi:10.48550/arXiv.2211.05844>.
- [4] Li, T.-H. (2024) ``Quantile crossing spectrum and spline autoregression estimation'', <doi:10.48550/arXiv.2412.02513>.
- [5] Li, T.-H. (2024) ``Spline autoregression method for estimation of quantile spectrum'', <doi:10.48550/arXiv.2412.17163>.

**Depends** R (>= 3.5)

**Imports** RhpcBLASctl, doParallel, fields, foreach, mgcv, nlme, parallel, quantreg, splines, stats, graphics, colorRamps, MASS

**License** GPL (>= 2)

**URL** <https://github.com/IBM/qfa>, <https://github.com/th12019/QFA>

**NeedsCompilation** yes

**Encoding** UTF-8

**RoxygenNote** 7.3.2

**Author** Ta-Hsin Li [cre, aut]

**Repository** CRAN

**Date/Publication** 2025-01-08 08:40:06 UTC

## Contents

birthweight	2
engel	3
per	4
qacf	4
qcser	5
qdft	6
qdft2qacf	7
qdft2qper	8
qdft2qser	8
qfa.plot	9
qkl.divergence	10
qper	11
qper2	11
qser	12
qser2ar	13
qser2qacf	14
qser2sar	14
qspec.ar	15
qspec.lw	17
qspec.sar	18
qspec2qcoh	19
sar.eq.bootstrap	20
sar.eq.test	21
sar.gc.bootstrap	22
sar.gc.coef	23
sar.gc.test	24
sqdft	25
sqdft.fit	26
sqr	27
sqr.fit	28
sqr.fit.optim	29
tqr.fit	31
tsqr.fit	32
yearssn	33
<b>Index</b>	<b>34</b>

---

birthweight

*Birthweight data*

---

### Description

Infant birth weight data. Precare and Education should be treated as factors.

**Usage**

```
data(birthweight)
```

**Format**

An object of class `data.frame` with 50000 rows and 12 columns.

**Source**

natality2022us.csv, <<https://www.nber.org/research/data/vital-statistics-natality-birth-data>>

**References**

Koenker, R. (2005). Quantile Regression. Cambridge University Press.

---

engel

*Engel food expenditure data*

---

**Description**

The Engel food expenditure data from the R package `quantreg`.

**Usage**

```
data(engel)
```

**Format**

An object of class `data.frame` with 235 rows and 2 columns.

**References**

Koenker, R. (2005). Quantile Regression. Cambridge University Press.

---

per *Periodogram (PER)*

---

### Description

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

### Usage

```
per(y)
```

### Arguments

y vector or matrix of time series s (if matrix, nrow(y) = length of time series)

### Value

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)
```

---

qacf *Quantile Autocovariance Function (QACF)*

---

### Description

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

### Usage

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

### Arguments

y vector or matrix of time series (if matrix, 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)
```

---

qcser

*Quantile-Crossing Series (QC SER)*

---

**Description**

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

**Usage**

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

**Arguments**

y	vector or matrix of time series
tau	sequence of quantile levels in (0,1)
normalize	TRUE or FALSE (default): normalize QC SER to have unit variance

**Value**

A matrix or array of quantile-crossing 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)
y.qser <- qcser(y,tau)
dim(y.qser)
```

---

`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**

<code>y</code>	vector or matrix of time series (if matrix, <code>nrow(y)</code> = length of time series)
<code>tau</code>	sequence of quantile levels in (0,1)
<code>n.cores</code>	number of cores for parallel computing (default = 1)
<code>cl</code>	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=64)
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=64)
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=64)
y.qdft <- qdft(y2,tau,n.cores=n.cores,cl=cl)
parallel::stopCluster(cl)
```

qdft2qacf

*Quantile Autocovariance Function (QACF)***Description**

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

**Usage**

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

**Arguments**

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

**Value**

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

`qacf` matrix 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")
```

---

qdft2qper                      *Quantile Periodogram (QPER)*

---

### Description

This function computes quantile 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

### 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,]))
```

---

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()

**Value**

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.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,],type='l',xlab="TIME",ylab="QSER")
```

---

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	zlim for qper (default = range(qper))
rg.tau	ylim for tau (default = range(tau))
rg.freq	xlim for freq (default = c(0, 0.5))
color	colors (default = colorRamps::matlab.like2(1024))
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	<i>Kullback-Leibler Divergence of Quantile Spectral Estimate</i>
----------------	--

---

**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 (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 (QPER)*

---

**Description**

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

**Usage**

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

**Arguments**

y	vector or matrix of time series (if matrix, 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

**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)
```

