# Travel Mode - Multinomial Logit Model

## February 1, 2012

For multinomial models that include category–specific as well as global effects the function "mlogit" from the library "mlogit" can be used.

#### > library(mlogit)

The "Travel Mode"—data are stored in the "Edcat"—package and can be loaded by the following command.

```
> data(ModeChoice, package="Ecdat")
```

For the use of the function "mlogit" an appropriate data set has to be built. This is done by use of the function "mlogit.data".

```
> travel.long <- mlogit.data(ModeChoice, choice="mode", shape="long", alt.levels=c("air","
```

Now the model can be fitted. In the formula first the category–specific effects and then, separated by "|", the global effects are specified.

```
> travel.kat.id <- mlogit(mode ~ invt + gc|hinc, data=travel.long)
> summary(travel.kat.id)

Call:
mlogit(formula = mode ~ invt + gc | hinc, data = travel.long,
    method = "nr", print.level = 0)

Frequencies of alternatives:
    air train bus car
0.27619 0.30000 0.14286 0.28095
```

```
nr method 4 iterations, 0h:0m:0s g'(-H)^-1g = 0.00216 successive fonction values within tolerance limits
```

### Coefficients:

```
Estimate Std. Error t-value Pr(>|t|)
train:(intercept) 3.5250366 0.6549825 5.3819 7.371e-08 ***
bus:(intercept) 2.2782769 0.7176686 3.1746 0.001501 **
car:(intercept) 1.5334957 0.7065856 2.1703 0.029985 *
invt -0.0031266 0.0009548 -3.2746 0.001058 **
gc -0.0016225 0.0055279 -0.2935 0.769130
```

```
train:hinc
                 -0.0355771 0.0131492 -2.7056 0.006817 **
bus:hinc
car:hinc
                 Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Log-Likelihood: -250.17
McFadden R^2: 0.11839
Likelihood ratio test : chisq = 67.186 (p.value = 3.9423e-13)
  Now the same model is fitted with the package "VGAM".
> library(VGAM)
  At first the data need to be prepared adequately to be ready for use with
the function "vglm".
> travelmode <- matrix(ModeChoice$mode, byrow = T, ncol = 4); colnames(travelmode) = c("ai
> travelhinc <- matrix(ModeChoice$hinc, byrow = T, ncol = 4)
> travelhinc <- travelhinc[,1]
> travelinvt <- matrix(ModeChoice$invt, byrow = T, ncol = 4); colnames(travelinvt) = c("in
> travelgc <- matrix(ModeChoice$gc, byrow = T, ncol = 4); colnames(travelgc) = c("gcair","
> travelinvt <- sweep(travelinvt[,-1], 1, travelinvt[,1])</pre>
> travelgc <- sweep(travelgc[,-1], 1, travelgc[,1])</pre>
> Invt <- travelinvt[,1]</pre>
> Gc <- travelgc[,1]</pre>
> traveldat <- cbind(travelhinc, travelinvt, Invt, travelgc, Gc)
> traveldat <- as.data.frame(traveldat)</pre>
  Now the model can be fitted.
> fit <- vglm(travelmode ~ Invt + Gc + travelhinc,
             multinomial(parallel = FALSE ~ travelhinc, refLevel = 1),
             xij = list(Invt ~ invttrain + invtbus + invtcar,
                        Gc ~ gctrain + gcbus + gccar),
             form2 = ~ Invt + invttrain + invtbus + invtcar +
                       Gc + gctrain + gcbus + gccar + travelhinc,
             data = traveldat, trace = TRUE)
VGLM
       linear loop 1: deviance = 501.4629
VGLM
       linear loop 2: deviance = 500.3338
VGLM
       linear loop 3 : deviance = 500.3317
VGLM
       linear loop 4: deviance = 500.3317
> summary(fit)
Call:
vglm(formula = travelmode ~ Invt + Gc + travelhinc, family = multinomial(parallel = FALSE
   travelhinc, refLevel = 1), data = traveldat, form2 = ~Invt +
   invttrain + invtbus + invtcar + Gc + gctrain + gcbus + gccar +
```

