# Package 'RND' 

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RND-package Risk Neutral Density Extraction Package

## Description

This package is a collection of various functions to extract the implied risk neutral density from option.

## Details

| Package: | RND |
| :--- | :--- |
| Type: | Package |
| Version: | 1.2 |
| Date: | $2017-01-10$ |
| License: | GPL $(>=2)$ |

## Author(s)

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## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
###
### You should see that all methods extract the same density!
###
r=0.05
te = 60/365
s0}=100
sigma = 0.25
y}=0.0
call.strikes.bsm = seq(from = 500, to = 1500, by = 5)
market.calls.bsm = price.bsm.option(r =r, te = te, s0 = s0,
    k = call.strikes.bsm, sigma = sigma, y = y)$call
put.strikes.bsm = seq(from = 500, to = 1500, by = 5)
market.puts.bsm = price.bsm.option(r =r, te = te, s0 = s0,
    k = put.strikes.bsm, sigma = sigma, y = y)$put
###
### See where your results will be outputted to...
###
getwd()
###
### Running this may take a few minutes...
###
### MOE(market.calls.bsm, call.strikes.bsm, market.puts.bsm,
### put.strikes.bsm, s0, r , te, y, "bsm2")
###
```

approximate.max

Max Function Approximation

## Description

approximate. max gives a smooth approximation to the max function.

## Usage

approximate.max $(x, y, k=5)$

## Arguments

x
the first argument for the max function
y
the second argument fot the max function
k
a tuning parameter. The larger this value, the closer the function output to a true max function.

## Details

approximate. max approximates the max of $x$, and $y$ as follows:

$$
g(x, y)=\frac{1}{1+\exp (-k(x-y))}, \quad \max (x, y) \approx x g(x, y)+y(1-g(x, y))
$$

## Value

approximate maximum of $x$ and $y$

## Author(s)

Kam Hamidieh

## References

Melick, W. R. and Thomas, C.P. (1997) Recovering an asset's implied pdf from option proces: An application to crude oil during the gulf crisis. Journal of Financial and Quantitative Analysis, 32(1), 91-115

## Examples

```
#
# To see how the max function compares with approximate.max,
# run the following code.
#
i = seq(from = 0, to = 10, by = 0.25)
y = i - 5
max.values = pmax(0,y)
approximate.max.values = approximate.max (0, y,k=5)
matplot(i, cbind(max.values, approximate.max.values), lty = 1, type = "l",
col=c("black","red"), main = "Max in Black, Approximate Max in Red")
```

bsm.objective
BSM Objective Function

## Description

bsm. objective is the objective function to be minimized in extract.bsm. density.

## Usage

bsm.objective(s0, r, te, y, market.calls, call.strikes, call.weights = 1, market.puts, put.strikes, put.weights $=1$, lambda $=1$, theta)

## Arguments

s0 current asset value
r
risk free rate
te time to expiration
$y \quad$ dividend yield
market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
call.weights weights to be used for calls
market.puts market calls (cheapest to most expensive)
put.strikes strikes for the puts (smallest to largest)
put.weights weights to be used for calls
lambda Penalty parameter to enforce the martingale condition
theta initial values for the optimization. This must be a vector of length 2: first component is $\mu$, the lognormal mean of the underlying density, and the second component is $\sqrt{t} \sigma$ which is the time scaled volatility parameter of the underlying density.

## Details

This function evaluates the weighted squared differences between the market option values and values predicted by the Black-Scholes-Merton option pricing formula.

## Value

Objective function evalued at a specific set of values.

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
\(r=0.05\)
te \(=60 / 365\)
s0 \(=1000\)
sigma \(=0.25\)
\(\mathrm{y}=0.01\)
call.strikes \(=\operatorname{seq}(\) from \(=500\), to \(=1500\), by \(=25)\)
market.calls \(=\) price.bsm.option( \(r=r\), te \(=\) te, \(s 0=s 0\),
    \(\mathrm{k}=\) call.strikes, sigma = sigma, \(\mathrm{y}=\mathrm{y}) \$\) call
```

```
put.strikes = seq(from = 510, to = 1500, by = 25)
market.puts = price.bsm.option(r =r, te = te, s0 = s0,
    k = put.strikes, sigma = sigma, y = y)$put
###
### perfect initial values under BSM framework
###
mu.0 = log(s0) + ( r - y - 0.5 * sigma^2) * te
zeta.0 = sigma * sqrt(te)
mu.0
zeta.0
###
### The objective function should be *very* small
###
bsm.obj.val = bsm.objective(theta=c(mu.0, zeta.0), r = r, y=y, te = te, s0 = s0,
    market.calls = market.calls, call.strikes = call.strikes,
    market.puts = market.puts, put.strikes = put.strikes, lambda = 1)
bsm.obj.val
```

compute.implied.volatility

Compute Impied Volatility

## Description

compute.implied.volatility extracts the implied volatility for a call option.

## Usage

compute.implied.volatility(r, te, s0, k, y, call.price, lower, upper)

## Arguments

| $r$ | risk free rate |
| :--- | :--- |
| te | time to expiration |
| $s 0$ | current asset value |
| $k$ | strike of the call option |
| $y$ | dividend yield |
| call.price | call price |

lower lower bound of the implied volatility to look for
upper upper bound of the implied volatility to look for

## Details

The simple R uniroot function is used to extract the implied volatility.

## Value

sigma extratced implied volatility

## Author(s)

Kam Hamidieh

## References

J. Hull (2011) Options, Futures, and Other Derivatives and DerivaGem Package Prentice Hall, Englewood Cliffs, New Jersey, 8th Edition
R. L. McDonald (2013) Derivatives Markets Pearson, Upper Saddle River, New Jersey, 3rd Edition

## Examples

```
#
# Create prices from BSM with various sigma's
#
r=0.05
y = 0.02
te = 60/365
s0 = 400
sigma.range = seq(from = 0.1, to = 0.8, by = 0.05)
k.range = floor(seq(from = 300, to = 500, length.out = length(sigma.range)))
bsm.calls = numeric(length(sigma.range))
for (i in 1:length(sigma.range))
{
    bsm.calls[i] = price.bsm.option(r = r, te = te, s0 = s0, k = k.range[i],
                                    sigma = sigma.range[i], y = y)$call
}
bsm.calls
k.range
#
# Computed implied sigma's should be very close to sigma.range.
#
compute.implied.volatility(r = r, te = te, s0 = s0, k = k.range, y = y,
                                    call.price = bsm.calls, lower = 0.001, upper = 0.999)
sigma.range
```

dew Edgeworth Density

## Description

dew is the probability density function implied by the Edgeworth expansion method.

## Usage

dew (x, r, y, te, s0, sigma, skew, kurt)

## Arguments

| $x$ | value at which the denisty is to be evaluated |
| :--- | :--- |
| $r$ | risk free rate |
| $y$ | dividend yield |
| te | time to expiration |
| s0 | current asset value |
| sigma | volatility |
| skew | normalized skewness |
| kurt | normalized kurtosis |

## Details

This density function attempts to capture deviations from lognormal density by using Edgeworth expansions.

## Value

density value at $x$

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London
R. Jarrow and A. Rudd (1982) Approximate valuation for arbitrary stochastic processes. Journal of Finanical Economics, 10, 347-369
C.J. Corrado and T. Su (1996) S\&P 500 index option tests of Jarrow and Rudd's approximate option valuation formula. Journal of Futures Markets, 6, 611-629

## Examples

## \#

\# Look at a true lognorma density \& related dew
\#
$r=0.05$
$\mathrm{y}=0.03$
s0 $=1000$
sigma $=0.25$
te $=100 / 365$
strikes $=$ seq $($ from $=600$, to $=1400$, by $=1)$
v $\quad=\operatorname{sqrt}\left(\exp \left(\operatorname{sigma}^{\wedge} 2 *\right.\right.$ te $\left.)-1\right)$
ln. skew $=3 * v+v^{\wedge} 3$
ln. kurt $=16 * v^{\wedge} 2+15 * v^{\wedge} 4+6 * v^{\wedge} 6+v^{\wedge} 8$
skew. $4=\ln$. skew * 1.50
kurt. 4 = ln.kurt * 1.50
skew. $5=\ln$. skew * 0.50
kurt. $5=\ln$. kurt * 2.00
ew.density. $4=\operatorname{dew}(x=s t r i k e s, r=r, y=y, t e=t e, s 0=s 0$, sigma=sigma,
skew=skew.4, kurt=kurt.4)
ew.density. $5=\operatorname{dew}(x=$ strikes, $r=r, y=y, t e=t e, s 0=s 0$, sigma=sigma, skew=skew.5, kurt=kurt.5)
bsm.density $=\operatorname{dlnorm}(x=$ strikes, meanlog $=\log (s 0)+(r-y-(\operatorname{sigma} 2) / 2) * t e$, sdlog $=$ sigma*sqrt(te), log = FALSE)
matplot(strikes, cbind(bsm.density, ew.density.4, ew.density.5), type="l",
lty=c(1,1,1), col=c("black","red","blue"),
main="Black $=\mathrm{BSM}, \quad$ Red $=\mathrm{EW} 1.5$ Times, Blue $=\mathrm{EW} 0.50$ \& 2")
$\mathrm{dgb} \quad$ Generalized Beta Density

## Description

dgb is the probability density function of generalized beta distribution.

