

Medical Care - Comparison of different count data models

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To fit the count data from the medcare data set Poisson Models, Negative Binomial Models, Zero Inflated Models and Hurdle Models will be fitted. To fit a Negative Binomial Generalized Linear Model we will use the function "glm.nb" from the "MASS" package. The package "pscl" will be needed to fit Zero Inflated Models and Hurdle Models.

```
> library(catdata)
> data(medcare)
> library(MASS)
> library(pscl)
```

The data from the medcare data set are reduced to males with responses smaller than 30.

```
> medmale <- medcare[medcare$male==1,]
> medmale <-medmale[medmale$ofp<=30,]
```

Now we split the data 50 times randomly into a training and a test data set and fit a Poisson Modell, an Negative Binomial Modell, two Zero Inflated Model and two Hurdle Models (each time one model with only the Intercept for the mixture, one with predictors in the mixture). As a prediction we take the Median of the predicted distributions.

```
> set.seed(5)
> subs<-1:nrow(medmale)
> reps <- 50
> squerror1<-squerror3<-squerror4<-squerror5<-squerror6<-squerror7<-c()
> abseerror1<-abseerror3<-abseerror4<-abseerror5<-abseerror6<-abseerror7<-c()
> for(i in 1:reps){
+ learn <- sample(subs,600)
+ test <- subs[-learn]
+
+ #####
+ med=glm(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+ family=poisson,data=medmale[learn,])
+ l <- predict(med, newdata=medmale[test,], type="response")
+
+ a<-rep(0,length(medmale[test,1]))
+}
```

```

+ for(j in 1:length(medmale[test,1])){
+ while(ppois(a[j],lambda=1[j])<0.5){
+ a[j] <- a[j] + 1
+ }
+ }
+ diffs <- a - medmale[test,1]
+ squerror1[i]<-mean(diffs^2)
+ abserror1[i]<-mean(abs(diffs))
+
+ #####
+ med3=glm.nb(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+ data=medmale[learn,])
+ l <- predict(med3, newdata=medmale[test,], type="response")
+ a<-rep(0,length(medmale[test,1]))
+
+ for(j in 1:length(medmale[test,1])){
+
+ while((pnbinom(a[j],mu=1[j],size=med3$theta)<0.5){
+ a[j] <- a[j]+1
+ }
+ }
+
+ diffs <- a - medmale[test,1]
+ squerror3[i]<-mean(diffs^2)
+ abserror3[i]<-mean(abs(diffs))
+
+ #####
+ med4=zeroinfl(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school|1,
+ data=medmale[learn,])
+ pii <- 1-predict(med4, newdata=medmale[test,], type="zero")
+ mui <- predict(med4, newdata=medmale[test,], type="count")
+ a<-rep(0,length(medmale[test,1]))
+
+ for(j in 1:length(medmale[test,1])){
+ cdis <- 0
+
+ while(cdis<0.5){
+
+ cdis <- cdis + pii[j]*exp(-mui[j])*((mui[j]^a[j])/factorial(a[j])) + (1-pii[j])*
+ I(a[j]==0)
+ a[j] <- a[j]+1
+ }
+ a[j]<-a[j] - 1
+ }
+ diffs <- a - medmale[test,1]
+ squerror4[i]<-mean(diffs^2)
+ abserror4[i]<-mean(abs(diffs))
+
+ #####
+ med5=zeroinfl(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,

```

```

+           data=medmale[learn,])
+
+ pii <- 1-predict(med5, newdata=medmale[test,], type="zero")
+ mui <- predict(med5, newdata=medmale[test,], type="count")
+ a<-rep(0,length(medmale[test,1]))
+
+ for(j in 1:length(medmale[test,1])){
+   cdis <- 0
+
+   while(cdis<0.5){
+
+     cdis <- cdis + pii[j]*exp(-mui[j])*((mui[j]^a[j])/factorial(a[j])) + (1-pii[j])*
+       I(a[j]==0)
+     a[j] <- a[j]+1
+   }
+   a[j]<-a[j] - 1
+ }
+
+ diffs <- a - medmale[test,1]
+
+ squerror5[i]<-mean(diffs^2)
+ abserror5[i]<-mean(abs(diffs))
+
+ #####
+ med6=hurdle(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school|1,
+   data=medmale[learn,])
+ mui <- predict(med6, newdata=medmale[test,], type="count")
+ gammai <- predict(med6, newdata=medmale[test,], type="zero")
+ pii2<-1-gammai*(1-exp(-mui))
+
+
+ fa<-function(z,a){
+   ((z^a)/factorial(a))*exp(-z) }
+
+ a<-rep(0,length(medmale[test,1]))
+
+ for(j in 1:length(medmale[test,1])){
+   cdis <- pii2[j]
+
+   if(cdis<0.5){
+     while(cdis<0.5){
+       #while((1-cumdist)>eps){
+       cdis <- cdis + fa(z=mui[j],a=(a[j]+1))*gammai[j]
+       a[j]<-a[j]+1
+       #a <- a+1
+     }
+   }
+   else{a[j]<-0}
+ }
+
+

