

# 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 = l[j]) < 0.5) {
```

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+         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 = l[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) {
+         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]
+   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) {
+          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)
+    aberror7[i] <- mean(abs(diffs))
+  }

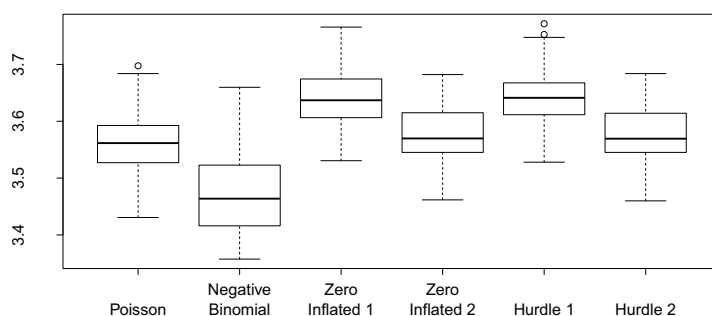
```

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

```

> par(mgp = c(0, 3, 0))
> boxplot(aberror1, aberror3, aberror4, aberror5,
+   aberror6, aberror7, 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)
> 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-05
> 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) {
+       lrps[i] <- lrps[i] + (ppois(a, lambda = l[j]) -
+         I(medmale[test, 1][j] <= a))^2
+     }
+   }
+   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])) {
+     for (a in 0:100) {
+       lrps3[i] <- lrps3[i] + (pnbinom(a,
+         mu = l[j], size = med3$theta) -
+         I(medmale[test, 1][j] <= a))^2
+     }
+   }
+   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])
+     for (a in 0:100) {
+       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)
+     }
+   }
+   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])
+     for (a in 0:100) {
+       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)
+     }
+   }
+   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)
+   }
+   for (j in 1:length(medmale[test, 1])) {
+     cumdist <- pii2[j]
+     for (a in 1:100) {
+       cumdist <- cumdist + fa(z = mui[j],
+       a = a) * gammai[j]
+       lrps6[i] <- lrps6[i] + as.numeric((cumdist -
+       I(medmale[test, 1][j] <= a))^2)
+     }
+   }
+   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) {

```

```

+         cumdist <- cumdist + fa(z = mui[j],
+           a = a) * gammai[j]
+         lrps7[i] <- lrps7[i] + as.numeric((cumdist -
+           I(medmale[test, 1][j] <= a))^2)
+       }
+     }
+ }
> 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)

```

