# Package 'temStaR'

2 changeCovMtx2Rho

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# Description

Change covariance matrix to Rho matrix.

# Usage

changeCovMtx2Rho(CovMtx, alpha, theta, betaVec, PDflag = TRUE)

chf\_NTS 3

chf\_NTS

chf\_NTS

# Description

chf\_NTS calculates Ch.F of the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If a time parameter value is given, it calculates Ch.F of the NTS profess  $\phi(u) = E[\exp(iu(X(t+s)-X(s)))] = \exp(t\log(E[\exp(iuX(1))]))$ , where X is the NTS process generated by the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ .

# Usage

```
chf_NTS(u, param)
```

#### **Arguments**

ntsparam

u An array of u

A vector of the NTS parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . For NTS process case it is a

vector of parameters  $(\alpha, \theta, \beta, \gamma, \mu, t)$ .

#### Value

Characteristic function of the NTS distribution

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)</pre>
u \leftarrow seq(from = -2*pi, to = 2*pi, length.out = 101)
phi <- chf_NTS(u, ntsparam)</pre>
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
#scaling annual parameters to one day
dt <- 1/250 #one day
ntsparam <- c(alpha, theta, beta, gamma, mu, dt)</pre>
u \leftarrow seq(from = -2*pi, to = 2*pi, length.out = 101)
phi <- chf_NTS(u, ntsparam)</pre>
```

4 chf\_stdNTS

chf\_stdNTS

chf\_stdNTS

# Description

chf\_stdNTS calculates Ch.F of the standard NTS distribution with parameters  $(\alpha, \theta, \beta)$ . If a time parameter value is given, it calculates Ch.F of the standard NTS profess  $\phi(u) = E[\exp(iu(X(t+s)-X(s)))] = \exp(t\log(E[\exp(iuX(1))]))$ , where X is the standard NTS process generated by the standard NTS distribution with parameters  $(\alpha, \theta, \beta)$ .

#### Usage

```
chf_stdNTS(u, param)
```

#### **Arguments**

An array of u

ntsparam

A vector of the standard NTS parameters  $(\alpha, \theta, \beta)$ . For the standard NTS process case it is a vector of parameters  $(\alpha, \theta, \beta, t)$ .

#### Value

Characteristic function of the standard NTS distribution

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)</pre>
u \leftarrow seq(from = -2*pi, to = 2*pi, length.out = 101)
phi <- chf_stdNTS(u, ntsparam)</pre>
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
#scaling annual parameters to one day
dt <- 1/250 #one day
ntsparam <- c(alpha, theta, beta, gamma, mu, dt)</pre>
u \leftarrow seq(from = -2*pi, to = 2*pi, length.out = 101)
phi <- chf_stdNTS(u, ntsparam)</pre>
```

copulaStdNTS 5

copulaStdNTS	copulaStdNTS
--------------	--------------

#### **Description**

copulaStdNTS calculates the stdNTS copula values

#### Usage

```
copulaStdNTS(u, st, subTS = NULL)
```

#### References

Y. S. Kim, D. Volkmann (2013), Normal Tempered Stable Copula, Applied Mathematics Letters, 26(7), 676-680 https://www.sciencedirect.com/science/article/pii/S0893965913000384

cvarGauss cvarGauss

#### **Description**

Calculate the CVaR for the normal distributed market model. Developer's version.

#### Usage

```
cvarGauss(eta, mu = 0, sigma = 1)
```

cvarmarginalmnts cvarmarginalmnts

#### **Description**

cvarmarginalmnts calculates the CVaR of the n-th element of the multivariate NTS distributed random variable.

#### Usage

```
cvarmarginalmnts(eta, n, st)
```

#### **Arguments**

eta the significant level for CVaR. Real value between 0 and 1.

n the n-th element to be calculated.

st Structure of parameters for the n-dimensional NTS distribution.

6 cvarnts

cvarnts

cvarnts

# **Description**

cvarnts calculates Conditional Value at Risk (CVaR, or expected shortfall ES) of the NTS market model with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it calculates CVaR of the standard NTS distribution with parameter  $(\alpha, \theta, \beta)$ 

## Usage

```
cvarnts(eps, ntsparam)
```

#### Arguments

eps the significant level for CVaR. Real value between 0 and 1.

ntsparam A vector of the NTS parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . A vector of the standard NTS

parameters  $(\alpha, \theta, \beta)$ .

#### Value

CVaR of the NTS distribution.

#### References

- Y. S. Kim, S. T. Rachev, M. L. Bianchi, and F. J. Fabozzi (2010), Computing VaR and AVaR in infinitely divisible distributions, Probability and Mathematical Statistics, 30 (2), 223-245.
- S. T. Rachev, Y. S. Kim, M. L. Bianchi, and F. J. Fabozzi (2011), Financial Models with Levy Processes and Volatility Clustering, John Wiley & Sons

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)</pre>
u \leftarrow c(0.01, 0.05)
q <- cvarnts(u, ntsparam)</pre>
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
\verb|ntsparam| <- c(alpha, theta, beta, gamma, mu)|\\
u \leftarrow c(0.01, 0.05)
q <- cvarnts(u, ntsparam)</pre>
#Annual based parameters
alpha <- 1.2
theta <- 1
```

dBeta 7

```
beta <- -0.2 gamma <- 0.3 mu <- 0.1 #scaling annual parameters to one day dt <- 1/250 #one day ntsparam <- c(alpha, theta, beta, gamma, mu, dt) u <- c(0.01, 0.05) q <- c(varnts(u, ntsparam)
```

dBeta

dBeta

# Description

The first derivative of the beta. Developer's version.

# Usage

```
dBeta(n, w, betaArray, covMtx)
```

dcopulaStdNTS

dcopulaStdNTS

# Description

 ${\tt dcopulaStdNTS}\ calculates\ density\ of\ the\ stdNTS\ copula.$ 

# Usage