## Usage

$$
\operatorname{dgb}(x, a, b, v, w)
$$

## Arguments

| x | value at which the denisty is to be evaluated |
| :--- | :--- |
| a | power parameter $>0$ |
| b | scale paramter $>0$ |
| v | first beta paramter $>0$ |
| w | second beta parameter $>0$ |

## Details

Let B be a beta random variable with parameters v and w , then $Z=b(B /(1-B))^{1 / a}$ is a generalized beta with parameters (a,b,v,w).

## Value

density value at x

## Author(s)

Kam Hamidieh

## References

R.M. Bookstaber and J.B. McDonald (1987) A general distribution for describing security price returns. Journal of Business, 60, 401-424
X. Liu and M.B. Shackleton and S.J. Taylor and X. Xu (2007) Closed-form transformations from risk-neutral to real-world distributions Journal of Business, 60, 401-424
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

\#
\# Just simple plot of the density
\#
$x=\operatorname{seq}($ from $=500$, to $=1500$, length.out $=10000)$
a $=10$
b $=1000$
$v=3$
$w=3$
$d x=\operatorname{dgb}(x=x, a=a, b=b, v=v, w=w)$
plot(dx ~ x, type="l")
$\mathrm{dmln} \quad$ Density of Mixture Lognormal

## Description

mln is the probability density function of a mixture of two lognormal densities.

## Usage

dmln(x, alpha.1, meanlog.1, meanlog.2, sdlog.1, sdlog.2)

## Arguments

X
alpha. 1 proportion of the first lognormal. Second one is 1 - alpha. 1
meanlog. 1 mean of the log of the first lognormal
meanlog. 2 mean of the log of the second lognormal
sdlog. 1 standard deviation of the log of the first lognormal
sdlog. 2 standard deviation of the log of the second lognormal

## Details

mln is the density $\mathrm{f}(\mathrm{x})=$ alpha. $1 * \mathrm{~g}(\mathrm{x})+(1-$ alpha. 1$) * \mathrm{~h}(\mathrm{x})$, where g and h are densities of two lognormals with parameters (mean.log.1, sdlog.1) and (mean.log.2, sdlog.2) respectively.

## Value

out density value at $x$

## Author(s)

Kam Hamidieh

## References

B. Bahra (1996): Probability distribution of future asset prices implied by option prices. Bank of England Quarterly Bulletin, August 1996, 299-311
P. Soderlind and L.E.O. Svensson (1997) New techniques to extract market expectations from financial instruments. Journal of Monetary Economics, 40, 383-429
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# A bimodal risk neutral density!
#
mln.alpha.1 = 0.4
mln.meanlog. 1 = 6.3
mln.meanlog. 2 = 6.5
mln.sdlog.1 = 0.08
mln.sdlog.2 = 0.06
k = 300:900
dx = dmln(x = k, alpha.1 = mln.alpha.1, meanlog.1 = mln.meanlog.1,
    meanlog.2 = mln.meanlog.2,
    sdlog.1 = mln.sdlog.1, sdlog. 2 = mln.sdlog.2)
plot(dx ~ k, type="l")
```


## Description

mln .am is the probability density function of a mixture of three lognormal densities.

## Usage

dmln.am(x, u.1, u.2, u.3, sigma.1, sigma.2, sigma.3, p.1, p.2)

## Arguments

$x \quad$ value at which the denisty is to be evaluated
u. $1 \quad \log$ mean of the first lognormal
u. $2 \quad \log$ mean of the second lognormal
u. $3 \quad \log$ mean of the third lognormal
sigma. $1 \quad \log$ standard deviation of the first lognormal
sigma. $2 \quad \log$ standard deviation of the second lognormal
sigma. 3 log standard deviation of the third lognormal
p. 1 weight assigned to the first density
p. 2 weight assigned to the second density

## Details

mln is density $\mathrm{f}(\mathrm{x})=\mathrm{p} .1 * \mathrm{f} 1(\mathrm{x})+\mathrm{p} .2 * \mathrm{f} 2(\mathrm{x})+(1-\mathrm{p} .1-\mathrm{p} .2) * \mathrm{f} 3(\mathrm{x})$, where f 1 , f 2 , and f 3 are lognormal densities with log means u.1,u.2, and u. 3 and standard deviations sigma.1, sigma.2, and sigma. 3 respectively.

## Value

out density value at x

## Author(s)

Kam Hamidieh

## References

Melick, W. R. and Thomas, C. P. (1997). Recovering an asset's implied pdf from option prices: An application to crude oil during the gulf crisis. Journal of Financial and Quantitative Analysis, 32(1), 91-115.

## Examples

```
###
### Just look at a generic density and see if it integrates to 1.
###
u.1 = 4.2
u.2 = 4.5
u.3 = 4.8
sigma. }1=0.3
sigma.2 = 0.20
sigma. }3=0.1
p.1 = 0.25
p.2=0.45
x = seq(from = 0, to = 250, by = 0.01)
y = dmln.am(x = x, u.1 = u.1, u.2 = u.2, u. 3 = u.3, sigma.1 = sigma.1, sigma.2 = sigma.2,
    sigma. }3=\mathrm{ sigma. 3, p.1 = p.1, p.2 = p.2)
plot(y ~ x, type="l")
sum(y * 0.01)
###
### Yes, the sum is near 1.
###
```

dshimko Density Implied by Shimko Method

## Description

dshimko is the probability density function implied by the Shimko method.

## Usage

dshimko(r, te, s0, k, y, a0, a1, a2)

## Arguments

| $r$ | risk free rate |
| :--- | :--- |
| te | time to expiration |
| s0 | current asset value |
| $k$ | strike at which volatility to be computed |
| $y$ | dividend yield |
| a0 | constant term in the quadratic polynomial |
| a1 | coefficient term of $k$ in the quadratic polynomial |
| a2 | coefficient term of $k$ squared in the quadratic polynomial |

## Details

The implied volatility is modeled as: $\sigma(k)=a_{0}+a_{1} k+a_{2} k^{2}$

## Value

density value at x

## Author(s)

## Kam Hamidieh

## References

D. Shimko (1993) Bounds of probability. Risk, 6, 33-47
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# a0, a1, a2 values come from Shimko's paper.
#
r=0.05
y=0.02
a0 = 0.892
a1 = -0.00387
a2 = 0.00000445
te = 60/365
s0 = 400
k = seq(from = 250, to = 500, by = 1)
sigma = 0.15
#
# Does it look like a proper density and intergate to one?
#
dx = dshimko(r = r, te = te, s0 = s0, k = k, y = y, a0 = a0, a1 = a1, a2 = a2)
plot(dx ~ k, type="l")
#
# sum(dx) should be about 1 since dx is a density.
#
sum(dx)
```


## Description

ew. objective is the objective function to be minimized in ew. extraction.

## Usage

ew.objective(theta, r, y, te, s0, market.calls, call.strikes, call.weights = 1, lambda = 1)

## Arguments

| theta | initial values for the optimization |
| :--- | :--- |
| $r$ | risk free rate |
| $y$ | dividend yield |
| te | time to expiration |
| s0 | current asset value |
| market.calls | market calls (most expensive to cheapest) |
| call.strikes | strikes for the calls (smallest to largest) |
| call.weights | weights to be used for calls |
| lambda | Penalty parameter to enforce the martingale condition |

## Details

This function evaluates the weighted squared differences between the market option values and values predicted by Edgworth based expansion of the risk neutral density.

## Value

Objective function evalued at a specific set of values

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London
R. Jarrow and A. Rudd (1982) Approximate valuation for arbitrary stochastic processes. Journal of Finanical Economics, 10, 347-369
C.J. Corrado and T. Su (1996) S\&P 500 index option tests of Jarrow and Rudd's approximate option valuation formula. Journal of Futures Markets, 6, 611-629

## Examples

```
r=0.05
y = 0.03
s0 = 1000
sigma = 0.25
te = 100/365
k = seq(from=800, to = 1200, by = 50)
v = sqrt(exp(sigma^2 * te) - 1)
ln.skew = 3 * v + v^3
ln.kurt = 16 * v^2 + 15 * v^4 + 6 * v^6 + v^8
```

\#
\# The objective function should be close to zero.
\# Also the weights are automatically set to 1.
\#
market.calls.bsm $=$ price.bsm.option $(r=r$, te $=t e, ~ s 0=s 0, k=k$,
sigma=sigma, y=y)\$call
ew. objective(theta $=c($ sigma, ln.skew, ln.kurt), $r=r, y=y, ~ t e ~=~ t e, ~ s 0=s 0$,
market.calls = market.calls.bsm, call.strikes $=\mathrm{k}, \operatorname{lambda}=1$ )
extract.am.density Mixture of Lognormal Extraction for American Options

## Description

extract.am.density extracts the mixture of three lognormals from American options.

