# Package 'indirect' 

February 9, 2022

Type Package<br>Title Elicitation of Independent Conditional Means Priors for Generalised Linear Models

Version 0.2.1
Date 2022-02-09
Author Geoffrey R. Hosack
Maintainer Geoff Hosack [geoff.hosack@csiro.au](mailto:geoff.hosack@csiro.au)
Description Functions are provided to facilitate prior elicitation for Bayesian generalised linear models using independent conditional means priors. The package supports the elicitation of multivariate normal priors for generalised linear models. The approach can be applied to indirect elicitation for a generalised linear model that is linear in the parameters. The package is designed such that the facilitator executes functions within the R console during the elicitation session to provide graphical and numerical feedback at each design point. Various methodologies for eliciting fractiles (equivalently, percentiles or quantiles) are supported, including versions of the approach of Hosack et al. (2017) [doi:10.1016/j.ress.2017.06.011](doi:10.1016/j.ress.2017.06.011). For example, experts may be asked to provide central credible intervals that correspond to a certain probability. Or experts may be allowed to vary the probability allocated to the central credible interval for each design point. Additionally, a median may or may not be elicited.

License GPL-3
Depends R (>=3.1.0)
Imports MASS, gplots
Suggests tools, utils
RoxygenNote 7.1.2
NeedsCompilation no

## Repository CRAN

Date/Publication 2022-02-09 05:30:02 UTC

## $R$ topics documented:

checkX ..... 2
CNdiag ..... 2
designLink ..... 3
dGompertzNorm ..... 4
dLogitNorm ..... 5
elicitPt ..... 6
indirect ..... 7
makeSweave ..... 8
muSigma ..... 9
mV ..... 10
pdist ..... 11
plotDesignPoint ..... 12
saveRecord ..... 14
Index ..... 16
checkX Helper function that checks for sensible covariate matrix.

## Description

Helper function that checks for sensible covariate matrix.

## Usage

checkX(X)

## Arguments

$\mathrm{X} \quad$ numeric matrix of covariates, $n$ design points by $p$ covariates, for a given model and design points.

## Value

throws an error if not full rank.

CNdiag Function to check condition number diagnostic.

## Description

This function calculates the condition number of the rescaled $n x p$ design matrix $X$ such that each column has unit length.

## Usage

CNdiag(X)

## Arguments

$X \quad$ Design matrix

## Value

a scalar giving the condition number of the rescaled design matrix

## Examples

```
X <- matrix(rnorm(16), nrow = 4)
CNdiag(X)
```


## Description

This builds the structure that will store elicited data. The linear predictor has a normal prior $g(\theta) N(m, V), \theta$ is the elicitation target. Link functions $g($.$) : logit, log, cloglog, identity.$

## Usage

```
    designLink(
        design,
        link = "identity",
        target = "Target",
        CI.prob = 1/2,
        expertID = "Expert",
        facilitator = "Facilitator",
        rapporteur = "none",
        intro.comments = "This is a record of the elicitation session.",
        fit.method = "KL"
    )
```


## Arguments

design a dataframe with covariate values that will be displayed to the expert(s) during the elicitation session.
link character logit, log, cloglog, identity
target character, name of target parameter of elicitation exercise
CI.prob numeric, a fraction between 0 and 1 that defines probability attributed to central credible interval. For example, $1 / 2$ for a central credible interval of probability 0.5 , or $1 / 3$ for a central credible interval of probablity $0.333 \ldots$ The default is probability $1 / 2$.
expertID character, identifier for expert or group of experts
facilitator character, facilitator identifier

$$
\begin{array}{ll}
\text { rapporteur } & \text { character, rapporteur identifier. Default "none". } \\
\text { intro.comments } & \begin{array}{l}
\text { character, text with any prefacing comments. This may include, for example, } \\
\text { the definition of the target parameter for the elictation session. Beware of non- }
\end{array} \\
& \begin{array}{l}
\text { ASCII text and special characters, which may affect the ability to save the elic- } \\
\text { itation record with function saveRecord or create a summary report with func- } \\
\text { tion makeSweave if called by the function makeSweave may affect ability to } \\
\text { render by means of Sweave or knitr etc. }
\end{array} \\
\text { fit.method } & \begin{array}{l}
\text { character, method used to fit conditional means prior: KL (default), moment, SS } \\
\text { (see vignette and } m V \text { for more information on these options) }
\end{array}
\end{array}
$$

## Details

Assumption: at least two fractiles selected from the median, upper and lower bounds of hte central credible interval of probability CI.prob will be elicited at each design point. The probabilities assigned to the central credible intervals can vary across design points. The argument CI. prob can later be adjusted by design point during the elicitation exercise, see function elicitPt. In the first instance, it is set to a global value specified by CI. prob in function designLink with default value 0.5 .

