

# Package ‘Davies’

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**Title** The Davies Quantile Function

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 Davies

*The Davies distribution*


---

**Description**

Density, distribution function, quantile function and random generation for the Davies distribution.

**Usage**

```

ddavies(x, params, log=FALSE)
pdavies(x, params, log.p=FALSE, lower.tail=TRUE)
qdavies(p, params, lower.tail=TRUE)
rdavies(n, params)
ddavies.p(x, params, log=FALSE)

```

**Arguments**

x	quantile
p	vector of probabilities
n	number of observations. If <code>length(n) &gt; 1</code> , the length is taken to be the number required
<code>lower.tail</code>	logical; if TRUE (default), probabilities are $P(X \leq x)$ , otherwise $P(X > x)$
<code>log, log.p</code>	logical; if TRUE, probabilities are given as $\log(p)$
<code>params</code>	A three-member vector holding $C$ , $\lambda_1$ and $\lambda_2$

**Details**

The Davies distribution is defined in terms of its quantile function:

$$Cp^{\lambda_1}/(1-p)^{\lambda_2}$$

It does not have a closed-form probability density function or cumulative density function, so numerical solution is used.

Function `ddavies.p()` returns the density of the Davies function but as a function of the quantile.

**Value**

Function `ddavies()` gives the density, `pdavies()` gives the distribution function, `qdavies()` gives the quantile function, and `rdavies()` generates random deviates.

**Author(s)**

Robin K. S. Hankin

**References**

R. K. S. Hankin and A. Lee 2006. "A new family of non-negative distributions" *Australia and New Zealand Journal of Statistics*, 48(1):67–78

**See Also**

[Gld](#), [fit.davies.p](#), [least.squares](#), [skewness](#)

**Examples**

```
params <- c(10,0.1,0.1)
x <- seq(from=4,to=20,by=0.2)
p <- seq(from=1e-3,to=1-1e-3,len=50)

rdavies(n=5,params)
least.squares(rdavies(100,params))
plot(pdavies(x,params))

plot(p,qdavies(p,params))
plot(x,ddavies(x,params),type="b")
```

---

 davies.moment

*Moments of the Davies distribution*


---

**Description**

Moments of order statistics of random variables drawn from a Davies distribution

**Usage**

```
davies.moment(n=1 , i=1 , order=1 , params)
M(order,params)
mu(params)
expected.value(n,i,params)
expected.value.approx(n,i,params)
variance(params)
skewness(params)
kurtosis(params)
```

**Arguments**

params	A three-member vector holding $C$ , $\lambda_1$ and $\lambda_2$
n	The number of observations
i	Return information about the $i$ -th order statistic (ie $i = 1$ means the smallest, $i = n$ means the biggest)
order	The order (eg order=2 gives the square)

## Details

Function `davies.moment(n,i,order=r)` gives the  $r$ -th moment of the  $i$ -th order statistic of  $n$  observations. The following aliases are just convenience wrappers with  $n = i = 1$  (ie moments of one observation from a Davies distribution):

- `M()` gives the  $r$ -th moment for  $n = i = 1$
- `mu()` gives the first moment of a Davies distribution (ie the mean)
- `variance()` gives the second *central* moment of a Davies distribution
- `skewness()` gives the normalized skewness of a Davies distribution
- `kurtosis()` gives the normalized kurtosis of a Davies distribution

## Author(s)

Robin K. S. Hankin

## See Also

[expected.value](#), [expected.gld](#)

## Examples

```
params <- c(10,0.1,0.1)
davies.moment(n=100,i=99,2,params) # ie the second moment of the 99th smallest
                                   # observation of 100 drawn from a Davies
                                   # distribution with parameters p

mean(rdavies(1e6,params))-mu(params)

#now reproduce the S-K graph:

f <- function(x,y){c(skewness(c(1,x,y)),kurtosis(c(1,x,y)))}
g <- function(j,vector,pp,qq=1){points(t(sapply(vector,f,y=j)),type="l",col="black",lty=qq)}

vector <- c((0:300)/100 , (0:300)/10000 , seq(from=3,to=10,len=100))
vector <- sort(unique(vector))

plot(t(sapply((0:10)/10,f,y=0)),
     xlim=c(-3,3),ylim=c(0,10),
     type="n",xlab="skewness",ylab="kurtosis")
g(0.001,vector,"red",qq=1)
g(0.01,vector,"yellow",qq=2)
g(0.02,vector,"green",qq=3)
g(0.05,vector,"blue",qq=4)
g(0.1 ,vector,"purple",qq=5)
g(0.14,vector,"black",qq=6)

x <- seq(from=-3,to=3,len=30)
points(x,x^2+1,type="l",lwd=2)
```

```
leg.txt <- expression(lambda[2]==0.001,
                      lambda[2]==0.01,lambda[2]==0.02,lambda[2]==0.05,
                      lambda[2]==0.1,lambda[2]==0.14)
legend(-1.1,10,leg.txt,col="black",lty=1:6)
```

---

davies.start	<i>start value for Davies minimization routines</i>
--------------	---

---

### Description

Gives a “start” value for the optimization routines. Uses heuristics that seem to work.

