

Supplementary Material for:  
“Simple Ways to Interpret Effects in Modeling Ordinal Categorical Data”  
in *Statistica Neerlandica* (2018), by A. Agresti and C. Tarantola

The R-package **erer** of Sun (2016) has a function *ocME* that supplies marginal effects at the mean, using output from the *polr* function. This function works only for logit and probit link functions. We provide here an extension of it (called *ocMEM*) that handles also log-log and complementary log-log link functions. Furthermore we provide a function called *ocAME*, based on the *ocME* function for average marginal effects, that handles complementary log-log, logit, log-log and probit link functions.

One should first estimate the model using the function *polr* of the MASS library, and then pass the resulting object to *ocMEM()* or *ocAME()* to obtain the marginal effects estimates.

For the log-log link and cumulative log-log function we also provide the density and distribution functions (*p1gumbel* and *d1gumbel* for the log-log; *p1Gumbel* and *d1Gumbel* for the cumulative log-log ). These functions should be uploaded before using *ocME* and *ocAME*.

Table S1: R function to obtain the distribution function for the log-log link.

---

```
p1gumbel<-function (q, loc = 0, scale = 1, lower.tail = TRUE)
{
  q <- (q - loc)/scale
  p <- exp(-exp(-q))
  if (!lower.tail)
    1 - p
  else p
}
```

---

Table S2: R function to obtain the density function for the log-log link.

```
d1gumbel<-function (x, loc = 0, scale = 1, log = FALSE)
{
  x <- (x - loc)/scale
  d <- log(1/scale) - x - exp(-x)
  if (!log)
    exp(d)
  else d
}
```

Table S3: R function to obtain the distribution function for cumulative log-log link

```
p1Gumbel<-function (q, loc = 0, scale = 1, lower.tail = TRUE)
{
  q <- (q - loc)/scale
  p <- exp(-exp(q))
  if (lower.tail)
    1 - p
  else p
}
```

Table S4: R function to obtain the density function for the cumulative log-log link.

```
d1Gumbel<-function (x, loc = 0, scale = 1, log = FALSE)
{
  x <- -(x - loc)/scale
  d <- log(1/scale) - x - exp(-x)
  if (!log)
    exp(d)
  else d
}
```

Table S5: R function to obtain marginal effects at the mean.

---

```

ocMEM<-function (w, rev.dum = TRUE, digits = 3) {

# 1. Check inputs

  if (!inherits(w, "polr")) {
    stop("Need an ordered choice model from 'polr()'.\n")
  }
  if (w$method != "probit" & w$method != "logistic" & w$method != "loglog" & w$method != "cloglog") {
    stop("Need a probit or logit model.\n")
  }

# 2. Get data out

  lev <- w$lev
  J <- length(lev)
  x.name <- attr(x = w$terms, which = "term.labels")
  x2 <- w$model[, x.name]
  ww <- paste(~ 1, paste("+", x.name, collapse = " "), collapse = " ")
  x <- model.matrix(as.formula(ww), data = x2)[, -1]
  x.bar <- as.matrix(colMeans(x))
  b.est <- as.matrix(coef(w))
  K <- nrow(b.est)
  xb <- t(x.bar) %*% b.est
  z <- c(-10^6, w$zeta, 10^6)
  pfun <- switch(w$method, probit = pnorm, logistic = plogis, loglog = pigumbel, cloglog = p1Gumbel)
  dfun <- switch(w$method, probit = dnorm, logistic = dlogis, loglog = d1gumbel, cloglog = d1Gumbel)
  V2 <- vcov(w)
  V3 <- rbind(cbind(V2, 0, 0), 0, 0)
  ind <- c(1:K, nrow(V3) - 1, (K + 1):(K + J - 1), nrow(V3))
  V4 <- V3[ind, ]
  V5 <- V4[, ind]

# 3. Calculate marginal effects (ME) # 3.1 ME value

  f.xb<-matrix(0,nrow=1,ncol=2)
  f.xb[1,1]<- dfun(z[1] - xb) - dfun(z[2] - xb)
  f.xb[1,2]<- dfun(z[J] - xb) - dfun(z[J+1] - xb)
  me<- b.est %*% f.xb
}

