

# Package: mpred (via r-universe)

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**Type** Package  
**Title** Generic Predictive Margins  
**Version** 0.2.4.1  
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**Description** Provides a function to compute predictive margins.  
**License** GPL-2  
**LazyLoad** Yes  
**Depends** R (>= 3.3.3), stats  
**Imports** parallel, ordinal  
**Suggests** mclogit, magrittr, carData, MASS, ggplot2  
**RoxygenNote** 7.1.0  
**Repository** <https://melff.r-universe.dev>  
**RemoteUrl** <https://github.com/melff/mpred>  
**RemoteRef** HEAD  
**RemoteSha** 98040e126651ab91f5ac2b77fdee64f4201f9436

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cibeta	<i>Confidence/prediction interval based on a beta distribution</i>
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**Description**

Creates prediction intervals from a beta distribution with given mean and standard deviation. Prediction intervals stay between 0 and 1.

**Usage**

```
cibeta(mean, sd, level)
```

**Arguments**

mean	a scalar; the mean of the beta distribution
sd	a scalar; the standard deviation of the beta distribution
level	a scalar; the confidence level

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cinorm	<i>Confidence/prediction interval based on a normal distribution</i>
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**Description**

Creates prediction intervals from a normal distribution with given mean and standard deviation.

**Usage**

```
cinorm(mean, sd, level)
```

**Arguments**

mean	a scalar; the mean parameter of the normal distribution
sd	a scalar; the standard deviation parameter of the normal distribution
level	a scalar; the confidence level

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get_cifunc	<i>A generic function that returns functions for confidence/prediction intervals</i>
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### Description

Returns a function for confidence or prediction intervals appropriate for an fitted model object. For linear regression this will be `cinorm()`, for logistic regression (done by `glm`) this will be `cibeta`, etc.

### Usage

```
get_cifunc(obj)

## S3 method for class 'lm'
get_cifunc(obj)

## S3 method for class 'glm'
get_cifunc(obj)

## S3 method for class 'mblogit'
get_cifunc(obj)

## S3 method for class 'mclogit'
get_cifunc(obj)

## S3 method for class 'clm'
get_cifunc(obj)

## S3 method for class 'clmm'
get_cifunc(obj)
```

### Arguments

`obj` an fitted model object

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predmarg	<i>Generic function to produce predictive margins</i>
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### Description

Generic function to produce predictive margins

**Usage**

```

predmarg(
  obj,
  settings,
  data,
  subset,
  type = NULL,
  groups = NULL,
  setup = NULL,
  cifunc = get_cifunc(obj),
  level = 0.95,
  parallel = FALSE,
  mc.cores = if (.Platform$OS.type == "windows") 1L else max.cores,
  ...
)

```

**Arguments**

obj	a model object, e.g. returned by <code>lm</code> , <code>glm</code> , etc.
settings	an optional data frame of settings for independent variables.
data	an optional data frame for which the predictive margins are computed. If omitted, an attempt is made to obtain the data from the model object.
subset	an optional logical vector that defines a subset for which a predictive margin is computed
type	an optional character string that specifies the type of predictions, e.g. probabilities or cumulative probabilities. For future versions only.
groups	a variable that defines groups for which predictive margins are computed. This variable has to have the same number of observations as the data to which the model was fitted.
setup	an optional expression that is evaluated for each setting, i.e. individually for each row of the settings data frame. Can be used to modify independent variables.
cifunc	a function to compute prediction intervals. By default it is the chosen by the function of the same name.
level	level of confidence intervals of predictions.
parallel	logical value that determines whether predictions for individual settings are computed in parallel. (Does not yet work on windows.)
mc.cores	number of CPU cores used for parallel processing.
...	optional vectors of values of independent variables. These further arguments, if present, are used to create a data frame of settings, using <a href="#">expand.grid</a> .

**Details**

The generic function `predmarg` computes predictive margins for various settings of the independent variables. It is also possible to provide settings for independent variables that are included in the model, but that are used in the `setup` expression to transform independent variables. See the examples below.

**Value**

a data frame with the following variables:

pred	the mean prediction for the setting of the independent variables
var.pred	the (estimated) variance of the mean prediction
se.pred	the standard error of prediction, i.e. the square root of the variance of the mean prediction
lower	lower prediction interval computed with qfunc
upper	upper prediction interval computed with qfunc
...	the independent variables for which values are set to create the predictions are also included in the resulting data frame.