## Value

list of design with entries: theta, a $n x 4$ matrix with columns that give lower, median and upper quantiles followed by CI. prob and $n$ equal to the number of design points (scenarios); link, the link function used; target; expert facilitator; rapporteur; date; intro.comments; fit.method.

## Examples

```
X <- matrix(c(1, 1, 0, 1), nrow = 2) # design
Z <- designLink(design = X, link = "logit", target = "target",
    CI.prob = 1/2, expertID = "Expert", facilitator = "facilitator")
```


## Description

density for Gompertz transformed univariate Gaussian

## Usage

dGompertzNorm(x, mu, sigma)

## Arguments

x ,
numeric real
mu, numeric real
sigma, numeric real positive

## Value

tranformed density on support $(0,1)$

## Examples

mu <- -1
sigma <- 1
z <- rnorm(10000, mu, sigma)
hist(1 $-\exp (-\exp (z))$, freq $=$ FALSE $)$

integrate (function( $x$ ) dGompertzNorm ( $x, m u=m u$, sigma $=$ sigma), lower $=0$, upper $=1$ ) \# equals 1
dLogitNorm density for logit transformed univariate Gaussian

## Description

density for logit transformed univariate Gaussian

## Usage

dLogitNorm(x, mu, sigma)

## Arguments

| x, | numeric real |
| :--- | :--- |
| mu, | numeric real |
| sigma, | numeric real positive |

## Value

tranformed density on support $(0,1)$

## Examples

```
mu <- -1
sigma <- 1
z <- rnorm(10000, mu, sigma)
hist(exp(z)/(1 + exp(z)), freq = FALSE)
curve(dLogitNorm(x, mu = mu, sigma = sigma), col = 'red', add = TRUE, from = 0.01, to = 0.99)
integrate(function(x) dLogitNorm(x,mu=mu, sigma = sigma), lower = 0, upper = 1) # equals 1
```


## Description

Function to create or update elicitation at a given design point.

## Usage

elicitPt(
Z,
design.pt $=$ NULL,
lower.CI.bound = NA,
median = NA,
upper.CI.bound = NA,
CI.prob $=$ NULL,
comment = " "
)

## Arguments

Z
list of design with entries: theta, a $n x 4$ matrix with columns that give lower, median and upper quantiles of the central credible interval followed by the probability CI.prob allocated to the interval; link, the link function used; and target. This list object is created by designLink
design.pt single integer that denotes design point of interest
lower.CI.bound scalar that gives the lower bound of the central credible interval, default NA.
median scalar value, default NA
upper.CI.bound scalar that gives the upper bound of the central credible interval, default NA.
CI.prob numeric, a fraction between 0 and 1 that defines probability attributed to central credible interval. For example, $1 / 2$ for quartiles or $1 / 3$ for tertiles. Default NULL uses the initial CI. prob as defined by designLink.
comment character, ASCII text providing contributed commentary associated with elicitation design point. It is recommended to avoid special characters such as quotation marks etc.

## Value

Z, a list of design with entries: theta, a $n x 4$ matrix with columns that give lower, median and upper quantiles followed by CI.prob with updated entries for row specified by argument design.pt; link, the link function used; and target.

## Examples

```
X <- matrix(c(1, 1, 0, 1), nrow = 2) # design
Z <- designLink(design = X)
Z <- elicitPt(Z, design.pt = 1,
    lower.CI.bound = -1,
    median = 0,
    upper.CI.bound = 1,
    comment = "A completed elicitation scenario.")
```

indirect indirect: A package for assisting indirect elicitation of priors for gen-
eralised linear models.

## Description

The indirect package provides three categories of functions: elicitation functions, fitting functions and visualisation functions.

## Elicitation functions

These are the functions that are used to record expert opinion. This is where edits will be made and so on. The key function is designLink, which defines a list object that contains information about the design and elicitation. The elicitations are recorded and updated via function elicitPt.