```

```

colnames(me) <- paste("effect", lev[c(1,J)], sep = ".")  
  

# 3.2 ME standard error  
  

se <- matrix(0, nrow = K, ncol = 2)  

ind<-0  

for (j in c(1,J)) {  

  u1 <- c(z[j] - xb)  

  u2 <- c(z[j + 1] - xb)  

  if (w$method == "probit") {  

    s1 <- -u1  

    s2 <- -u2  

  }  

  else if (w$method == "logistic"){  

    s1 <- 1 - 2 * pfun(u1)  

    s2 <- 1 - 2 * pfun(u2)  

  }  

  else if (w$method == "loglog"){  

    s1 <- exp(-u1)-1  

    s2 <- exp(-u2)-1  

  }  

  else if (w$method == "cloglog"){  

    s1 <- (1-exp(u1))  

    s2 <- (1-exp(u2))  

  }  
  

  else {  

    stop("Specified link not available.")  

  }  

  d1 <- dfun(u1) * (diag(1, K, K) - s1 * (b.est %*% t(x.bar)))  

  d2 <- -1 * dfun(u2) * (diag(1, K, K) - s2 * (b.est %*%  

    t(x.bar)))  

  q1 <- dfun(u1) * s1 * b.est  

  q2 <- -1 * dfun(u2) * s2 * b.est  
  

  if(j==1)  

  {  

    dr <- cbind( d2, 0, q2)  

  }  

  else  

  {dr <- cbind( d1, q1, 0)

```

```

}

V <- V5[c(1:K, K + j, K + j + 1), c(1:K, K + j, K + j +
1)]
cova <- dr %*% V %*% t(dr)
ind<-ind+1
se[, ind] <- sqrt(diag(cova))
}

colnames(se) <- paste("SE", lev[c(1,J)], sep = ".")
rownames(se) <- colnames(x)

# 4. Revise ME and standard error for dummy variable

if (rev.dum) {
  for (k in 1:K) {ind<-0
    if (identical(sort(unique(x[, k])), c(0, 1))) {
      for (j in c(1,J)) {
        x.d1 <- x.bar
        x.d1[k, 1] <- 1
        x.d0 <- x.bar
        x.d0[k, 1] <- 0
        ua1 <- z[j] - t(x.d1) %*% b.est
        ub1 <- z[j + 1] - t(x.d1) %*% b.est
        ua0 <- z[j] - t(x.d0) %*% b.est
        ub0 <- z[j + 1] - t(x.d0) %*% b.est
        ind<-ind+1
        me[k, ind] <- pfun(ub1) - pfun(ua1) - (pfun(ub0) -
          pfun(ua0))
        d1 <- (dfun(ua1) - dfun(ub1)) %*% t(x.d1) -
          (dfun(ua0) - dfun(ub0)) %*% t(x.d0)
        q1 <- -dfun(ua1) + dfun(ua0)
        q2 <- dfun(ub1) - dfun(ub0)
        dr <- cbind(d1, q1, q2)
        V <- V5[c(1:K, K + j, K + j + 1), c(1:K, K +
          j, K + j + 1)]
        se[k, ind] <- sqrt(c(dr %*% V %*% t(dr)))
      }
    }
  }
}

#5. Output

```

```

z.value <- me/se
p.value <- 2 * (pnorm(abs(z.value),lower.tail = FALSE))
out <- list()
  out[[1]] <- round(cbind(effect = me[, 1], std.error = se[, 1], z.value = z.value[, 1], p.value = p.value[, 1]), digits)
  out[[2]] <- round(cbind(effect = me[, 2], std.error = se[, 2], z.value = z.value[, 2], p.value = p.value[, 2]), digits)
names(out) <- paste("ME", lev[c(1,J)], sep = ".")
result <- out
class(result) <- "ocMEM"
return(result)
}
-----
```

Table S6: R function to obtain avarage marginal effects.

---

```

ocAME<-function (w, rev.dum = TRUE, digits = 3) {

# 1. Check inputs

  if (!inherits(w, "polr")) {
    stop("Need an ordered choice model from 'polr()' .\n")
  }
  if (w$method != "probit" & w$method != "logistic" & w$method != "loglog" & w$method != "cloglog") {
    stop("Need a probit or logit model.\n")
  }

# 2. Get data out

  lev <- w$lev
  J <- length(lev)
  x.name <- attr(x = w$terms, which = "term.labels")
  x2 <- w$model[, x.name]
  ww <- paste(~ 1, paste("+", x.name, collapse = " "), collapse = " ")
  x <- model.matrix(as.formula(ww), data = x2)[, -1]
  b.est <- as.matrix(coef(w))
  K <- nrow(b.est)
  xb <- x %*% b.est
  z <- c(-10^6, w$zeta, 10^6)
  pfun <- switch(w$method, probit = pnorm, logistic = plogis, loglog = p1gumbel, cloglog = p1Gumbel)
  dfun <- switch(w$method, probit = dnorm, logistic = dlogis, loglog = d1gumbel, cloglog = d1Gumbel)
  V2 <- vcov(w)
  V3 <- rbind(cbind(V2, 0, 0), 0, 0)
  ind <- c(1:K, nrow(V3) - 1, (K + 1):(K + J - 1), nrow(V3))
  V4 <- V3[ind, ]
  V5 <- V4[, ind]

# 3. Calculate average marginal effects (AME)

# 3.1 AME value

  vec1<-rep(1, nrow(x))
  f.xb<-matrix(0,nrow(x),2)
  f.xb[,1]<-dfun(vec1%*% t(z[1]) - xb) -dfun(vec1%*% t(z[2]) - xb)
  f.xb[,2]<-dfun(vec1%*% t(z[J]) - xb) -dfun(vec1%*% t(z[J+1]) - xb)
}