```
dcopulaStdNTS(u, st, subTS = NULL)
```

# References

Y. S. Kim, D. Volkmann (2013), Normal Tempered Stable Copula, Applied Mathematics Letters, 26(7), 676-680 https://www.sciencedirect.com/science/article/pii/S0893965913000384

8 dinvCdf\_stdNTS\_int

dCVaRstdNTS\_numint

 $dCVaRstdNTS\_numint$ 

# Description

Calculate the marginal contribution to CVaR for the multivariate stdNTS Model. Developer's version.

#### Usage

```
dCVaRstdNTS_numint(
  eta,
  alpha,
  theta,
  beta,
  cv = NULL,
  v = NULL,
  N = 20,
  rho = 1e-04
)
```

dCVaR\_numint

dCVaR\_numint

# **Description**

The first derivative of CVaR for the beta parameter of the stdNTS. Developer's version.

# Usage

```
dCVaR_numint(eta, alpha, theta, beta, N = 200, rho = 0.1)
```

dinvCdf\_stdNTS\_int

 $dinvCdf\_stdNTS\_int$ 

# Description

The first derivative of inverse CDF for the beta parameter of the stdNTS. Developer's version.

# Usage

```
dinvCdf_stdNTS_int(eta, x = NULL, alpha, theta, beta)
```

dmarginalmnts 9

dmarginalmnts

dmarginalmnts

#### **Description**

dmarginalmnts calculates the marginal density of the n-th element of the multivariate NTS distributed random variable.

#### Usage

```
dmarginalmnts(x, n, st)
```

#### **Arguments**

x the x such that  $f(x) = \frac{d}{dx}P(X_n < x)$ 

n the n-th element to be calculated.

st Structure of parameters for the n-dimensional NTS distribution.

dmnts dmnts

# Description

dmnts calculates the density of the multivariate NTS distribution:  $f(x_1, \cdots, x_n) = \frac{d^n}{dx_1 \cdots dx_n} P(x_n < R_1, \cdots, x_n < R_n)$ . The multivariate NTS random vector  $R = (R_1, \cdots, R_n)$  is defined

 $R = \mu + diag(\sigma)X$ ,

where

X follows  $stdNTS_n(\alpha, \theta, \beta, \Sigma)$ 

#### Usage

```
dmnts(x, st, subTS = NULL)
```

#### **Arguments**

x array of the  $(x_1, \dots, x_n)$ 

st Structure of parameters for the n-dimensional NTS distribution.

st\$ndim: dimension

 $\mbox{st}\mbox{\mbox{\it mu}}$  :  $\mu$  mean vector (column vector) of the input data.

 ${\tt st\$sigma}: \sigma$  standard deviation vector (column vector) of the input data.

st\$alpha :  $\alpha$  of the std NTS distribution (X). st\$theta :  $\theta$  of the std NTS distribution (X).

st\$beta :  $\beta$  vector (column vector) of the std NTS distribution (X).

st\$Rho :  $\rho$  matrix of the std NTS distribution (X).

numofsample number of samples.

10 dnts

#### Value

Simulated NTS random vectors

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

#### **Examples**

```
library("temStaR")
library(mvtnorm)
strPMNTS <- list(ndim = 2,</pre>
              mu = c(0.5, -1.5),
              sigma = c(2, 3),
              alpha = 0.1,
              theta = 3,
              beta = c(0.1, -0.3),
              Rho = matrix( data = c(1.0, 0.75, 0.75, 1.0),
                            nrow = 2, ncol = 2)
dmnts(c(0.6, -1.0), st = strPMNTS)
strPMNTS <- list(ndim = 2,</pre>
                 mu = c(0, 0, 0),
                 sigma = c(1, 1, 1),
                 alpha = 0.1,
                 theta = 3,
                 beta = c(0.1, -0.3, 0),
                 Rho = matrix(
                     data = c(1.0, 0.75, 0.1, 0.75, 1.0, 0.2, 0.1, 0.2, 1.0),
                     nrow = 3, ncol = 3)
pmnts(c(0,0,0), st = strPMNTS)
dmnts(c(0,0,0), st = strPMNTS)
```

dnts

dnts

#### **Description**

dnts calculates pdf of the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it calculates pdf of the standard NTS distribution with parameter  $(\alpha, \theta, \beta)$  If a time parameter value is given, it calculates pdf of the NTS profess f(x)dx = d(P((X(t+s) - X(s)) < x)), where X is the NTS process generated by the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ .

#### Usage

```
dnts(xdata, ntsparam)
```

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#### **Arguments**

xdata An array of x  $\begin{array}{ll} \text{A vector of the NTS parameters } (\alpha, \theta, \beta, \gamma, \mu). \text{ For the NTS process case it is a} \\ \text{vector of parameters } (\alpha, \theta, \beta, \gamma, \mu, t). \text{ A vector of the standard NTS parameters} \\ (\alpha, \theta, \beta). \end{array}$ 

#### Value

Density of NTS distribution

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)</pre>
x \leftarrow seq(from = -6, to = 6, length.out = 101)
d <- dnts(x, ntsparam)</pre>
plot(x,d,type = 'l')
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)</pre>
x \leftarrow seq(from = -2, to = 2, by = 0.01)
d <- dnts(x, ntsparam)</pre>
plot(x,d,type = 'l')
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
#scaling annual parameters to one day
dt <- 1/250 #one day
ntsparam <- c(alpha, theta, beta, gamma, mu, dt)</pre>
x <- seq(from = -0.02, to = 0.02, length.out = 101)
d <- dnts(x, ntsparam)</pre>
plot(x,d,type = 'l')
```

12 fitmnts

fitmnts fitmnts

#### **Description**

```
fitmnts fit parameters of the n-dimensional NTS distribution.
```

```
r=\mu+diag(\sigma)X where X \mbox{ follows } stdNTS_n(\alpha,\theta,\beta,\Sigma)
```

#### Usage

```
\code{res <- fitmnts(returndata, n)}
\code{res <- fitmnts(returndata, n, alphaNtheta = c(alpha, theta))}
\code{res <- fitmnts(returndata, n, stdflag = TRUE ) }
\code{res <- fitmnts(returndata, n, alphaNtheta = c(alpha, theta), stdflag = TRUE)}</pre>
```

# **Arguments**

returndata Raw data to fit the parameters. The data must be given as a matrix form. Each

column of the matrix contains a sequence of asset returns. The number of row

of the matrix is the number of assets.

n Dimension of the data. That is the number of assets.

alphaNtheta If  $\alpha$  and  $\theta$  are given, then put those numbers in this parameter. The function fixes

those parameters and fits other remaining parameters. If you set alphaNtheta

= NULL, then the function fits all parameters including  $\alpha$  and  $\theta$ .

stdflag If you want only standard NTS parameter fit, set this value be TRUE.

