

# Smoothing and regression splines

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## Load data

1. Consider the nonparametric regression of `cnt` as a function of `instant`. Estimate the regression function  $m(\text{instant})$  of `cnt` as a function of `instant` using a cubic regression spline estimated with the R function `smooth.splines` and choosing the smoothing parameter by Generalized Cross Validation.

```
sm.sp <- smooth.spline(x = instant, y = cnt,
                       cv = FALSE, all.knots = FALSE)
sm.sp

## Call:
## smooth.spline(x = instant, y = cnt, cv = FALSE, all.knots = FALSE)
##
## Smoothing Parameter spar= 0.24802 lambda= 1.005038e-07 (11 iterations)
## Equivalent Degrees of Freedom (Df): 93.34091
## Penalized Criterion (RSS): 470358491
## GCV: 845608.3

# Number of knots
sm.sp$fit$nk-2

## [1] 134
```

a) Which is the value of the chosen penalty parameter  $\lambda$ ?

The value of  $\lambda$  is 1.005038e-07.

b) Which is the corresponding equivalent number of degrees of freedom  $df$ ?

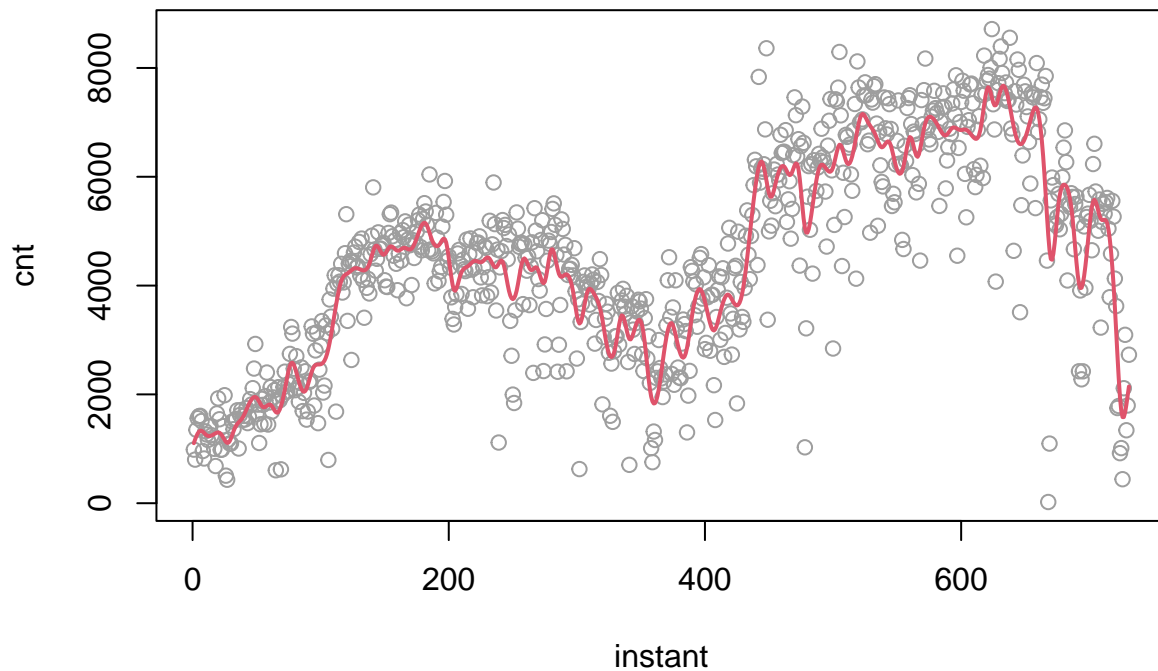
The corresponding equivalent number of degrees of freedom ( $df$ ) is 93.34091.

c) How many knots have been used?

We have used 134 knots.

d) Give a graphic with the scatter plot and the estimated regression function  $\hat{m}(\text{instant})$ .

```
plot(instant, cnt, col=8)
#abline(v=sm.sp$fit$min+sm.sp$fit$knot*sm.sp$fit$range, col=8, lty=2)
lines(sm.sp, col=2, lwd=2)
```



2. The script `IRWLS_logistic_regression.R` includes the definition of the function `logistic.IRWLS.splines` performing nonparametric logistic regression using splines with a IRWLS procedure. The basic syntax is the following: `logistic.IRWLS.splines(x=..., y=..., x.new=..., df=..., plts=TRUE)` where the arguments are the explanatory variable `x`, the 0-1 response variable `y`, the vector `x.new` of new values of variable `x` where we want to predict the probability of `y` being 1 given that `x` is equal to `x.new`, the equivalent number of parameters (or model degrees of freedom) `df`, and the logical `plts` indicating if plots are desired or not. Define a new variable `cnt.5000` taking the value 1 for days such that the number of total rental bikes is larger than or equal to 5000, on 0 otherwise.