```

```

f.xb1<-apply(f.xb,2,mean)
me <- b.est %*% matrix(data = f.xb1, nrow = 1)
colnames(me) <- paste("effect", lev[c(1,J)], sep = ".")  
  

# 3.2 AME standard error  
  

se <- matrix(0, nrow = K, ncol = 2)
ind<-0
for (j in c(1,J)) {
  temp<-matrix(0, nrow = K, ncol = length(b.est)+2)

  for (k in 1:length(xb)){
    u1 <- c(z[j] - xb[k])
    u2 <- c(z[j + 1] - xb[k])

    if (w$method == "probit") {
      s1 <- -u1
      s2 <- -u2
    }
    else if (w$method == "logistic"){
      s1 <- 1 - 2 * pfun(u1)
      s2 <- 1 - 2 * pfun(u2)
    }
    else if (w$method == "loglog"){
      s1 <- exp(-u1)-1
      s2 <- exp(-u2)-1
    }
    else if (w$method == "cloglog"){
      s1 <- (1-exp(u1))
      s2 <- (1-exp(u2))
    }
    else {
      stop("Specified link not available.")
    }
    d1 <- dfun(u1) * (diag(1, K, K) - s1 * (b.est %*% t(x[,])))
    d2 <- -1 * dfun(u2) * (diag(1, K, K) - s2 * (b.est %*%
      t(x[,])))
    q1 <- dfun(u1) * s1 * b.est
    q2 <- -1 * dfun(u2) * s2 * b.est
    if(j==1)
    {
  
```

```

drtemp <- cbind( d2, 0, q2)
}
else
{drtemp <- cbind( d1, q1, 0)
}
temp<-temp+drtemp
}
dr<-temp/length(xb)
V <- V5[c(1:K, K + j, K + j + 1), c(1:K, K + j, K + j +
1)]
cova <- dr %*% V %*% t(dr)
ind<-ind+1
se[, ind] <- sqrt(diag(cova))
}

colnames(se) <- paste("SE", lev[c(1,J)], sep = ".")
rownames(se) <- colnames(x)

```

# 4. Revise AME and standard error for dummy variable.

```

if (rev.dum) {
  for (k in 1:K) {ind<-0
    if (identical(sort(unique(x[, k])), c(0, 1))) {
      for (j in c(1,J)) {
        x.d1 <- x
        x.d1[,k] <- 1
        x.d0 <- x
        x.d0[,k] <- 0
        ua1 <-vec1*z[j] - x.d1 %*% b.est
        ub1 <-vec1*z[j + 1] - x.d1 %*% b.est
        ua0 <-vec1*z[j] - x.d0 %*% b.est
        ub0 <-vec1*z[j + 1] - x.d0 %*% b.est
        ind<-ind+1
        me[k, ind] <- mean(pfun(ub1) - pfun(ua1) - (pfun(ub0) -
          pfun(ua0)))
        temp<-0
      for (g in 1: nrow(x)){
        d1 <- (dfun(ua1[g]) - dfun(ub1[g])) %*% t(x.d1[g,]) -
          (dfun(ua0[g]) - dfun(ub0[g])) %*% t(x.d0[g,])
        q1 <- -dfun(ua1[g]) + dfun(ua0[g])
        q2 <- dfun(ub1[g]) - dfun(ub0[g])
        drtemp <- cbind(d1, q1, q2)
      }
    }
  }
}

```

```

        temp<-temp+drtemp
    }
    dr<-temp/nrow(x)
    V <- V5[c(1:K, K + j, K + j + 1), c(1:K, K +
        j, K + j + 1)]
    se[k, ind] <- sqrt(c(dr %*% V %*% t(dr)))
}
}
}
}

# 5. Output

z.value <- me/se
p.value <- 2 * (pnorm(abs(z.value),lower.tail = FALSE))
out <- list()
out[[1]] <- round(cbind(effect = me[, 1], std.error = se[, 1], z.value = z.value[, 1], p.value = p.value[, 1]), digits)
out[[2]] <- round(cbind(effect = me[, 2], std.error = se[, 2], z.value = z.value[, 2], p.value = p.value[, 2]), digits)
names(out) <- paste("ME", lev[c(1,J)], sep = ".")
result<-out
class(result) <- "ocAME"
return(result)
}
-----
```