### Usage

```
davies.start(x, threeps=c(0.1,0.5,0.9), small = 0.01)
```

### Arguments

x	dataset to be used
threeps	a three-element vector representing the quantiles to be balanced. The default values balance the first and ninth deciles and the median. These seem to work for me pretty well; YMMV
small	a “small” value to be used for $\lambda_1$ and $\lambda_2$ because using exactly zero is inappropriate

### Details

Returns a “start” value of the parameters for use in one of the Davies fitting routines `maximum.likelihood()` or `least.squares()`.

Uses three heuristic methods (one assuming  $\lambda_1 = \lambda_2$ , one with  $\lambda_1 = 0$ , and one with  $\lambda_2 = 0$ ). Returns the best one of the three, as measured by `objective()`.

### Author(s)

Robin K. S. Hankin

### See Also

[least.squares](#), [maximum.likelihood](#), [objective](#)

**Examples**

```

d <- rchisq(40,1)
davies.start(d)
least.squares(d)

params <- c(10 , 0.1 , -0.1)
x <- rdavies(100 , params)
davies.start(x)

f <- function(threeps){objective(davies.start(x, threeps),x)}

(jj<-optim(c(0.1,0.5,0.9),f))
davies.start(x,jj$par)
least.squares(x)

#not bad at all.

```

---

expected.gld

*expected value of the Generalized Lambda Distribution*

---

**Description**

Returns the expected value of the Generalized Lambda Distribution

**Usage**

```

expected.gld(n=1, i=1, params)
expected.gld.approx(n=1, i=1, params)

```

**Arguments**

n	Number of observations
i	Order statistic: $i = 1$ means the smallest of $n$ , and $n = i$ means the largest
params	The four parameters of a GLD distribution

**Details**

expected.gld and expected.approx return the exact and approximate values of the expected value of a Generalized Lambda Distribution RV.

Exploits the fact that the gld quantile function is the sum of a constant and two davies quantile functions

**Author(s)**

Robin K. S. Hankin

## References

A. Ozturk and R. F. Dale, “Least squares estimation of the parameters of the generalized lambda distribution”, *Technometrics* 1985, 27(1):84 [it does not appear to be possible, as of R-2.9.1, to render the diacritic marks in the first author’s names in a nicely portable way]

## See Also

[Gld](#), [expected.value](#)

## Examples

```
params <- c(4.114,0.1333,0.0193,0.1588)
mean(rgld(1000,params))
expected.gld(n=1,i=1,params)
expected.gld.approx(n=1,i=1,params)

f <- function(n){apply(matrix(rgld(n+n,params),2,n),2,min)}
#ie f(n) gives the smaller of 2 rgld RVs, n times.

mean(f(1000))
expected.gld(n=2,i=1,params)
expected.gld.approx(n=2,i=1,params)

plot(1:100,expected.gld.approx(n=100,i=1:100,params)-expected.gld(n=100,i=1:100,params))
# not bad, eh? ...yyeeeeesss, but the parameters given by Ozturk give
# an almost zero second derivative for d(qgld)/dp, so the good agreement
# isn't surprising really. Observe that the error is minimized at about
# p=0.2, where the point of inflection is.
```

---

fit.davies.p

*Fits and plots Davies distributions to datasets*

---

## Description

A convenience wrapper (and pretty-printer) for `maximum.likelihood()` and `least.squares()`. Given a dataset, it draws an empirical quantile function (`fit.davies.p()`) or PDF (`fit.davies.q()`) and superimposes the dataset.

## Usage

```
fit.davies.p(x , print.fit=FALSE, use.q=TRUE , params=NULL, small=1e-5 , ...)
fit.davies.q(x , print.fit=FALSE, use.q=TRUE , params=NULL, ...)
```

**Arguments**

<code>x</code>	dataset to be fitted and plotted
<code>print.fit</code>	Boolean with TRUE meaning print details of the fit
<code>use.q</code>	Boolean with TRUE meaning use <code>least.squares()</code> (rather than <code>maximum.likelihood()</code> )
<code>params</code>	three-element vector holding the three parameters of the davies dataset. If NULL, determine the parameters using the method indicated by <code>use.q</code>
<code>small</code>	small positive number showing range of quantiles to plot
<code>...</code>	Additional parameters passed to <code>plot()</code>