```

```

+ diffs <- a - medmale[test,1]
+
+ squerror6[i]<-mean(diffs^2)
+ abserror6[i]<-mean(abs(diffs))
+
+ #####
+ med7=hurdle(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+             data=medmale[learn,])
+ mui <- predict(med7, newdata=medmale[test,], type="count")
+ gammai <- predict(med7, newdata=medmale[test,], type="zero")
+ pii2<-1-gammai*(1-exp(-mui))
+
+ fa<-function(z,a){
+ ((z^a)/factorial(a))*exp(-z) }
+
+ a<-rep(0,length(medmale[test,1]))
+
+ for(j in 1:length(medmale[test,1])){
+ cdis <- pii2[j]
+
+ if(cdis<0.5){
+ while(cdis<0.5){
+ #while((1-cumdists)>eps){
+ cdis <- cdis + fa(z=mui[j],a=(a[j]+1))*gammai[j]
+ a[j]<-a[j]+1
+ }
+ }
+ else{a[j]<-0}
+ }
+
+ diffs <- a - medmale[test,1]
+
+ squerror7[i]<-mean(diffs^2)
+ abserror7[i]<-mean(abs(diffs))
+
+ }

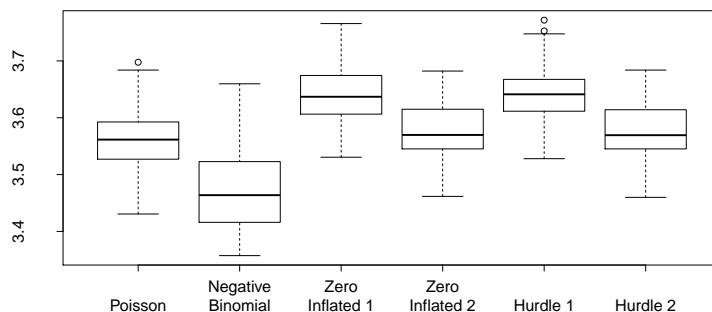
```

Here the absolut 50 prediction errors of the respective models are plotted.

```

> par(mgp=c(0,3,0))
> boxplot(abserror1,abserror3,abserror4,abserror5,abserror6,abserror7,
+ names=c("Poisson","Negative\nBinomial","Zero\nInflated 1","Zero\nInflated 2",
+         "Hurdle 1","Hurdle 2"),cex=1.3,cex.axis=1.6)