---

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**

y	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

**Examples**

```
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,]))
```

---

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**

y	vector or matrix of time series (if matrix, 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)

```

---

qser2ar

*Autoregression (AR) Model of Quantile Series*


---

**Description**

This function fits an autoregression (AR) model to quantile series (QSER) separately for each quantile level using `stats::ar()`.

**Usage**

```
qser2ar(y.qser, p = NULL, order.max = NULL, method = c("none", "gamm", "sp"))
```

**Arguments**

<code>y.qser</code>	matrix or array of pre-calculated QSER, e.g., using <code>qser()</code>
<code>p</code>	order of AR model (default = NULL: selected by AIC)
<code>order.max</code>	maximum order for AIC if <code>p = NULL</code> (default = NULL: determined by <code>stats::ar()</code> )
<code>method</code>	quantile smoothing method: "gamm", "sp", or "NA" (default)

**Value**

a list with the following elements:

<code>A</code>	matrix or array of AR coefficients
<code>V</code>	vector or matrix of residual covariance
<code>p</code>	order of AR model
<code>n</code>	length of time series
<code>residuals</code>	matrix or array of residuals

---

qser2qacf

*ACF of Quantile Series (QSER) or Quantile-Crossing Series (QCACF)*


---

**Description**

This function creates the ACF of quantile series or quantile-crossing series

**Usage**

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

**Arguments**

y.qser            matrix or array of quantile-crossing series

**Value**

A matrix or array of ACF

**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)
y.qacf <- qser2qacf(y.qser)
dim(y.qacf)
```

---

qser2sar

*Spline Autoregression (SAR) Model of Quantile Series*


---

**Description**

This function fits spline autoregression (SAR) model to quantile series (QSER).

**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, e.g., using qser()
tau	sequence of quantile levels where y.qser is calculated
d	subsampling rate of quantile levels (default = 1)
p	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
p	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

---

qspec.ar

*Autoregression (AR) Estimator of Quantile Spectrum*


---

**Description**

This function computes autoregression (AR) estimate of quantile spectrum from time series or quantile series (QSER).

**Usage**

```
qspec.ar(
  y,
  tau,
  y.qser = NULL,
  p = NULL,
  order.max = NULL,
  freq = NULL,
  method = c("none", "gamm", "sp"),
  n.cores = 1,
  cl = NULL
)
```

**Arguments**

<code>y</code>	vector or matrix of time series (if matrix, <code>nrow(y)</code> = length of time series)
<code>tau</code>	sequence of quantile levels in (0,1)
<code>y.qser</code>	matrix or array of pre-calculated QSER (default = NULL: compute from <code>y</code> and <code>tau</code> );
<code>p</code>	order of AR model (default = NULL: automatically selected by AIC)
<code>order.max</code>	maximum order for AIC if <code>p</code> = NULL (default = NULL: determined by <code>stats::ar()</code> )
<code>freq</code>	sequence of frequencies in [0,1) (default = NULL: all Fourier frequencies)
<code>method</code>	quantile smoothing method: "gamm" for <code>mgcv::gamm()</code> , "sp" for <code>stats::smooth.spline()</code> , or "none" (default) if <code>y.qser</code> is supplied, <code>y</code> and <code>tau</code> can be left unspecified
<code>n.cores</code>	number of cores for parallel computing of QDFT if <code>y.qser</code> = NULL (default = 1)
<code>cl</code>	pre-existing cluster for repeated parallel computing of QDFT (default = NULL)

**Value**

a list with the following elements:

<code>spec</code>	matrix or array of AR quantile spectrum
<code>freq</code>	sequence of frequencies
<code>fit</code>	object of AR model
<code>qser</code>	matrix or array of quantile series if <code>y.qser</code> = 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)
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)$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.ar[1,1,sel.f,]))
y.qser <- qcser(y1,tau)
y.qspec.ar <- qspec.ar(y.qser=y.qser,p=1)$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.ar[sel.f,]))
y.qspec.arqs <- qspec.ar(y.qser=y.qser,p=1,method="sp")$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.arqs[sel.f,]))

```



---

qspec.lw

*Lag-Window (LW) Estimator of Quantile Spectrum*


---

### Description

This function computes lag-window (LW) estimate of quantile spectrum with or without quantile smoothing from time series or quantile autocovariance function (QACF).

### Usage