#### Value

Structure of parameters for the n-dimensional NTS distribution.

resmu:  $\mu$  mean vector of the input data.

 $\operatorname{res} sigma : \sigma$  standard deviation vector of the input data.

res\$alpha :  $\alpha$  of the std NTS distribution (X). res\$theta :  $\theta$  of the std NTS distribution (X).

res\$beta :  $\beta$  vector of the std NTS distribution (X).

res\$Rho :  $\rho$  matrix of the std NTS distribution (X), which is correlation matrix of epsilon.

res\$CovMtx : Covariance matrix of return data r.

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

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#### **Examples**

```
library(functional)
library(nloptr)
library(pracma)
library(spatstat)
library(Matrix)
library(quantmod)
library(temStaR)
getSymbols("^GSPC", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr1 <- as.numeric(GSPC$GSPC.Adjusted)</pre>
getSymbols("^DJI", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr2 <- as.numeric(DJI$DJI.Adjusted)</pre>
returndata <- matrix(data = c(diff(log(pr1)),diff(log(pr2))),</pre>
                      ncol = 2, nrow = (length(pr1)-1))
res <- fitmnts( returndata = returndata, n=2 )</pre>
#Fix alpha and theta.
#Estimate alpha dna theta from DJIA and use those parameter for IBM, INTC parameter fit.
getSymbols("^DJI", src="yahoo", from = "2016-1-1", to = "2020-08-31")
prDJ <- as.numeric(DJI$DJI.Adjusted)</pre>
ret <- diff(log(prDJ))</pre>
ntsparam <- fitnts(ret)</pre>
getSymbols("IBM", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr1 <- as.numeric(IBM$IBM.Adjusted)</pre>
getSymbols("INTC", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr2 <- as.numeric(INTC$INTC.Adjusted)</pre>
returndata <- matrix(data = c(diff(log(pr1)), diff(log(pr2))),\\
                      ncol = 2, nrow = (length(pr1)-1))
res <- fitmnts( returndata = returndata,</pre>
                n = 2
                 alphaNtheta = c(ntsparam["alpha"], ntsparam["theta"]) )
```

fitmnts\_par

fitmnts

#### **Description**

fitmnts fit parameters of the n-dimensional NTS distribution. A parallel version of fitmnts()

# Usage

```
fitmnts_par(
  returndata,
  n,
  alphaNtheta = NULL,
  stdflag = FALSE,
  parallelSocketCluster = NULL,
  PDflag = TRUE
)
```

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#### **Examples**

```
library(functional)
library(nloptr)
library(pracma)
library(spatstat)
library(Matrix)
library(foreach)
library(doParallel)
library(quantmod)
library(temStaR)
getSymbols("^GSPC", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr1 <- as.numeric(GSPC$GSPC.Adjusted)</pre>
getSymbols("^DJI", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr2 <- as.numeric(DJI$DJI.Adjusted)</pre>
returndata <- matrix(data = c(diff(log(pr1)),diff(log(pr2))),</pre>
                      ncol = 2, nrow = (length(pr1)-1))
numofcluster <- detectCores()</pre>
cl <- makePSOCKcluster(numofcluster)</pre>
registerDoParallel(cl)
res <- fitmnts_par( returndata = returndata, n=2, parallelSocketCluster = cl )</pre>
stopCluster(cl)
```

fitnts

fitnts

# Description

fitnts fit parameters  $(\alpha, \theta, \beta, \gamma, \mu)$  of the NTS distribution. This function using the curvefit method between the empirical cdf and the NTS cdf.

#### Usage

```
\code{fitnts(rawdat)}
\code{fitnts(rawdat), ksdensityflag = 1}
\code{fitnts(rawdat, initialparam = c(alpha, theta, beta, gamma, mu))}
\code{fitnts(rawdat, initialparam = c(alpha, theta, beta, gamma, mu)), ksdensityflag = 1}
\code{fitnts(rawdat, initialparam = c(alpha, theta, beta, gamma, mu)), maxeval = 100, ksdensityflag
```

# Arguments

rawdat Raw data to fit the parameters.

initialparam A vector of initial NTS parameters. This function uses the nloptr package. If

it has a good initial parameter then estimation performs better. If users do not know a good initial parameters, then just set it as initialparam=NaN, that is default. The function cffitnts() may be helpful to find the initial parameters.

maxeval Maximum evaluation number for nloptr. The iteration stops on this many func-

tion evaluations.

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ksdensityflag

This function fit the parameters using the curvefit method between the empirical cdf and the NTS cdf. If ksdensityflag = 1 (default), then the empirical cdf is calculated by the kernel density estimation. If ksdensityflag = 0, then the empirical cdf is calculated by the empirical cdf.