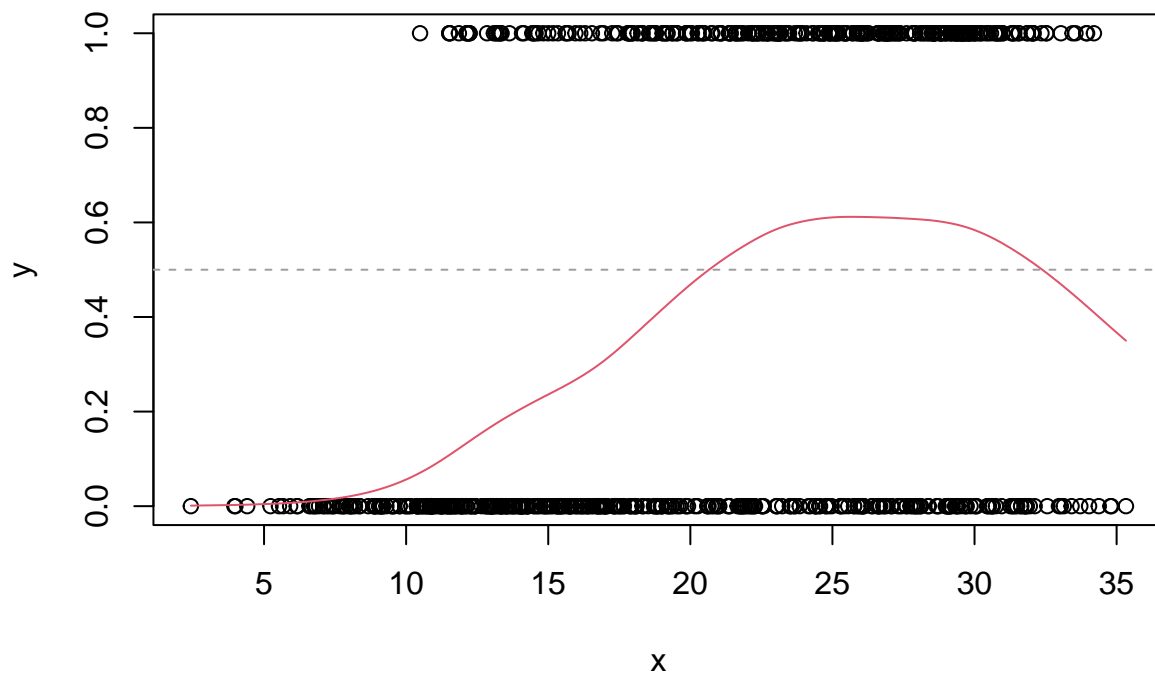
```
source("IRWLS_logistic_regression.R")
cnt.5000 <- as.numeric(cnt >= 5000)
```

a) Use the function `logistic.IRWLS.splines` to fit the non-parametric binary regression `cnt.5000` as a function of the temperature, using `df=6`. In which range of temperatures is  $\Pr(\text{cnt} \geq 5000 | \text{temp})$  larger than 0,5?

```
# Sort data according to x
x <- temp
y <- cnt.5000
sx <- sort(x, index.return = TRUE)
x <- sx$x
y <- y[sx$ix]

IRWLS.sp <- logistic.IRWLS.splines(x=x, y=y, x.new = x, df=6)

plot(x, y)
lines(x, IRWLS.sp$fitted.values, col=2)
abline(h=0.5, col=8, lty=2)
```



```
x.05 <- x[as.numeric(IRWLS.sp$predicted.values >= 0.5) == 1]
x.min <- min(x.05)
x.max <- max(x.05)
print(sprintf("min: %f, max: %f", x.min, x.max))
```

```
## [1] "min: 20.739153, max: 32.355847"
```

Looking at the returned prediction, the temperatures between 21.7° and 32.4° have  $\Pr(\text{cnt} \geq 5000 | \text{temp})$  larger than 0.5. We can also check it looking at the plot.