```
qspec.lw(
  y,
  tau,
  y.qacf = NULL,
  M = NULL,
  method = c("none", "gamm", "sp"),
  spar = "GCV",
  n.cores = 1,
  cl = NULL
)
```

### Arguments

y	vector or matrix of time series (if matrix, <code>nrow(y)</code> = 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
M	bandwidth parameter of lag window (default = NULL: quantile periodogram)
method	quantile smoothing method: "gamm" for <code>mgcv::gamm()</code> , "sp" for <code>stats::smooth.spline()</code> , or "none" (default)
spar	smoothing parameter in <code>smooth.spline()</code> 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

A list with the following elements:

spec	matrix or array of spectral estimate
spec.lw	matrix or array of spectral estimate without quantile smoothing
lw	lag-window sequence
qacf	matrix or array of quantile autocovariance function if y.qacf = 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.qacf <- qacf(cbind(y1,y2),tau)
y.qper.lw <- qspec.lw(y.qacf=y.qacf,M=5)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.lw[1,1,sel.f,]))
y.qper.lwqs <- qspec.lw(y.qacf=y.qacf,M=5,method="sp",spar=0.9)$spec
qfa.plot(ff[sel.f],tau,Re(y.qper.lwqs[1,1,sel.f,]))

```

---

qspec.sar

*Spline Autoregression (SAR) Estimator of Quantile Spectrum*


---

**Description**

This function computes spline autoregression (SAR) estimate of quantile spectrum.

**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**

y	vector or matrix of time series (if matrix, <code>nrow(y)</code> = length of time series)
y.qser	matrix or array of pre-calculated QSER (default = NULL: compute from y and tau); if y.qser is supplied, y can be left unspecified
tau	sequence of quantile levels in (0,1)
d	subsampling rate of quantile levels (default = 1)
p	order of SAR model (default = NULL: automatically selected by AIC)
order.max	maximum order for AIC if p = NULL (default = NULL: determined by <code>stats::ar()</code> )

spar	penalty parameter alla <code>smooth.spline</code> (default = NULL: automatically selected)
method	criterion for penalty parameter selection: "GCV", "AIC" (default), or "BIC"
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 <code>y.qser = NULL</code> (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 SAR quantile spectrum
freq	sequence of frequencies
fit	object of SAR model
qser	matrix or array of quantile series if <code>y.qser = NULL</code>

**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$spec[1,1,sel.f,]))
# compute from quantile series
y.qser <- qser(cbind(y1,y2),tau)
y.sar <- qspec.sar(y.qser=y.qser,tau=tau,p=1)
qfa.plot(ff[sel.f],tau,Re(y.sar$spec[1,1,sel.f,]))

```

---

qspec2qcoh

*Quantile Coherence Spectrum*


---

**Description**

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

**Usage**

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

**Arguments**

qspect	array of quantile spectrum
k	index of first series (default = 1)
kk	index of second series (default = 2)

**Value**

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

**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.qacf <- qacf(cbind(y1,y2),tau)
y.qper.lw <- qspec.lw(y.qacf=y.qacf,M=5)$spec
y.qcoh <- qspec2qcoh(y.qper.lw,k=1,kk=2)
qfa.plot(ff[sel.f],tau,y.qcoh)

```

---

sar.eq.bootstrap	<i>Bootstrap Simulation of SAR Coefficients for Testing Equality of Granger-Causality in Two Samples</i>
------------------	--

---

**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**

y.qser	matrix or array of QSER from qser() or qspec.sar()\$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	method of residual calculation: "ar" (default) or "sar"
n.cores	number of cores for parallel computing (default = 1)
mthreads	if FALSE, set RhpcBLASctl::blas_set_num_threads(1) (default = TRUE)
seed	seed for random sampling (default = 1234567)

**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=64)
y21 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y12 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y22 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
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)

```

---

sar.eq.test

*Wald Test and Confidence Band for Equality of 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.l	matrix of lower limits of 95% confidence band for A1 - A2

**Examples**