#### Value

Estimated parameters

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

#### **Examples**

```
library("temStaR")
library("quantmod")
getSymbols("^GSPC", src="yahoo", from = "2010-1-1", to = "2020-12-31")
pr <- as.numeric(GSPC$GSPC.Adjusted)</pre>
ret <- diff(log(pr))</pre>
ntsparam <- fitnts(ret)</pre>
Femp = ecdf(ret)
x = seq(from=min(ret), to = max(ret), length.out = 100)
cemp = Femp(x)
ncdf = pnts(x, c(ntsparam))
plot(x,ncdf,type = 'l', col = "red")
points(x,cemp, type = '1', col = "blue")
a = density(ret)
p = dnts(x,ntsparam)
plot(x,p,type = 'l', col = "red")
lines(a,type = 'l', col = "blue")
```

fitstdnts

fitstdnts

# Description

fitstdnts fit parameters  $(\alpha, \theta, \beta)$  of the standard NTS distribution. This function using the curvefit method between the empirical cdf and the standard NTS cdf.

#### Usage

```
\code{fitstdnts(rawdat)}
\code{fitstdnts(rawdat), ksdensityflag = 1}
\code{fitstdnts(rawdat, initialparam = c(alpha, theta, beta))}
\code{fitstdnts(rawdat, initialparam = c(alpha, theta, beta)), ksdensityflag = 1}
\code{fitstdnts(rawdat, initialparam = c(alpha, theta, beta)), maxeval = 100, ksdensityflag = 1}
```

#### **Arguments**

rawdat Raw data to fit the parameters.

initialparam A vector of initial standard NTS parameters. This function uses the nloptr

package. If it has a good initial parameter then estimation performs better. If users do not know a good initial parameters, then just set it as initialparam=NaN,

that is default.

maxeval Maximum evaluation number for nloptr. The iteration stops on this many func-

tion evaluations.

ksdensityflag This function fit the parameters using the curvefit method between the empirical

cdf and the standard NTS cdf. If ksdensityflag = 1 (default), then the empirical cdf is calculated by the kernel density estimation. If ksdensityflag = 0,

then the empirical cdf is calculated by the empirical cdf.

#### Value

Estimated parameters

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

# **Examples**

```
library("temStaR")
library("quantmod")
getSymbols("^GSPC", src="yahoo", from = "2010-1-1", to = "2020-12-31")
pr <- as.numeric(GSPC$GSPC.Adjusted)</pre>
ret <- diff(log(pr))</pre>
stdret <- (ret-mean(ret))/sd(ret)</pre>
stdntsparam <- fitstdnts(stdret)</pre>
Femp = ecdf(stdret)
x = seq(from=min(stdret), to = max(stdret), length.out = 100)
cemp = Femp(x)
ncdf = pnts(x, c(stdntsparam))
plot(x,ncdf,type = 'l', col = "red")
lines(x,cemp, type = '1', col = "blue")
a = density(stdret)
p = dnts(x,stdntsparam)
plot(x,p,type = 'l', col = "red", ylim = c(0, max(a$y, p)))
lines(a,type = 'l', col = "blue")
```

fitstdntsFixAlphaThata

fitstdntsFixAlphaThata

#### **Description**

Fit beta of stdNTS distribution with fixed alpha and theta.

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#### Usage

```
fitstdntsFixAlphaThata(
  rawdat,
  alpha,
  theta,
  initialparam = NaN,
  maxeval = 100,
  ksdensityflag = 1
)
```

gensamplepathnts

gensamplepathnts

## **Description**

gensamplepathnts generate sample paths of the NTS process with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it generate sample paths of the standard NTS process with parameters  $(\alpha, \theta, \beta)$ .

# Usage

```
gensamplepathnts(npath, ntimestep, ntsparam, dt)
```

#### **Arguments**

npath Number of sample paths  $\begin{array}{ll} \text{number of sample paths} \\ \text{ntimestep} & \text{number of time step} \\ \text{ntsparam} & \text{A vector of the NTS parameters } (\alpha, \theta, \beta, \gamma, \mu). \text{ A vector of the standard NTS parameters } (\alpha, \theta, \beta). \\ \text{dt} & \text{the time length of one time step by the year fraction. "dt=1" means 1-year.} \end{array}$ 

#### Value

Structure of the sample path. Matrix of sample path. Column index is time.

```
library("temStaR")
#standard NTS process sample path
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)
npath <- 5
ntimestep <- 250
dt <- 1/250
simulation <- gensamplepathnts(npath, ntimestep, ntsparam, dt)
matplot(colnames(simulation), t(simulation), type = 'l')
#NTS process sample path
alpha <- 1.2</pre>
```

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```
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)
npath <- 5
ntimestep <- 250
dt <- 1/250
simulation <- gensamplepathnts(npath, ntimestep, ntsparam, dt)
matplot(colnames(simulation), t(simulation), type = 'l')</pre>
```

getGammaVec

getGammaVec

#### **Description**

beta to gamma in StdNTS

#### Usage

```
getGammaVec(alpha, theta, betaVec)
```

getPortNTSParam

getPortNTSParam

#### **Description**

Portfolio return with capital allocation weight is  $R_p = \langle w, r \rangle$ , which is a weighted sum of of elements in the N-dimensional NTS random vector.  $R_p$  becomes an 1-dimensional NTS random variable. getPortNTSParam find the parameters of  $R_p$ .

#### Usage

```
\code{res <- setPortfolioParam(strPMNTS,w)}
\code{res <- setPortfolioParam(strPMNTS,w, FALSE)}</pre>
```

# **Arguments**

strPMNTS

Structure of parameters for the n-dimensional NTS distribution.

strPMNTS\$ndim: dimension

strPMNTSmu :  $\mu$  mean vector (column vector) of the input data.

 ${\tt strPMNTS\$sigma}: \sigma {\tt standard} {\tt deviation} {\tt vector} ({\tt column} {\tt vector}) {\tt of} {\tt the} {\tt input}$ 

data.

strPMNTS\$alpha :  $\alpha$  of the std NTS distribution (X).

strPMNTS\$theta:  $\theta$  of the std NTS distribution (X).

strPMNTS\$beta:  $\beta$  vector (column vector) of the std NTS distribution (X).

res\$Rho :  $\rho$  matrix (Correlation) of the std NTS distribution (X).

resSigma: Covariance  $\Sigma$  matrix of return data r.

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 $\label{eq:capital allocation weight vector.} % \begin{tabular}{ll} & Capital allocation weight vector. \\ & If stdform is FALSE, then the return parameter has the following representation \\ & R_p = < w, r > = \mu + diag(\sigma)X, \\ & \text{where} \\ & X \text{ follows } stdNTS_1(\alpha,\theta,\beta,1). \\ & If \text{ stdform is TRUE, then the return parameter has the following representation} \\ & R_p = < w, r > \text{ follows } NTS_1(\alpha,\theta,\beta,\gamma,\mu,1) \\ \end{tabular}$ 

#### Value

The weighted sum follows 1-dimensional NTS.

