## Package 'qfa'

April 8, 2025

```
Type Package
Title Quantile-Frequency Analysis (QFA) of Time Series
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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>.
       [6] Li, T.-H., and Megiddo, N. (2025) `Spline quantile regression,"
          <doi:10.48550/arXiv.2501.03883>.
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Imports RhpcBLASctl,
      doParallel,
      fields,
      foreach,
      mgcv,
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      quantreg,
      splines,
      stats,
      graphics,
      colorRamps,
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NeedsCompilation yes
```

2 birthweight

# **Encoding** UTF-8 **RoxygenNote** 7.3.2

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## Description

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

engel 3

## Usage

```
data(birthweight)
```

#### **Format**

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

#### **Source**

natality2022us.csv, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data">natality2022us.csv</a>, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data">natality2022us.csv</a>, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data">natality2022us.csv</a>, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data">natality2022us.csv</a>, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data">natality2022us.csv</a>, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data">natality2022us.csv</a>, <a href="https://www.nber.org/research/data/vital-statistics-natality-birth-data/vital-statisti

#### 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**

У

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

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#### 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

У	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)</pre>
```

qcser 5

a	CS	se	r

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 vector or matrix 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

#### **Examples**

```
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)
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

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

tau sequence of quantile levels in (0,1)

 $n.\,cores \qquad \qquad number \ of \ cores \ for \ parallel \ computing \ (default=1)$ 

cl pre-existing cluster for repeated parallel computing (default = NULL)

## Value

matrix or array of quantile discrete Fourier transform of y

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#### **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)</pre>
```

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.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>
```

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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,]))</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()

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#### 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='1',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='1',xlab="TIME",ylab="QSER")</pre>
```

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
)
```

```
freq sequence of frequencies in (0,0.5) at which quantile spectrum is evaluated 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 zlim for qper (default = range(qper)) ylim for tau (default = range(tau)) rg. freq xlim for freq (default = c(0,0.5)) color colors (default = colorRamps::matlab.like2(1024))
```

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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	Kullback-Leibler Divergence of Quantile Spectral Estimate
qki.divergence	Kuilback-Leibier 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 (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)
```

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## **Arguments**

У	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)</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

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#### **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,]))</pre>
```

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	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)</pre>
```

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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**

y.qser matrix or array of pre-calculated QSER, e.g., using qser()
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 quantile 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 series

residuals 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

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## **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)</pre>
```

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

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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,
   c1 = NULL
)
```

## Arguments

У	vector or matrix of time series (if matrix, $nrow(y) = length$ of time series)
tau	sequence of quantile levels in $(0,1)$
y.qser	matrix or array of pre-calculated QSER (default = NULL: compute from y and tau);
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	quantile smoothing method: "gamm" for mgcv::gamm(), "sp" for stats::smooth.spline(), or "none" (default) if y.qser is supplied, y and tau can be left unspecified
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

freq sequence of frequencies

fit object of AR model

qser matrix or array of quantile series if y.qser = NULL
```

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#### **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)$spec
qfa.plot(ff[sel.f],tau,Re(y.qspec.ar[1,1,sel.f,]))
y.qspec.ar <- qspec.ar(y.qser=y.qser,p=1)$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,]))</pre>
```

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
)
```

```
vector or matrix of time series (if matrix, nrow(y) = length of time series)
У
tau
                  sequence of quantile levels in (0,1)
                  matrix or array of pre-calculated QACF (default = NULL: compute from y and
y.qacf
                  tau); if y. qacf is supplied, y and tau can be left unspecified
М
                  bandwidth parameter of lag window (default = NULL: quantile periodogram)
                  quantile smoothing method: "gamm" for mgcv::gamm(), "sp" for stats::smooth.spline(),
method
                  or "none" (default)
                  smoothing parameter in smooth.spline() if method = "sp" (default = "GCV")
spar
                  number of cores for parallel computing (default = 1)
n.cores
                  pre-existing cluster for repeated parallel computing (default = NULL)
cl
```

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#### 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,]))</pre>
```

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,
   c1 = NULL
)
```

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## **Arguments**

У	vector or matrix of time series (if matrix, nrow(y) = 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)
р	order of SAR model (default = NULL: automatically selected by AIC)
order.max	$maximum\ order\ for\ AIC\ if\ p\ =\ NULL\ (default\ =\ NULL:\ determined\ by\ {\tt stats::ar()})$
spar	penalty parameter alla smooth.spline (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 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 SAR quantile 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$spec[1,1,sel.f,]))
# compute from quantile series
y.qser <- qser(cbind(y1,y2),tau)
y.sar <- qspec.sar(y.qser=y.qser,tau=tau,p=1)
qfa.plot(ff[sel.f],tau,Re(y.sar$spec[1,1,sel.f,]))</pre>
```

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		· ~	1-
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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**

qspec 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**

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.

sar.eq.test

#### 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)</pre>
```

 ${\sf sar.eq.test}$ 

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).