## Usage

extract.am.density(initial.values $=r e p(N A, 10), r, t e, ~ s 0, ~ m a r k e t . c a l l s$, call.weights = NA, market.puts, put.weights = NA, strikes, lambda = 1, hessian.flag = F, cl = list(maxit = 10000))

## Arguments

initial.values initial values for the optimization
$r$ risk free rate
te time to expiration
s0 current asset value
market.calls market calls (most expensive to cheapest)
call.weights weights to be used for calls. Set to 1 by default.
market.puts market calls (cheapest to most expensive)
extract.am.density

| put.weights | weights to be used for puts. Set to 1 by default. |
| :--- | :--- |
| strikes | strikes (smallest to largest) |
| lambda | Penalty parameter to enforce the martingale condition |
| hessian.flag | If FALSE then no Hessian is produced |
| cl | List of parameter values to be passed to the optimization function |

## Details

The extracted density is in the form of $\mathrm{f}(\mathrm{x})=\mathrm{p} .1 * \mathrm{f} 1(\mathrm{x})+\mathrm{p} .2 * \mathrm{f} 2(\mathrm{x})+(1-\mathrm{p} .1-\mathrm{p} .2) * \mathrm{f} 3(\mathrm{x})$, where $\mathrm{f} 1, \mathrm{f} 2$, and f 3 are lognormal densities with $\log$ means $\mathrm{u} .1, \mathrm{u} .2$, and u .3 and standard deviations sigma. 1 , sigma. 2 , and sigma. 3 respectively.
For the description of w. 1 and w. 2 see equations (7) \& (8) of Melick and Thomas paper below.

## Value

w. 1 First weight, a number between 0 and 1, to weigh the option price bounds
w. 2 Second weight, a number between 0 and 1, to weigh the option price bounds
u. $1 \quad \log$ mean of the first lognormal
u. $2 \quad \log$ mean of the second lognormal
u. $3 \quad \log$ mean of the third lognormal
sigma. $1 \quad \log$ sd of the first lognormal
sigma. $2 \quad \log$ sd of the second lognormal
sigma. $3 \quad \log$ sd of the third lognormal
p. 1 weight assigned to the first density
p. 2 weight assigned to the second density
converge.result
Captures the convergence result
hessian Hessian Matrix

## Author(s)

Kam Hamidieh

## References

Melick, W. R. and Thomas, C. P. (1997). Recovering an asset's implied pdf from option prices: An application to crude oil during the gulf crisis. Journal of Financial and Quantitative Analysis, 32(1), 91-115.

## Examples

```
###
### Try with synthetic data first.
###
r=0.01
te=60/365
w.1 = 0.4
w.2 = 0.25
u.1 = 4.2
u.2 = 4.5
u.3 = 4.8
sigma. }1=0.3
sigma.2 = 0.20
sigma. 3 = 0.15
p.1 = 0.25
p.2 = 0.45
theta = c(w.1,w.2,u.1,u.2,u.3,sigma.1,sigma.2, sigma.3,p.1,p.2)
p.3 = 1 - p.1 - p.2
###
### Generate some synthetic American calls & puts
###
expected.f0 = sum(c(p.1, p.2, p.3) * exp(c(u.1,u.2,u.3) +
    (c(sigma.1, sigma.2, sigma.3)^2)/2) )
expected.f0
strikes = 50:150
```

market.calls = numeric(length(strikes))
market.puts $=$ numeric(length(strikes))
for (i in 1:length(strikes))
\{
if ( strikes[i] < expected.f0) \{
market.calls[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.1, u. $1=u .1$,
u. 2 = u. $2, ~ u .3=u .3$, sigma. $1=$ sigma. 1 , sigma. $2=$ sigma. 2 ,
sigma. $3=$ sigma. 3 , p. $1=$ p. 1, p. $2=$ p.2) $\$$ call.value
market.puts[i] = price.am.option(k = strikes[i], r=r, te =te, w = w. $2, \mathrm{u} .1=\mathrm{u} .1$,
u. 2 = u. 2, u. $3=$ u. 3 , sigma. $1=$ sigma. 1 , sigma. $2=$ sigma. 2 ,
sigma. $3=$ sigma. 3 , p. $1=$ p. 1, p. $2=$ p. 2$) \$$ put.value
\} else \{
market.calls[i] = price.am.option(k = strikes[i], r = r, te = te, w = w. $2, u .1=u .1$,
u. $2=$ u. 2, u. $3=$ u. 3 , sigma. $1=$ sigma. 1 , sigma. $2=$ sigma. 2 ,
sigma. 3 = sigma. 3 , p. $1=\mathrm{p} .1, \mathrm{p} .2=\mathrm{p} .2) \$ \mathrm{call}$. value
market. puts[i] = price.am.option(k = strikes[i], r=r, te =te, w=w.1, u.1 = u.1,

```
                        u.2 = u.2, u. 3 = u.3, sigma. }1=\mathrm{ sigma.1, sigma. 2 = sigma.2,
                sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$put.value
    }
}
###
### ** IMPORTANT **: The code that follows may take 1-2 minutes.
### Copy and paste onto R console the commands
### that follow the greater sign >.
###
### Try the optimization with exact inital values.
### They should be close the actual initials.
###
#
# > s0 = expected.f0 * exp(-r * te)
# > s0
#
# > extract.am.density(initial.values = theta, r = r, te = te, s0 = s0,
# market.calls = market.calls, market.puts = market.puts, strikes = strikes,
# lambda = 1, hessian.flag = FALSE)
#
# > theta
#
###
### Now try without our the correct initial values...
###
#
# > optim.obj.no.init = extract.am.density( r = r, te = te, s0 = s0,
# market.calls = market.calls, market.puts = market.puts, strikes = strikes,
# lambda = 1, hessian.flag = FALSE)
#
# > optim.obj.no.init
# > theta
#
###
### We do get different values but how do the densities look like?
###
#
###
### plot the two densities side by side
###
#
# > x = 10:190
#
# > y. 1 = dmln.am(x = x, p. 1 = theta[9], p. 2 = theta[10],
# u.1 = theta[3], u.2 = theta[4], u. 3 = theta[5],
# sigma. }1=\mathrm{ theta[6], sigma. 2 = theta[7], sigma. 3 = theta[8] )
#
# > o = optim.obj.no.init
#
# > y.2 = dmln.am(x = x, p.1 = o$p.1, p. 2 = o$p.2,
```

```
# u.1 = o$u.1, u.2 = o$u.2, u.3 = o$u.3,
# sigma.1 = o$sigma.1, sigma.2 = o$sigma.2, sigma. 3 = o$sigma.3 )
#
# > matplot(x, cbind(y.1,y.2), main = "Exact = Black, Approx = Red", type="l", lty=1)
#
###
### Densities are close.
###
```

extract.bsm.density Extract Lognormal Density

## Description

bsm. extraction extracts the parameters of the lognormal density as implied by the BSM model.

## Usage

extract.bsm.density(initial.values $=c(N A, N A), r, y, t e, s 0$, market.calls, call.strikes, call.weights $=1$, market.puts, put.strikes, put.weights $=1$, lambda $=1$, hessian.flag $=\mathrm{F}, \mathrm{cl}=\operatorname{list}($ maxit $=10000)$ )

## Arguments

initial.values initial values for the optimization
$r \quad$ risk free rate
$y \quad$ dividend yield
te time to expiration
s0 current asset value
market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
call.weights weights to be used for calls
market.puts market calls (cheapest to most expensive)
put.strikes strikes for the puts (smallest to largest)
put.weights weights to be used for puts
lambda Penalty parameter to enforce the martingale condition
hessian.flag if F , no hessian is produced
cl list of parameter values to be passed to the optimization function

## Details

If initial.values are not specified then the function will attempt to pick them automatically. cl is a list that can be used to pass parameters to the optim function.
extract.bsm.density

## Value

Let S_T with the lognormal random variable of the risk neutral density.

```
mu mean of log(S_T)
zeta sd of log(S_T)
converge.result
    Did the result converge?
hessian Hessian matrix
```


## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London
J. Hull (2011) Options, Futures, and Other Derivatives and DerivaGem Package Prentice Hall, Englewood Cliffs, New Jersey, 8th Edition
R. L. McDonald (2013) Derivatives Markets Pearson, Upper Saddle River, New Jersey, 3rd Edition

## Examples

```
#
# Create some BSM Based options
#
r = 0.05
te = 60/365
s0 = 1000
sigma = 0.25
y =0.01
call.strikes = seq(from = 500, to = 1500, by = 25)
market.calls = price.bsm.option(r =r, te = te, s0 = s0,
    k = call.strikes, sigma = sigma, y = y)$call
put.strikes = seq(from = 510, to = 1500, by = 25)
market.puts = price.bsm.option(r =r, te = te, s0 = s0,
                        k = put.strikes, sigma = sigma, y = y)$put
#
# Get extract the parameter of the density
#
extract.bsm.density(r = r, y = y, te = te, s0 = s0, market.calls = market.calls,
    call.strikes = call.strikes, market.puts = market.puts,
    put.strikes = put.strikes, lambda = 1, hessian.flag = FALSE)
```

```
# The extracted parameters should be close to these actual values:
#
actual.mu = log(s0) + ( r - y - 0.5 * sigma^2) * te
actual.zeta = sigma * sqrt(te)
actual.mu
actual.zeta
```

extract.ew.density
Extract Edgeworth Based Density

## Description

ew. extraction extracts the parameters for the density approximated by the Edgeworth expansion method.

