

# Travel Mode - Multinomial Logit Model

February 5, 2020

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="Edcat")
```

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", "train", "bus", "car"))
```

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")
```

Frequencies of alternatives:

```
      air  train    bus    car
0.27619 0.30000 0.14286 0.28095
```

nr method

4 iterations, 0h:0m:0s

$g'(-H)^{-1}g = 0.00216$

successive function values within tolerance limits

Coefficients :

	Estimate	Std. Error	z-value	Pr(> z )	
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        -0.0569409  0.0124103 -4.5882  4.471e-06 ***
bus:hinc          -0.0355771  0.0131492 -2.7056  0.006817 **
car:hinc          -0.0023652  0.0104475 -0.2264  0.820898
---
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)

```

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$mode, byrow = T, ncol = 4)
> colnames(travelmode) <- c("air", "train", "bus", "car")
> travelhinc <- matrix(ModeChoice$hinc, byrow = T, ncol = 4)
> travelhinc <- travelhinc[,1]
> travelinvnt <- matrix(ModeChoice$invnt, byrow = T, ncol = 4)
> colnames(travelinvnt) <- c("invntair", "invnttrain", "invntbus", "invntcar")
> travelgc <- matrix(ModeChoice$gc, byrow = T, ncol = 4)
> colnames(travelgc) <- c("gcair", "gctrain", "gcbus", "gccar")
> travelinvnt <- sweep(travelinvnt[, -1], 1, travelinvnt[, 1])
> travelgc <- sweep(travelgc[, -1], 1, travelgc[, 1])
> Invt <- travelinvnt[, 1]
> Gc <- travelgc[, 1]
> traveldat <- cbind(travelhinc, travelinvnt, Invt, travelgc, Gc)
> traveldat <- as.data.frame(traveldat)

```

Now the model can be fitted.

```

> fit <- vglm(travelmode ~ Invt + Gc + travelhinc,
+             multinomial(parallel = FALSE ~ travelhinc, refLevel = 1),
+             xij = list(Invt ~ invnttrain + invntbus + invntcar,
+                       Gc ~ gctrain + gcbus + gccar),
+             form2 = ~ Invt + invnttrain + invntbus + invntcar +
+                       Gc + gctrain + gcbus + gccar + travelhinc,
+             data = traveldat, trace = TRUE)

```

```

VGLM   linear loop  1 : deviance = 501.46294
VGLM   linear loop  2 : deviance = 500.33384
VGLM   linear loop  3 : deviance = 500.33167
VGLM   linear loop  4 : deviance = 500.33167

```

```
> summary(fit)
```

Call:

```
vglm(formula = travelmode ~ Invt + Gc + travelhinc, family = multinomial(parallel = FALSE
```

```

travelhinc, refLevel = 1), data = travel.dat, form2 = ~Invt +
invtttrain + invtbus + invtcar + Gc + gctrain + gcbus + gccar +
travelhinc, xij = list(Invt ~ invtttrain + invtbus + invtcar,
Gc ~ gctrain + gcbus + gccar), trace = TRUE)

Pearson residuals:
      Min      1Q  Median      3Q      Max
log(mu[,2]/mu[,1]) -2.782 -0.5490 -0.2578  0.7185  3.302
log(mu[,3]/mu[,1]) -2.139 -0.4076 -0.2213 -0.1314  4.308
log(mu[,4]/mu[,1]) -2.005 -0.6589 -0.2444  0.8660  5.128

Coefficients:
      Estimate Std. Error z value Pr(>|z|)
(Intercept):1  3.5250538  0.6549818   5.382 7.37e-08 ***
(Intercept):2  2.2782935  0.7176672   3.175  0.00150 **
(Intercept):3  1.5334984  0.7065854   2.170  0.02998 *
Invt          -0.0031266  0.0009548  -3.275  0.00106 **
Gc            -0.0016225  0.0055279  -0.294  0.76913
travelhinc:1  -0.0569415  0.0124103  -4.588 4.47e-06 ***
travelhinc:2  -0.0355781  0.0131491  -2.706  0.00682 **
travelhinc:3  -0.0023652  0.0104474  -0.226  0.82090
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Names of linear predictors: log(mu[,2]/mu[,1]), log(mu[,3]/mu[,1]),
log(mu[,4]/mu[,1])

Residual deviance: 500.3317 on 622 degrees of freedom

Log-likelihood: -250.1658 on 622 degrees of freedom

Number of Fisher scoring iterations: 4

No Hauck-Donner effect found in any of the estimates

Reference group is level 1 of the response

> summary(travel.kat.id)

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

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 function values within tolerance limits

Coefficients :

	Estimate	Std. Error	z-value	Pr(> z )	
train:(intercept)	3.5250366	0.6549825	5.3819	7.371e-08	***
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car:(intercept)	1.5334957	0.7065856	2.1703	0.029985	*
inv	-0.0031266	0.0009548	-3.2746	0.001058	**
gc	-0.0016225	0.0055279	-0.2935	0.769130	
train:hinc	-0.0569409	0.0124103	-4.5882	4.471e-06	***
bus:hinc	-0.0355771	0.0131492	-2.7056	0.006817	**
car:hinc	-0.0023652	0.0104475	-0.2264	0.820898	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Log-Likelihood: -250.17

McFadden R<sup>2</sup>: 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	z-value	Pr(> z )
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
inv	-0.003126577	0.0009547997	-3.2745899	1.058154e-03
gc	-0.001622510	0.0055279076	-0.2935124	7.691305e-01
train:hinc	-0.056940856	0.0124103406	-4.5881784	4.471305e-06
bus:hinc	-0.035577091	0.0131492259	-2.7056415	6.817260e-03
car:hinc	-0.002365193	0.0104474472	-0.2263895	8.208985e-01

> summary(fit)\$coef3

	Estimate	Std. Error	z value	Pr(> z )
(Intercept):1	3.525053759	0.6549818022	5.381911	7.369921e-08
(Intercept):2	2.278293544	0.7176672465	3.174582	1.500525e-03
(Intercept):3	1.533498382	0.7065854109	2.170294	2.998455e-02
Inv	-0.003126583	0.0009547997	-3.274596	1.058133e-03
Gc	-0.001622528	0.0055279012	-0.293516	7.691278e-01
travelhinc:1	-0.056941460	0.0124102999	-4.588242	4.469941e-06
travelhinc:2	-0.035578093	0.0131491108	-2.705741	6.815211e-03
travelhinc:3	-0.002365198	0.0104474488	-0.226390	8.208981e-01