DEEP LEARNING

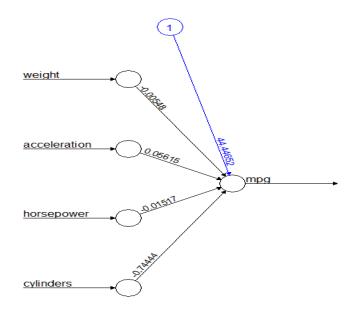
1. Artificial neural network with no hidden layers is just linear regression

Artificial neural networks are available in package neuralnet.

> library(neuralnet)

Create the training and testing data.

> plot(nn0)



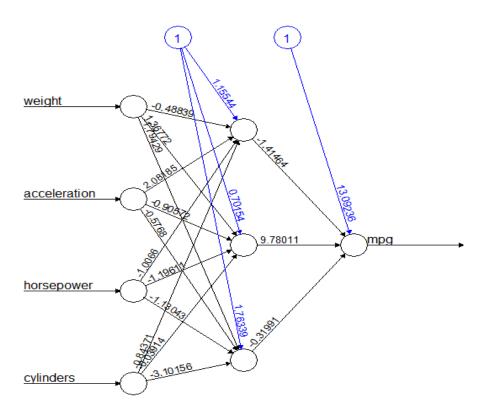
Error: 1720.963134 Steps: 8565

This is the linear regression! Weights are slopes, and the intercept is shown in blue.

2. Artificial neural network

Now, introduce 3 hidden nodes.

> nn3 = neuralnet(mpg ~ weight+acceleration+horsepower+cylinders, data=Auto.train, hidden=3)
> plot(nn3)



Error: 6036.459375 Steps: 148

3. Prediction power

Which ANN gives a more accurate prediction? Use the test data for comparison.

> Predict0 = compute(nn0, subset(Auto.test, select=c(weight, acceleration, horsepower, cylinders)))

This prediction consists of X-variables "neurons" and predicted Y-variable "net.result".

```
> names(Predict0)
[1] "neurons" "net.result"
> mean( (Auto.test$mpg - Predict0$net.result)^2 )
[1] 18.83350028
```

Prediction MSE of this ANN is 18.83.

> Predict3 = compute(nn3, subset(Auto.test, select=c(weight,acceleration,horsepower,cylinders)))

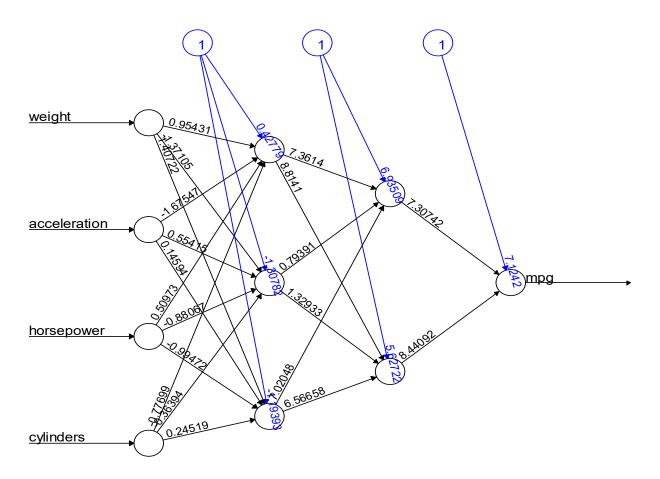
> mean((Auto.test\$mpg - Predict3\$net.result)^2) [1] 61.84868054

Its 3-node competitor has a much higher prediction MSE.

4. Multilayer structure

The number of hidden nodes can be given as a vector. Its components show the number of hidden nodes at each hidden layer.

> nn3.2 = neuralnet(mpg \sim weight+acceleration+horsepower+cylinders, data=Auto.train, hidden=c(3,2) > plot(nn3.2)



Error: 6036.459375 Steps: 101

5. Artificial neural network for classification

> library(nnet)

Prepare our categorical variables ECO and ECO4

```
> ECO = ifelse( mpg > 22.75, "Economy", "Consuming" )
> ECO4 = rep("Economy",n)
> ECO4[mpg < 29] = "Good"
> ECO4[mpg < 22.75] = "OK"
> ECO4[mpg < 17] = "Consuming"
> Auto.train = Auto[Z,]
> Auto.test = Auto[-Z,]
Train an artificial neural network to classify cars into "Economy" and "Consuming".
> nn.class = nnet( as.factor(ECO) ~ weight + acceleration + horsepower + cylinders, data=Auto.train, size=3)
# weights:
             19
initial
          value 169.471585
       value 138.379332
final
converged
> summary(nn.class)
a 4-3-1 network with 19 weights
options were - entropy fitting
 b->h1 i1->h1 i2->h1 i3->h1 i4->h1
 -0.11
         -0.35
                  0.12
                          0.32
                                  0.16
 b->h2 i1->h2 i2->h2 i3->h2 i4->h2
  0.59
        -0.37
                -0.56
                          0.67
                                  0.44
 b->h3 i1->h3 i2->h3 i3->h3 i4->h3
  0.50
          0.16
                  0.41 - 0.49
                                  0.02
 b->o h1->o h2->o h3->o
-0.12 - 0.27
             0.03 0.02
This ANN has p=4 inputs, one layer of M=3 hidden nodes, and a single (K=1) output. We need to estimate
M(p+1)+K(M+1) = (3)(5)+(1)(4) = 19 weights.
Classification into K > 2 categories is similar.
> nn.class = nnet( as.factor(ECO4) ~ weight+acceleration+horsepower+cylinders, data=Auto.train, size=3)
# weights: 31
initial value 333.799856
final value 276.956630
converged
> summary(nn.class)
a 4-3-4 network with 31 weights
options were - softmax modelling
 b->h1 i1->h1 i2->h1 i3->h1 i4->̈h1
                          0.23
                                  0.11
  0.39
          0.69 - 0.58
 b->h2 i1->h2 i2->h2 i3->h2 i4->h2
        -0.67
 -0.49
                -0.65 -0.68
                                  0.11
 b->h3 i1->h3 i2->h3 i3->h3 i4->h3
          0.24
                  0.59
                          0.43
                                  0.00
 b->01 h1->01 h2->01 h3->01
  0.98
        -0.25
                  0.32
                        -0.12
 b->o2 h1->o2 h2->o2 h3->o2
  0.13
          0.41
                  0.19
                        -0.02
 b->o3 h1->o3 h2->o3 h3->o3
  0.20
          0.28
                  0.34
                         -0.01
 b->04 h1->04 h2->04 h3->04
```

Here K = 4 categories, so we are estimating (3)(5) + (4)(4) = 31 weights.

0.41

-0.24

0.42

-0.09