## Usage

extract.ew.density(initial.values $=c(N A, N A, N A), r, y, t e, s 0$, market.calls, call.strikes, call.weights $=1$, lambda $=1$, hessian.flag = F , cl = list(maxit = 10000))

## Arguments

initial.values initial values for the optimization
$r$
risk free rate
$y \quad$ dividend yield
te time to expiration
s0 current asset value
market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
call.weights weights to be used for calls
lambda Penalty parameter to enforce the martingale condition
hessian.flag if F , no hessian is produced
cl list of parameter values to be passed to the optimization function

## Details

If initial.values are not specified then the function will attempt to pick them automatically. cl in form of a list can be used to pass parameters to the optim function.
extract.ew.density

## Value

| sigma | volatility of the underlying lognormal |
| :--- | :--- |
| skew | normalized skewness |
| kurt | normalized kurtosis |
| converge.result |  |$\quad$|  | Did the result converge? |
| :--- | :--- |
| hessian | Hessian matrix |

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London
R. Jarrow and A. Rudd (1982) Approximate valuation for arbitrary stochastic processes. Journal of Finanical Economics, 10, 347-369
C.J. Corrado and T. Su (1996) S\&P 500 index option tests of Jarrow and Rudd's approximate option valuation formula. Journal of Futures Markets, 6, 611-629

## Examples

```
#
# ln.skew & ln.kurt are the normalized skewness and kurtosis of a true lognormal.
#
r=0.05
y = 0.03
s0 = 1000
sigma = 0.25
te = 100/365
strikes = seq(from=600, to = 1400, by = 1)
v = sqrt(exp(sigma^2 * te) - 1)
ln.skew = 3 * v + v^3
ln.kurt = 16* v^2 + 15* v^4 + 6 * v^6 + v^8
#
# Now "perturb" the lognormal
#
new.skew = ln.skew * 1.50
new.kurt = ln.kurt * 1.50
#
# new.skew & new.kurt should not be extracted.
# Note that weights are automatically set to 1.
#
```

```
market.calls = price.ew.option(r = r, te = te, s0 = s0, k=strikes, sigma=sigma,
    y=y, skew = new.skew, kurt = new.kurt)$call
ew.extracted.obj = extract.ew.density(r = r, y = y, te = te, s0 = s0,
    market.calls = market.calls, call.strikes = strikes,
    lambda = 1, hessian.flag = FALSE)
ew.extracted.obj
```

extract.gb.density Generalized Beta Extraction

## Description

extract.gb.density extracts the generalized beta density from market options.

## Usage

extract.gb.density(initial.values $=c(N A, N A, N A, N A), r$, te, $y, ~ s 0$, market.calls, call.strikes, call.weights $=1$, market.puts, put.strikes, put.weights = 1, lambda = 1, hessian.flag = F, cl = list(maxit = 10000))

## Arguments

initial.values initial values for the optimization
r
te time to expiration
$y \quad$ dividend yield
s0 current asset value
market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
call.weights weights to be used for calls
market.puts market calls (cheapest to most expensive)
put.strikes strikes for the puts (smallest to largest)
put.weights weights to be used for puts
lambda Penalty parameter to enforce the martingale condition
hessian.flag if F , no hessian is produced
cl list of parameter values to be passed to the optimization function

## Details

This function extracts the generalized beta density implied by the options.
extract.gb.density

## Value

| a | extracted power parameter |
| :--- | :--- |
| b | extracted scale paramter |
| v | extracted first beta paramter |
| w | extracted second beta parameter |
| converge.result |  |
|  | Did the result converge? |
| hessian | Hessian matrix |

## Author(s)

Kam Hamidieh

## References

R.M. Bookstaber and J.B. McDonald (1987) A general distribution for describing security price returns. Journal of Business, 60, 401-424
X. Liu and M.B. Shackleton and S.J. Taylor and X. Xu (2007) Closed-form transformations from risk-neutral to real-world distributions Journal of Business, 60, 401-424
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# create some GB based calls and puts
#
r = 0.03
te = 50/365
k = seq(from = 800, to = 1200, by = 10)
a = 10
b = 1000
v = 2.85
w = 2.85
y = 0.01
s0 = exp((y-r)*te) * b * beta(v + 1/a,w - 1/a)/beta(v,w)
s0
call.strikes = seq(from = 800, to = 1200, by = 10)
market.calls = price.gb.option(r = r, te = te, y = y, s0 = s0,
    k = call.strikes, a = a, b = s0, v = v, w = w)$call
put.strikes = seq(from = 805, to = 1200, by = 10)
market.puts = price.gb.option(r = r, te = te, y = y, s0 = s0,
    k = put.strikes, a = a, b = s0, v = v, w = w)$put
```

```
#
# Extraction...should match the a,b,v,w above. You will also get warning messages.
# Weigths are automatically set to 1.
#
extract.gb.density(r=r, te=te, y = y, s0=s0, market.calls = market.calls,
    call.strikes = call.strikes, market.puts = market.puts,
    put.strikes = put.strikes, lambda = 1, hessian.flag = FALSE)
```

extract.mln.density Extract Mixture of Lognormal Densities

## Description

mln . extraction extracts the parameters of the mixture of two lognormals densities.

## Usage

extract.mln.density(initial.values $=c(N A, N A, N A, N A, N A), r, y, t e, ~ s 0$, market.calls, call.strikes, call.weights $=1$, market.puts, put.strikes, put.weights = 1, lambda = 1, hessian.flag = F, cl = list(maxit = 10000))

## Arguments

initial.values initial values for the optimization
$r$ risk free rate
$y \quad$ dividend yield
te time to expiration
s0 current asset value
market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
call.weights weights to be used for calls
market.puts market calls (cheapest to most expensive)
put.strikes strikes for the puts (smallest to largest)
put.weights weights to be used for puts
lambda Penalty parameter to enforce the martingale condition
hessian.flag if F , no hessian is produced
cl list of parameter values to be passed to the optimization function

## Details

mln is the density $\mathrm{f}(\mathrm{x})=$ alpha. $1 * \mathrm{~g}(\mathrm{x})+(1-\operatorname{alpha} .1) * \mathrm{~h}(\mathrm{x})$, where g and h are densities of two lognormals with parameters (mean.log.1, sdlog.1) and (mean.log.2, sdlog.2) respectively.

## Value

| alpha. 1 | extracted proportion of the first lognormal. Second one is $1-$ alpha. 1 |
| :--- | :--- |
| meanlog. 1 | extracted mean of the log of the first lognormal |
| meanlog. 2 | extracted mean of the log of the second lognormal |
| sdlog. 1 | extracted standard deviation of the log of the first lognormal |
| sdlog. 2 | extracted standard deviation of the log of the second lognormal |
| converge.result | Did the result converge? |
| hessian | Hessian matrix |

## Author(s)

Kam Hamidieh

## References

F. Gianluca and A. Roncoroni (2008) Implementing Models in Quantitative Finance: Methods and Cases
B. Bahra (1996): Probability distribution of future asset prices implied by option prices. Bank of England Quarterly Bulletin, August 1996, 299-311
P. Soderlind and L.E.O. Svensson (1997) New techniques to extract market expectations from financial instruments. Journal of Monetary Economics, 40, 383-4
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# Create some calls and puts based on mln and
# see if we can extract the correct values.
#
r = 0.05
y = 0.02
te = 60/365
meanlog.1 = 6.8
meanlog.2 = 6.95
sdlog.1 = 0.065
sdlog.2 = 0.055
alpha.1 = 0.4
call.strikes = seq(from = 800, to = 1200, by = 10)
market.calls = price.mln.option(r = r, y = y, te = te, k = call.strikes,
    alpha.1 = alpha.1, meanlog.1 = meanlog.1, meanlog. 2 = meanlog.2,
        sdlog.1 = sdlog.1, sdlog.2 = sdlog.2)$call
```

```
s0 = price.mln.option(r = r, y = y, te = te, k = call.strikes, alpha.1 = alpha.1,
    meanlog.1 = meanlog.1, meanlog.2 = meanlog.2,
    sdlog.1 = sdlog.1, sdlog.2 = sdlog.2)$s0
s0
put.strikes = seq(from = 805, to = 1200, by = 10)
market.puts = price.mln.option(r = r, y = y, te = te, k = put.strikes,
    alpha.1 = alpha.1, meanlog.1 = meanlog.1,
    meanlog.2 = meanlog.2, sdlog.1 = sdlog.1,
    sdlog.2 = sdlog.2)$put
###
### The extracted values should be close to the actual values.
###
extract.mln.density(r = r, y = y, te = te, s0 = s0, market.calls = market.calls,
    call.strikes = call.strikes, market.puts = market.puts,
    put.strikes = put.strikes, lambda = 1, hessian.flag = FALSE)
```

extract.rates Extract Risk Free Rate and Dividend Yield

## Description

extract. rates extracts the risk free rate and the dividend yield from European options.

## Usage

extract.rates(calls, puts, s0, k, te)

## Arguments

| calls | market calls (most expensive to cheapest) |
| :--- | :--- |
| puts | market puts (cheapest to most expensive) |
| s0 | current asset value |
| k | strikes for the calls (smallest to largest) |
| te | time to expiration |

## Details

The extraction is based on the put-call parity of the European options. Shimko (1993) - see below shows that the slope and intercept of the regression of the calls minus puts onto the strikes contains the risk free and the dividend rates.