## Fitting functions

These are generally helper functions except for the function muSigma that is used for estimating the mean vector and covariance matrix of the unknown coefficients for the multivariate normal prior. Helper functions include $m V$ for the elicited moments of conditional means priors.

## Visualisation functions

These are functions for visualisation. The core function is plotDesignPoint.

## References

Hosack, G. R., Hayes, K. R., \& Barry, S. C. (2017). Prior elicitation for Bayesian generalised linear models with application to risk control option assessment. Reliability Engineering and System Safety, 167:351-361. doi:10.1016/j.ress.2017.06.011

## Description

Creates a Sweave file that can be used to generate a pdf document of the summary report.

## Usage

```
    makeSweave(
        filename.rds = "",
        reportname = "",
        title = "Elicitation record",
        contact.details = "none",
        fitted.fractiles = TRUE,
        cumul.prob.bounds = c(0.05, 0.95)
    )
```


## Arguments

filename.rds character, filename of the record saved as an RDS object, see ?saveRDS.
reportname character, filename without extension to be used for the generated Sweave (.Rnw) file. The Sweave file supports the creation of report (.pdf) documentation and accompanying files such as the . tex file generated by using Sweave followed by tools: :texi2pdf().
title character, a title for the report
contact. details
character, an email address or other mechanism by which the expert may contact the facilitator or rapporteur
fitted.fractiles
logical or numeric vector. A logical value of FALSE will not plot any fitted fractiles from the fitted subjective probability distribution. A logical value of TRUE will plot the fitted fractiles that correspond to the final iteration of the raw elicited fractiles. Alternatively, a numeric vector can specify arbitrary fractiles for plotting from the fitted distribution, e.g., $c(1 / 10,1 / 4,1 / 2,3 / 4,9 / 10)$
cumul.prob.bounds
numeric vector that specifies the upper and lower plot bounds determined by this credible interval. The default is the 0.90 central credible interval, $c(0.05,0.95)$

## Examples

```
## Not run:
X <- matrix(c(1, 1, 0, 1), nrow = 2) # design
Z <- designLink(design = X)
Z <- elicitPt(Z, design.pt = 1,
    lower.CI.bound = -1,
```

```
    median = 0,
    upper.CI.bound = 1,
    comment = "A completed elicitation scenario.")
    tmp.rds <- tempfile(pattern = "record", fileext =".rds")
    saveRecord(Z, file = tmp.rds)
    tmpReport <- tempfile(pattern = "report")
    makeSweave(filename.rds = tmp.rds, reportname = tmpReport)
    setwd(tempdir())
    utils::Sweave(paste0(tmpReport, ".Rnw"))
    tools::texi2pdf(paste0(tmpReport, ".tex"))
    ## End(Not run)
```

    muSigma \(\quad\) Function to estimate mean and covariance for unknown parameters \(\beta\).
    
## Description

Function to estimate mean and covariance for unknown parameters $\beta$.

## Usage

muSigma(Z, X = NULL, fit.method = "KL", wls.method = "default")

## Arguments

Z list of design points and link function that is an output of function designLink
$X \quad$ model matrix for model formula and design points. The covariates must correspond to the description of design points in $Z$, but can be transformed etc. If NULL then $X$ will be coerced by applying as .matrix() to Z\$design. The matrix $X$ should be full rank when subsetted to the elicited design points. If a column of $X$ has the name offset then this column is treated as an offset during estimation
fit.method character, moment, KL. See mV. Default is KL.
wls.method character giving the numerical solution method: QR, using the QR decomposition, SVD, using the singular value decomposition, or option default that uses solve()

## Value

list of mu, numeric vector of location parameters for the normal prior; Sigma, the covariance matrix; and log.like, a scalar

## Examples

```
X <- matrix(c(1, 1, 0, 1), nrow = 2) # design
Z <- designLink(design = X)
Z <- elicitPt(Z, design.pt = 1,
    lower.CI.bound = -1,
    median = 0,
    upper.CI.bound = 1,
    comment = "The first completed elicitation scenario.")
Z <- elicitPt(Z, design.pt = 2,
    lower.CI.bound = -2,
    median = 1,
    upper.CI.bound = 2,
    comment = "The second completed elicitation scenario.")
prior <- muSigma(Z, X, fit.method = "KL")
prior$mu
prior$Sigma
```

mV
Helper function that translates elicited quantiles of target into independent conditional means normal prior for a defined inverse link function.

