Brauchle_HW_9

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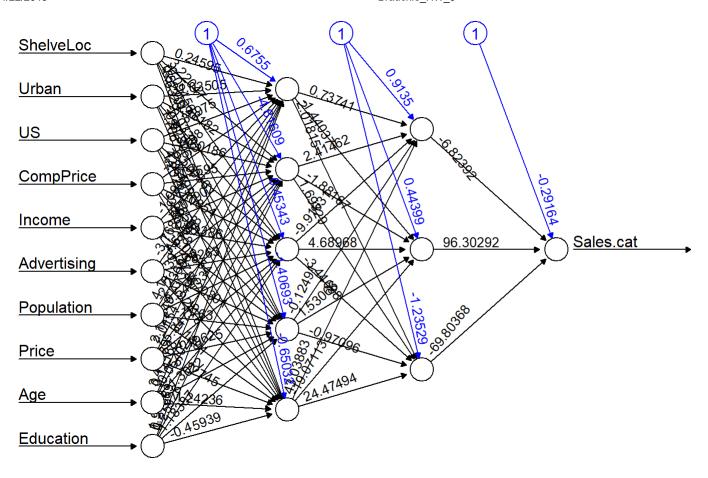
4/21/2019

```
library(ISLR)
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(neuralnet)
##
## Attaching package: 'neuralnet'
## The following object is masked from 'package:dplyr':
##
##
       compute
```

```
set.seed(1)
#categorize Sales
carseats <- Carseats %>% mutate(Sales.cat = ifelse(Sales>8,1,0))
# make categorical variables numeric to use in neural net
carseats[,c(7,10,11)] <- lapply(carseats[,c(7,10,11)], as.numeric)</pre>
# scale numeric variables
numeric carseats <- carseats %>% dplyr::select if(is.numeric) %>% dplyr::select(-Sales, -Sales.c
at, -7, -10, -11)
maxs <- apply(numeric carseats, 2, max)</pre>
mins <- apply(numeric_carseats, 2, min)</pre>
# Use scale() and convert the resulting matrix to a data frame
scaled.data <- as.data.frame(scale(numeric carseats, center=mins, scale=maxs-mins))</pre>
carseats_scaled <- cbind(carseats[,c(7,10,11, 12)], scaled.data)</pre>
#create train and test set
subset <- sample(1:nrow(carseats scaled),round(0.75*nrow(carseats scaled)))</pre>
train <- carseats scaled[subset,]</pre>
test <- carseats scaled[-subset,]
#make formula of names to plug into neural net
feats <- names(dplyr::select(carseats scaled, -Sales.cat))</pre>
# Concatenate strings
f <- paste(feats,collapse=' + ')</pre>
f <- paste('Sales.cat ~',f)</pre>
# Convert to formula
f <- as.formula(f)</pre>
f
```

```
## Sales.cat ~ ShelveLoc + Urban + US + CompPrice + Income + Advertising +
## Population + Price + Age + Education
```

```
nn <- neuralnet(f, train, hidden=c(5,3), linear.output=FALSE)
plot(nn, rep = "best")</pre>
```



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This neural network has 3 layers and two hiiden layers. The first hidden layer has 5 nodes and the second has 3 nodes. This neural network predicts whether a location sold more than 8,000 carseats or not using various weights for each node.

```
#create data frame
best_cv <- data.frame(matrix(ncol = 3, nrow = 0))

#loop to get best number of Layers and size of each Layer
k <- c(1, 3:10)
for(i in k){
    for(j in 1:i){# j < i
        nn <- neuralnet(f,data=train,hidden=c(i,j),linear.output=F)
        print(i)
        predicted.nn.values <- neuralnet::compute(nn, test[,-c(4)])
        predicted.nn.values$net.result <- sapply(predicted.nn.values$net.result,round,digits=0)
        cv.error <- mean(predicted.nn.values$net.result==test$Sales.cat)
        best_cv <- rbind(best_cv, c(i, j, cv.error))
    }
}</pre>
```

```
## [1] 1
## [1] 3
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```

```
x <- c("i_test", "j_test", "err")
colnames(best_cv) <- x

best_cv[which.max(best_cv$err),]</pre>
```

```
## i_test j_test err
## 23 7 4 0.91
```

The cross-validation shows that the best accuracy rate comes when using a neural network with 7 layers and a size of 4 nodes for each layer, with an accuracy of 91%.