## Value

risk.free.rate
extracted risk free rate
dividend.yield
extracted dividend rate

## Author(s)

Kam Hamidieh

## References

D. Shimko (1993) Bounds of probability. Risk, 6, 33-47

## Examples

\#
\# Create calls and puts based on BSM
\#
$r=0.05$
te $=60 / 365$
s0 $=1000$
$k \quad=\operatorname{seq}($ from $=900$, to $=1100$, by $=25)$
sigma $=0.25$
$y=0.01$
bsm.obj $=$ price.bsm.option $(r=r$, te $=$ te, $s 0=s 0, k=k$, sigma $=$ sigma, $y=y)$
calls = bsm.obj\$call
puts $=$ bsm.obj\$put
\#
\# Extract rates should give the values of $r$ and $y$ above:
\#
rates $=$ extract.rates(calls = calls, puts $=$ puts, $k=k, s 0=s 0$, te $=t e)$ rates

```
extract.shimko.density
```

Extract Risk Neutral Density based on Shimko's Method

## Description

shimko.extraction extracts the implied risk neutral density based on modeling the volatility as a quadratic function of the strikes.

## Usage

extract.shimko.density(market.calls, call.strikes, r, y, te, s0, lower, upper)

## Arguments

market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
r
risk free rate
$y \quad$ dividend yield
te time to expiration
s0 current asset value
lower lower bound for the search of implied volatility
upper upper bound for the search of implied volatility

## Details

The correct values for range of search must be specified.

## Value

implied.curve.obj
variable that holds a0, a1, and a2 which are the constant terms of the quadratic polynomial
shimko.density
density evaluated at the strikes
implied.volatilities
implied volatilities at each call.strike

## Author(s)

Kam Hamidieh

## References

D. Shimko (1993) Bounds of probability. Risk, 6, 33-47
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# Test the function shimko.extraction. If BSM holds then a1 = a2 = 0.
#
r=0.05
y=0.02
```

```
te = 60/365
s0 = 1000
k = seq(from = 800, to = 1200, by = 5)
sigma}=0.2
bsm.calls = price.bsm.option(r = r, te = te, s0 = s0, k = k,
                            sigma = sigma, y = y)$call
extract.shimko.density(market.calls = bsm.calls, call.strikes = k, r = r, y = y, te = te,
                s0 = s0, lower = -10, upper = 10)
#
# Note: a0 is about equal to sigma, and a1 and a2 are close to zero.
#
```

fit.implied.volatility.curve
Fit Implied Quadratic Volatility Curve

## Description

fit.implied.volatility.curve estimates the coefficients of the quadratic equation for the implied volatilities.

## Usage

fit.implied.volatility.curve(x, k)

## Arguments

| $x$ | a set of implied volatilities |
| :--- | :--- |
| $k$ | range of strikes |

## Details

This function estimates volatility $\sigma$ as a quadratic function of strike $k$ with the coefficents $a_{0}, a_{1}, a_{2}$ : $\sigma(k)=a_{0}+a_{1} k+a_{2} k^{2}$

## Value

a0 constant term in the quadratic ploynomial
a1 coefficient term of k in the quadratic ploynomial
a2 coefficient term of $k$ squared in the quadratic polynomial
summary.obj statistical summary of the fit

## Author(s)

Kam Hamidieh

## References

D. Shimko (1993) Bounds of probability. Risk, 6, 33-47
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

\#
\# Suppose we see the following implied volatilities and strikes:
\#
implied.sigma $=c(0.11,0.08,0.065,0.06,0.05)$
strikes $\quad=c(340,360,380,400,410)$
tmp $\quad=$ fit.implied.volatility.curve( $x=$ implied.sigma, $k=$ strikes)
tmp
strike.range $=340: 410$
plot(implied.sigma ~ strikes)
lines(strike.range, tmp\$a0 + tmp\$a1 * strike.range + tmp\$a2 * strike.range^2)

```
gb.objective Generalized Beta Objective
```


## Description

gb. objective is the objective function to be minimized in extract.gb. density.

## Usage

gb.objective(theta, r, te, y, s0, market.calls, call.strikes, call.weights = 1, market.puts, put.strikes, put.weights = 1, lambda = 1)

## Arguments

theta initial values for optimization
$r$ risk free rate
te time to expiration
$y \quad$ dividend yield
s0 current asset value
market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
call. weights weights to be used for calls
market.puts market calls (cheapest to most expensive)
put.strikes strikes for the puts (smallest to largest)
put.weights weights to be used for puts
lambda Penalty parameter to enforce the martingale condition

## Details

This is the function minimized by extract.gb. desnity function.

## Value

obj value of the objective function

## Author(s)

Kam Hamidieh

## References

R.M. Bookstaber and J.B. McDonald (1987) A general distribution for describing security price returns. Journal of Business, 60, 401-424
X. Liu and M.B. Shackleton and S.J. Taylor and X. Xu (2007) Closed-form transformations from risk-neutral to real-world distributions Journal of Business, 60, 401-424
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# The objective should be very small!
# Note the weights are automatically
# set to 1.
#
r = 0.03
te = 50/365
k = seq(from = 800, to = 1200, by = 10)
a = 10
b}=100
v = 2.85
w = 2.85
y = 0.01
s0 = exp((y-r)*te) * b * beta(v + 1/a,w - 1/a)/beta(v,w)
s0
call.strikes = seq(from = 800, to = 1200, by = 10)
market.calls = price.gb.option(r = r, te = te, s0 = s0, y = y,
    k = call.strikes, a = a, b = b, v = v, w = w)$call
put.strikes = seq(from = 805, to = 1200, by = 10)
```

```
market.puts = price.gb.option(r = r, te = te, s0 = s0, y = y,
    k = put.strikes, a = a, b = b, v = v, w = w)$put
gb.objective(theta=c(a,b,v,w),r = r, te = te, y = y, s0 = s0,
    market.calls = market.calls, call.strikes = call.strikes,
    market.puts = market.puts, put.strikes = put.strikes, lambda = 1)
```

get.point.estimate

Point Estimation of the Density

## Description

get. point.estimate estimates the risk neutral density by center differentiation.

## Usage

get.point.estimate(market.calls, call.strikes, r, te)

## Arguments

| market.calls | market calls (most expensive to cheapest) |
| :--- | :--- |
| call.strikes | strikes for the calls (smallest to largest) |
| $r$ | risk free rate |
| te | time to expiration |

## Details

This is a non-parametric estimate of the risk neutral density. Due to center differentiation, the density values are not estimated at the highest and lowest strikes.

## Value

point.estimates
values of the estimated density at each strike

## Author(s)

Kam Hamidieh

## References

J. Hull (2011) Options, Futures, and Other Derivatives and DerivaGem Package Prentice Hall, Englewood Cliffs, New Jersey, 8th Edition

## Examples

```
###
### Recover the lognormal density based on BSM
###
r=0.05
te}=60/36
s0}=100
k = seq(from = 500, to = 1500, by = 1)
sigma = 0.25
y}=0.0
```

```
bsm.calls = price.bsm.option(r = r, te = te, s0 = s0, k = k, sigma = sigma, y = y)$call
density.est = get.point.estimate(market.calls = bsm.calls,
    call.strikes = k, r = r , te = te)
len = length(k)-1
### Note, estimates at two data points (smallest and largest strikes) are lost
plot(density.est ~ k[2:len], type = "l")
```

mln.am. objective Objective function for the Mixture of Lognormal of American Options

## Description

mln .am. objective is the objective function to be minimized in extract.am. density.

## Usage

mln.am.objective(theta, s0, r, te, market.calls, call.weights = NA, market.puts, put.weights = NA, strikes, lambda = 1)

## Arguments

| theta | initial values for the optimization |
| :--- | :--- |
| s0 | current asset value |
| $r$ | risk free rate |
| te | time to expiration |
| market.calls | market calls (most expensive to cheapest) |
| call.weights | weights to be used for calls |
| market.puts | market calls (cheapest to most expensive) |
| put.weights | weights to be used for calls |
| strikes | strikes for the calls (smallest to largest) |
| lambda | Penalty parameter to enforce the martingale condition |

## Details

mln is density $\mathrm{f}(\mathrm{x})=\mathrm{p} .1 * \mathrm{f} 1(\mathrm{x})+\mathrm{p} .2 * \mathrm{f} 2(\mathrm{x})+(1-\mathrm{p} .1-\mathrm{p} .2) * \mathrm{f} 3(\mathrm{x})$, where f 1 , f 2 , and f 3 are lognormal densities with log means u.1,u.2, and u. 3 and standard deviations sigma.1, sigma.2, and sigma. 3 respectively.

## Value

obj Value of the objective function

## Author(s)

Kam Hamidieh

## References

Melick, W. R. and Thomas, C. P. (1997). Recovering an asset's implied pdf from option prices: An application to crude oil during the gulf crisis. Journal of Financial and Quantitative Analysis, 32(1), 91-115.

