

Local Poisson Regression

*

Andreu Meca, Jan Leyva, Victor Duque, Geraldo Gariza

3/22/2021

1. Bandwidth choice for the local Poisson regression

Modify the functions `h.cv.sm.binomial` and `loglik.CV` to obtain a bandwidth choice method for the local Poisson regression based on the leave-one-out cross-validation (loo-CV) estimation of the expected likelihood of an independent observation. Remember that the loo-CV estimation of the expected log-likelihood of an independent observation, when using `h` as bandwidth, is

$$l_{cv}(h) = \frac{1}{n} \sum_{i=1}^n \log(\hat{Pr}_h^{-i}(Y = y_i | X = x_i))$$

Where $\hat{Pr}_h^{-i}(Y = y_i | X = x_i)$ is an estimation of

$$Pr(Y = y_i | X = x_i) = e^{-\lambda_i} \frac{\lambda_i^{y_i}}{y_i!},$$

should be estimated by maximum local likelihood using `h` as bandwidth (for instance, using the function `sm.poisson` from the R package `sm`).

```
# Changing the loglik function with sm.poisson

data <- read.csv("HDI.2017.subset.csv", sep = ";", dec = ',')

loglik.CV_poisson <- function(x,y,h){
  n <- length(x)-1
  pred <- sapply(1:n,
    function(i,x,y,h){
      sm.poisson(x=x[-i],y=y[-i],h=h,eval.points=x[i],
        display="none")$estimate
    }, x,y,h)
  return((sum(log(exp(-pred)*(pred^y)/factorial(y))))/n)
}

# Changing the h.cv with method= CV_poisson

h.cv.sm_poisson <- function(x,y,rg.h=NULL,l.h=10,
  method=loglik.CV_poisson){
```

```

cv.h <- numeric(l.h)
if (is.null(rg.h)){
  hh <- c(h.select(x,y,method="cv"),
          h.select(x,y,method="aicc"))#,hcv(x,y))
  rg.h <- range(hh)*c(1/1.1, 1.5)
}
i <- 0
gr.h <- exp( seq(log(rg.h[1]), log(rg.h[2]), l=1.h) )
for (h in gr.h){
  i <- i+1
  cv.h[i] <- method(x,y,h)
}
return(list(h = gr.h,
            cv.h = cv.h,
            h.cv = gr.h[which.max(cv.h)]))
}

```

2. Local Poisson regression for Country Development Data

Consider the country development dataset (file HDI.2017.subset.csv) containing information on development indicators measured in 179 countries (Source: Human DevelopmentData(1990-2017)). Variable le.fm always takes non-negative values. Define le.fm.r as the rounded value of le.fm: le.fm.r <- round(le.fm) Fit a local Poisson regression modeling le.fm.r as a function of Life.expec. Use sm.poisson from the R package sm with the bandwidth obtained by loo-CV.

```
library(sm)
```

```
## Package 'sm', version 2.2-5.6: type help(sm) for summary information
```

```
data <- read.csv("HDI.2017.subset.csv", sep = ";", dec = ',')
```

```
le.fm.r <- round(data$le.fm)
```

```

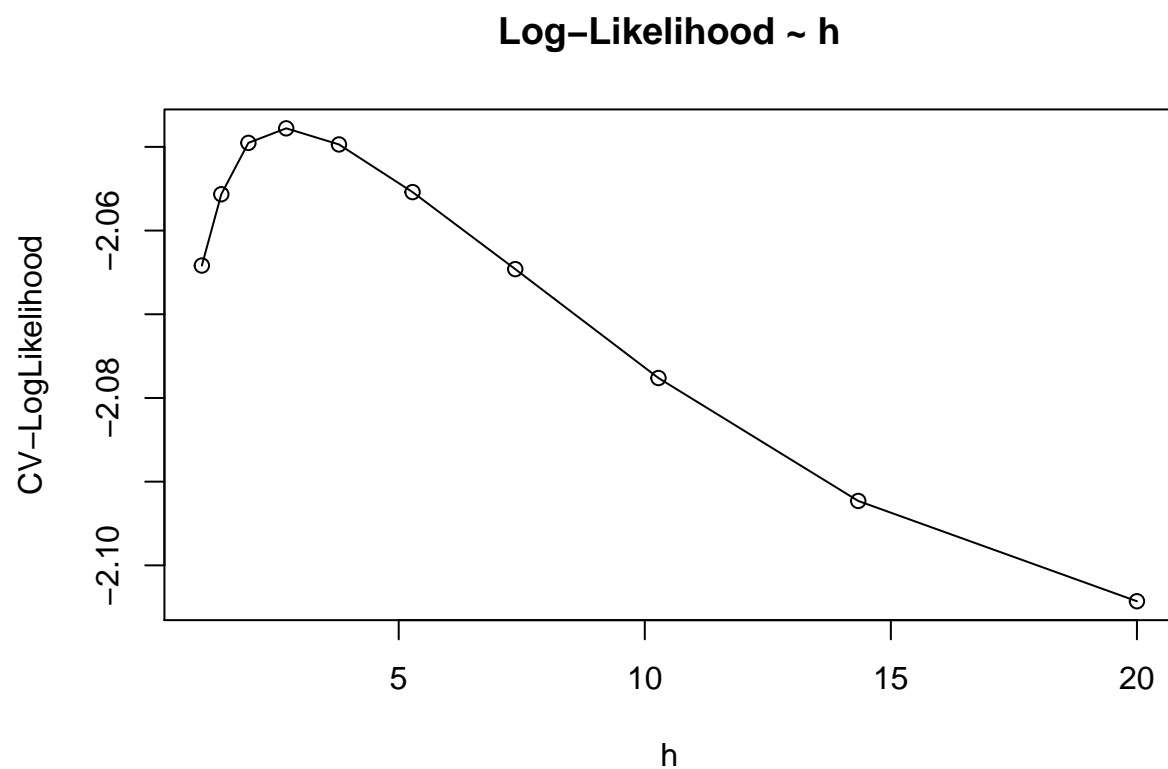
## 2. Local Poisson regression for Country Development Data -----
h.CV.loglik <- h.cv.sm_poisson(y = le.fm.r, x = data$Life.expec, rg.h=c(1,20),
                              method=loglik.CV_poisson)

