

# Package ‘GTDL’

October 12, 2022

**Type** Package

**Title** The Generalized Time-Dependent Logistic Family

**Version** 1.0.0

**Date** 2022-03-25

**Author** Jalmar Carrasco [aut, cre],  
Luciano Santana [aut],  
Lizandra Fabio [aut]

**Maintainer** Jalmar Carrasco <carrascojalmar@gmail.com>

**Description** Computes the probability density, survival function,  
the hazard rate functions and generates random samples from the  
GTDL distribution given by Mackenzie, G. (1996) <[doi:10.2307/2348408](https://doi.org/10.2307/2348408)>.  
The likelihood estimates, the randomized quantile (Louzada, F., et al.  
(2020) <[doi:10.1109/ACCESS.2020.3040525](https://doi.org/10.1109/ACCESS.2020.3040525)>)  
residuals and the normally transformed randomized survival  
probability (Li,L., et al. (2021) <[doi:10.1002/sim.8852](https://doi.org/10.1002/sim.8852)>)  
residuals are obtained for the GTDL model.

**License** GPL (>= 3)

**Encoding** UTF-8

**LazyData** TRUE

**RoxygenNote** 7.1.1

**Imports** survival,

**Suggests** stats,

**Depends** R (>= 2.10)

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2022-03-28 07:50:12 UTC

## R topics documented:

artset1987	.....	2
------------	-------	---

fGTDL	3
mle1.GTDL	4
mle2.GTDL	5
nrsp.GTDL	7
random.quantile.GTDL	8
tumor	10

**Index****11****artset1987***Artset1987 data***Description**

Times to failure of 50 devices put on life test at time 0.

**Usage**

```
data(artset1987)
```

**Format**

This data frame contains the following columns:

- t: Times to failure

**References**

- Aarset, M. V. (1987). How to Identify a Bathtub Hazard Rate. *IEEE Transactions on Reliability*, 36, 106–108.

**Examples**

```
data(artset1987)
head(artset1987)
```

**Description**

Density function, survival function, failure function and random generation for the GTDL distribution.

**Usage**

```
dGTDL(t, param, log = FALSE)

hGTDL(t, param)

sGTDL(t, param)

rGTDL(n, param)
```

**Arguments**

t	vector of integer positive quantile.
param	parameters (alpha and gamma are scalars, lambda non-negative).
log	logical; if TRUE, probabilities p are given as log(p).
n	number of observations.

**Details**

- Density function

$$f(t | \boldsymbol{\theta}) = \lambda \left( \frac{\exp\{\alpha t + \mathbf{X}^\top \boldsymbol{\beta}\}}{1 + \exp\{\alpha t + \mathbf{X}^\top \boldsymbol{\beta}\}} \right) \times \left( \frac{1 + \exp\{\alpha t + \mathbf{X}^\top \boldsymbol{\beta}\}}{1 + \exp\{\mathbf{X}^\top \boldsymbol{\beta}\}} \right)^{-\lambda/\alpha}$$

- Survival function

$$S(t | \boldsymbol{\theta}) = \left( \frac{1 + \exp\{\alpha t + \mathbf{X}^\top \boldsymbol{\beta}\}}{1 + \exp\{\mathbf{X}^\top \boldsymbol{\beta}\}} \right)^{-\lambda/\alpha}$$

- Failure function

$$h(t | \boldsymbol{\theta}) = \lambda \left( \frac{\exp\{\alpha t + \mathbf{X}^\top \boldsymbol{\beta}\}}{1 + \exp\{\alpha t + \mathbf{X}^\top \boldsymbol{\beta}\}} \right)$$

**Value**

dGTDL gives the density function, hGTDL gives the failure function, sGTDL gives the survival function and rGTDL generates random samples.

Invalid arguments will return an error message.

## Source

[d-p-q-r]GTDL are calculated directly from the definitions.

## References

- Mackenzie, G. (1996). Regression Models for Survival Data: The Generalized Time-Dependent Logistic Family. *Journal of the Royal Statistical Society. Series D (The Statistician)*. 45. 21-34.

## Examples

```
library(GTDL)
t <- seq(0,20,by = 0.1)
lambda <- 1.00
alpha <- -0.05
gamma <- -1.00
param <- c(lambda,alpha,gamma)
y1 <- hGTDL(t,param)
y2 <- sGTDL(t,param)
y3 <- dGTDL(t,param,log = FALSE)
tt <- as.matrix(cbind(t,t,t))
yy <- as.matrix(cbind(y1,y2,y3))
matplot(tt,yy,type="l",xlab="time",ylab="",lty = 1:3,col=1:3,lwd=2)

y1 <- hGTDL(t,c(1,0.5,-1.0))
y2 <- hGTDL(t,c(1,0.25,-1.0))
y3 <- hGTDL(t,c(1,-0.25,1.0))
y4 <- hGTDL(t,c(1,-0.50,1.0))
y5 <- hGTDL(t,c(1,-0.06,-1.6))
tt <- as.matrix(cbind(t,t,t,t,t))
yy <- as.matrix(cbind(y1,y2,y3,y4,y5))
matplot(tt,yy,type="l",xlab="time",ylab="Hazard function",lty = 1:3,col=1:3,lwd=2)
```

## Description

Estimate of the parameters.