Gc ~ gctrain + gcbus + gccar), trace = TRUE)

travelhinc, xij = list(Invt ~ invttrain + invtbus + invtcar,

```
Min
                              1Q Median
log(mu[,2]/mu[,1]) -2.7819 -0.54900 -0.25782 0.71847 3.3016
log(mu[,3]/mu[,1]) -2.1395 -0.40759 -0.22128 -0.13141 4.3081
log(mu[,4]/mu[,1]) -2.0047 -0.65888 -0.24440 0.86597 5.1282
Coefficients:
                  Value Std. Error t value
(Intercept):1 3.5250538 0.6549818 5.38191
(Intercept):2 2.2782935 0.7176672 3.17458
(Intercept):3 1.5334984 0.7065854 2.17029
Invt
             -0.0031266 0.0009548 -3.27460
Gc
             travelhinc:1 -0.0569415 0.0124103 -4.58824
travelhinc:2 -0.0355781 0.0131491 -2.70574
travelhinc:3 -0.0023652 0.0104474 -0.22639
Number of linear predictors: 3
Names of linear predictors:
log(mu[,2]/mu[,1]), log(mu[,3]/mu[,1]), log(mu[,4]/mu[,1])
Dispersion Parameter for multinomial family:
Residual Deviance: 500.3317 on 622 degrees of freedom
Log-likelihood: -250.1658 on 622 degrees of freedom
Number of Iterations: 4
> summary(travel.kat.id)
Call:
mlogit(formula = mode ~ invt + gc | hinc, data = travel.long,
   method = "nr", print.level = 0)
Frequencies of alternatives:
   air train
                 bus
0.27619 0.30000 0.14286 0.28095
nr method
4 iterations, 0h:0m:0s
g'(-H)^-1g = 0.00216
successive fonction values within tolerance limits
Coefficients :
                   Estimate Std. Error t-value Pr(>|t|)
train:(intercept) 3.5250366 0.6549825 5.3819 7.371e-08 ***
bus:(intercept)
                  2.2782769 0.7176686 3.1746 0.001501 **
```

Pearson Residuals:

```
car:(intercept)
           1.5334957 0.7065856 2.1703 0.029985 *
           invt
gc
           train:hinc
           bus:hinc
           -0.0355771 0.0131492 -2.7056 0.006817 **
           car:hinc
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Log-Likelihood: -250.17
McFadden R^2: 0.11839
Likelihood ratio test : chisq = 67.186 (p.value = 3.9423e-13)
```

At last we compare the coefficients of the two fitted models.

#### > summary(travel.kat.id)\$CoefTable

```
Estimate
                                Std. Error
                                                          Pr(>|t|)
                                              t-value
train:(intercept) 3.525036582 0.6549824780 5.3818792 7.371222e-08
bus:(intercept)
                  2.278276953 0.7176686437 3.1745527 1.500676e-03
car:(intercept)
                  1.533495711 0.7065856497 2.1702899 2.998489e-02
invt
                 -0.003126577 0.0009547997 -3.2745899 1.058154e-03
                 -0.001622510 0.0055279076 -0.2935124 7.691305e-01
gc
                 -0.056940856 0.0124103406 -4.5881784 4.471305e-06
train:hinc
                 -0.035577091 0.0131492259 -2.7056415 6.817260e-03
bus:hinc
car:hinc
                 -0.002365193 0.0104474472 -0.2263895 8.208985e-01
```

#### > summary(fit)@coef3

```
ValueStd. Errort value(Intercept):13.5250537590.65498180225.381911(Intercept):22.2782935440.71766724653.174582(Intercept):31.5334983820.70658541092.170294Invt-0.0031265830.0009547997-3.274596Gc-0.0016225280.0055279012-0.293516travelhinc:1-0.0569414600.0124102999-4.588242travelhinc:2-0.0355780930.0131491108-2.705741travelhinc:3-0.0023651980.0104474488-0.226390
```