

```

glm.mq.poisson<-
function(x,y,offs=rep(1,nrow(x)),case.weights=rep(1,nrow(x)),maxit=100,acc=
1e-04,
var.weights=rep(1,nrow(x)),weights.x=FALSE,q=0.5,k=1.6)
{
#This is the robust glm for count data by Cantoni

      glm.rob.poisson <-
function(X,y,weights.on.x=FALSE,chuber=1.345,offset=offs)
      {
# Preliminary definitions
      mytol <- .Machine$double.eps^.25
      nb <- length(y)
      assign("nb",nb)

      basepsi <- function(x)
      {
          x*pmin(1,chuber/abs(x))
      }
      assign("basepsi",basepsi)

      basepsiprime <- function(x)
      {
          1*(abs(x)<chuber)
      }
      assign("basepsiprime",basepsiprime)

# Initializing....
      beta.old <-
as.vector(glm(y~X-1,family=poisson,offset=log(offset))$coeff)
      Xwith <- X

      eta <- Xwith%*%beta.old
      probab <- offset*exp(eta)
      mu <- probab
      V <- mu
      deriv.mu <- offset*exp(eta)
      r.stand <- (y-mu)/sqrt(V)
      ifelse (weights.on.x,w.x <- sqrt(1-hat(X)),w.x <-
rep(1,length=nb))
      assign("w.x",w.x)
      assign("Xwith",Xwith)
      assign("y",y)

# pmin() on the argument of pbinom() is necessary, due to the
# bad behavior of pbinom() when evaluated at values greater than size

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#
# pmax() is used to avoid the warning messages due to the generation of NA
# in the ifelse procedure

g.objective <- function(beta)
{
  eta <- Xwith%*%beta
  probab <- offset*exp(eta)
  mu <- probab
  V <- probab
  r.stand <- (y-mu)/sqrt(V)
  deriv.mu <- offset*exp(eta)
  jinf <- floor(mu-chuber*sqrt(V))
  jsup <- floor(mu+chuber*sqrt(V))

  if(chuber==Inf)
  {
    esp.cond <- rep(1,nb)
  }
  if(chuber!=Inf)
  {
    esp.cond <- -chuber*ppois(jinf,mu) +
chuber*(1-ppois(jsup,mu)) +
mu/sqrt(V)*(ppois(jinf,mu)-ppois(jinf-1,mu)
-(ppois(jsup,mu) -
ppois(jsup-1,mu)))
  }
  a.const <- apply(Xwith*as.vector(1/nb/
sqrt(V)*w.x*esp.cond*deriv.mu), 2,sum)
  apply(Xwith*as.vector(1/nb/
sqrt(V)*w.x*basepsi(r.stand)*deriv.mu),2,sum)-a.const
}
assign("g.objective",g.objective)

grad.g <- function(beta)
{
  delta <- .Machine$double.eps^.5
  Ident <- diag(1,length(beta))
  1/delta*(apply(beta+delta*Ident,2,g.objective)-
as.vector(g.objective(beta)))
}
tmp.times<-0

# Main
repeat
{tmp.times<-tmp.times+1
  g.old <- g.objective(beta.old)
  grad.g.old <- grad.g(beta.old)
  csi <- solve(grad.g.old,-g.old)
  beta.new <- as.vector(beta.old+csi)
}

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                                if(abs(max(beta.old-beta.new))/abs(max(beta.old)) <
mytol) break
                                if (tmp.times>100)break
                                beta.old <- beta.new
                                NULL
                                }

                                eta <- Xwith%*%beta.old
                                fit <- offset*exp(eta)
                                list(coef=beta.old,fitted.values=fit)
                                }

```

#Stopping rule

```

    irls.delta <- function(old, new) abs(max(old-new))/abs(max(old))
    if (qr(x)$rank < ncol(x))
    stop("X matrix is singular")
    if (length(case.weights) != nrow(x))
    stop("Length of case.weights must equal number of observations")
    if (any(case.weights < 0))
    stop("Negative case.weights are not allowed")
    n<-length(case.weights)
    ifelse (weights.x,w.x <- sqrt(1-hat(x)),w.x <- rep(1,length=n))
    assign("w.x",w.x)

```

#We fit the glm.rob for computing the starting values

```

    temp.rob <-glm.rob.poisson
(X=x,y=y,weights.on.x=weights.x,cluster=k,offset=offs)
    resid.init <- y-temp.rob$fitted.values
    fit.init <- temp.rob$fitted.values
    phi.init<-1
    done <- FALSE
    conv <- NULL
    qest <- matrix(0, nrow = ncol(x), ncol = length(q))
    qfit <- matrix(0, nrow = nrow(x), ncol = length(q))
    qres <- matrix(0, nrow = nrow(x), ncol = length(q))
    qvar <- matrix(0, nrow = ncol(x), ncol = length(q))
    qphi<-NULL
    for(i in 1:length(q)) {