**Examples**

```
library(magrittr)

# Simple linear regression

fm <- lm(weight ~ poly(height, 2), data = women)
pm <- predmarg(fm,
              height=seq(from=58, to=72,
                         length=10))

str(pm)

plot(pred~height, data=pm,
     type="l")

with(women, points(height, weight))
with(pm, lines(height, lower, lty=2))
with(pm, lines(height, upper, lty=2))

# Logistic regression

library(carData)
Chile %<>% within({
  vote2 <- factor(vote, levels=c("N", "Y"))
  vote2 <- as.integer(vote2=="Y")
})

glm.Chile.1 <- glm(vote2~sex+age+income+education,
                  data=Chile,
                  family=binomial)
summary(glm.Chile.1)

pm.Chile.1.income <- predmarg(glm.Chile.1,
                             income=seq(from=2500, to=200000, length=20))

plot(pred~income, data=pm.Chile.1.income,
     type="l")
```

```

# Baseline category logit

library(mclogit)
library(MASS)

mb.Chile <- mblogit(vote~statusquo,
                   data=Chile)
pm.mb.Chile <- predmarg(mb.Chile,
                       statusquo=seq(from=-2, to=2, length=20))
str(pm.mb.Chile)

library(ggplot2)
(ggplot(pm.mb.Chile,
        aes(x=statusquo,
            y=pred,
            fill=response)
        ) + geom_area())

(ggplot(pm.mb.Chile,
        aes(x=statusquo,
            y=pred,
            ymin=lower,
            ymax=upper)
        ) + geom_line() +geom_ribbon(alpha=.25) + facet_grid(~response))

mb.hs <- mblogit(Sat~Infl+Type+Cont,weights=Freq,
                 data=housing)

pm.mb.hs <- predmarg(mb.hs,
                    Infl=levels(Infl),
                    Type=levels(Type))

dodge <- position_dodge(width=.8)
(ggplot(pm.mb.hs)
 +facet_wrap(~Type)
 +geom_bar(
   aes(fill=response,
       x=Infl,
       y=pred),
   stat='identity',position=dodge,width=.8)
 +geom_errorbar(
   aes(x=Infl,
       ymin=lower,
       ymax=upper,group=response),
   position=dodge,width=.4))

# The following requires the most current 'mclogit' version on GitHub
# and fails with the CRAN version
# # Baseline category logit with random effects

```

```

#
# # Some artificial data
# exadata <- local({
#   B <- cbind(c(-.5,.3),
#             c(.5,-.5))
#   set.seed(42)
#   x <- rnorm(n=60)
#   X <- cbind(1,x)
#   Eta <- X %*% B
#   j <- rep(1:10,6)
#   jf <- as.factor(j)
#   u1 <- rnorm(n=10,sd=.8)
#   u2 <- rnorm(n=10,sd=.8)
#   Eta <- Eta + cbind(u1[j],x*u2[j])
#   expEta <- cbind(1,exp(Eta))
#   sum.expEta <- rowSums(expEta)
#   pi <- expEta/sum.expEta
#   Y <- t(apply(pi,1,rmultinom,n=1,size=300))
#   res <- data.frame(Y,x,j,jf)
#   names(res)[1:3] <- paste0("y",1:3)
#   res
# })
#
# # Baseline logit model with random intercepts and random slopes
# mbrsl <- mblogit(cbind(y1,y2,y3)~x,data=exadata,
#                 random = ~1+x|j)
# summary(mbrsl)
#
# # Predictive margins for values of x
# pm.mbrsl <- predmarg(mbrsl,x=seq(from=min(x),to=max(x),length=24))
# (ggplot(pm.mbrsl,
#         aes(x=x,
#             y=pred,
#             fill=response
#             )
#         ) + geom_area())
#
# # Predictive margins for the random effects
# pm.mbrsl.j <- predmarg(mbrsl,j=1:10)
# (ggplot(pm.mbrsl.j,
#         aes(x=j,
#             y=pred,
#             fill=response
#             )
#         ) + geom_bar(position="fill",stat="identity"))
#
# pm.mbrsl.jx <- predmarg(mbrsl,
#                       j=1:10,
#                       x=seq(from=min(x),to=max(x),
#                             length=24))
#
# (ggplot(pm.mbrsl.jx,
#         aes(x=x,

```

```
#           y=pred,  
#           fill=response  
#         )  
#       ) + geom_area()  
#     + facet_wrap(~j)
```



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