```
\begin{split} R_p = < w, r > &= \mu + diag(\sigma)X, \\ \text{where} \\ X \text{ follows } stdNTS_1(\alpha, \theta, \beta, 1). \\ \text{Hence we obtain} \\ \text{res$mu} : \mu \text{ mean of } R_p. \\ \text{res$sigma} : \sigma \text{ standard deviation of } R_p. \\ \text{res$alpha} : \alpha \text{ of } X. \\ \text{res$theta} : \theta \text{ of } X. \\ \text{res$beta} : \beta \text{ of } X. \end{split}
```

#### References

Proposition 2.1 of Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

20 ipnts

importantSamplining importantSamplining

#### **Description**

importantSamplining do the important sampling for the TS Subordinator.

#### Usage

```
importantSamplining(alpha, theta)
```

ipnts

ipnts

#### **Description**

ipnts calculates inverse cdf of the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it calculates inverse cdf of the standard NTS distribution with parameter  $(\alpha, \theta, \beta)$ 

#### Usage

```
ipnts(u, ntsparam, maxmin = c(-10, 10), du = 0.01)
```

#### **Arguments**

u Real value between 0 and 1

ntsparam A vector of the NTS parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . A vector of the standard NTS

parameters  $(\alpha, \theta, \beta)$ .

#### Value

Inverse cdf of the NTS distribution. It is the same as qnts function.

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)
u <- seq(from = 0.01, to = 0.99, length.out = 99)
q <- ipnts(u, ntsparam)
plot(u,q,type = 'l')
alpha <- 1.2</pre>
```

mctCVaR\_MNTS 21

```
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)</pre>
u \leftarrow seq(from = 0.01, to = 0.99, length.out = 99)
q <- ipnts(u, ntsparam)</pre>
plot(x,q,type = 'l')
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
#scaling annual parameters to one day
dt <- 1/250 #one day
ntsparam \leftarrow c(alpha, theta, beta, gamma, mu, dt)
u \leftarrow seq(from = 0.01, to = 0.99, length.out = 99)
q <- ipnts(u, ntsparam)</pre>
plot(x,q,type = 'l')
```

mctCVaR\_MNTS

 $mctCVaR\_MNTS$ 

# Description

Calculate the marginal contribution to CVaR for the multivariate NTS market model: the random vector r is

```
r = \mu + diag(\sigma)X where X \text{ follows } stdNTS_N(\alpha,\theta,\beta,\Sigma)
```

#### Usage

```
\code{mctCVaR_MNTS(eta, n, w, stmnts)}
```

#### **Arguments**

n	The targer stock to calculate the mctCVaR
eta	Significant level of CVaR.
W	The capital allocation rate vector for the current portfolio
stmnts	Structure of parameters for the N-dimensional NTS distribution.
CVaRstd	$\mbox{CVaR}$ Value of StdNTS residual. If NULL, the function automatically find it. The default value is NULL
dCVaRstd	The first derivative of the stdNTS CVaR for beta. If NULL, the function automatically find it. The default value is NULL

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iCDFstd The inverst cdf of stdNTS at the significant level eta. If NULL, the function automatically find it. The default value is NULL st\$ndim: Dimension of the model. Here st\$ndim=N. stmu :  $\mu$  mean vector (column vector) of the input data.  $st\$sigma: \sigma$  standard deviation vector (column vector) of the input data. stalpha:  $\alpha$  of the std NTS distribution (X). st\$theta:  $\theta$  of the std NTS distribution (X). st\$beta :  $\beta$  vector (column vector) of the std NTS distribution (X). st\$Rho :  $\rho$  matrix of the std NTS distribution (X), which is correlation matrix of epsilon.

stCovMtx: Covariance matrix of return data r.

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

```
library(functional)
library(nloptr)
library(pracma)
library(spatstat)
library(Matrix)
library(quantmod)
library(mvtnorm)
library("temStaR")
#Fix alpha and theta.
#Estimate alpha dna theta from DJIA and use those parameter for IBM, INTL parameter fit.
getSymbols("^DJI", src="yahoo", from = "2020-8-25", to = "2020-08-31")
prDJ <- as.numeric(DJI$DJI.Adjusted)</pre>
ret <- diff(log(prDJ))</pre>
ntsparam <- fitnts(ret)</pre>
getSymbols("IBM", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr1 <- as.numeric(IBM$IBM.Adjusted)</pre>
getSymbols("INTL", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr2 <- as.numeric(INTL$INTL.Adjusted)</pre>
returndata <- matrix(data = c(diff(log(pr1)), diff(log(pr2))),</pre>
                      ncol = 2, nrow = (length(pr1)-1))
st <- fitmnts( returndata = returndata,</pre>
                 n = 2
                 alphaNtheta = c(ntsparam["alpha"], ntsparam["theta"]) )
w \leftarrow c(0.3, 0.7)
eta <- 0.01
mctCVaR_MNTS(1, eta, w, st) #MCT-CVaR for IBM
mctCVaR_MNTS(2, eta, w, st) #MCT-CVaR for INTL
```

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mctStdDev mctStdDev
---------------------

#### **Description**

Morginal contribution to Risk for Standard Deviation.