```

y11 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y21 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
y12 <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)
y22 <- stats::arima.sim(list(order=c(1,0,0), ar=-0.5), n=64)
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))
test <- sar.eq.test(A1,A1.sim,A2,A2.sim,sel.lag=NULL,sel.tau=NULL)

```

---

sar.gc.bootstrap      *Bootstrap Simulation of SAR Coefficients for Granger-Causality Analysis*

---

**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**

y.qser	matrix or array of QSER from qser() or qspec.sar()\$qser
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))
nsim	number of bootstrap samples (default = 1000)
method	method of residual calculation: "ar" (default) or "sar"
n.cores	number of cores for parallel computing (default = 1)
mthreads	if FALSE, set RnpcBLASctl::blas_set_num_threads(1) (default = TRUE)
seed	seed for random sampling (default = 1234567)

**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=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.sim <- sar.gc.bootstrap(y.sar$qser,y.sar$fit,index=c(1,2),nsim=5)

```

---

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)

**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)
y.sar <- qspec.sar(cbind(y1,y2), tau=tau, p=1)
A <- sar.gc.coef(y.sar$fit, index=c(1,2))
```

---

sar.gc.test

*Wald Test and Confidence Band for Granger-Causality Analysis*

---

**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**

`A` 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.l` 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=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$fit,index=c(1,2))
A.sim <- sar.gc.bootstrap(y.sar$qser,y.sar$fit,index=c(1,2),nsim=5)
y.gc <- sar.gc.test(A,A.sim)

```

sqdft

*Spline Quantile Discrete Fourier Transform (SQDFT) of Time Series***Description**

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

**Usage**

```

sqdft(
  y,
  tau,
  spar = NULL,
  d = 1,
  weighted = FALSE,
  method = c("AIC", "BIC", "SIC"),
  ztol = 1e-05,
  n.cores = 1,
  cl = NULL
)

```

**Arguments**

<code>y</code>	vector or matrix of time series (if matrix, <code>nrow(y)</code> = length of time series)
<code>tau</code>	sequence of quantile levels in (0,1)
<code>spar</code>	smoothing parameter: if <code>spar=NULL</code> , smoothing parameter is selected by <code>method</code>
<code>d</code>	subsampling rate of quantile levels (default = 1)
<code>weighted</code>	if TRUE, penalty function is weighted (default = FALSE)
<code>method</code>	criterion for smoothing parameter selection when <code>spar=NULL</code> ("AIC", "BIC", or "SIC")
<code>ztol</code>	zero tolerance parameter used to determine the effective dimensionality of the fit
<code>n.cores</code>	number of cores for parallel computing (default = 1)
<code>cl</code>	pre-existing cluster for repeated parallel computing (default = NULL)

**Value**

A list with the following elements:

coefficients	matrix of regression coefficients
qdft	matrix or array of the spline quantile discrete Fourier transform of y
crit	criteria for smoothing parameter selection: (AIC,BIC,SIC)

**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,spar=NULL,d=4,metho="AIC")$qdft
```

---

sqdft.fit	<i>Spline Quantile Discrete Fourier Transform (SQDFT) of Time Series Given Smoothing Parameter</i>
-----------	--

---

**Description**

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

**Usage**

```
sqdft.fit(
  y,
  tau,
  spar = 1,
  d = 1,
  weighted = FALSE,
  ztol = 1e-05,
  n.cores = 1,
  cl = NULL
)
```

**Arguments**

y	vector or matrix of time series (if matrix, nrow(y) = length of time series)
tau	sequence of quantile levels in (0,1)
spar	smoothing parameter
d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)
ztol	zero tolerance parameter used to determine the effective dimensionality of the fit
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:

coefficients	matrix of regression coefficients
qdft	matrix or array of the spline quantile discrete Fourier transform of y
crit	criteria for smoothing parameter selection: (AIC,BIC,SIC)

**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.fit(y,tau,spar=1,d=4)$qdft
```

