

# On the usage of the geepack

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## 1 Introduction

This note contains a few extra examples. We illustrate the usage of a the `waves` argument and the `zcor` argument together with a fixed working correlation matrix for the `geeglm()` function.

## 2 Citing geepack

The primary reference for the `geepack` package is

Halekoh, U., Højsgaard, S., Yan, J. (2006) *The R Package geepack for Generalized Estimating Equations* (2006) Journal of Statistical Software  
<https://www.jstatsoft.org/article/view/v015i02>

```

> library(geepack)
> citation("geepack")

To cite geepack in publications use:

Højsgaard, S., Halekoh, U. & Yan J. (2006) The R Package geepack for
Generalized Estimating Equations Journal of Statistical Software, 15,
2, pp1--11

Yan, J. & Fine, J.P. (2004) Estimating Equations for Association
Structures Statistics in Medicine, 23, pp859--880.

Yan, J (2002) geepack: Yet Another Package for Generalized Estimating
Equations R-News, 2/3, pp12-14.

To see these entries in BibTeX format, use 'print(<citation>,
bibtex=TRUE)', 'toBibtex(..)', or set
'options(citation.bibtex.max=999)'.

```

If you use `geepack` in your own work, please do cite the above reference.

### 3 Simulating a dataset

To illustrate the usage of the `waves` argument and the `zcor` argument together with a fixed working correlation matrix for the `geeglm()` we simulate some data suitable for a regression model.

```

> library(geepack)
> timeorder <- rep(1:5, 6)
> tvar       <- timeorder + rnorm(length(timeorder))
> idvar     <- rep(1:6, each=5)
> uuu       <- rep(rnorm(6), each=5)
> yvar      <- 1 + 2*tvar + uuu + rnorm(length(tvar))
> simdat    <- data.frame(idvar, timeorder, tvar, yvar)
> head(simdat,12)

  idvar timeorder      tvar      yvar
1     1         1 1.7529009 6.2151324
2     1         2 2.2067282 5.2547923
3     1         3 2.3776130 6.1238307
4     1         4 2.4357724 6.1646994
5     1         5 5.4654916 13.1490039
6     2         1 0.8323012 2.8649181
7     2         2 2.3.8662108 9.4942795
8     2         3 2.6178176 7.3706183
9     2         4 2.8523845 7.9387815
10    2         5 5.5576360 12.7722539
11    3         1 0.9673407 0.9353348
12    3         2 3.7615635 8.5728429

```

Notice that clusters of data appear together in `simdat` and that observations are ordered (according to `timeorder`) within clusters.

We can fit a model with an AR(1) error structure as

```

> mod1 <- geeglm(yvar~tvar, id=idvar, data=simdat, corstr="ar1")
> mod1

Call:
geeglm(formula = yvar ~ tvar, data = simdat, id = idvar, corstr = "ar1")

Coefficients:
(Intercept)      tvar
1.108898     2.050621

Degrees of Freedom: 30 Total (i.e. Null); 28 Residual

Scale Link:           identity
Estimated Scale Parameters: [1] 0.9235482

Correlation: Structure = ar1   Link = identity
Estimated Correlation Parameters:
alpha
0.1219607

Number of clusters: 6  Maximum cluster size: 5

```

This works because observations are ordered according to time within each subject in the dataset.

## 4 Using the waves argument

If observations were not ordered according to cluster and time within cluster we would get the wrong result:

```

> set.seed(123)
> ## library(doBy)
> simdatPerm <- simdat[sample(nrow(simdat)),]
> ## simdatPerm <- orderBy(~idvar, simdatPerm)
> simdatPerm <- simdatPerm[order(simdatPerm$idvar),]
> head(simdatPerm)

  idvar timeorder      tvar      yvar
3      1          3 2.377613 6.123831
5      1          5 5.465492 13.149004
4      1          4 2.435772 6.164699
1      1          1 1.752901 6.215132
2      1          2 2.206728 5.254792
10     2          5 5.557636 12.772254

```

Notice that in `simdatPerm` data is ordered according to subject but the time ordering within subject is random.