#### Usage

```
mctStdDev(n, w, covMtx)
```

#### **Arguments**

n The targer stock to calculate the mctCVaR

w The capital allocation rate vector for the current portfolio

CovMtx Covariance matrix of return data.

#### **Description**

Calculate the marginal contribution to VaR for the multivariate NTS market model: the random vector r is

```
r = \mu + diag(\sigma)X
```

where

X follows  $stdNTS_N(\alpha, \theta, \beta, \Sigma)$ 

# Usage

```
\code{mctVaR_MNTS(n, eta, w, st)}
```

#### **Arguments**

11	The target stock to calculate the flict valv
eta	Significant level of VaR.
	FF1 1 11 11 1 1 1 1 1 1 1 1 1 1 1 1 1 1

W The capital allocation rate vector for the current portfolio

The terrest steels to coloulete the metVeD

iCDFstd The inverst cdf of stdNTS at the significant level eta. If NULL, the function

automatically find it. The default value is NULL stndim: Dimension of the model. Here st<math>ndim=N. st $mu: \mu$  mean vector (column vector) of the input data.

 ${\tt st\$sigma}: \sigma$  standard deviation vector (column vector) of the input data.

st\$alpha :  $\alpha$  of the std NTS distribution (X). st\$theta :  $\theta$  of the std NTS distribution (X).

st\$beta :  $\beta$  vector (column vector) of the std NTS distribution (X).

st\$Rho :  $\rho$  matrix of the std NTS distribution (X), which is correlation matrix

of epsilon.

st\$CovMtx: Covariance matrix of return data r.

st Structure of parameters for the N-dimensional NTS distribution.

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#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

#### **Examples**

```
library(functional)
library(nloptr)
library(pracma)
library(spatstat)
library(Matrix)
library(quantmod)
library(mvtnorm)
library(temStaR)
#Fix alpha and theta.
#Estimate alpha dna theta from DJIA and use those parameter for IBM, INTL parameter fit.
getSymbols("^DJI", src="yahoo", from = "2020-8-25", to = "2020-08-31")
prDJ <- as.numeric(DJI$DJI.Adjusted)</pre>
ret <- diff(log(prDJ))</pre>
ntsparam <- fitnts(ret)</pre>
getSymbols("IBM", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr1 <- as.numeric(IBM$IBM.Adjusted)</pre>
getSymbols("INTL", src="yahoo", from = "2016-1-1", to = "2020-08-31")
pr2 <- as.numeric(INTL$INTL.Adjusted)</pre>
returndata <- matrix(data = c(diff(log(pr1)), diff(log(pr2))),</pre>
                      ncol = 2, nrow = (length(pr1)-1))
st <- fitmnts( returndata = returndata,</pre>
                n = 2,
                 alphaNtheta = c(ntsparam["alpha"], ntsparam["theta"]) )
w \leftarrow c(0.3, 0.7)
eta <- 0.01
mctVaR\_MNTS(1, eta, w, st) \#MCT-VaR for IBM
mctVaR_MNTS(2, eta, w, st) #MCT-VaR for INTL
```

moments\_NTS

moments\_NTS

#### **Description**

moments\_NTS calculates mean, variance, skewness, and excess kurtosis of the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ .

#### Usage

```
moments_NTS(param)
```

#### **Arguments**

param

A vector of the NTS parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ .

moments\_stdNTS 25

#### Value

First 4 moments (Mean, Variance, Skewness, Excess Kurtosis) of NTS distribution. The mean is always the same as the parameter  $\mu$ .

#### References

Kim, Y.S, K-H Roh, R. Douady (2020) Tempered Stable Processes with Time Varying Exponential Tails https://arxiv.org/pdf/2006.07669.pdf

#### **Examples**

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)
moments_NTS(param = ntsparam)</pre>
```

moments\_stdNTS

moments\_stdNTS

#### **Description**

moments\_stdNTS calculates mean, variance, skewness, and excess kurtosis of the standard NTS distribution with parameters  $(\alpha, \theta, \beta)$ .

#### Usage

```
moments_stdNTS(param)
```

#### **Arguments**

param

A vector of the standard NTS parameters  $(\alpha, \theta, \beta)$ .

#### Value

First 4 moments (Mean, Variance, Skewness, Excess Kurtosis) of NTS distribution. Of course, the mean and variance are always 0 and 1, respectively.

#### References

Kim, Y.S, K-H Roh, R. Douady (2020) Tempered Stable Processes with Time Varying Exponential Tails https://arxiv.org/pdf/2006.07669.pdf

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)
moments_stdNTS(param = ntsparam)</pre>
```

26 pmnts

pmarginalmnts	pmarginalmnts
---------------	---------------

#### **Description**

pmarginalmnts calculates the marginal cdf of the n-th element of the multivariate NTS distributed random variable.

#### Usage

```
pmarginalmnts(x, n, st)
```

#### **Arguments**

```
x the x such that F(x) = P(X_n < x) n the n-th element to be calculated. Structure of parameters for the n-dimensional NTS distribution.
```

pmnts	pmnts
-------	-------

# Description

```
pmnts calculates the cdf values of the multivariate NTS distribution: F(x_1,\cdots,x_n)=P(x_n< R_1,\cdots,x_n< R_n). The multivariate NTS random vector R=(R_1,\cdots,R_n) is defined R=\mu+diag(\sigma)X, where X follows stdNTS_n(\alpha,\theta,\beta,\Sigma)
```

#### Usage

```
pmnts(x, st, subTS = NULL)
```

#### **Arguments**

```
x array of the (x_1,\cdots,x_n) st Structure of parameters for the n-dimensional NTS distribution. st$ndim: dimension st$mu: \mu mean vector (column vector) of the input data. st$sigma: \sigma standard deviation vector (column vector) of the input data. st$alpha: \alpha of the std NTS distribution (X). st$theta: \theta of the std NTS distribution (X). st$beta: \beta vector (column vector) of the std NTS distribution (X). st$Rho: \rho matrix of the std NTS distribution (X). numof sample number of samples.
```

