

Car-Accident-Kernel-Regression.R

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2019-11-30

```
# Non-parametric Regression - Kernel Regression
# A data frame giving a series of measurements of head acceleration
# in a simulated motorcycle accident, used to test crash helmets.

# First, we should plot the data to see whether we should use Non-parametric
# regression
# Clearly, we could see there is no linear relationship in this plot.
# So we use Kernel Regression for estimation.
library(MASS)
mcycle

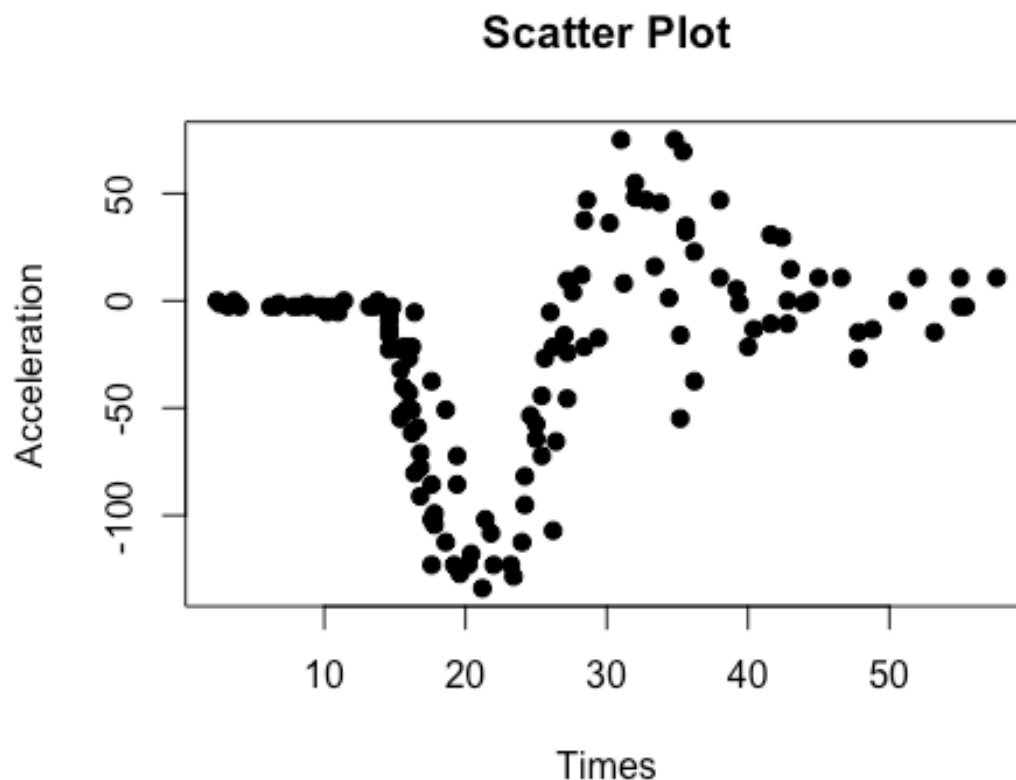
##      times  accel
## 1      2.4    0.0
## 2      2.6   -1.3
## 3      3.2   -2.7
## 4      3.6    0.0
## 5      4.0   -2.7
## 6      6.2   -2.7
## 7      6.6   -2.7
## 8      6.8   -1.3
## 9      7.8   -2.7
## 10     8.2   -2.7
## 11     8.8   -1.3
## 12     8.8   -2.7
## 13     9.6   -2.7
## 14    10.0   -2.7
## 15    10.2   -5.4
## 16    10.6   -2.7
## 17    11.0   -5.4
## 18    11.4    0.0
## 19    13.2   -2.7
## 20    13.6   -2.7
## 21    13.8    0.0
## 22    14.6  -13.3
## 23    14.6   -5.4
## 24    14.6   -5.4
## 25    14.6   -9.3
## 26    14.6  -16.0
## 27    14.6  -22.8
## 28    14.8   -2.7
## 29    15.4  -22.8
## 30    15.4  -32.1
```