## Description

The default for fit.method is option KL. This option uses an objective function that minimises a discretised directed divergence from a cumulative distribution implied by raw elicited fractiles to a normal conditional mean prior for the linear predictor. An alterative method moment assigns the location parameter of the normal conditional mean prior to the elicited median on the linear predictor scale. The variance parameter is estimated as $V=\left(\left(g\left(f_{u}\right)-g\left(f_{l}\right) /(\text { qnorm }(u)-q n o r m(l))\right)^{2}\right.$, where $l$ is the probability associated with the fractile $f_{l}$ that defines the lower bound for the central credible interval and $u$ is the probability associated with the fractile $f_{u}$ that defines the upper bound for the central credible interval. This is also used to initialise the optimisation for the KL method. Another optimsation method that minimises the sum of squares is also available as method SS. See the vignette for more details on the choice of objective function for KL and SS.

## Usage

$m V(Z, f i t . m e t h o d=$ "KL")

## Arguments

Z
list object that contains matrix theta of elicitations and character link, see plotDesignPoint
fit.method character, moment, KL, SS. Default is KL.

## Value

A list with vector of means m and diagonal covariance matrix V .

```
pdist
```

Helper function that gives the probability distribution function for design point.

## Description

Helper function that gives the probability distribution function for design point.

## Usage

```
pdist(x, Z, design.pt = NULL, fit.method = "KL")
```


## Arguments

$x$ numeric: coordinate
Z list of design points and link function, see designLink
design.pt integer: design point
fit.method character: method for fit in mV , default is KL

## Examples

```
# design matrix: two scenarios
X <- matrix(c(1, 1, 0, 1), nrow = 2)
rownames(X) <- c("scenario1", "scenario2")
colnames(X) <- c("covariate1", "covariate2")
#' # logit link
# central credible intervals with probability = 1/2
Z <- designLink(design = X, link = "logit", CI.prob = 0.5)
#' # lower and upper quartiles and median
Z <- indirect::elicitPt(Z, design.pt = 1,
    lower.CI.bound = 0.2,
    median = 0.4,
    upper.CI.bound = 0.6,
    comment = "Completed.")
indirect::plotDesignPoint(Z, design.pt = 1,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1),
    fitted.fractiles = TRUE, fitted.curve = TRUE)
# probability that target is below 0.1 and
# probability that target is below 0.9
indirect::pdist(c(0.1, 0.9), Z, design.pt = 1)
```


## Description

Plot elicited data, fitted marginals or model output

## Usage

```
plotDesignPoint(
    Z,
    X = NULL,
    design.pt = NULL,
    elicited.fractiles = TRUE,
    fitted.fractiles = FALSE,
    fitted.curve = FALSE,
    CI.prob = NULL,
    estimated.probs = NULL,
    modelled.fractiles = FALSE,
    modelled.curve = FALSE,
    cumul.prob.bounds = c(0.05, 0.95),
    theta.bounds = NULL,
    ylim.max = NULL,
    xlog = FALSE,
    design.table = TRUE,
    n.pts = 101
)
```


## Arguments

Z list object that contains matrix theta of elicitations, character link and character target as initialised by designLink and updated by elicitPt
X design matrix (can be NULL, unless modelled output is requested)
design.pt single integer that denotes design point of interest
elicited.fractiles
logical, plot vertical lines for elicited fractiles?
fitted.fractiles
logical, plot vertical lines for fitted conditional mean prior fractiles for this design point? Alternatively, a numeric vector of arbitrary fractiles to be plotted from the fitted elicitation distribution. If TRUE then the fractiles corresponding to the median, upper and lower level central CI are plotted
fitted.curve, logical plot fitted conditional mean prior density for this design point?
CI.prob numeric scalar, locally specified probability assigned to the elicited central credible interval of the current design point. Defaults to NULL in which case the global value initially assigned by designLink or as updated by elicitPt is used
estimated.probs
numeric vector of values for which estimated probabilities are to be estimated from the fitted elicitation distribution for the target theta. Default is NULL. The result is output to the console.
modelled.fractiles
logical, plot vertical lines for modelled fractiles from the conditional mean prior distribution fit to all design points? This option requires a design matrix $X$ of full column rank.
modelled.curve logical, plot modelled conditional mean prior density for the entire model? This option requires a design matrix $X$ of full column rank.
cumul.prob.bounds
numeric vector of length two, giving plot bounds by cumulative probability. This argument is ignored if there is not enough data to fit a parametric distribution or if theta. bounds is not NULL
theta.bounds numeric vector giving support of response for plotting purposes (can be NULL). This will overwrite cumul. prob.bounds, if applicable
ylim.max numeric maximum value of $y$-axis (can be NULL)
$x$ log $\quad \operatorname{logical} \log x$-axis
design.table logical include design dataframe, elicited fractiles and modelled or fitted fractiles
n.pts numeric giving number of point to evalate density curve (if plotted)

