

Neural-Network-for-Classification.R

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```
#Neural Network for Classification
```

```
#We'll use ISLR's built in College Data Set which has several features of a college and a  
#categorical column indicating whether or not the School is Public or Private.
```

```
library(ISLR)
```

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

```
data(College)
```

```
str(College) #all variables are numerical besides Private
```

```
## 'data.frame':   777 obs. of  18 variables:  
## $ Private      : Factor w/ 2 levels "No","Yes": 2 2 2 2 2 2 2 2 2 2 ...  
## $ Apps         : num  1660 2186 1428 417 193 ...  
## $ Accept       : num  1232 1924 1097 349 146 ...  
## $ Enroll       : num   721  512  336  137  55 158 103 489 227 172 ...  
## $ Top10perc    : num    23  16  22  60  16  38  17  37  30  21 ...  
## $ Top25perc    : num    52  29  50  89  44  62  45  68  63  44 ...  
## $ F.Undergrad  : num  2885 2683 1036 510 249 ...  
## $ P.Undergrad  : num   537 1227  99  63 869 ...  
## $ Outstate     : num  7440 12280 11250 12960 7560 ...  
## $ Room.Board   : num  3300 6450 3750 5450 4120 ...  
## $ Books        : num   450  750  400  450  800  500  500  450  300  660 ...  
## $ Personal     : num  2200 1500 1165 875 1500 ...  
## $ PhD          : num    70  29  53  92  76  67  90  89  79  40 ...  
## $ Terminal     : num    78  30  66  97  72  73  93 100  84  41 ...  
## $ S.F.Ratio    : num   18.1 12.2 12.9 7.7 11.9 9.4 11.5 13.7 11.3 11.5 ...  
## $ perc.alumni  : num    12  16  30  37  2  11  26  37  23  15 ...  
## $ Expend       : num  7041 10527 8735 19016 10922 ...  
## $ Grad.Rate    : num    60  56  54  59  15  55  63  73  80  52 ...
```

```
#DATA PREPROCESSING
```

```
# Create vector of column Max and Min values
```

```
maxs = apply(College[, 2:18], 2, max)
```

```
# apply(x, margin, function).
```

```
#If margin = 1, function is applied on the rows. If margin = 2, function is applied on the columns
```

```
mins = apply(College[, 2:18], 2, min)
```

```
#normalizing data
```

```
scaled.data = as.data.frame(scale(College[, 2:18], center = mins, scale = maxs - mins))
```

```
#SPLIT DATA
```

```
set.seed(1234)
```

```
ind = sample(2, nrow(College), replace = T, prob = c(0.7, 0.3))
```

```

TrainData = College[ind == 1, ]
TestData = College[ind == 2, ]

library(nnet)
nn = nnet(Private ~ ., data=TrainData, linout=F, size=10, decay=0.01, maxit=1000)

## # weights: 191
## initial value 485.887287
## iter 10 value 247.762510
## iter 20 value 174.812331
## iter 30 value 119.555050
## iter 40 value 117.766013
## iter 50 value 109.355874
## iter 60 value 105.428478
## iter 70 value 104.587145
## iter 80 value 100.003301
## iter 90 value 99.254662
## iter 100 value 99.069631
## iter 110 value 97.800280
## iter 120 value 96.287717
## iter 130 value 95.060649
## iter 140 value 94.493642
## iter 150 value 92.765077
## iter 160 value 92.464401
## iter 170 value 91.896218
## iter 180 value 90.482196
## iter 190 value 90.260756
## iter 200 value 90.130014
## iter 210 value 89.849747
## iter 220 value 89.589006
## iter 230 value 89.284073
## iter 240 value 89.013106
## iter 250 value 88.720591
## iter 260 value 88.388993
## iter 270 value 88.370376
## iter 280 value 88.299777
## iter 290 value 88.273862
## iter 300 value 88.259755
## iter 310 value 88.237144
## iter 320 value 88.217238
## iter 330 value 88.078857
## iter 340 value 88.019195
## iter 350 value 87.837137
## iter 360 value 87.776408
## iter 370 value 87.690785
## iter 380 value 87.487479
## iter 390 value 87.392482
## iter 400 value 87.303044
## iter 410 value 87.218736
## iter 420 value 87.028616
## iter 430 value 86.845645
## iter 440 value 86.665065
## iter 450 value 86.619367
## iter 460 value 86.605685

```

```
## iter 470 value 86.550946
## iter 480 value 86.431257
## iter 490 value 86.377424
## iter 500 value 86.235417
## iter 510 value 85.775320
## iter 520 value 85.717657
## iter 530 value 85.602845
## iter 540 value 85.395966
## iter 550 value 85.308607
## iter 560 value 84.589309
## iter 570 value 83.462937
## iter 580 value 82.013836
## iter 590 value 81.392742
## iter 600 value 81.147700
## iter 610 value 79.974230
## iter 620 value 78.664159
## iter 630 value 78.466769
## iter 640 value 77.553235
## iter 650 value 77.468005
## iter 660 value 76.498758
## iter 670 value 72.680294
## iter 680 value 70.825021
## iter 690 value 70.274280
## iter 700 value 69.980880
## iter 710 value 69.105281
## iter 720 value 68.734577
## iter 730 value 68.528781
## iter 740 value 67.405952
## iter 750 value 67.194740
## iter 760 value 66.906339
## iter 770 value 66.703103
## iter 780 value 66.630486
## iter 790 value 66.426696
## iter 800 value 66.129554
## iter 810 value 65.887121
## iter 820 value 65.453291
## iter 830 value 65.208523
## iter 840 value 65.123731
## iter 850 value 64.821433
## iter 860 value 64.616198
## iter 870 value 64.520869
## iter 880 value 64.039870
## iter 890 value 63.733603
## iter 900 value 63.382449
## iter 910 value 63.199916
## iter 920 value 62.951946
## iter 930 value 62.858501
## iter 940 value 62.766967
## iter 950 value 61.931618
## iter 960 value 61.465205
## iter 970 value 61.223511
## iter 980 value 60.904719
## iter 990 value 60.478556
## iter1000 value 60.317879
```