##	31	15.4	-53.5
##	32	15.4	-54.9
##	33	15.6	-40.2
##	34	15.6	-21.5
##	35	15.8	-21.5
##	36	15.8	-50.8
##	37	16.0	-42.9
##	38	16.0	-26.8
##	39	16.2	-21.5
##	40	16.2	-50.8
##	41	16.2	-61.7
##	42	16.4	-5.4
##	43	16.4	-80.4
##	44	16.6	-59.0
##	45	16.8	-71.0
##	46	16.8	-91.1
##	47	16.8	-77.7
##	48	17.6	-37.5
##	49	17.6	-85.6
##	50	17.6	-123.1
##	51	17.6	-101.9
##	52	17.8	-99.1
##	53	17.8	-104.4
##	54	18.6	-112.5
##	55	18.6	-50.8
##	56	19.2	-123.1
##	57	19.4	-85.6
##	58	19.4	-72.3
##	59	19.6	-127.2
##	60	20.2	-123.1
##	61	20.4	-117.9
##	62	21.2	-134.0
##	63	21.4	-101.9
##	64	21.8	-108.4
##	65	22.0	-123.1
##	66	23.2	-123.1
##	67	23.4	-128.5
##	68	24.0	-112.5
##	69	24.2	-95.1
##	70	24.2	-81.8
##	71	24.6	-53.5
##	72	25.0	-64.4
##	73	25.0	-57.6
##	74	25.4	-72.3
##	75	25.4	-44.3
##	76	25.6	-26.8
##	77	26.0	-5.4
##	78	26.2	-107.1
##	79	26.2	-21.5
##	80	26.4	-65.6

## 81	27.0	-16.0
## 82	27.2	-45.6
## 83	27.2	-24.2
## 84	27.2	9.5
## 85	27.6	4.0
## 86	28.2	12.0
## 87	28.4	-21.5
## 88	28.4	37.5
## 89	28.6	46.9
## 90	29.4	-17.4
## 91	30.2	36.2
## 92	31.0	75.0
## 93	31.2	8.1
## 94	32.0	54.9
## 95	32.0	48.2
## 96	32.8	46.9
## 97	33.4	16.0
## 98	33.8	45.6
## 99	34.4	1.3
## 100	34.8	75.0
## 101	35.2	-16.0
## 102	35.2	-54.9
## 103	35.4	69.6
## 104	35.6	34.8
## 105	35.6	32.1
## 106	36.2	-37.5
## 107	36.2	22.8
## 108	38.0	46.9
## 109	38.0	10.7
## 110	39.2	5.4
## 111	39.4	-1.3
## 112	40.0	-21.5
## 113	40.4	-13.3
## 114	41.6	30.8
## 115	41.6	-10.7
## 116	42.4	29.4
## 117	42.8	0.0
## 118	42.8	-10.7
## 119	43.0	14.7
## 120	44.0	-1.3
## 121	44.4	0.0
## 122	45.0	10.7
## 123	46.6	10.7
## 124	47.8	-26.8
## 125	47.8	-14.7
## 126	48.8	-13.3
## 127	50.6	0.0
## 128	52.0	10.7
## 129	53.2	-14.7
## 130	55.0	-2.7

```
## 131  55.0  10.7
## 132  55.4  -2.7
## 133  57.6  10.7

cycleDF <- as.data.frame(mcycle)
X <- cycleDF$times
Y <- cycleDF$accel
plot(X, Y, main = "Scatter Plot", xlab = "Times", ylab = "Acceleration", pch
= 19)
```



```
# Cross Validation
# Leave-one-out cross validation
n <- dim.data.frame(cycleDF)[1] # n: sample size

h_seq <- seq(from = 1.5, to = 10.0, by = 0.5)

# bandwidths we are using
CV_err_h <- rep(NA, length(h_seq))

for (j in 1:length(h_seq)){
  h_using <- h_seq[j]
  CV_err <- rep(NA, n)
  for (i in 1:n) {
```

