

# Package ‘MargCond’

April 9, 2018

**Type** Package

**Title** Joint Marginal-Conditional Model

**Version** 1.0.0

**Date** 2018-04-06

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**Depends** gee, lme4, MASS, Matrix

**Description** Fits joint marginal conditional models for multivariate longitudinal data, as in Proudfoot, Faig, Natarajan, and Xu (2018) <doi:10.1002/sim.7552>. Development of this package was supported by the UCSD Altman Translational Research Institute, NIH grant UL1TR001442. The content is solely the responsibility of the authors and does not necessarily represent the official views of the NIH.

**License** GPL-2

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2018-04-09 11:00:52 UTC

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MargCond	<i>Function to fit joint marginal-conditional models for longitudinal multivariate data.</i>
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## Description

Produces an object of class "MargCond" which is a marginal-conditional multivariate model.

**Usage**

```
MargCond(formula, data, ID, tol = 1e-04, max.iter = 50,
          corstr = "independence", silent = F)
```

**Arguments**

formula	a two-sided linear formula object similar to those in <code>lmer</code> .
data	a data frame in which to interpret the variables occurring in the formula.
ID	a vector which identifies the clusters. The length of ID should be the same as the number of observations. Data are assumed to be sorted so that observations on a cluster are contiguous rows for all entities in the formula.
tol	the tolerance used in the fitting algorithm.
max.iter	the maximum number of iterations for the ES algorithm.
corstr	a character string specifying the correlation structure. The following are permitted: "independence", "fixed", "stat_M_dep", "non_stat_M_dep", "exchangeable", "AR-M" and "unstructured"
silent	a logical variable controlling whether an indication at each iteration is printed.

**Details**

The joint marginal-conditional model

Care should be taken when specifying the random effects structure (see the singular models section of <https://bbolker.github.io/mixedmodels-misc/glmmFAQ.html>). As initial estimates for the expectation-substitution algorithm are taken from the univariate mixed model fits, we recommend that these models be fit separately first and examined to ensure that they are not over parameterized.

**Value**

An object of class "MargCond" representing the fit.

An object of class "MargCond" is a list containing the following components:

coefficients	a named vector of coefficients.
sigma	a named vector of outcome error standard deviations.
SE	a vector of coefficient, random effect, and error standard deviations.
residuals	the residuals, that is response minus fitted values.
working.correlation	the working correlation returned by the GEE step at convergence.
rand.eff	the random effect covariance matrix.
outcomes	vector of outcome names
Call	the matched call.
v.cov	the scaled covariance matrix of theta
obs	the total number of observations

groups	the total number of clusters
converge	logical indicator of whether the expectation-substitution algorithm converged (i.e. the difference between each element of theta from the previous iteration is smaller than tol, and the number of iterations is less than max.iter).

## References

Proudfoot J. A., Faig W., Natarajan L., and Xu R. (2018) A joint marginal-conditional model for multivariate longitudinal data. *Statistics in Medicine*. <https://doi.org/10.1002/sim.7552>

## See Also

[gee](#), [lmer](#).

## Examples

```
set.seed(2112)
NN = 80
n_times = 1:3

## Simulating some data
simdat <- simDat(n = NN,
  fixed_effects = list(c(1, 1, 2), c(1.5, 1, 3)),
  rand_effects = list(1, 1),
  error_var = c(4, 4),
  error_structure = 'normal',
  rho = .35,
  times = n_times,
  X = cbind(rep(1, NN * length(n_times)),
    rnorm(NN * length(n_times), 0, 2),
    rbinom(NN * length(n_times), 1, .5)),
  Z = cbind(rep(1, NN * length(n_times))))

## Adding random missing values
aa <- sample(1:nrow(simdat), 10, replace = TRUE)
bb <- sample(1:7, 10, replace = TRUE)
for (i in 1:length(aa)) {
  simdat[aa[i], bb[i]] <- NA
}

## A fit for this simulated multivariate longitudinal data,
## including a random intercept and exchangeable correlation
## structure.
summary(MargCond(c(outcome1, outcome2) ~ X2 + X3 + (1 | ID),
  data = simdat, ID = simdat$ID, corstr = "exchangeable"))
```



```
                                rbinom(NN * length(n_times), 1, .5)),
Z = cbind(rep(1, NN * length(n_times))))

## Adding random missing values
aa <- sample(1:nrow(simdat), 10, replace = TRUE)
bb <- sample(1:7, 10, replace = TRUE)
for (i in 1:length(aa)) {
  simdat[aa[i], bb[i]] <- NA
}
```

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