

# Neural-Network.R

patriciamaya

2020-11-03

```
#Neural Network Model
# we are predicting sales of child car seats at 400 different stores.

library(ISLR)

## Warning: package 'ISLR' was built under R version 4.0.2

data("Carseats")
attach(Carseats)
data<- Carseats
##?Carseats

# ** NORMALIZE DATASET
#we can only normalize numerical variables
num_cols <- unlist(lapply(data, is.numeric))
num_cols

##      Sales    CompPrice      Income Advertising Population      Price
##      TRUE      TRUE      TRUE      TRUE      TRUE      TRUE
## ShelfLoc      Age Education      Urban      US
##      FALSE      TRUE      TRUE      FALSE      FALSE

dataNum <- data[,num_cols] #only numerical variables
str(dataNum)

## 'data.frame':   400 obs. of  8 variables:
## $ Sales      : num  9.5 11.22 10.06 7.4 4.15 ...
## $ CompPrice  : num  138 111 113 117 141 124 115 136 132 132 ...
## $ Income     : num   73 48 35 100 64 113 105 81 110 113 ...
## $ Advertising: num   11 16 10 4 3 13 0 15 0 0 ...
## $ Population : num  276 260 269 466 340 501 45 425 108 131 ...
## $ Price      : num  120 83 80 97 128 72 108 120 124 124 ...
## $ Age        : num   42 65 59 55 38 78 71 67 76 76 ...
## $ Education  : num   17 10 12 14 13 16 15 10 10 17 ...

#to normalize numerical variables we will use min - max transformation
#x in [min, max]
#x' = (x-min) / (max-min), so x' in [0,1]

#min & max of all columns
mins<- apply(dataNum, 2, min) #1-rows, 2 -columns
maxs<- apply(dataNum, 2, max)

scaled.data<-as.data.frame(scale(dataNum, center = mins, scale= maxs - mins))
summary(scaled.data) #we can see now min is 0 and max is 1.
```

```

##      Sales      CompPrice      Income      Advertising
## Min.   :0.0000   Min.   :0.0000   Min.   :0.0000   Min.   :0.0000
## 1st Qu.:0.3313   1st Qu.:0.3878   1st Qu.:0.2197   1st Qu.:0.0000
## Median :0.4604   Median :0.4898   Median :0.4848   Median :0.1724
## Mean   :0.4607   Mean   :0.4895   Mean   :0.4814   Mean   :0.2288
## 3rd Qu.:0.5728   3rd Qu.:0.5918   3rd Qu.:0.7071   3rd Qu.:0.4138
## Max.   :1.0000   Max.   :1.0000   Max.   :1.0000   Max.   :1.0000
##      Population      Price      Age      Education
## Min.   :0.0000   Min.   :0.0000   Min.   :0.0000   Min.   :0.0000
## 1st Qu.:0.2585   1st Qu.:0.4551   1st Qu.:0.2682   1st Qu.:0.2500
## Median :0.5251   Median :0.5569   Median :0.5364   Median :0.5000
## Mean   :0.5107   Mean   :0.5497   Mean   :0.5150   Mean   :0.4875
## 3rd Qu.:0.7786   3rd Qu.:0.6407   3rd Qu.:0.7455   3rd Qu.:0.7500
## Max.   :1.0000   Max.   :1.0000   Max.   :1.0000   Max.   :1.0000

#add to scaled df the variables that were factors (categorical)
data <- data.frame(scaled.data, data[!num_cols] )
str(data)

## 'data.frame':   400 obs. of  11 variables:
## $ Sales      : num  0.584 0.69 0.618 0.455 0.255 ...
## $ CompPrice  : num  0.622 0.347 0.367 0.408 0.653 ...
## $ Income     : num  0.525 0.273 0.141 0.798 0.434 ...
## $ Advertising: num  0.379 0.552 0.345 0.138 0.103 ...
## $ Population : num  0.533 0.501 0.519 0.914 0.661 ...
## $ Price      : num  0.575 0.353 0.335 0.437 0.623 ...
## $ Age        : num  0.309 0.727 0.618 0.545 0.236 ...
## $ Education  : num  0.875 0 0.25 0.5 0.375 0.75 0.625 0 0 0.875 ...
## $ ShelfLoc   : Factor w/ 3 levels "Bad","Good","Medium": 1 2 3 3 1 1 3 2 3 3 ...
## $ Urban      : Factor w/ 2 levels "No","Yes": 2 2 2 2 2 1 2 2 1 1 ...
## $ US         : Factor w/ 2 levels "No","Yes": 2 2 2 2 1 2 1 2 1 2 ...

# ** CONSTRUCT NN MODEL
set.seed(123)
indx<- sample(2, nrow(data), replace=T, prob=c(0.8,0.2))
train <- data[indx==1, ]
test  <- data[indx==2, ]

#library(neuralnet)
library(nnet)
nn <- nnet(Sales ~ . , data= train, linout= T , size= 10, decay=0.01)

## # weights:  131
## initial  value 118.897428
## iter  10 value 2.787889
## iter  20 value 1.469418
## iter  30 value 1.355768
## iter  40 value 1.301456
## iter  50 value 1.280936
## iter  60 value 1.268716
## iter  70 value 1.254081
## iter  80 value 1.240087
## iter  90 value 1.234015
## iter 100 value 1.230016
## final   value 1.230016
## stopped after 100 iterations