**Value**

If `print.fit` is TRUE, return the optimal parameters

**Author(s)**

Robin K. S. Hankin

**See Also**

[least.squares](#), [maximum.likelihood](#)

**Examples**

```
fit.davies.q(rchisq(100,1))
fit.davies.p(exp(rnorm(100)))

data(x00m700p4)
fit.davies.q(x00m700p4)
```

---

Gld

*The Generalized Lambda Distribution*

---

**Description**

Density, distribution function, quantile function and random generation for the Generalized Lambda Distribution

**Usage**

```
dgld(x, params)
dgld.p(x, params)
pgld(q, params)
qgld(p, params)
rgld(n, params)
```

**Arguments**

x, q	vector of quantiles
p	vector of probabilities
n	In function <code>rgld()</code> , the number of observations. If <code>length(n) &gt; 1</code> , the length is taken to be the number required
params	vector of parameters: <code>params[1]==lambda1</code> et seq

**Details**

The Generalized Lambda distribution has quantile function

$$f(x) = \lambda_1 + (p^{\lambda_3} - (1-p)^{\lambda_4})/\lambda_2$$

**Value**

Function `dgld()` gives the density, `dgld.p()` gives the density in terms of the quantile, `pgld()` gives the distribution function, `qgld()` gives the quantile function, and `rgld()` generates random deviates.

**References**

- M. J. Wichura 1988. "Algorithm AS 241: The Percentage Points of the Normal Distribution". *Applied Statistics*, **37**, 477–484.
- A. Ozturk and R. F. Dale 1985. "Least squares estimation of the parameters of the generalized lambda distribution". *Technometrics* 27(1):84

**See Also**

[Davies, expected.gld](#)

**Examples**

```
params <- c(4.114,0.1333,0.0193,0.1588) #taken straight from some paper
gld.rv <- rgld(100,params)

hist(gld.rv)
fit.davies.q(gld.rv) #remember the Davies distn has 3 DF and the GLD 4...
```

---

least.squares	<i>Finds the optimal Davies distribution for a dataset</i>
---------------	--

---

### Description

Finds the best-fit Davies distribution using either the least-squares criterion (`least.squares()`) or maximum likelihood (`maximum.likelihood()`)

### Usage

```
least.squares(data, do.print = FALSE, start.v = NULL)
maximum.likelihood(data, do.print = FALSE, start.v = NULL)
```

### Arguments

<code>data</code>	dataset to be fitted
<code>do.print</code>	Boolean with TRUE meaning print a GFM
<code>start.v</code>	A suitable starting vector of parameters $c(C, \lambda_1, \lambda_2)$ , with default NULL meaning to use <code>start()</code>

### Details

Uses `optim()` to find the best-fit Davies distribution to a set of data.

**Function `least.squares()` does not match that of Hankin and Lee 2006.**

### Value

Returns the parameters  $C, \lambda_1, \lambda_2$  of the best-fit Davies distribution to the dataset `data`

### Note

BUGS:

Function `least.squares()` does not use the same methodology of Hankin and Lee 2006, and its use is discouraged pending implementation.

Quite apart from that, it can be screwed with bad value for `start.v`. Function `maximum.likelihood()` can be very slow. It might be possible to improve this by using some sort of hot-start for `optim()`.

### Author(s)

Robin K. S. Hankin

### See Also

[davies.start](#), [optim](#), [objective](#), [likelihood](#)

**Examples**

```
p <- c(10 , 0.1 , 0.1)
d <- rdavies(10,p)

maximum.likelihood(d) # quite slow
least.squares(d)      # much faster but not recommended
```

---

likelihood	<i>likelihood for the Davies distribution</i>
------------	---

---

**Description**

Likelihood of observing data, on the hypothesis of their coming from a Davies distribution of parameters `params`.

Function `neg.log.likelihood()` gives minus the loglikelihood

**Usage**

```
likelihood(params, data)
```

**Arguments**

<code>params</code>	Parameters of the Davies distribution
<code>data</code>	dataset for which the likelihood is computed

**Author(s)**

Robin K. S. Hankin

**See Also**

[Davies](#)

**Examples**

```
p1 <- c(10, 0.1, 0.1)
p2 <- c(10, 0.4, 0.1)
d <- rdavies(100,p1)
likelihood(p1,d)
likelihood(p2,d) #should be smaller.
neg.log.likelihood(p1,rstupid(100)) #should be large negative.
```

---

`objective`*The objective function for fitting the Davies distribution*

---

**Description**

The “distance” of a dataset from a particular Davies distribution

**Usage**

```
objective(params, dataset)
objective.approx(params, dataset)
```

**Arguments**

<code>params</code>	A three-member vector holding $C$ , $\lambda_1$ and $\lambda_2$
<code>dataset</code>	The dataset to be considered

**Details**

Used by the `fit.davies.p()` and `fit.davies.q()` functions

**Value**

`objective` returns the “distance” of a dataset from a particular Davies distribution as measured by the sums of the squares of the differences between observed (`dataset`) and expected (`expected.value()`) values.