## Usage

```
mle1.GTDL(start, t, method = "BFGS")
```

### Arguments

- start            Initial values for the parameters to be optimized over.
- t                non-negative random variable representing the failure time and leave the snapshot failure rate, or danger.
- method          The method to be used.

### Value

Returns a list of summary statistics of the fitted GTDL distribution.

### References

- Aarset, M. V. (1987). How to Identify a Bathtub Hazard Rate. IEEE Transactions on Reliability, 36, 106–108.
- Mackenzie, G. (1996) Regression Models for Survival Data: The Generalized Time-Dependent Logistic Family. Journal of the Royal Statistical Society. Series D (The Statistician). 45. 21-34.

### See Also

[optim](#)

### Examples

```
# times data (from Aarset, 1987)
data(artset1987)
mod <- mle1.GTDL(c(1,-0.05,-1),t = artset1987)
```

mle2.GTDL

*Maximum likelihood estimates of the GTDL model*

### Description

Maximum likelihood estimates of the GTDL model

### Usage

```
mle2.GTDL(t, start, formula, censur, method = "BFGS")
```

## Arguments

<code>t</code>	non-negative random variable representing the failure time and leave the snapshot failure rate, or danger.
<code>start</code>	Initial values for the parameters to be optimized over.
<code>formula</code>	The structure matrix of covariates of dimension n x p.
<code>censur</code>	censoring status 0=censored, a=fail.
<code>method</code>	The method to be used.

## Value

Returns a list of summary statistics of the fitted GTDL model.

## References

- Mackenzie, G. (1996) Regression Models for Survival Data: The Generalized Time-Dependent Logistic Family. Journal of the Royal Statistical Society. Series D (The Statistician). (45). 21-34.

## See Also

[optim](#)

## Examples

```
### Example 1

require(survival)
data(lung)

lung <- lung[-14,]
lung$sex <- ifelse(lung$sex==2, 1, 0)
lung$ph.ecog[lung$ph.ecog==3]<-2
t1 <- lung$time
start1 <- c(0.03,0.05,-1,0.7,2,-0.1)
formula1 <- ~lung$sex+factor(lung$ph.ecog)+lung$age
censur1 <- ifelse(lung$status==1,0,1)
fit.model1 <- mle2.GTDL(t = t1,start = start1,
                         formula = formula1,
                         censur = censur1)
fit.model1

### Example 2

data(tumor)
t2 <- tumor$time
start2 <- c(1,-0.05,1.7)
formula2 <- ~tumor$group
censur2 <- tumor$censored
fit.model2 <- mle2.GTDL(t = t2,start = start2,
```

```

    formula = formula2,
    censur = censur2)
fit.model2

```

**nrsp.GTDL**

*Normally-transformed randomized survival probability residuals for the GTDL model*

## Description

Normally-transformed randomized survival probability residuals for the GTDL model

## Usage

```
nrsp.GTDL(t, formula, pHat, censur)
```

## Arguments

- |         |  |
|---------|--|
| t       | non-negative random variable representing the failure time and leave the snapshot failure rate, or danger. |
| formula | The structure matrix of covariates of dimension n x p.   |
| pHat    | Estimate of the parameters from the GTDL model.  |
| censur  | Censoring status 0=censored, a=fail.   |

## Value

Normally-transformed randomized survival probability residuals

## References

- Li, L., Wu, T., e Cindy, F. (2021). Model diagnostics for censored regression via randomized survival probabilities. *Statistics in Medicine*, 40, 1482–1497.
- de Oliveira, L. E. F., dos Santos L. S., da Silva, P. H. F., Fabio, L. C., Carrasco, J. M. F.(2022). Análise de resíduos para o modelo logístico generalizado dependente do tempo (GTDL). Submitted.