```

#We define the starting values

```

        resid <- resid.init
        fit<-fit.init
        phi<-phi.init
        a.j<- case.weights
        w<-case.weights
        coef<-temp.rob$coef

        for (iiter in 1:maxit) {

            resid.old <- resid

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coef.old<-coef

# We define the probability mu=t*exp(xb)
  probab<-fit
  mu <- probab
  deriv.mu <- mu
#We define the variance
  V <- phi*probab

#We define the scale
  scale<- c(sqrt(V))

#We standardize the residuals
  r.stand <- (y-mu)/sqrt(V)

#we compute i1 and i2
  jinf <- floor(mu-k*sqrt(V))
  jsup <- floor(mu+k*sqrt(V))

#We compute the values of a_j(b)
  if(k==Inf)
  {
    a.j <- rep(1,n)
  }
  if(k!=Inf)
  {
    a.j <- (-k)*ppois(jinf,mu) + k*(1-
ppois(jsup,mu)) +
    mu/sqrt(V)*(ppois(jinf,mu)-ppois(jinf-1,mu)
-(ppois(jsup,mu) -
    ppois(jsup-1,mu))) }

  a.j<-2*a.j*(q[i]*(r.stand>0)+(1-q[i])*(r.stand<=0))

#we define a part of w_j
  w<-diag(c(mu)/scale)*diag(c(w.x))

#we compute psi_q(res)
  tmp <- psi.huber((resid)/scale,k=k) *
case.weights*((resid)/scale)
  tmp1 <- 2 * (1 - q[i]) * tmp
  tmp1[resid > 0] <- 2 * q[i] * tmp[resid > 0]
  tmp <- tmp1

#we compute psi_q(r )-E(psi_q(r ))
  A<-(tmp-a.j)

```

```

        if(k==Inf)
        {
            esp.carre.cond <- rep(1,n)
        }
        if(k!=Inf)
    {
        esp.carre.cond <-k*(ppois(jinf,mu)-
ppois(jinf-1,mu) +(ppois(jsup,mu) - ppois(jsup-1,mu)))+(mu^2/
V^(3/2))*(ppois(jinf-1,mu)-ppois(jinf-2,mu)-(ppois(jinf,mu)-
ppois(jinf-1,mu))-(ppois(jsup-1,mu) - ppois(jsup-2,mu))+(ppois(jsup,mu) -
ppois(jsup-1,mu)))+(mu/V^(3/2))*(ppois(jsup-1,mu) - ppois(jinf,mu))
        }
        b.j<-2*esp.carre.cond*(q[i]*(r.stand>0)+(1-
q[i])*(r.stand<=0))
        B<-diag(c(V*b.j))

#We estimate betas
        temp <- coef+solve(t(x)%*%W%*%B%*%x)%*%t(x)%*%W%*%A

        coef <- temp
        eta <- x%*%coef
        fit <- offs*exp(eta)
        resid <- y-fit

        convi <- irls.delta(coef.old, coef)
        conv <- c(conv, convi)
        done <- (convi <= acc)
        if (done)
            break
    }
    if (!done)
        warning(paste("MQPoisson failed to converge in", maxit,
"steps at q = ", q[i]))

# Asymptotic estimated variance of the robust estimator

        probab<-fit
        mu <- probab
        deriv.mu<-mu

#We define the variance
        V <- phi*probab
        r.stand <- (y-mu)/sqrt(V)
        scale<- c(sqrt(V))

        jinf <- floor(mu-k*sqrt(V))
        jsup <- floor(mu+k*sqrt(V))

        if(k==Inf)
        {

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        esp.cond <- rep(1,n)
    }
    else
    {
        esp.cond <- -k*ppois(jinf,mu) + k*(1-
        ppois(jsup,mu)) + mu/sqrt(V)*(ppois(jinf,mu)-ppois(jinf-1,mu) -
        (ppois(jsup,mu) - ppois(jsup-1,mu)))
    }

    esp.cond<-2*esp.cond*(q[i]*(r.stand>0)+(1-
    q[i]))*(r.stand<=0))
    a.const <- apply(x*as.vector(1/n/
    sqrt(V)*w.x*esp.cond*deriv.mu), 2,sum)

    if(k==Inf)
    {
        esp.carre.cond <- 1
    }
    else
    {
        esp.carre.cond <- k^2*(ppois(jinf,mu)+1-
        ppois(jsup,mu))+1/V*(mu^2*(2*ppois(jinf-1,mu)-ppois(jinf-2,mu)-
        ppois(jinf,mu)-2*ppois(jsup-1,mu)+ppois(jsup-2,mu)+ppois(jsup,mu))
        +mu*(ppois(jsup-1,mu)-ppois(jinf-1,mu)))
    }

    esp.carre.cond<-4*esp.carre.cond*(q[i]*(r.stand>0)+(1-
    q[i]))*(r.stand<=0))^2
    matQaux <- as.vector(esp.carre.cond/V*w.x^2*deriv.mu^2)
    matQ1 <- (1/n)*t(x)%*(matQaux*x)
    matQ2 <- a.const%*t(a.const)
    matQ <- matQ1-matQ2

    if(k==Inf)
    {
        esp.psi.score <- 1/sqrt(V)
    }
    else
    {
        esp.psi.score <- k*(ppois(jinf,mu)-ppois(jinf-1,mu)
        +(ppois(jsup,mu) - ppois(jsup-1,mu)))+(mu^2/V^(3/2))*(ppois(jinf-1,mu)-
        ppois(jinf-2,mu)-(ppois(jinf,mu)-ppois(jinf-1,mu))-(ppois(jsup-1,mu) -
        ppois(jsup-2,mu))+(ppois(jsup,mu) - ppois(jsup-1,mu)))+(mu/
        V^(3/2))*(ppois(jsup-1,mu) - ppois(jinf,mu))
    }

    esp.psi.score<-2*esp.psi.score*(q[i]*(r.stand>0)+(1-
    q[i]))*(r.stand<=0))
    matMaux <- as.vector(esp.psi.score/sqrt(V)*w.x*deriv.mu^2)
    matM <- 1/n*t(x)%*(matMaux*x)
    matMinv <- solve(matM)

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as.var <- 1/n*matMinv%*%matQ%*%matMinv

      qest[, i] <- coef
      qfit[, i] <- fit
      qres[,i] <- y-fit
      qvar[,i]<-as.numeric(round(diag(as.var),4))
    }
    list(fitted.values=qfit, var.beta=qvar,residuals=qres, q.values=q,
coefficients=qest,matQ=matQ,matM=matM)
  }

```