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#### 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 ${\tt sar.eq.bootstrap()}$ 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

#### **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,se1.lag=NULL,se1.tau=NULL)</pre>
```

```
{\it sar.gc.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.

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#### 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	$if \ FALSE, set \ RhpcBLASctl :: blas\_set\_num\_threads(1) \ (default = TRUE)$
seed	seed for random sampling (default = 1234567)

#### Value

array of simulated bootstrap samples of selected SAR coefficients

## **Examples**

```
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))
```

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## **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))</pre>
```

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.1	matrix of lower limits of 95% confidence band of A

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#### **Examples**

```
\label{eq:condition} \begin{array}{lll} 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) \end{array}
```

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"),
   ztol = 1e-05,
   n.cores = 1,
   cl = NULL
)
```

## Arguments

У	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: if spar=NULL, smoothing parameter is selected by method
d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)
method	crietrion for smoothing parameter selection when spar=NULL ("AIC" or "BIC")
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)
```

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#### **Examples**

```
\label{eq:condition} $$y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)$$ tau <- seq(0.1,0.9,0.05)$$ y.sqdft <- sqdft(y,tau,spar=NULL,d=4,metho="AIC")$qdft
```

sqdft.fit

Spline Quantile Discrete Fourier Transform (SQDFT) of Time Series Given Smoothing Parameter

## 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

у	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 Fouror BICier transform of y
crit	criteria for smoothing parameter selection: (AIC,BIC)

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#### **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"),
   ztol = 1e-05
)
```

formula	a formula object, with the response on the left of a $\sim$ 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" or "BIC")
ztol	a zero tolerance parameter used to determine the effective dimensionality of the fit

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#### 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,])</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(
    X,
    y,
    tau,
    spar = 1,
    d = 1,
    weighted = FALSE,
    mthreads = TRUE,
    ztol = 1e-05
)
```

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#### Value

A list with the following elements:

```
coefficients matrix of regression coefficients

crit sequence critera for smoothing parameter select: (AIC,BIC)

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)
)
```

```
Χ
                   vecor or matrix of explanatory variables (including intercept)
                   vector of dependent variable
У
                   sequence of quantile levels in (0,1)
tau
                   smoothing parameter
spar
d
                   subsampling rate of quantile levels (default = 1)
weighted
                   if TRUE, penalty function is weighted (default = FALSE)
                   optimization method: "BFGS" (default), "ADAM", or "GRAD"
method
                   matrix of regression coefficients from quantreg::rq(y~X) for initialization
beta.rq
                   (default = NULL)
theta0
                   initial value of spline coefficients (default = NULL)
```

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spar0 smoothing parameter for stats::smooth.spline() to smooth beta.rq for initilaiztion (default = NULL) sg.rate vector of sampling rates for quantiles and observations in stochastic gradient version of GRAD and ADAM mthreads if FALSE, set RhpcBLASctl::blas\_set\_num\_threads(1) (default = TRUE) control a list of control parameters maxit: max number of iterations (default = 100) stepsize: stepsize for ADAM and GRAD (default = 0.01) warmup: length of warmup phase for ADAM and GRAD (default = 70) stepupdate: frequency of update for ADAM and GRAD (default = 20) stepredn: stepsize discount factor for ADAM and GRAD (default = 0.2) line.search.type: line search option (1,2,3,4) for GRAD (default = 1) line.search.max: max number of line search trials for GRAD (default = 1) seed: seed for stochastic version of ADAM and GRAD (default = 1000) trace: -1 return results from all iterations, 0 (default) return final result

#### Value

A list with the following elements:

beta matrix of regression coefficients

all.beta coefficients from all iterations for GRAD and ADAM

spars smoothing parameters from stats::smooth.spline() for initialization

fit 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)
}</pre>
```

tqr.fit Trigonometric Quantile Regression (TQR)

#### **Description**

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

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#### 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**

```
\label{eq:comparison} $$y <- stats::arima.sim(list(order=c(1,0,0), ar=0.5), n=64)$$ 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
)
```

```
y time series

f0 frequency in [0,1)

tau sequence of quantile levels in (0,1)

spar smoothing parameter
```

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d	subsampling rate of quantile levels (default = 1)
weighted	if TRUE, penalty function is weighted (default = FALSE)
mthreads	$if\ FALSE, set\ RhpcBLASctl:: \mathtt{blas\_set\_num\_threads} (1)\ (default = TRUE)$
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')</pre>
```

yearssn

Yearly sunspot numbers

## 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.

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