```

```

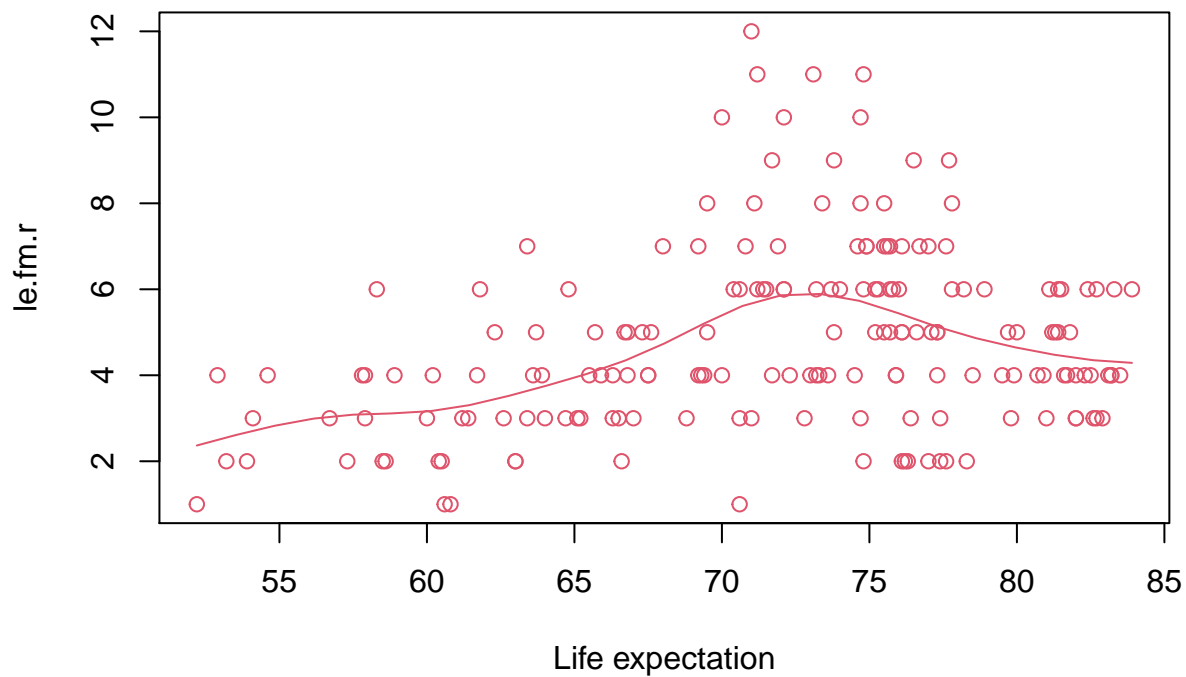
plot(h.CV.loglik$h, h.CV.loglik$cv.h, main="Log-Likelihood ~ h",
     xlab="h", ylab="CV-LogLikelihood")
lines(h.CV.loglik$h, h.CV.loglik$cv.h)

```



We have calculated the maximum likelihood of the h by leave one out cross-validation, in order to obtain the best value of h : 2.7144176.

```
sm.poisson(y = le.fm.r, x = data$Life.expec, h = h.CV.loglik$h.cv, xlab = "Life expectation")
```



After that, We have used the \hat{h} to represent the Local Poisson Regression with the function *sm.poisson* of the package *sm*. We can observe with values around of 65 and 80 **life expectation** We get the highest values of the response variable **le.fm.r**.