# Package 'temStaR'

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2 changeCovMtx2Rho

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Description

changeCovMtx2Rho

Change coverance matrix to Rho matrix.

# Usage

changeCovMtx2Rho(CovMtx, alpha, theta, betaVec)

change Cov Mtx 2Rho

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 <- cvarnts(u, ntsparam)
```

dBeta

dBeta

### **Description**

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

# Usage

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

dcopulaStdNTS

dcopulaStdNTS

#### **Description**

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

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)
```

8 dmnts

dinvCdf\_stdNTS

 $dinvCdf\_stdNTS$ 

#### **Description**

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

#### Usage

```
dinvCdf_stdNTS(eta, alpha, theta, beta)
```

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

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)
```

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#### **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). 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

```
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)
```

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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)
```

#### **Arguments**

xdata An array of x

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 the standard NTS parameters

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

#### 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 \leftarrow c(alpha, theta, beta)
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')
```

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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)
x <- seq(from = -0.02, to = 0.02, length.out = 101)
d <- dnts(x, ntsparam)
plot(x,d,type = '1')</pre>
```

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**

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  $\theta$ .

#### Value

Structure of parameters for the n-dimensional NTS distribution.

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

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.

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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)
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 = "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("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))),</pre>
                      ncol = 2, nrow = (length(pr1)-1))
res <- fitmnts( returndata = returndata,</pre>
                 n = 2
                 alphaNtheta = c(ntsparam["alpha"], ntsparam["theta"]) )
```

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}
```

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```
\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.

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

```
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 = '1', 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")
```

14 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.

initial param 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

```
library("temStaR")
library("quantmod")
getSymbols("^GSPC", src="yahoo", from = "2010-1-1", to = "2020-12-31")
pr <- as.numeric(GSPC$GSPC.Adjusted)
ret <- diff(log(pr))
stdret <- (ret-mean(ret))/sd(ret)
stdntsparam <- fitstdnts(stdret)

Femp = ecdf(stdret)
x = seq(from=min(stdret), to = max(stdret), length.out = 100)</pre>
```

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```
cemp = Femp(x)
ncdf = pnts(x, c(stdntsparam))
plot(x,ncdf,type = 'l', col = "red")
lines(x,cemp, type = 'l', 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")
```

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

#### 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
theta <- 1
beta <- -0.2</pre>
```

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```
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

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

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

data.

strPMNTSalpha:  $\alpha$  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.

Capital allocation weight vector.

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stdform

If stdform is FALSE, then the return parameter has the following representation  $R_p=< w, r>=\mu+diag(\sigma)X,$ 

where

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

If stdform is TRUE, then the return parameter has the following representation  $R_p = < w, r > \text{follows } NTS_1(\alpha, \theta, \beta, \gamma, \mu, 1)$ 

#### Value

The weighted sum follows 1-dimensional NTS.

$$R_p = \langle w, r \rangle = \mu + diag(\sigma)X,$$

where

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

Hence we obtain

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

res $sigma: \sigma$  standard deviation of  $R_p$ .

resalpha:  $\alpha$  of X. restheta:  $\theta$  of X. restheta:  $\theta$  of X.

#### 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

18 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>
```

mctCVaRmnts 19

```
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')
```

mctCVaRmnts

mctCVaRmnts

# Description

Calculate the marginal contribution to CVaR for the multivariate NTS market model: the random vector  $\boldsymbol{r}$  is

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

# Usage

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

#### **Arguments**

eta	Significant level of CVaR.
n	The targer stock to calculate the mctCVaR
W	The capital allocation rate vector for the current portfolio
st	Structure of parameters for the N-dimensional NTS distribution.
	st\$ndim: Dimension of the model. Here st\$ndim=N.
	$st\mbox{\$mu}$ : $\mu$ mean vector (column vector) of the input data.
	$\mathtt{st\$sigma}: \sigma$ 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).

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```
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.
```

#### 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
mctVaRmnts(eta, 1, w, st) #MCT-VaR for IBM
mctVaRmnts(eta, 2, w, st) #MCT-VaR for INTL
mctCVaRmnts(eta, 1, w, st) #MCT-CVaR for IBM
mctCVaRmnts(eta, 2, w, st) #MCT-CVaR for INTL
```

mctCVaRnts

mctCVaRnts

#### **Description**

Calculate the marginal contribution to CVaR for the multivariate NTS market model. Developer's version.

mctStdDev 21

#### Usage