---

sqr

*Spline Quantile Regression (SQR) by formula*


---

**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(
  formula,
  tau = seq(0.1, 0.9, 0.2),
  spar = NULL,
  d = 1,
  data,
  subset,
  na.action,
  model = TRUE,
  weighted = FALSE,
  mthreads = TRUE,
  method = c("AIC", "BIC", "SIC"),
  ztol = 1e-05
)
```

**Arguments**

formula	a formula object, with the response on the left of a ~ operator, and the terms, separated by + operators, on the right.
tau	sequence of quantile levels in (0,1)
spar	smoothing parameter: if spar=NULL, smoothing parameter is selected by method
d	subsampling rate of quantile levels (default = 1)

data	a data.frame in which to interpret the variables named in the formula
subset	an optional vector specifying a subset of observations to be used
na.action	a function to filter missing data (see rq in the 'quantreg' package)
model	if TRUE then the model frame is returned (needed for calling summary subsequently)
weighted	if TRUE, penalty function is weighted (default = FALSE)
mthreads	if FALSE, set RhpcBLASctl::blas_set_num_threads(1) (default = TRUE)
method	a criterion for smoothing parameter selection if spar=NULL ("AIC", "BIC", or "SIC")
ztol	a zero tolerance parameter used to determine the effective dimensionality of the fit

**Value**

object of SQR fit

**Examples**

```
library(quantreg)
data(engel)
engel$income <- engel$income - mean(engel$income)
tau <- seq(0.1,0.9,0.05)
fit <- rq(foodexp ~ income,tau=tau,data=engel)
fit.sqr <- sqr(foodexp ~ income,tau=tau,spar=0.5,data=engel)
par(mfrow=c(1,1),pty="m",lab=c(10,10,2),mar=c(4,4,2,1)+0.1,las=1)
plot(tau,fit$coef[2,],xlab="Quantile Level",ylab="Coeff1")
lines(tau,fit.sqr$coef[2,])
```

---

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 `rqf.nb.f` in the "quantreg" package with the kind permission of Dr. R. Koenker.

**Usage**

```
sqr.fit(
  X,
  y,
  tau,
  spar = 1,
  d = 1,
  weighted = FALSE,
```

```

    mthreads = TRUE,
    ztol = 1e-05
  )

```

### Arguments

X	design matrix (nrow(X) = length(y))
y	response vector
tau	sequence of quantile levels in (0,1)
spar	smoothing parameter
d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)
mthreads	if FALSE, set RnpcBLASctl::blas_set_num_threads(1) (default = TRUE)
ztol	zero tolerance parameter used to determine the effective dimensionality of the fit

### Value

A list with the following elements:

coefficients	matrix of regression coefficients
crit	sequence criteria for smoothing parameter select: (AIC,BIC,SIC)
np	sequence of number of effective parameters
fid	sequence of fidelity measure as quasi-likelihood
nit	number of iterations

---

sqr.fit.optim

*Spline Quantile Regression (SQR) by Gradient Algorithms*


---

### Description

This function computes spline quantile regression by a gradient algorithm BFGS, ADAM, or GRAD.

### Usage

```

sqr.fit.optim(
  X,
  y,
  tau,
  spar = 0,
  d = 1,
  weighted = FALSE,
  method = c("BFGS", "ADAM", "GRAD"),
  beta.rq = NULL,

```

```

theta0 = NULL,
spar0 = NULL,
sg.rate = c(1, 1),
mthreads = TRUE,
control = list(trace = 0)
)

```

### Arguments

<code>X</code>	vector or matrix of explanatory variables (including intercept)
<code>y</code>	vector of dependent variable
<code>tau</code>	sequence of quantile levels in (0,1)
<code>spar</code>	smoothing parameter
<code>d</code>	subsampling rate of quantile levels (default = 1)
<code>weighted</code>	if TRUE, penalty function is weighted (default = FALSE)
<code>method</code>	optimization method: "BFGS" (default), "ADAM", or "GRAD"
<code>beta.rq</code>	matrix of regression coefficients from <code>quantreg::rq(y~X)</code> for initialization (default = NULL)
<code>theta0</code>	initial value of spline coefficients (default = NULL)
<code>spar0</code>	smoothing parameter for <code>stats::smooth.spline()</code> to smooth <code>beta.rq</code> for initialization (default = NULL)
<code>sg.rate</code>	vector of sampling rates for quantiles and observations in stochastic gradient version of GRAD and ADAM
<code>mthreads</code>	if FALSE, set <code>RhpcBLASctl::blas_set_num_threads(1)</code> (default = TRUE)
<code>control</code>	a list of control parameters <ul style="list-style-type: none"> <li><code>maxit</code>: max number of iterations (default = 100)</li> <li><code>stepsize</code>: stepsize for ADAM and GRAD (default = 0.01)</li> <li><code>warmup</code>: length of warmup phase for ADAM and GRAD (default = 70)</li> <li><code>stepupdate</code>: frequency of update for ADAM and GRAD (default = 20)</li> <li><code>stepredn</code>: stepsize discount factor for ADAM and GRAD (default = 0.2)</li> <li><code>line.search.type</code>: line search option (1,2,3,4) for GRAD (default = 1)</li> <li><code>line.search.max</code>: max number of line search trials for GRAD (default = 1)</li> <li><code>seed</code>: seed for stochastic version of ADAM and GRAD (default = 1000)</li> <li><code>trace</code>: -1 return results from all iterations, 0 (default) return final result</li> </ul>

### Value

A list with the following elements:

<code>beta</code>	matrix of regression coefficients
<code>all.beta</code>	coefficients from all iterations for GRAD and ADAM
<code>spars</code>	smoothing parameters from <code>stats::smooth.spline()</code> for initialization
<code>fit</code>	object from the optimization algorithm

**Examples**