Fitting the model as before gives

```

> mod2 <- geeglm(yvar~tvar, id=idvar, data=simdatPerm, corstr="ar1")
> mod2

Call:
geeglm(formula = yvar ~ tvar, data = simdatPerm, id = idvar,
       corstr = "ar1")

Coefficients:
(Intercept)      tvar
1.124689     2.048263

Degrees of Freedom: 30 Total (i.e. Null);  28 Residual

Scale Link:           identity
Estimated Scale Parameters: [1] 0.9238069

Correlation: Structure = ar1   Link = identity
Estimated Correlation Parameters:
alpha
-0.1104243

Number of clusters: 6  Maximum cluster size: 5

```

Likewise if clusters do not appear contiguously in data we also get the wrong result (the clusters are not recognized):

```

> ## simdatPerm2 <- orderBy(~timeorder, data=simdat)
> simdatPerm2 <- simdat[order(simdat$timeorder),]
> geeglm(yvar~tvar, id=idvar, data=simdatPerm2, corstr="ar1")

Call:
geeglm(formula = yvar ~ tvar, data = simdatPerm2, id = idvar,
       corstr = "ar1")

Coefficients:
(Intercept)      tvar
1.099137     2.050805

Degrees of Freedom: 30 Total (i.e. Null);  28 Residual

Scale Link:           identity
Estimated Scale Parameters: [1] 0.9234633

Correlation: Structure = ar1   Link = identity
Estimated Correlation Parameters:
alpha
0

Number of clusters: 30  Maximum cluster size: 1

```

To obtain the right result we must give the **waves** argument:

```

> wav <- simdatPerm$timeorder
> wav

[1] 3 5 4 1 2 5 4 3 2 1 5 4 1 3 2 4 3 5 2 1 2 4 5 3 1 3 2 1 5 4

> mod3 <- geeglm(yvar~tvar, id=idvar, data=simdatPerm, corstr="ar1", waves=wav)
> mod3

Call:
geeglm(formula = yvar ~ tvar, data = simdatPerm, id = idvar,
       waves = wav, corstr = "ar1")

Coefficients:
(Intercept)      tvar
1.108898     2.050621

Degrees of Freedom: 30 Total (i.e. Null); 28 Residual

Scale Link:           identity
Estimated Scale Parameters: [1] 0.9235482

Correlation: Structure = ar1   Link = identity
Estimated Correlation Parameters:
alpha
0.1219607

Number of clusters: 6  Maximum cluster size: 5

```

## 5 Using a fixed correlation matrix and the `zcor` argument

Suppose we want to use a fixed working correlation matrix:

```

> cor.fixed <- matrix(c(1      , 0.5   , 0.25,  0.125, 0.125,
+                      0.5   , 1     , 0.25,  0.125, 0.125,
+                      0.25,  0.25 , 1     , 0.5   , 0.125,
+                      0.125, 0.125, 0.5   , 1     , 0.125,
+                      0.125, 0.125, 0.125, 0.125, 1      ), 5, 5)

> cor.fixed

 [,1] [,2] [,3] [,4] [,5]
[1,] 1.000 0.500 0.250 0.125 0.125
[2,] 0.500 1.000 0.250 0.125 0.125
[3,] 0.250 0.250 1.000 0.500 0.125
[4,] 0.125 0.125 0.500 1.000 0.125
[5,] 0.125 0.125 0.125 0.125 1.000

```

Such a working correlation matrix has to be passed to `geeglm()` as a vector in the `zcor` argument. This vector can be created using the `fixed2Zcor()` function:

```

> zcor <- fixed2Zcor(cor.fixed, id=simdatPerm$idvar, waves=simdatPerm$timeorder)
> zcor

[1] 0.125 0.500 0.250 0.250 0.125 0.125 0.125 0.125 0.500 0.125 0.125
[13] 0.125 0.125 0.500 0.125 0.125 0.250 0.250 0.500 0.125 0.125 0.125
[25] 0.125 0.500 0.125 0.250 0.500 0.250 0.500 0.125 0.125 0.125 0.250
[37] 0.250 0.125 0.125 0.500 0.125 0.125 0.250 0.500 0.125 0.500 0.125
[49] 0.125 0.250 0.250 0.125 0.500 0.500 0.125 0.125 0.125 0.125 0.125

```

Notice that `zcor` contains correlations between measurements within the same cluster. Hence if a cluster contains only one observation, then there will be generated no entry in `zcor` for that cluster. Now we can fit the model with:

```

> mod4 <- geeglm(yvar~tvar, id=idvar, data=simdatPerm, corstr="fixed", zcor=zcor)
> mod4

Call:
geeglm(formula = yvar ~ tvar, data = simdatPerm, id = idvar,
       zcor = zcor, corstr = "fixed")

Coefficients:
(Intercept)      tvar
1.075672     2.061962

Degrees of Freedom: 30 Total (i.e. Null); 28 Residual

Scale Link:           identity
Estimated Scale Parameters: [1] 0.9240214

Correlation: Structure = fixed   Link = identity
Estimated Correlation Parameters:
alpha:1
1

Number of clusters: 6  Maximum cluster size: 5

```

## 6 When do GEE's work best?

GEEs work best when you have relatively many relatively small clusters in your data.