## Examples

```
r = 0.01
te = 60/365
w. }1=0.
w.2 = 0.25
u.1 = 4.2
u.2 = 4.5
u.3 = 4.8
sigma.1 = 0.30
sigma.2 = 0.20
sigma. }3=0.1
p.1 = 0.25
p.2 = 0.45
theta = c(w.1,w.2,u.1,u.2,u.3,sigma.1,sigma.2,sigma.3,p.1,p.2)
p.3 = 1 - p. 1-p.2
p. }
expected.f0 = sum(c(p.1, p.2, p.3) * exp(c(u.1,u.2,u.3) +
                                    (c(sigma.1, sigma.2, sigma.3)^2)/2) )
expected.f0
strikes = 30:170
market.calls = numeric(length(strikes))
market.puts = numeric(length(strikes))
for (i in 1:length(strikes))
{
    if ( strikes[i] < expected.f0) {
```

```
        market.calls[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.1, u.1 = u.1,
            u.2 = u.2, u.3 = u.3, sigma.1 = sigma.1, sigma.2 = sigma.2,
                        sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$call.value
        market.puts[i] = price.am.option(k = strikes[i], r = r, te = te, w = w. 2, u.1 = u.1,
                        u.2 = u.2, u.3 = u.3, sigma. 1 = sigma.1, sigma. 2 = sigma.2,
                            sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$put.value
} else {
    market.calls[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.2, u.1 = u.1,
                        u.2 = u.2, u. 3 = u.3, sigma. 1 = sigma.1, sigma. 2 = sigma.2,
                        sigma.3 = sigma.3, p.1 = p.1, p. 2 = p.2)$call.value
market.puts[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.1, u.1 = u.1,
                        u.2 = u.2, u. 3 = u.3, sigma. 1 = sigma.1, sigma. 2 = sigma.2,
                        sigma. 3 = sigma. 3, p.1 = p.1, p.2 = p.2)$put.value
    }
}
###
### Quickly look at the option values...
###
par(mfrow=c(1, 2))
plot(market.calls ~ strikes, type="l")
plot(market.puts ~ strikes, type="l")
par(mfrow=c(1,1))
###
### ** IMPORTANT **: The code that follows may take a few seconds.
### Copy and paste onto R console the commands
### that follow the greater sign >.
###
###
### Next try the objective function. It should be zero.
### Note: Let weights be the defaults values of 1.
###
#
# > s0 = expected.f0 * exp(-r * te)
# > s0
#
# > mln.am.objective(theta, s0 =s0, r = r, te = te, market.calls = market.calls,
# market.puts = market.puts, strikes = strikes, lambda = 1)
#
###
### Now directly try the optimization with perfect initial values.
###
#
#
# > optim.obj.with.synthetic.data = optim(theta, mln.am.objective, s0 = s0, r=r, te=te,
#
#
    market.calls = market.calls, market.puts = market.puts, strikes = strikes,
        lambda = 1, hessian = FALSE , control=list(maxit=10000) )
```

```
#
# > optim.obj.with.synthetic.data
#
# > theta
#
###
### It does take a few seconds but the optim converges to exact theta values.
###
```

```
mln.objective Objective function for the Mixture of Lognormal
```


## Description

mln . objective is the objective function to be minimized in extract.mln. density.

## Usage

mln.objective(theta, $r, y, t e, ~ s 0, ~ m a r k e t . c a l l s, ~ c a l l . s t r i k e s, ~ c a l l . w e i g h t s, ~$ market.puts, put.strikes, put.weights, lambda = 1)

## Arguments

| theta | initial values for the optimization |
| :--- | :--- |
| $r$ | risk free rate |
| y | dividend yield |
| te | time to expiration |
| s0 | current asset value |
| market.calls | market calls (most expensive to cheapest) |
| call.strikes | strikes for the calls (smallest to largest) |
| call.weights | weights to be used for calls |
| market.puts | market calls (cheapest to most expensive) |
| put.strikes | strikes for the puts (smallest to largest) |
| put.weights | weights to be used for puts |
| lambda | Penalty parameter to enforce the martingale condition |

## Details

mln is the density $\mathrm{f}(\mathrm{x})=$ alpha. $1 * \mathrm{~g}(\mathrm{x})+(1-\operatorname{alpha} .1) * \mathrm{~h}(\mathrm{x})$, where g and h are densities of two lognormals with parameters (mean.log.1, sdlog.1) and (mean.log.2, sdlog.2) respectively.

## Value

obj value of the objective function

## Author(s)

Kam Hamidieh

## References

F. Gianluca and A. Roncoroni (2008) Implementing Models in Quantitative Finance: Methods and Cases
B. Bahra (1996): Probability distribution of future asset prices implied by option prices. Bank of England Quarterly Bulletin, August 1996, 299-311
P. Soderlind and L.E.O. Svensson (1997) New techniques to extract market expectations from financial instruments. Journal of Monetary Economics, 40, 383-429
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# The mln objective function should be close to zero.
# The weights are automatically set to 1.
#
r = 0.05
te = 60/365
y = 0.02
meanlog.1 = 6.8
meanlog.2 = 6.95
sdlog.1 = 0.065
sdlog.2 = 0.055
alpha.1 = 0.4
# This is the current price implied by parameter values:
s0 = 981.8815
call.strikes = seq(from = 800, to = 1200, by = 10)
market.calls = price.mln.option(r=r, y = y, te = te, k = call.strikes,
                        alpha.1 = alpha.1, meanlog.1 = meanlog.1, meanlog.2 = meanlog.2,
        sdlog.1 = sdlog.1, sdlog.2 = sdlog.2)$call
put.strikes = seq(from = 805, to = 1200, by = 10)
market.puts = price.mln.option(r = r, y = y, te = te, k = put.strikes,
                                alpha.1 = alpha.1, meanlog.1 = meanlog.1, meanlog. 2 = meanlog.2,
    sdlog.1 = sdlog.1, sdlog.2 = sdlog.2)$put
mln.objective(theta=c(alpha.1,meanlog.1, meanlog.2 , sdlog.1, sdlog.2),
    r = r, y = y, te = te, s0 = s0,
    market.calls = market.calls, call.strikes = call.strikes,
    market.puts = market.puts, put.strikes = put.strikes, lambda = 1)
```


## Description

MOE function extracts the risk neutral density based on all models and summarizes the results.

## Usage

MOE(market.calls, call.strikes, market.puts, put.strikes, call.weights = 1, put.weights = 1, lambda = 1, s0, r, te, y, file.name = "myfile")

## Arguments

market.calls market calls (most expensive to cheapest)
call.strikes strikes for the calls (smallest to largest)
market.puts market calls (cheapest to most expensive)
put.strikes strikes for the puts (smallest to largest)
call.weights Weights for the calls (must be in the same order of calls)
put.weights Weights for the puts (must be in the same order of puts)
lambda Penalty parameter to enforce the martingale condition
s0 Current asset value
$r \quad$ risk free rate
te time to expiration
$y \quad$ dividend yield
file.name File names where analysis is to be saved. SEE DETAILS!

## Details

The MOE function in a few key strokes extracts the risk neutral density via various methods and summarizes the results.
This function should only be used for European options.
NOTE: Three files will be produced: filename will have the pdf version of the results. file.namecalls.csv will have the predicted call values. file.nameputs.csv will have the predicted put values.

## Value

bsm.mu mean of $\log (\mathrm{S}(\mathrm{T})$ ), when $\mathrm{S}(\mathrm{T})$ is lognormal
bsm.sigma $\quad \mathrm{SD}$ of $\log (\mathrm{S}(\mathrm{T})$ ), when $\mathrm{S}(\mathrm{T})$ is lognormal
gb.a extracted power parameter, when $S(T)$ is assumed to be a GB rv
gb.b extracted scale paramter, when $S(T)$ is assumed to be a GB rv
gb.v extracted first beta paramter, when $S(T)$ is assumed to be a GB rv

| gb.w | extracted second beta parameter, when $\mathrm{S}(\mathrm{T})$ is assumed to be a GB rv |
| :---: | :---: |
| mln.alpha. 1 | extracted proportion of the first lognormal. Second one is 1 -alpha. 1 in mixture of lognormals |
| mln.meanlog. 1 | extracted mean of the log of the first lognormal in mixture of lognormals |
| mln.meanlog. 2 | extracted mean of the log of the second lognormal in mixture of lognormals |
| mln.sdlog. 1 | extracted standard deviation of the log of the first lognormal in mixture of lognormals |
| mln.sdlog. 2 | extracted standard deviation of the $\log$ of the second lognormal in mixture of lognormals |
| ew.sigma | volatility when using the Edgeworth expansions |
| ew.skew | normalized skewness when using the Edgeworth expansions |
| ew.kurt | normalized kurtosis when using the Edgeworth expansions |
| a0 | extracted constant term in the quadratic polynomial of Shimko method |
| a1 | extracted coefficient term of $k$ in the quadratic polynomial of Shimko method |
| a2 | extracted coefficient term of $k$ squared in the quadratic polynomial of Shimko method |

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
###
### You should see that all methods extract the same density!
###
r = 0.05
te = 60/365
s0 = 1000
sigma = 0.25
y = 0.02
strikes = seq(from = 500, to = 1500, by = 5)
bsm.prices = price.bsm.option(r =r, te = te, s0 = s0,
                k = strikes, sigma = sigma, y = y)
calls = bsm.prices$call
puts = bsm.prices$put
###
```

```
### See where your results will go...
###
getwd()
###
### Running this may take 1-2 minutes...
###
### MOE(market.calls = calls, call.strikes = strikes, market.puts = puts,
### put.strikes = strikes, call.weights = 1, put.weights = 1,
### lambda = 1, s0 = s0, r = r, te = te, y = y, file.name = "myfile")
###
### You may get some warning messages. This happens because the
### automatic initial value selection sometimes picks values
### that produce NaNs in the generalized beta density estimation.
### These messages are often inconsequential.
###
```

oil.2012.10.01

West Texas Intermediate Crude Oil Options on 2013-10-01

## Description

This dataset contains West Texas Intermediate (WTI) crude oil options with 43 days to expiration at the end of the business day October 1, 2012. On October 1, 2012, WTI closed at 92.44.