```

data(engel)
y <- engel$foodexp
X <- cbind(rep(1,length(y)),engel$income-mean(engel$income))
tau <- seq(0.1,0.9,0.05)
fit.rq <- quantreg::rq(y ~ X[,2],tau)
fit.sqr <- sqr(y ~ X[,2],tau,d=2,spar=0.2)
fit <- sqr.fit.optim(X,y,tau,spar=0.2,d=2,method="BFSG",beta.rq=fit.rq$coef)
fit <- sqr.fit.optim(X,y,tau,spar=0.2,d=2,method="BFSG",beta.rq=fit.rq$coef)
par(mfrow=c(1,2),pty="m",lab=c(10,10,2),mar=c(4,4,2,1)+0.1,las=1)
for(j in c(1:2)) {
  plot(tau,fit.rq$coef[j,],type="n",xlab="QUANTILE LEVEL",ylab=paste0("COEFF",j))
  points(tau,fit.rq$coef[j,],pch=1,cex=0.5)
  lines(tau,fit.sqr$coef[j,],lty=1); lines(tau,fit$beta[j,],lty=2,col=2)
}

```

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	vector of 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')

```

tsqr.fit

*Trigonometric Spline Quantile Regression (TSQR) of Time Series***Description**

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

**Usage**

```
tsqr.fit(
  y,
  f0,
  tau,
  spar = 1,
  d = 1,
  weighted = FALSE,
  mthreads = TRUE,
  prepared = TRUE,
  ztol = 1e-05
)
```

**Arguments**

y	time series
f0	frequency in [0,1)
tau	sequence of quantile levels in (0,1)
spar	smoothing parameter
d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)
mthreads	if FALSE, set RnpcBLASctl::blas_set_num_threads(1) (default = TRUE)
prepared	if TRUE, intercept is removed and coef of cosine is doubled when f0 = 0.5
ztol	zero tolerance parameter used to determine the effective dimensionality of the fit

**Value**

object of `sqr.fit()` (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)
fit.sqr <- tsqr.fit(y,f0=0.1,tau=tau,spar=1,d=4)
plot(tau,fit$coef[1,],type='p',xlab='QUANTILE LEVEL',ylab='TQR COEF')
lines(tau,fit.sqr$coef[1,],type='l')
```



---

yearssn	<i>Yearly sunspot numbers</i>
---------	-------------------------------

---

**Description**

Sunspot numbers from 1700 to 2007.

**Usage**

```
data(yearssn)
```

**Format**

An object of class `data.frame` with 308 rows and 2 columns.

**References**

Li, T.-H. (2014). *Time Series with Mixed Spectra*. CRC Press.

# Index

## \* datasets

birthweight, 2  
engel, 3  
yearsasn, 33

birthweight, 2

engel, 3

per, 4

qacf, 4

qcser, 5

qdft, 6

qdft2qacf, 7

qdft2qper, 8

qdft2qser, 8

qfa.plot, 9

qk1.divergence, 10

qper, 11

qper2, 11

qser, 12

qser2ar, 13

qser2qacf, 14

qser2sar, 14

qspec.ar, 15

qspec.lw, 17

qspec.sar, 18

qspec2qcoh, 19

sar.eq.bootstrap, 20

sar.eq.test, 21

sar.gc.bootstrap, 22

sar.gc.coef, 23

sar.gc.test, 24

sqdft, 25

sqdft.fit, 26

sqr, 27

sqr.fit, 28

sqr.fit.optim, 29

tqr.fit, 31

tsqr.fit, 32

yearsasn, 33