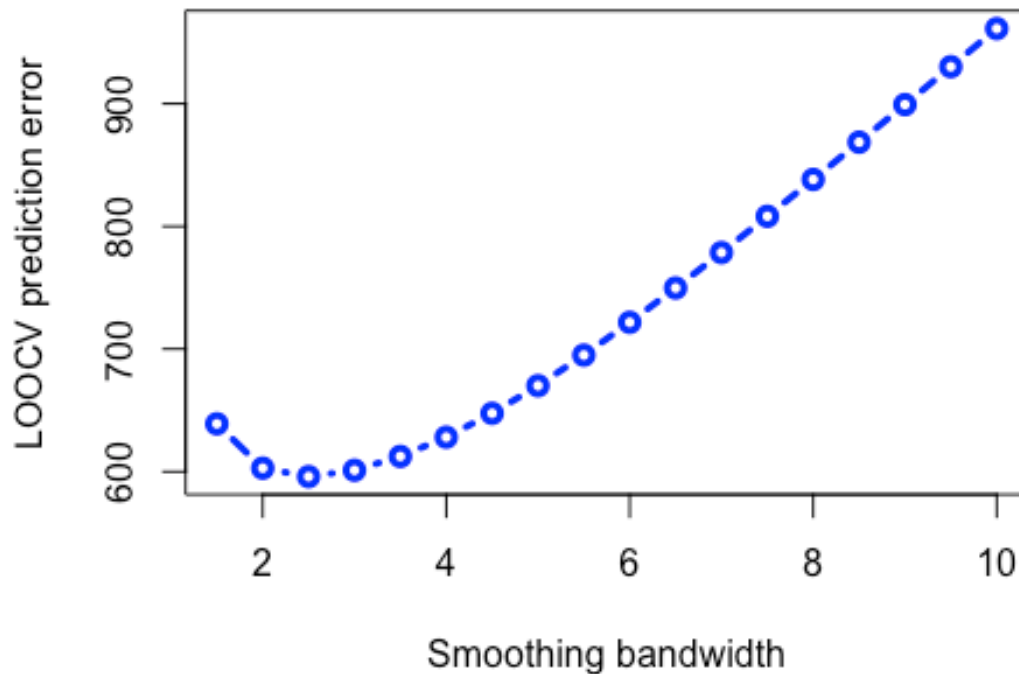
```

X_val <- X[i]
Y_val <- Y[i]
# Validation set
X_tr <- X[-i]
Y_tr <- Y[-i]
# Training set
Y_val_predict <- ksmooth(x = X_tr, y = Y_tr,
                        kernel = "normal", bandwidth = h_using, x.points
= X_val)
  CV_err[i] <- (Y_val - Y_val_predict$y)^2
}
  CV_err_h[j] <- mean(CV_err)
}
CV_err_h

## [1] 638.9920 602.8414 595.9586 601.1110 612.3813 628.1086 647.5628
670.0977
## [9] 695.0158 721.7479 749.7548 778.6409 808.2178 838.2577 868.6444
899.2966
## [17] 930.1937 961.2848

# Plot which bandwidth we will use
plot(x=h_seq, y=CV_err_h, type="b", lwd=3, col="blue", xlab="Smoothing
bandwidth", ylab="LOOCV prediction error")

```



```
which(CV_err_h == min(CV_err_h))
## [1] 3
h_seq[which(CV_err_h == min(CV_err_h))]
## [1] 2.5
# We choose 2.5 as our bandwidth in this dataset
Kreg <- ksmooth(x=X,y=Y,kernel = "normal",bandwidth = 0.9)
plot(X,Y,pch=20, xlab = "Times", ylab = "Accelaration", main = "Kenelized
Scatter Plot")
lines(Kreg, lwd=4, col="purple")
```

Kenelized Scatter Plot