## Usage

data(oil.2012.10.01)

## Format

A data frame with 332 observations on the following 7 variables.
type a factor with levels $C$ for call option $P$ for put option
strike option strike
settlement option settlement price
openint option open interest
volume trading volume
delta option delta
impliedvolatility option implied volatility

## Source

CME posts sample data at: http://www.cmegroup.com/market-data/datamine-historical-data/endofday.html

## Examples

```
data(oil.2012.10.01)
```

```
pgb CDF of Generalized Beta
```


## Description

pgb is the cumulative distribution function (CDF) of a genaralized beta random variable.

## Usage

pgb(x, a, b, v, w)

## Arguments

X
a power parameter $>0$
b scale paramter $>0$
$v \quad$ first beta paramter $>0$
w second beta parameter $>0$

## Details

Let $B$ be a beta random variable with parameters $v$ and $w$. Then $Z=b *(B /(1-B))^{\wedge}(1 / a)$ is a generalized beta random variable with parameters (a,b,v,w).

## Value

out $\quad$ CDF value at x

## Author(s)

Kam Hamidieh

## References

R.M. Bookstaber and J.B. McDonald (1987) A general distribution for describing security price returns. Journal of Business, 60, 401-424
X. Liu and M.B. Shackleton and S.J. Taylor and X. Xu (2007) Closed-form transformations from risk-neutral to real-world distributions Journal of Business, 60, 401-424
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# What does the cdf of a GB look like?
#
a = 1
b}=1
v}=
w = 2
x = seq(from = 0, to = 500, by = 0.01)
y = pgb(x = x, a = a, b = b, v = v, w = w)
plot(y ~ x, type = "l")
abline(h=c(0,1), lty=2)
```

price.am.option Price American Options on Mixtures of Lognormals

## Description

price.am.option gives the price of a call and a put option at a set strike when the risk neutral density is a mixture of three lognormals.

## Usage

price.am.option(k, r, te, w, u.1, u.2, u.3, sigma.1, sigma.2, sigma.3, p.1, p.2)

## Arguments

k
$r$ risk free rate
te time to expiration
w Weight, a number between 0 and 1, to weigh the option price bounds
u. $1 \quad \log$ mean of the first lognormal
u. $2 \quad \log$ mean of the second lognoral
u. $3 \quad \log$ mean of the second lognoral
sigma. $1 \quad \log$ sd of the first lognormal
sigma. $2 \quad \log$ mean of the second lognormal
sigma. $3 \quad \log$ mean of the third lognormal
p. 1 weight assigned to the first density
p. 2 weight assigned to the second density

## Details

mln is density $\mathrm{f}(\mathrm{x})=\mathrm{p} .1 * \mathrm{f} 1(\mathrm{x})+\mathrm{p} .2 * \mathrm{f} 2(\mathrm{x})+(1-\mathrm{p} .1-\mathrm{p} .2) * \mathrm{f} 3(\mathrm{x})$, where $\mathrm{f} 1, \mathrm{f} 2$, and f 3 are lognormal densities with $\log$ means u.1,u.2, and u. 3 and standard deviations sigma.1, sigma.2, and sigma. 3 respectively.
Note: Different weight values, w, need to be assigned to whether the call or put is in the money or not. See equations (7) \& (8) of Melick and Thomas paper below.

## Value

call.value American call value
put.value American put value
expected.f0 Expected mean value of asset at expiration
prob.f0.gr.k Probability asset values is greater than strike
prob.f0.ls.k Probability asset value is less than strike
expected.f0.f0.gr.k
Expected value of asset given asset exceeds strike
expected.f0.f0.ls.k
Expected value of asset given asset is less than strike

## Author(s)

Kam Hamidieh

## References

Melick, W. R. and Thomas, C. P. (1997). Recovering an asset's implied pdf from option prices: An application to crude oil during the gulf crisis. Journal of Financial and Quantitative Analysis, 32(1), 91-115.

## Examples

```
###
### Set a few parameters and create some
### American options.
###
r = 0.01
te = 60/365
w. 1 = 0.4
w.2 = 0.25
u.1 = 4.2
u.2 = 4.5
u.3 = 4.8
sigma.1 = 0.30
sigma.2 = 0.20
sigma.3 = 0.15
p.1 = 0.25
```

```
p.2 = 0.45
theta = c(w.1,w.2,u.1,u.2,u.3,sigma.1,sigma.2,sigma.3,p.1,p.2)
p.3 = 1 - p.1 - p.2
p.3
expected.f0 = sum(c(p.1, p.2, p.3)* exp(c(u.1,u.2,u.3) +
                                    (c(sigma.1, sigma.2, sigma.3)^2)/2) )
expected.f0
strikes = 30:170
market.calls = numeric(length(strikes))
market.puts = numeric(length(strikes))
for (i in 1:length(strikes))
{
    if ( strikes[i] < expected.f0) {
        market.calls[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.1, u.1 = u.1,
                        u.2 = u.2, u.3 = u.3, sigma.1 = sigma.1, sigma.2 = sigma.2,
                        sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$call.value
        market.puts[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.2, u.1 = u.1,
                        u.2 = u.2, u.3 = u.3, sigma. 1 = sigma.1, sigma. 2 = sigma.2,
                        sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$put.value
    } else {
        market.calls[i] = price.am.option(k = strikes[i], r = r, te = te, w = w.2, u.1 = u.1,
                        u.2 = u.2, u.3 = u.3, sigma.1 = sigma.1, sigma.2 = sigma.2,
                        sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$call.value
        market.puts[i] = price.am.option(k = strikes[i], r=r, te = te, w = w.1, u.1 = u.1,
                        u.2 = u.2, u.3 = u.3, sigma. 1 = sigma.1, sigma. 2 = sigma.2,
                        sigma.3 = sigma.3, p.1 = p.1, p.2 = p.2)$put.value
        }
}
###
### Quickly look at the option values...
###
par(mfrow=c(1,2))
plot(market.calls ~ strikes, type="l")
plot(market.puts ~ strikes, type="l")
par(mfrow=c (1,1))
```


## Description

bsm. option.price computes the BSM European option prices.

## Usage

price.bsm.option(s0, k, r, te, sigma, y)

## Arguments

| s0 | current asset value |
| :--- | :--- |
| k | strike |
| r | risk free rate |
| te | time to expiration |
| sigma | volatility |
| y | dividend yield |

## Details

This function implements the classic Black-Scholes-Merton option pricing model.

## Value

d1
value of $(\log (s 0 / k)+(r-y+(\operatorname{sigma} \wedge 2) / 2) * t e) /(\operatorname{sigma} * \operatorname{sqrt}(t e))$
d2
value of d1 - sigma * sqrt(te)
call call price
put put price

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London
J. Hull (2011) Options, Futures, and Other Derivatives and DerivaGem Package Prentice Hall, Englewood Cliffs, New Jersey, 8th Edition
R. L. McDonald (2013) Derivatives Markets Pearson, Upper Saddle River, New Jersey, 3rd Edition

## Examples

\#
\# call should be 4.76, put should be 0.81, from Hull 8th, page 315, 316
\#
$r=0.10$
te $=0.50$

```
s0 = 42
k = 40
sigma = 0.20
    y = 0
    bsm.option = price.bsm.option(r = r, te = te, s0 = s0, k = k, sigma = sigma, y = y)
    bsm.option
    #
    # Make sure put-call parity holds, Hull 8th, page 351
    #
    (bsm.option$call - bsm.option$put) - (s0 * exp(-y*te) - k * exp(-r*te))
```

price.ew.option Price Options with Edgeworth Approximated Density

## Description

price. ew. option computes the option prices based on Edgeworth approximated densities.

## Usage

price.ew.option(r, te, s0, k, sigma, y, skew, kurt)

## Arguments

| $r$ | risk free rate |
| :--- | :--- |
| te | time to expiration |
| $s 0$ | current asset value |
| k | strike |
| sigma | volatility |
| y | dividend rate |
| skew | normalized skewness |
| kurt | normalized kurtosis |

## Details

Note that this function may produce negative prices if skew and kurt are not well estimated from the data.

Value

| call | Edgeworth based call |
| :--- | :--- |
| put | Edgeworth based put |

## Author(s)

Kam Hamidieh

## References

E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London
R. Jarrow and A. Rudd (1982) Approximate valuation for arbitrary stochastic processes. Journal of Finanical Economics, 10, 347-369
C.J. Corrado and T. Su (1996) S\&P 500 index option tests of Jarrow and Rudd's approximate option valuation formula. Journal of Futures Markets, 6, 611-629

## Examples