## Examples

```

### Example 1

require(survival)
data(lung)
lung <- lung[-14,]
lung$sex <- ifelse(lung$sex==2, 1, 0)
lung$ph.ecog[lung$ph.ecog==3]<-2
t1 <- lung$time

```

```

formula1 <- ~lung$sex+factor(lung$ph.ecog)+lung$age
censur1 <- ifelse(lung$status==1,0,1)
start1 <- c(0.03,0.05,-1,0.7,2,-0.1)
fit.model1 <- mle2.GTDL(t = t1,start = start1,
                         formula = formula1,
                         censur = censur1)
r1 <- nrsp.GTDL(t = t1,formula = formula1 ,pHat = fit.model1$Coefficients[,1],
                  censur = censur1)
r1

### Example 2

data(tumor)
t2 <- tumor$time
formula2 <- ~tumor$group
censur2 <- tumor$censored
start2 <- c(1,-0.05,1.7)
fit.model2 <- mle2.GTDL(t = t2,start = start2,
                         formula = formula2,
                         censur = censur2)
r2 <- nrsp.GTDL(t = t2,formula = formula2, pHat = fit.model2$Coefficients[,1],
                  censur = censur2)
r2

```

**random.quantile.GTDL** *Randomized quantile residuals for the GTDL model*

## Description

Randomized quantile residuals for the GTDL model

## Usage

```
random.quantile.GTDL(t, formula, pHat, censur)
```

## Arguments

- |                |  |
|----------------|--|
| <b>t</b>       | non-negative random variable representing the failure time and leave the snapshot failure rate, or danger. |
| <b>formula</b> | The structure matrix of covariates of dimension n x p.   |
| <b>pHat</b>    | Estimate of the parameters from the GTDL model.  |
| <b>censur</b>  | censoring status 0=censored, a=fail.   |

## Details

The randomized quantile residual (Dunn and Smyth, 1996), which follow a standard normal distribution is used to assess departures from the GTDL model.

**Value**

Randomized quantile residuals

**References**

- Dunn, P. K. e Smyth, G. K. (1996). Randomized quantile residuals. *Journal of Computational and Graphical Statistics*, 5, 236–244.
- Louzada, F., Cuminato, J. A., Rodriguez, O. M. H., Tomazella, V. L. D., Milani, E. A., Ferreira, P. H., Ramos, P. L., Bochio, G., Perissini, I. C., Junior, O. A. G., Mota, A. L., Alegría, L. F. A., Colombo, D., Oliveira, P. G. O., Santos, H. F. L., e Magalhães, M. V. C. (2020). Incorporation of frailties into a non-proportional hazard regression model and its diagnostics for reliability modeling of downhole safety valves. *IEEE Access*, 8, 219757 – 219774.
- de Oliveira, L. E. F., dos Santos L. S., da Silva, P. H. F., Fabio, L. C., Carrasco, J. M. F. (2022). Análise de resíduos para o modelo logístico generalizado dependente do tempo (GTDL). Submitted.

**Examples**

```
### Example 1

require(survival)
data(lung)
lung <- lung[-14,]
lung$sex <- ifelse(lung$sex==2, 1, 0)
lung$ph.ecog[lung$ph.ecog==3]<-2
t1 <- lung$time
formula1 <- ~lung$sex+factor(lung$ph.ecog)+lung$age
censur1 <- ifelse(lung$status==1,0,1)
start1 <- c(0.03,0.05,-1,0.7,2,-0.1)
fit.model1 <- mle2.GTDL(t = t1,start = start1,
                         formula = formula1,
                         censur = censur1)
r1 <- random.quantile.GTDL(t = t1,formula = formula1 ,pHat = fit.model1$Coefficients[,1],
                           censur = censur1)
r1

### Example 2

data(tumor)
t2 <- tumor$time
formula2 <- ~tumor$group
censur2 <- tumor$censored
start2 <- c(1,-0.05,1.7)
fit.model2 <- mle2.GTDL(t = t2,start = start2,
                         formula = formula2,
                         censur = censur2)
r2 <- random.quantile.GTDL(t = t2,formula = formula2, pHat = fit.model2$Coefficients[,1],
                           censur = censur2)
r2
```

---

tumor                    *Tumor data*

---

### Description

Times (in days) of patients in ovarian cancer study

### Usage

```
data(tumor)
```

### Format

This data frame contains the following columns:

- time: survival time in days
- censored: censored = 0, dead = 1
- group: large tumor = 0, small tumor = 1

### References

- Colosimo, E. A and Giolo, S. R. Análise de Sobrevivência Aplicada. Edgard Blucher: São Paulo. 2006.

### Examples

```
data(tumor)
head(tumor)
```

# Index

artset1987, [2](#)  
dGTDL (fGTDL), [3](#)  
fGTDL, [3](#)  
fires (fGTDL), [3](#)  
hGTDL (fGTDL), [3](#)  
mle1.GTDL, [4](#)  
mle2.GTDL, [5](#)  
nrsp.GTDL, [7](#)  
optim, [5](#), [6](#)  
random.quantile.GTDL, [8](#)  
rGTDL (fGTDL), [3](#)  
sGTDL (fGTDL), [3](#)  
tumor, [10](#)