```

```

#linout -- stands for linear out-put and is used to determine whether the target variable is continuous
#linout = T if numerical variable, F if categorical variables.
#size = X , sets the num of neurons in hidden layer
#decay = , regularization term [0,1] -> if small: large network-may overfit, if large: smaller network
summary(nn)

```

```

## a 11-10-1 network with 131 weights
## options were - linear output units  decay=0.01
##  b->h1  i1->h1  i2->h1  i3->h1  i4->h1  i5->h1  i6->h1  i7->h1  i8->h1  i9->h1
##  -0.16  -0.61   0.29   0.07  -0.01   0.66  -0.05   0.06   0.38   0.14
## i10->h1 i11->h1
##   0.14   0.85
##  b->h2  i1->h2  i2->h2  i3->h2  i4->h2  i5->h2  i6->h2  i7->h2  i8->h2  i9->h2
##  -0.06   0.05   0.12  -0.04  -0.07  -0.45  -0.10   0.00  -0.17   0.17
## i10->h2 i11->h2
##  -0.07  -0.19
##  b->h3  i1->h3  i2->h3  i3->h3  i4->h3  i5->h3  i6->h3  i7->h3  i8->h3  i9->h3
##   0.27   0.30  -0.96  -0.80  -0.25   1.04   0.26   0.37  -0.44   0.25
## i10->h3 i11->h3
##  -0.03  -0.32
##  b->h4  i1->h4  i2->h4  i3->h4  i4->h4  i5->h4  i6->h4  i7->h4  i8->h4  i9->h4
##   0.04   0.03   0.04   0.01   0.00   0.00   0.02   0.03   0.01   0.03
## i10->h4 i11->h4
##   0.03   0.03
##  b->h5  i1->h5  i2->h5  i3->h5  i4->h5  i5->h5  i6->h5  i7->h5  i8->h5  i9->h5
##  -0.03   0.88  -0.70  -0.36  -0.96  -0.64  -0.57   0.05   0.66   0.21
## i10->h5 i11->h5
##   0.41   0.38
##  b->h6  i1->h6  i2->h6  i3->h6  i4->h6  i5->h6  i6->h6  i7->h6  i8->h6  i9->h6
##   0.12   0.39  -0.20   0.22   0.71  -0.58  -0.19   0.43   0.22   0.67
## i10->h6 i11->h6
##  -0.04  -0.05
##  b->h7  i1->h7  i2->h7  i3->h7  i4->h7  i5->h7  i6->h7  i7->h7  i8->h7  i9->h7
##  -0.06   0.01  -0.03  -0.04  -0.03  -0.11   0.00  -0.04  -0.06   0.01
## i10->h7 i11->h7
##  -0.05  -0.13
##  b->h8  i1->h8  i2->h8  i3->h8  i4->h8  i5->h8  i6->h8  i7->h8  i8->h8  i9->h8
##   0.20   0.05   0.07   0.12   0.05   0.26   0.17   0.12   0.10  -0.02
## i10->h8 i11->h8
##   0.30   0.33
##  b->h9  i1->h9  i2->h9  i3->h9  i4->h9  i5->h9  i6->h9  i7->h9  i8->h9  i9->h9
##  -0.22  -0.43  -0.61   0.03   0.18   0.09  -0.01  -0.40  -0.15  -0.08
## i10->h9 i11->h9
##   0.13   0.27
##  b->h10 i1->h10 i2->h10 i3->h10 i4->h10 i5->h10 i6->h10 i7->h10
##   0.37   0.43   0.23   0.31   0.10  -0.66   0.64  -0.41
## i8->h10 i9->h10 i10->h10 i11->h10
##   0.11  -0.28  -0.07   0.56
##  b->o  h1->o  h2->o  h3->o  h4->o  h5->o  h6->o  h7->o  h8->o  h9->o h10->o
##   0.24 -1.19  0.47 -1.23  0.01  1.14  1.30  0.01 -0.24 -0.62  0.90