## Value

a plot to the current device. See dev. cur() to check.

## Examples

```
# design matrix: two scenarios
X <- matrix(c(1, 1, 0, 1), nrow = 2)
rownames(X) <- c("scenario1", "scenario2")
colnames(X) <- c("covariate1", "covariate2")
# logit link
# central credible intervals with probability = 1/2
Z <- designLink(design = X, link = "logit", CI.prob = 0.5)
# 1st design point
# no elicited fractiles
indirect::plotDesignPoint(Z, design.pt = 1)
# elicited median
Z <- indirect::elicitPt(Z, design.pt = 1,
    lower.CI.bound = NA,
    median = 0.4,
    upper.CI.bound = NA,
    CI.prob = NULL)
indirect::plotDesignPoint(Z, design.pt = 1,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1))
# lower and upper quartiles and median
```

```
Z <- indirect::elicitPt(Z, design.pt = 1,
    lower.CI.bound = 0.2,
    median = 0.4,
    upper.CI.bound = 0.6,
    comment = "Completed.")
indirect::plotDesignPoint(Z, design.pt = 1,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1),
    fitted.fractiles = TRUE, fitted.curve = TRUE)
indirect::plotDesignPoint(Z, design.pt = 1,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1),
    fitted.fractiles = c(1/10, 1/4, 1/2, 3/4, 9/10),
    fitted.curve = TRUE)
# second design point
# central credible intervals with probability = 1/3
# elicit upper and lower tertiles
Z <- elicitPt(Z, design.pt = 2,
    lower.CI.bound = 0.1,
    upper.CI.bound = 0.3,
    CI.prob = 1/3,
    comment = "Switched to tertiles.")
indirect::plotDesignPoint(Z, design.pt = 2,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1))
indirect::plotDesignPoint(Z, design.pt = 2,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1),
    fitted.fractiles = TRUE, fitted.curve = TRUE)
indirect::plotDesignPoint(Z, design.pt = 2,
    elicited.fractiles = TRUE, theta.bounds = c(0, 1),
    fitted.fractiles = c(1/10, 1/3, 1/2, 2/3, 9/10),
    fitted.curve = TRUE)
```

    saveRecord Function to save elicitation record.
    
## Description

Function to save elicitation record.

## Usage

```
saveRecord(
    designLink.obj,
    conclusion.comments = "This concludes the elicitation record.",
    file = ""
)
```


## Arguments

designLink.obj list object initally created by function designLink and subsequently updated by function elicitPt

## conclusion. comments

character, comments to conclude session. Beware of non-ASCII text and special characters, which may affect ability to save or generate a Sweave document by using makeSweave
file character providing filename.

## Value

an RDS file is created with filename file. A timestamp is added to designLink.obj using Sys.time().

## Examples

```
## Not run:
X <- matrix(c(1, 1, 0, 1), nrow = 2) # design
Z <- designLink(design = X)
tmp <- tempfile(pattern = "report", fileext =".rds")
saveRecord(Z, file = tmp)
## End(Not run)
```


## Index

```
checkX, 2
CNdiag, 2
designLink, 3, 4, 6, 7, 11, 12, 14
dGompertzNorm,4
dLogitNorm, 5
elicitPt,4, 6, 7, 12, 14
indirect,7
makeSweave, 4, 8, 15
muSigma, 7, }
mV, 4, 7, 9, 10,11
pdist,11
plotDesignPoint, 7, 10,12
saveRecord, 4, 14
Sweave, 4, }
```