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#### Value

Simulated NTS random vectors

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

# **Examples**

```
library(mvtnorm)
library(temStaR)
strPMNTS <- list(ndim = 2,</pre>
              mu = c(0.5, -1.5),
              sigma = c(2, 3),
              alpha = 0.1,
              theta = 3,
              beta = c(0.1, -0.3),
              Rho = matrix( data = c(1.0, 0.75, 0.75, 1.0),
                            nrow = 2, ncol = 2)
pmnts(c(0.6, -1.0), st = strPMNTS)
strPMNTS <- list(ndim = 2,</pre>
                 mu = c(0, 0, 0),
                 sigma = c(1, 1, 1),
                 alpha = 0.1,
                 theta = 3,
                 beta = c(0.1, -0.3, 0),
                 Rho = matrix(
                     data = c(1.0, 0.75, 0.1, 0.75, 1.0, 0.2, 0.1, 0.2, 1.0),
                     nrow = 3, ncol = 3)
pmnts(c(0,0,0), st = strPMNTS)
dmnts(c(0,0,0), st = strPMNTS)
```

pnts

pnts

# Description

pnts calculates cdf of the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it calculates cdf of the standard NTS distribution with parameter  $(\alpha, \theta, \beta)$  If a time parameter value is given, it calculates cdf of the profess F(x) = P((X(t+s) - X(s)) < x), where X is the NTS process generated by the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ .

#### Usage

```
pnts(xdata, ntsparam, dz = 2^-8, m = 2^12)
```

28 pnts

#### **Arguments**

xdata An array of x  $\begin{array}{ll} \text{A vector of the NTS parameters } (\alpha, \theta, \beta, \gamma, \mu). \text{ For the NTS process case it is a} \\ \text{vector of parameters } (\alpha, \theta, \beta, \gamma, \mu, t). \text{ A vector of the standard NTS parameters} \\ (\alpha, \theta, \beta). \end{array}$ 

#### Value

Cumulative probability of the NTS distribution

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)</pre>
x \leftarrow seq(from = -6, to = 6, length.out = 101)
p <- pnts(x, ntsparam)</pre>
plot(x,p,type = 'l')
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)</pre>
x \leftarrow seq(from = -2, to = 2, by = 0.01)
p <- pnts(x, ntsparam)</pre>
plot(x,p,type = 'l')
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
#scaling annual parameters to one day
dt <- 1/250 #one day
ntsparam <- c(alpha, theta, beta, gamma, mu, dt)</pre>
x <- seq(from = -0.02, to = 0.02, length.out = 101)
p <- pnts(x, ntsparam)</pre>
plot(x,p,type = 'l')
```

portfolioCVaRmnts 29

portfolioCVaRmnts

portfolioCVaRmnts

#### **Description**

Calculate portfolio conditional value at risk (expected shortfall) on the NTS market model

#### Usage

```
portfolioCVaRmnts(strPMNTS, w, eta)
```

# **Arguments**

strPMNTS Structure of parameters for the n-dimensional NTS distribution.

strPMNTS\$ndim: dimension

 $trPMNTSmu: \mu$  mean vector (column vector) of the input data.

strPMNTSsigma :  $\sigma$  standard deviation vector (column vector) of the input

data.

strPMNTS\$alpha :  $\alpha$  of the std NTS distribution (X). strPMNTS\$theta :  $\theta$  of the std NTS distribution (X).

strPMNTS\$beta :  $\beta$  vector (column vector) of the std NTS distribution (X).

res\$Rho :  $\rho$  matrix (Correlation) of the std NTS distribution (X).

 $\verb"res$Sigma": Covariance $\Sigma$ matrix of return data $r$.$ 

w Capital allocation weight vector.

eta significanlt level

#### Value

portfolio value at risk on the NTS market model

portfolioVaRmnts

portfolioVaRmnts

# **Description**

Calculate portfolio value at risk on the NTS market model

# Usage

```
portfolioVaRmnts(strPMNTS, w, eta)
```

30 qmarginalmnts

#### **Arguments**

strPMNTS Structure of parameters for the n-dimensional NTS distribution.

strPMNTS\$ndim: dimension

 $strPMNTS$mu: \mu mean vector (column vector) of the input data.$ 

strPMNTSsigma :  $\sigma$  standard deviation vector (column vector) of the input

data.

strPMNTS\$alpha :  $\alpha$  of the std NTS distribution (X). strPMNTS\$theta :  $\theta$  of the std NTS distribution (X).

strPMNTS\$beta :  $\beta$  vector (column vector) of the std NTS distribution (X).

res\$Rho :  $\rho$  matrix (Correlation) of the std NTS distribution (X).

resSigma: Covariance  $\Sigma$  matrix of return data r.

w Capital allocation weight vector.

eta significanlt level

#### Value

portfolio value at risk on the NTS market model

qmarginalmnts	qmarginalmnts
4 921142	4

#### **Description**

qmarginalmnts calculates the quantile value of the n-th element of the multivariate NTS distributed random variable.

# Usage

```
qmarginalmnts(u, n, st)
```

#### **Arguments**

u vector of probabilities.

n the n-th element to be calculated.

st Structure of parameters for the n-dimensional NTS distribution.

qnts 31

qnts qnts

#### **Description**

qnts calculates quantile of the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it calculates quantile of the standard NTS distribution with parameter  $(\alpha, \theta, \beta)$  If a time parameter value is given, it calculates quantile of NTS profess. That is it finds x such that u = P((X(t+s) - X(s)) < x), where X is the NTS process generated by the NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ .

# Usage

```
qnts(u, ntsparam)
```

#### **Arguments**

u vector of probabilities.

ntsparam

A vector of the NTS parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . For the NTS process case it is a vector of parameters  $(\alpha, \theta, \beta, \gamma, \mu, t)$ . A vector of standard NTS parameters  $(\alpha, \theta, \beta)$ .

#### Value

The quantile function of the NTS distribution