```

```

#b- bias, i-input, h1-first neuron in hidden layer, o-output layer

```

```

#VISUALIZE FUNCTION

```

```
#To plot the neural network using nnet we need to use devtools  
library(devtools)
```

```
## Warning: package 'devtools' was built under R version 4.0.2
```

```
## Loading required package: usethis
```

```
## Warning: package 'usethis' was built under R version 4.0.2
```

```
source_url('https://gist.githubusercontent.com/fawda123/7471137/raw/466c1474d0a505ff044412703516c34f1a4')
```

```
## SHA-1 hash of file is 74c80bd5ddbc17ab3ae5ece9c0ed9beb612e87ef
```

```
plot.nnet(nn)
```

```
## Loading required package: scales
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```

```
## Loading required package: reshape
```

```
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,  
## logical.return = TRUE, : there is no package called 'reshape'
```





## 16 0.36510071  
## 20 0.47949243  
## 21 0.37672243  
## 24 0.33998799  
## 31 0.85072100  
## 32 0.49142524  
## 50 0.62531793  
## 59 0.30645896  
## 65 0.45070825  
## 67 0.52876758  
## 68 0.52758249  
## 87 0.47567004  
## 88 0.58881321  
## 89 0.38807777  
## 104 0.32470792  
## 106 0.48077012  
## 107 0.07396386  
## 111 0.50008265  
## 114 0.35281685  
## 118 0.50693334  
## 126 0.61310370  
## 132 0.38617805  
## 137 0.33489335  
## 139 0.61057279  
## 145 0.62355666  
## 151 0.61260162  
## 173 0.72499878  
## 179 0.66872340  
## 181 0.29475259  
## 189 0.41005701  
## 190 0.67722007  
## 193 0.38196235  
## 195 0.47328812  
## 202 0.37110482  
## 206 0.27318504  
## 219 0.62010379  
## 220 0.69399831  
## 222 0.37954722  
## 230 0.56275418  
## 238 0.45319213  
## 240 0.25188148  
## 248 0.14627399  
## 249 0.37547949  
## 260 0.31928753  
## 261 0.55768443  
## 262 0.36999990  
## 264 0.39162290  
## 271 0.76358756  
## 277 0.41105809  
## 294 0.73626270  
## 296 0.21634086  
## 297 0.61138054  
## 316 0.38732564  
## 317 0.90016385

```
## 320 0.41363693
## 321 0.36277931
## 327 0.41305825
## 330 0.69933616
## 334 0.36472243
## 340 0.61921694
## 347 0.45716790
## 352 0.59472339
## 356 0.49085516
## 360 0.25779825
## 363 0.22656155
## 373 0.45003937
## 376 0.36902141
## 380 0.40198550
## 386 0.38874874
## 391 0.37259571
## 400 0.57126469
```

```
#EVALUATE
```

```
#MEAN SQUARED ERROR
```

```
mse <- mean((nn.preds - test$Sales)^2)
```