```
mctCVaRnts(
  eta,
  n,
  w,
  covMtx,
  alpha,
  theta,
  betaArray,
  muArray,
  CVaR = NULL,
  dCVaR = NULL
)
```

 ${\tt mctStdDev}$ 

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.

mctVaRmnts mctVaRmnts

#### **Description**

Calculate the marginal contribution to VaR for the multivariate NTS market model: the random vector  $\boldsymbol{r}$  is

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

# Usage

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

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#### **Arguments**

eta	Significant level of CVaR.
n	The targer stock to calculate the mctCVaR
W	The capital allocation rate vector for the current portfolio
st	Structure of parameters for the N-dimensional NTS distribution.
	st\$ndim: Dimension of the model. Here st\$ndim=N.
	$st\mbox{mu}$ : $\mu$ 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), which is correlation matrix of epsilon.
	st $CovMtx$ : 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

```
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>
                 alphaNtheta = c(ntsparam["alpha"], ntsparam["theta"]) )
w \leftarrow c(0.3, 0.7)
eta <- 0.01
mctVaRmnts(eta, 1, w, st) #MCT-VaR for IBM
mctVaRmnts(eta, 2, w, st) #MCT-VaR for INTL
```

mctVaRnts 23

```
mctCVaRmnts(eta, 1, w, st) #MCT-CVaR for IBM
mctCVaRmnts(eta, 2, w, st) #MCT-CVaR for INTL
```

mctVaRnts

mctVaRnts

# Description

Calculate the marginal contribution to VaR for the multivariate NTS market model. Developer's version

#### Usage

```
mctVaRnts(
   eta,
   n,
   w,
   covMtx,
   alpha,
   theta,
   betaArray,
   muArray,
   icdf = NULL,
   dicdf = NULL
)
```

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  $(\alpha, \theta, \beta, \gamma, \mu)$ .

# Value

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

24 moments\_stdNTS

#### 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>
```

pmarginalmnts 25

pmarginal	lmnts
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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.
```

26 pnts

#### 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)
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)
```

pnts 27

#### **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')
```

28 portfolioVaRmnts

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)
```

qmarginalmnts 29

#### **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	inalmnts	almnts qmarginalmnts
---------------	---------------	----------	----------------------

#### **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.

30 qnts

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
```

rmnts 31

```
#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 Cholesky decomposition.

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

# Usage

```
rmnts(strMnts, numofsample)
```

# Arguments

numofsample number of samples.

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

strPMNTS\$ndim: dimension

strPMNTS\$mu :  $\mu$  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).

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

#### 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

32 rmnts\_subord

#### **Examples**

rmnts\_subord

rmnts\_subord

# Description

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

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

#### Usage

```
rmnts_subord(strPMNTS, numofsample, rW = NaN, rTau = NaN)
```

#### **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.

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).

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

#### **Examples**

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)
```

#### **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  $(\alpha, \theta, \beta, \gamma, \mu, t)$ .

# 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)</pre>
```

34 setPortfolioParam

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

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.  \begin{split} & \mathsf{strPMNTS\$ndim}: \text{dimension} \\ & \mathsf{strPMNTS\$mu}: \mu \text{ mean vector (column vector) of the input data.} \\ & \mathsf{strPMNTS\$sigma}: \sigma \text{ standard deviation vector (column vector) of the input data.} \\ & \mathsf{strPMNTS\$alpha}: \alpha \text{ of the std NTS distribution (X).} \\ & \mathsf{strPMNTS\$theta}: \theta \text{ of the std NTS distribution (X).} \\ & \mathsf{strPMNTS\$beta}: \beta \text{ vector (column vector) of the std NTS distribution (X).} \\ & \mathsf{strPMNTS\$Rho}: \Sigma \text{ matrix of the std NTS distribution (X).} \\ & \mathsf{w} & \mathsf{Capital allocation weight vector.} \end{split}
```

setPortfolioParam 35

#### Value

```
The weighted sum follows 1-dimensional NTS. R_p = < w, r >= \mu + diag(\sigma)X, where X \text{ follows } stdNTS_1(\alpha, \theta, \beta, 1). Hence we obtain \operatorname{res\$mu}: \mu \text{ mean of } R_p. \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.
```

#### 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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