```
library("temStaR")
alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)</pre>
u \leftarrow c(0.01, 0.05, 0.25, 0.5, 0.75, 0.95, 0.99)
q <- qnts(u, ntsparam)</pre>
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
ntsparam <- c(alpha, theta, beta, gamma, mu)</pre>
u \leftarrow c(0.01, 0.05, 0.25, 0.5, 0.75, 0.95, 0.99)
q <- qnts(u, ntsparam)</pre>
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
```

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```
#scaling annual parameters to one day dt <- 1/250 #one day ntsparam <- c(alpha, theta, beta, gamma, mu, dt) u <- c(0.01,0.05,0.25,0.5, 0.75, 0.95, 0.99) q <- qnts(u, ntsparam)
```

rmnts

rmnts

#### **Description**

```
rmnts generates random vector following the n dimensional NTS distribution using subordination.
```

```
r=\mu+diag(\sigma)X, where X \text{ follows } stdNTS_n(\alpha,\theta,\beta,\Sigma)
```

#### Usage

```
rmnts(strPMNTS, numofsample, rW = NULL, rTau = NULL)
```

#### **Arguments**

strPMNTS Structure of parameters for the n-dimensional NTS distribution.

strPMNTS\$ndim: dimension

 $trPMNTS\$  :  $\mu$  mean vector (column vector) of the input data.

 ${\tt strPMNTS\$sigma}: \sigma {\tt standard} {\tt deviation} {\tt vector} ({\tt column} {\tt vector}) {\tt of} {\tt the} {\tt input}$ 

data.

 ${\tt strPMNTS\$alpha}$  :  $\alpha$  of the std NTS distribution (X).

 ${\tt strPMNTS\$theta}: \theta \mbox{ of the std NTS distribution } (X).$ 

 ${\tt strPMNTS\$beta:}\ \beta\ {\tt vector}\ ({\tt column}\ {\tt vector})\ {\tt of}\ {\tt the}\ {\tt std}\ {\tt NTS}\ {\tt distribution}\ ({\tt X}).$ 

 ${\tt strPMNTS\$Rho}: \rho$  matrix of the std NTS distribution (X).

numofsample

number of samples.

#### Value

Simulated NTS random vectors

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

rnts 33

```
)
gensim <- rmnts( strPMNTS, 100 )
plot(gensim)</pre>
```

rnts

rnts

#### **Description**

rnts generates random numbers following NTS distribution with parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . If only three parameters are given, it generates random numbers of standard NTS distribution with parameter  $(\alpha, \theta, \beta)$  If a time parameter value is given, it generates random numbers of increments of NTS profess for time interval t.

#### Usage

```
rnts(n, ntsparam, u = NULL)
```

#### **Arguments**

n number of random numbers to be generated.

ntsparam A vector of NTS parameters  $(\alpha, \theta, \beta, \gamma, \mu)$ . For NTS process case it is a vector of parameters  $(\alpha, \theta, \beta, \gamma, \mu, t)$ . A vector of standard NTS parameters  $(\alpha, \theta, \beta)$ .

#### Value

NTS random numbers

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

```
library("temStaR")

alpha <- 1.2
theta <- 1
beta <- -0.2
ntsparam <- c(alpha, theta, beta)
r <- rnts(100, ntsparam) #generate 100 NTS random numbers
plot(r)

alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
ntsparam <- c(alpha, theta, beta, gamma, mu)
r <- rnts(100, ntsparam) #generate 100 NTS random numbers
plot(r)</pre>
```

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```
#Annual based parameters
alpha <- 1.2
theta <- 1
beta <- -0.2
gamma <- 0.3
mu <- 0.1
#scaling annual parameters to one day
dt <- 1/250 #one day
ntsparam <- c(alpha, theta, beta, gamma, mu, dt)
r <- rnts(100, ntsparam) #generate 100 NTS random numbers
plot(r)</pre>
```

setPortfolioParam

setPortfolioParam

# Description

Please use getPortNTSParam instead of setPortfolioParam.

Portfolio return with capital allocation weight is  $R_p = \langle w, r \rangle$ , which is a weighted sum of of elements in the N-dimensional NTS random vector.  $R_p$  becomes an 1-dimensional NTS random variable. setPortfolioParam find the parameters of  $R_p$ .

#### Usage

```
\code{res <- setPortfolioParam(strPMNTS,w)}</pre>
```

# **Arguments**

strPMNTS Structure of parameters for the n-dimensional NTS distribution.

strPMNTS\$ndim: dimension

strPMNTSmu :  $\mu$  mean vector (column vector) of the input data.

 ${\tt strPMNTS\$sigma}: \sigma$  standard deviation vector (column vector) of the input

data.

 ${\tt strPMNTS\$alpha}: \alpha$  of the std NTS distribution (X).

 ${\tt strPMNTS\$theta}: \theta$  of the std NTS distribution (X).

strPMNTS\$beta :  $\beta$  vector (column vector) of the std NTS distribution (X).

 ${\tt strPMNTS\$Rho}: \Sigma \ matrix \ of \ the \ std \ NTS \ distribution \ (X).$ 

w Capital allocation weight vector.

# Value

```
The weighted sum follows 1-dimensional NTS.
```

$$R_p = < w, r > = \mu + diag(\sigma)X,$$

where

X follows  $stdNTS_1(\alpha, \theta, \beta, 1)$ .

Hence we obtain

res\$mu :  $\mu$  mean of  $R_p$ .

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```
\label{eq:constraints} \begin{split} \operatorname{res\$sigma}: & \sigma \text{ standard deviation of } R_p. \\ \operatorname{res\$alpha}: & \alpha \text{ of } X. \\ \operatorname{res\$theta}: & \theta \text{ of } X. \\ \operatorname{res\$beta}: & \beta X. \end{split}
```

#### References

Kim, Y. S. (2020) Portfolio Optimization on the Dispersion Risk and the Asymmetric Tail Risk https://arxiv.org/pdf/2007.13972.pdf

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