b) Choose the parameter `df` by k-fold log-likelihood cross validation with `k = 5` and using `df.v = 3:15` as the set of possible values for `df`.

```
k.fold.ll.cv <- function(x,y,df,k=5){
  n <- length(x)
  Ik <- floor((0:(n-1))/(n/k))+1
  cum_sum <- 0

  for (i in (1:k)){
    y.i <- y[Ik==i]
    pred <- logistic.IRWLS.splines(x[Ik!=i], y[Ik!=i],
                                   x.new=x[Ik==i], df=df)$predicted.values
    #cum_sum <- cum_sum + sum((y.i-pred)^2)
    cum_sum <- cum_sum + sum(y.i*log(pred/(1-pred)) + log(1-pred))
  }
  k.cv <- cum_sum/n
  return(k.cv)
}
```

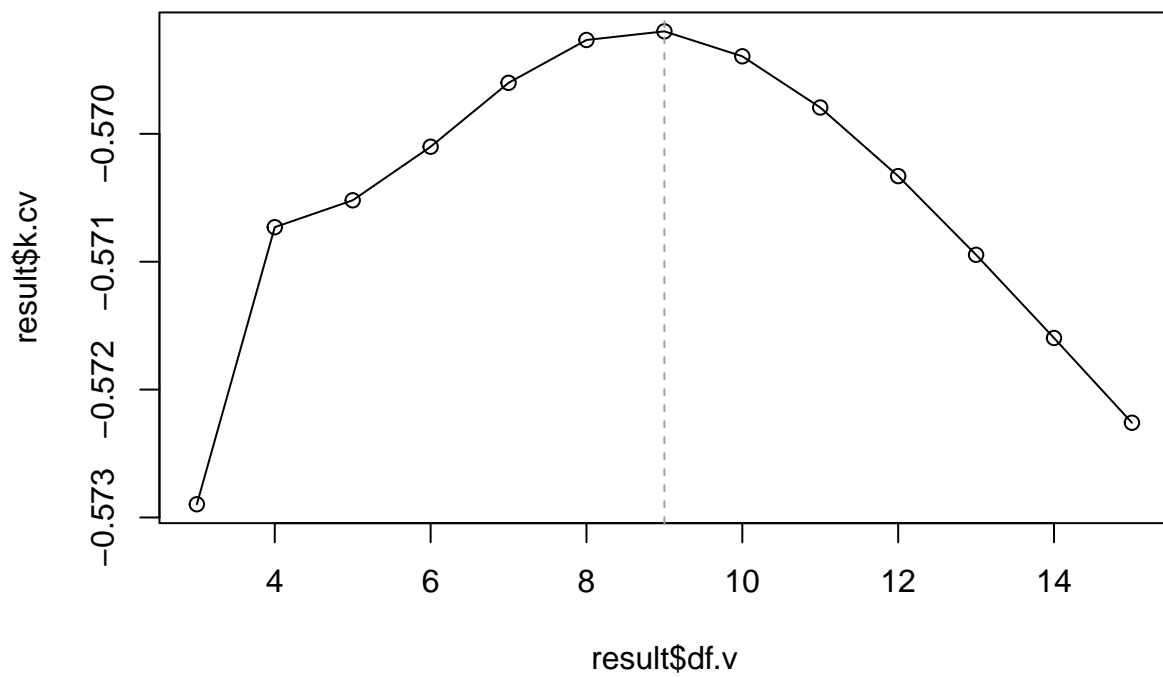
```
df.k.fold.ll.cv <- function(x,y,df.v,k=5){
  n <- length(x)
  perm <- sample(1:n)
  xperm <- x[perm]
  yperm <- y[perm]

  k.cv <- df.v*0
  for (i in (1:length(df.v))){
    df <- df.v[i]
    k.cv[i] <- k.fold.ll.cv(x=xperm, y=yperm, df, k)
  }
  return(list(k=k,df.v=df.v,k.cv=k.cv))
}
```

```
df.v <- 3:15
result <- df.k.fold.ll.cv(x,y,df.v)

selected_df <- result$df.v[which.max(result$k.cv)]

plot(result$df.v, result$k.cv)
lines(result$df.v, result$k.cv)
abline(v = selected_df, col="8", lty=2)
```



```
IRWLS.sp <- logistic.IRWLS.splines(x=x, y=y, x.new = x, df=selected_df)

plot(x, y)
lines(x, IRWLS.sp$fitted.values, col=2)
abline(h=0.5, col=8, lty=2)
```