```
#
# Here, the prices must match EXACTLY the BSM prices:
#
r = 0.05
y = 0.03
s0 = 1000
sigma = 0.25
te = 100/365
k = seq(from=800, to = 1200, by = 50)
v = sqrt(exp(sigma^2 * te) - 1)
ln.skew = 3 * v + v^3
ln.kurt = 16 * v^2 + 15 * v^4 + 6 * v^6 + v^8
ew.option.prices = price.ew.option(r = r, te = te, s0 = s0, k=k, sigma=sigma,
                                    y=y, skew = ln.skew, kurt = ln.kurt)
bsm.option.prices = price.bsm.option(r = r, te = te, s0 = s0, k=k, sigma=sigma, y=y)
ew.option.prices
bsm.option.prices
###
### Now ew prices should be different as we increase the skewness and kurtosis:
###
new.skew = ln.skew * 1.10
new.kurt = ln.kurt * 1.10
new.ew.option.prices = price.ew.option(r = r, te = te, s0 = s0, k=k, sigma=sigma,
                                    y=y, skew = new.skew, kurt = new.kurt)
new.ew.option.prices
bsm.option.prices
```

price.gb.option Generalized Beta Option Pricing

## Description

price.gb.option computes the price of options.

## Usage

price.gb.option(r, te, s0, k, y, a, b, v, w)

## Arguments

| $r$ | risk free interest rate |
| :--- | :--- |
| te | time to expiration |
| s0 | current asset value |
| k | strike |
| y | dividend yield |
| a | power parameter $>0$ |
| b | scale paramter $>0$ |
| v | first beta paramter $>0$ |
| w | second beta parameter $>0$ |

## Details

This function is used to compute European option prices when the underlying has a generalized beta (GB) distribution. Let B be a beta random variable with parameters v and w . Then $\mathrm{Z}=\mathrm{b}$ $*(B /(1-B))^{\wedge}(1 / a)$ is a generalized beta random variable with parameters with $(a, b, v, w)$.

## Value

| prob. 1 | Probability that a GB random variable with parameters $(\mathrm{a}, \mathrm{b}, \mathrm{v}+1 / \mathrm{a}, \mathrm{w}-1 / \mathrm{a})$ will be <br> above the strike |
| :--- | :--- |
| prob. 2 | Probability that a GB random variable with parameters $(\mathrm{a}, \mathrm{b}, \mathrm{v}, \mathrm{w})$ will be above <br> the strike |
| call | call price |
| put | put price |

## Author(s)

Kam Hamidieh

## References

R.M. Bookstaber and J.B. McDonald (1987) A general distribution for describing security price returns. Journal of Business, 60, 401-424
X. Liu and M.B. Shackleton and S.J. Taylor and X. Xu (2007) Closed-form transformations from risk-neutral to real-world distributions Journal of Business, 60, 401-424
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# A basic GB option pricing....
#
r=0.03
te = 50/365
s0 = 1000.086
k = seq(from = 800, to = 1200, by = 10)
y = 0.01
a = 10
b}=100
v = 2.85
w}=2.8
```

price.gb.option( $r=r$, te $=t e, ~ s 0=s 0, k=k, y=y, a=a, b=b, v=v, w=w)$
price.mln.option Price Options on Mixture of Lognormals

## Description

mln.option.price gives the price of a call and a put option at a strike when the risk neutral density is a mixture of two lognormals.

## Usage

price.mln.option(r, te, y, k, alpha.1, meanlog.1, meanlog.2, sdlog.1, sdlog.2)

## Arguments

| $r$ | risk free rate |
| :--- | :--- |
| te | time to expiration |
| $y$ | dividend yield |
| $k$ | strike |


| alpha. 1 | proportion of the first lognormal. Second one is 1-alpha. 1 |
| :--- | :--- |
| meanlog. 1 | mean of the log of the first lognormal |
| meanlog. 2 | mean of the log of the second lognormal |
| sdlog. 1 | standard deviation of the log of the first lognormal |
| sdlog. 2 | standard deviation of the log of the second lognormal |

## Details

mln is the density $\mathrm{f}(\mathrm{x})=$ alpha. $1 * \mathrm{~g}(\mathrm{x})+(1-\operatorname{alpha} .1) * \mathrm{~h}(\mathrm{x})$, where g and h are densities of two lognormals with parameters (mean.log.1, sdlog.1) and (mean.log.2, sdlog.2) respectively.

## Value

call call price
put put price
s0 current value of the asset as implied by the mixture distribution

## Author(s)

Kam Hamidieh

## References

F. Gianluca and A. Roncoroni (2008) Implementing Models in Quantitative Finance: Methods and Cases
B. Bahra (1996): Probability distribution of future asset prices implied by option prices. Bank of England Quarterly Bulletin, August 1996, 299-311
P. Soderlind and L.E.O. Svensson (1997) New techniques to extract market expectations from financial instruments. Journal of Monetary Economics, 40, 383-429
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

```
#
# Try out a range of options
#
r = 0.05
te = 60/365
k = 700:1300
y = 0.02
meanlog.1 = 6.80
meanlog.2 = 6.95
sdlog.1 = 0.065
sdlog.2 = 0.055
alpha.1 = 0.4
```

```
mln.prices = price.mln.option(r = r, y = y, te = te, k = k, alpha.1 = alpha.1,
    meanlog.1 = meanlog.1, meanlog.2 = meanlog.2, sdlog.1 = sdlog.1, sdlog.2 = sdlog.2)
par(mfrow=c(1,2))
plot(mln.prices$call ~ k)
plot(mln.prices$put ~ k)
par(mfrow=c(1,1))
```

```
price.shimko.option
Price Option based on Shimko's Method
```


## Description

price.shimko. option prices a European option based on the extracted Shimko volatility function.

## Usage

price.shimko.option(r, te, s0, k, y, a0, a1, a2)

## Arguments

| $r$ | risk free rate |
| :--- | :--- |
| te | time to expiration |
| $s 0$ | current asset value |
| $k$ | strike |
| $y$ | dividend yield |
| a0 | constant term in the quadratic polyynomial |
| a1 | coefficient term of $k$ in the quadratic polynomial |
| a2 | coefficient term of $k$ squared in the quadratic polynomial |

## Details

This function may produce negative option values when nonsensical values are used for a 0 , a 1 , and a2.

## Value

| call | call price |
| :--- | :--- |
| put | put price |

Author(s)
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## References

D. Shimko (1993) Bounds of probability. Risk, 6, 33-47
E. Jondeau and S. Poon and M. Rockinger (2007): Financial Modeling Under Non-Gaussian Distributions Springer-Verlag, London

## Examples

$r=0.05$
$\mathrm{y}=0.02$
te $=60 / 365$
$\mathrm{s} 0=1000$
$\mathrm{k}=950$
sigma $=0.25$
a0 $=0.30$
a1 $=-0.00387$
a2 $=0.00000445$
\#
\# Note how Shimko price is the same when a0 = sigma, a1=a2=0 but substantially
\# more when a0, a1, a2 are changed so the implied volatilies are very high!
\#
price.bsm.option(r = r, te $=$ te, $s 0=s 0, k=k$, sigma $=$ sigma, $y=y) \$ c a l l$
price.shimko.option( $r=r$, te $=t e, s 0=s 0, k=k, y=y$,
$a 0=$ sigma, $a 1=0, a 2=0) \$ c a l l$
price.shimko.option(r = r, te = te, s0 = s0, k = k, y = y,
$a 0=a 0, a 1=a 1, a 2=a 2) \$ c a l l$
sp500.2013.04.19 S\&P 500 Index Options on 2013-04-19

## Description

This dataset contains S\&P 500 options with 62 days to expiration at the end of the business day April 19, 2013. On April 19, 2013, S\&P 500 closed at 1555.25.

## Usage

data(sp500.2013.04.19)

## Format

A data frame with 171 observations on the following 19 variables.
bidsize.c call bid size
bid.c call bid price
ask.c call ask price
asksize.c call ask size
chg.c change in call price
impvol.c call implied volatility
vol.c call volume
openint.c call open interest
delta.c call delta
strike option strike
bidsize.p put bid size
bid.p put bid price
ask.p put ask price
asksize.p put ask size
chg.p change in put price
impvol.p put implied volatility
vol.p put volume
openint.p put open interest
delta.p put delta

## Source

http://www.cboe.com/DelayedQuote/QuoteTableDownload.aspx

## Examples

data(sp500.2013.04.19)
sp500.2013.06.24 S\&P 500 Index Options on 2013-06-24

## Description

This dataset contains S\&P 500 options with 53 days to expiration at the end of the business day June 24, 2013. On June 24, 2013, S\&P 500 closed at 1573.09.

## Usage

data(sp500.2013.06.24)

## Format

A data frame with 173 observations on the following 9 variables.
bid.c call bid price
ask.c call ask price
vol.c call volume
openint.c call open interest
strike option strike
bid.p put bid price
ask.p put ask price
vol.p put volume
openint.p put open interest

## Source

http://www.cboe.com/DelayedQuote/QuoteTableDownload.aspx

## Examples

data(sp500.2013.06.24)
vix.2013.06.25 VIX Options on 2013-06-25

## Description

This dataset contains VIX options with 57 days to expiration at the end of the business day June 25, 2013. On June 25, 2013, VIX closed at 18.21.

## Usage

data(vix.2013.06.25)

## Format

A data frame with 35 observations on the following 13 variables.
last.c closing call price
change.c change in call price from previous day
bid.c call bid price
ask.c call ask price
vol.c call volume
openint.c call open interest
strike option strike
last.p closing put price
change.p change in put price from previous day
bid.p put bid price
ask.p put ask price
vol.p put volume
openint.p put open interest

## Source

http://www.cboe.com/DelayedQuote/QuoteTableDownload.aspx

## Examples

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