`objective.approx()` uses `expected.approx()` rather than `expected()` to calculate expectations, as per equation 6.

**Author(s)**

Robin K. S. Hankin

**See Also**

[fit.davies.p](#), [fit.davies.q](#)

**Examples**

```
params <- c(10, 0.1, 0.1)
x <- rdavies(100, params)
objective(params, x)
objective.approx(params, x)

objective(least.squares(x), x)
objective(davies.start(x), x)
```

---

ozturk

*Parameters used in a paper by Ozturk*

---

### Description

A four-element vector giving the parameters used by Ozturk.

### Usage

```
data(x00m700p4)
```

### References

A. Ozturk and R. F. Dale 1985. "Least squares estimation of the parameters of the generalized lambda distribution". *Technometrics* 27(1):84; see discussion under `expected.gld.Rd`.

### See Also

[expected.gld](#)

### Examples

```
data(ozturk)
hist(rgld(100,ozturk))
```

---

plotcf

*p-value investigation*

---

### Description

Plots sorted p-values showing which ones would have been rejected

### Usage

```
plotcf(y, q=0.05)
```

### Arguments

y	dataset
q	p-value of critical region

### Details

Sorts p-values and plots the order statistic. Useful for investigating a statistical test by using it when the null hypothesis is KNOWN to be true, just to check if the probability of rejection really is alpha.

Also can be used when  $H_0$  is wrong, showing what beta is.

**Author(s)**

Robin K. S. Hankin

**Examples**

```
f.H0.T <- function(n,free=5){t.test(rt(n,df=free))$p.value}
f.H0.F <- function(n,free=5){t.test(rf(n,df1=free,df2=free))$p.value}

plotcf(sapply(rep(10,100),f.H0.T)) # should reject about 5: thus
# probability of a type I error is
# about 0.05 (as it should be; this
# is an exact test)
plotcf(sapply(rep(10,100),f.H0.F)) # should reject about 80: thus
# probability of a type II error is
# about 0.2 for this H_A.
```

---

 rstupid

*A stupid PDF*


---

**Description**

a contrived PDF that cannot be closely approximated by a Davies distribution

**Usage**

```
rstupid(n, a = 1, b = 2, c = 3, d = 4)
```

**Arguments**

n	Number of observations
a	start of first uniform bit
b	end of first uniform bit
c	start of second uniform bit
d	end of second uniform bit

**Details**

The stupid distribution is composed of two separate uniform distributions: one from  $a$  to  $b$ , and one from  $c$  to  $d$ . It is specifically designed to be NOT fittable to any Davies distribution.

You could probably come up with a more stupid distribution if you tried.

**Author(s)**

Robin K. S. Hankin

**See Also**[Davies](#)**Examples**

```
stupid <- rstupid(500)
fit.davies.q(stupid)
```

---

twolines.vert	<i>Order statistic comparison</i>
---------------	-----------------------------------

---

**Description**

Plots two lines and shades the bit in between them

**Usage**

```
twolines.vert(p, y1, y2, ...)
```

**Arguments**

p	vector of quantiles
y1	First set of ordinates
y2	Second set of ordinates
...	Extra arguments, passed to <code>segments()</code> , for the vertical lines

**Details**

Plots p against y1, and p against y2, and shades the bit in between using vertical lines. This is useful for comparing two order statistics

**Author(s)**

Robin K. S. Hankin

**See Also**[Davies, qqplot](#)**Examples**

```
twolines.vert(1:100, sort(rnorm(100)), sort(rnorm(100)))
params <- c(10, 0.1, 0.1)
twolines.vert(1:100, sort(rdavies(100, params)), sort(rdavies(100, params)))
```

---

`x00m700p4`*Peak concentration for 100 instantaneous releases*

---

**Description**

This data set gives the peak concentration for 100 independent instantaneous releases of neutral-buoyancy gas in a windtunnel

**Usage**

```
data(x00m700p4)
```

**Format**

A vector containing 100 observations

**References**

D. J. Hall and others 1991. *Repeat variability in instantaneously released heavy gas clouds—some wind tunnel model experiments*. Technical Report LR 804 (PA), Warren Spring Laboratory, Gunnels Wood Road, Stevenage, Hertfordshire SG1 2BX.

**Examples**

```
data(x00m700p4)  
fit.davies.q(x00m700p4)
```

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