```



Now the ranked probability score is computed. Here the difference between the predicted and the empirical cumulative distribution function is measured.

```
> library(catdata)
> data(medcare)
> #attach(medcare)
>
> library(MASS)
> library(psc1)
> medmale <- medcare[medcare$male==1,]
> medmale <- medmale[medmale$ofp<=30,]
> set.seed(5)
> subs<-1:nrow(medmale)
> reps <- 50
> eps <- 1e-5
> lrps <- lrps3 <- lrps4 <- lrps5 <- lrps6 <- lrps7 <- rep(0,reps)
> for(i in 1:reps){
+   learn <- sample(subs,600)
+   test <- subs[-learn]
+
+   med=glm(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+           family=poisson,data=medmale[learn,])
+
+   l <- predict(med, newdata=medmale[test,], type="response")
+
+   for(j in 1:length(medmale[test,1])){
+     for(a in 0:100){
+       #while((1-ppois(a,lambda=l[j]))>eps){
+       lrps[i] <- lrps[i] + (ppois(a,lambda=l[j]) - I(medmale[test,1][j] <= a))^2
+       #a <- a+1
+     }
+   }
+
+   med3=glm.nb(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+               data=medmale[learn,])
```

```

+ l <- predict(med3, newdata=medmale[test,], type="response")
+
+ for(j in 1:length(medmale[test,1])){
+ #a <- 0
+ for (a in 0:100){
+ #while((1-pnbinom(a,mu=l[j],size=med3$theta))>eps){
+ lrps3[i] <- lrps3[i] + (pnbinom(a,mu=l[j],size=med3$theta) -
+ I(medmale[test,1][j] <= a))^2
+ #a <- a+1
+ }
+ }
+
+
+ med4=zeroinfl(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school|1,
+ data=medmale[learn,])
+ pii <- 1-predict(med4, newdata=medmale[test,], type="zero")
+ mui <- predict(med4, newdata=medmale[test,], type="count")
+
+
+ for(j in 1:length(medmale[test,1])){
+ a <- 0
+ cumdist<-(1-pii[j])
+ #cumdist <- 0
+ for (a in 0:100){
+ #while((1-cumdist)>eps){
+ cumdist <- cumdist + pii[j]*exp(-mui[j])*(mui[j]^a)*(factorial(a)^(-1))
+ lrps4[i] <- lrps4[i] + as.numeric((cumdist - I(medmale[test,1][j] <= a))^2)
+ #a <- a+1
+ }
+ }
+
+
+ med5=zeroinfl(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+ data=medmale[learn,])
+ pii <- 1-predict(med5, newdata=medmale[test,], type="zero")
+ mui <- predict(med5, newdata=medmale[test,], type="count")
+
+ for(j in 1:length(medmale[test,1])){
+ a <- 0
+ cumdist<-(1-pii[j])
+ #cumdist <- 0
+ for (a in 0:100){
+ #while((1-cumdist)>eps){
+ cumdist <- cumdist + pii[j]*exp(-mui[j])*(mui[j]^a)*(factorial(a)^(-1))
+ lrps5[i] <- lrps5[i] + as.numeric((cumdist - I(medmale[test,1][j] <= a))^2)
+ #a <- a+1
+ }
+ }
+
+
+

```

```

+ med6=hurdle(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school|1,
+             data=medmale[learn,])
+ mui <- predict(med6, newdata=medmale[test,], type="count")
+ gammai <- predict(med6, newdata=medmale[test,], type="zero")
+ pii2<-1-gammai*(1-exp(-mui))
+
+ #rr<-coef(med6,model="zero")
+
+ #pii<-exp(rr)/(1+exp(rr))
+
+ #ga<-(1-pii)/(1-exp(-mui))
+
+ fa<-function(z,a){
+ ((z^a)/factorial(a))*exp(-z) }
+
+ for(j in 1:length(medmale[test,1])){
+   cumdist <- pii2[j]
+
+   for(a in 1:100){
+     #while((1-cumdist)>eps){
+     cumdist <- cumdist + fa(z=mui[j],a=a)*gammai[j]
+     lrps6[i] <- lrps6[i] + as.numeric((cumdist - I(medmale[test,1][j] <= a))^2)
+
+     #a <- a+1
+   }
+ }
+
+
+ med7=hurdle(ofp ~ hosp+healthpoor+healthexcellent+numchron+age+married+school,
+             data=medmale[learn,])
+ mui <- predict(med7, newdata=medmale[test,], type="count")
+ gammai <- predict(med7, newdata=medmale[test,], type="zero")
+ pii2<-1-gammai*(1-exp(-mui))
+
+ fa<-function(z,a){
+ ((z^a)/factorial(a))*exp(-z) }
+
+ for(j in 1:length(medmale[test,1])){
+   cumdist <- pii2[j]
+
+   for(a in 1:100){
+     #while((1-cumdist)>eps){
+     cumdist <- cumdist + fa(z=mui[j],a=a)*gammai[j]
+     lrps7[i] <- lrps7[i] + as.numeric((cumdist - I(medmale[test,1][j] <= a))^2)
+
+     #a <- a+1
+   }
+ }
+
+

```

```

+ }
> lrps <-lrps/length(medmale[test,1])
> lrps3 <-lrps3/length(medmale[test,1])
> lrps4 <-lrps4/length(medmale[test,1])
> lrps5 <-lrps5/length(medmale[test,1])
> lrps6 <-lrps6/length(medmale[test,1])
> lrps7 <-lrps7/length(medmale[test,1])

```

Now the respective ranked probability scores are plotted.

```

> par(mgp=c(0,3,0))
> boxplot(lrps,lrps3,lrps4,lrps5,lrps6,lrps7,
+ names=c("Poisson","Negative\nBinomial","Zero\nInflated 1","Zero\nInflated 2",
+         "Hurdle 1","Hurdle 2"),cex=1.3,cex.axis=1.6)
>

```