```

## final value 60.317879
## stopped after 1000 iterations

#The weights will be learned with a weight updating rate of 0.01 (the parameter decay).
#The parameter linout (linear out-put) indicates that the target variable is continuous or not.
#The maxit parameter sets the maximum number of iterations of the weight convergence algorithm.
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')

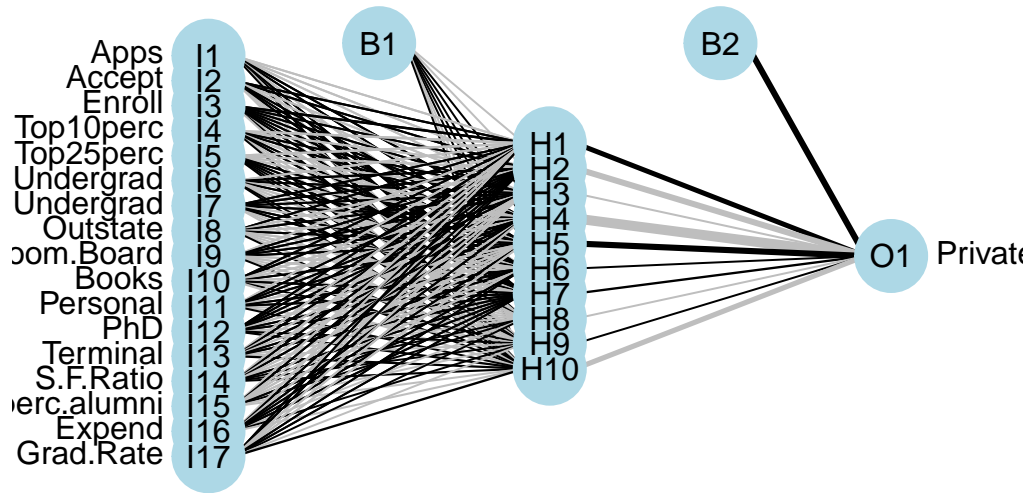
## SHA-1 hash of file is 74c80bd5ddbc17ab3ae5ece9c0ed9beb612e87ef
plot.nnet(nn)

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

```

[illegible]


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



```
summary(nn)
```

```
## a 17-10-1 network with 191 weights
## options were - entropy fitting decay=0.01
## b->h1 i1->h1 i2->h1 i3->h1 i4->h1 i5->h1 i6->h1 i7->h1 i8->h1 i9->h1
## -0.01 -0.03 0.14 0.41 -0.32 -0.31 -0.17 0.03 0.04 -0.02
## i10->h1 i11->h1 i12->h1 i13->h1 i14->h1 i15->h1 i16->h1 i17->h1
## 0.05 -0.04 0.15 -0.16 -0.34 -0.08 0.00 0.01
## b->h2 i1->h2 i2->h2 i3->h2 i4->h2 i5->h2 i6->h2 i7->h2 i8->h2 i9->h2
## 0.00 -0.04 0.09 0.00 -0.04 -0.05 0.03 0.08 -0.10 0.09
## i10->h2 i11->h2 i12->h2 i13->h2 i14->h2 i15->h2 i16->h2 i17->h2
## 0.23 0.01 0.03 0.01 0.01 0.01 0.00 0.01
## b->h3 i1->h3 i2->h3 i3->h3 i4->h3 i5->h3 i6->h3 i7->h3 i8->h3 i9->h3
## 0.00 -0.01 -0.01 0.00 0.00 0.00 0.01 0.00 0.00 0.02
## i10->h3 i11->h3 i12->h3 i13->h3 i14->h3 i15->h3 i16->h3 i17->h3
## 0.00 0.00 0.00 0.00 0.00 0.00 -0.01 0.00
## b->h4 i1->h4 i2->h4 i3->h4 i4->h4 i5->h4 i6->h4 i7->h4 i8->h4 i9->h4
## 0.00 0.10 0.00 0.03 0.02 0.00 -0.04 0.01 -0.03 0.00
```

```
## i10->h4 i11->h4 i12->h4 i13->h4 i14->h4 i15->h4 i16->h4 i17->h4
## -0.04 0.03 0.12 0.07 0.02 -0.01 0.03 -0.01
## b->h5 i1->h5 i2->h5 i3->h5 i4->h5 i5->h5 i6->h5 i7->h5 i8->h5 i9->h5
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.01 0.00
## i10->h5 i11->h5 i12->h5 i13->h5 i14->h5 i15->h5 i16->h5 i17->h5
## 0.00 0.00 0.00 0.00 0.00 0.00 0.01 0.00
## b->h6 i1->h6 i2->h6 i3->h6 i4->h6 i5->h6 i6->h6 i7->h6 i8->h6 i9->h6
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## i10->h6 i11->h6 i12->h6 i13->h6 i14->h6 i15->h6 i16->h6 i17->h6
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## b->h7 i1->h7 i2->h7 i3->h7 i4->h7 i5->h7 i6->h7 i7->h7 i8->h7 i9->h7
## 0.00 0.04 0.02 0.02 0.00 0.00 -0.02 0.02 -0.03 -0.03
## i10->h7 i11->h7 i12->h7 i13->h7 i14->h7 i15->h7 i16->h7 i17->h7
## 0.02 0.01 0.00 0.00 0.00 0.00 0.01 0.01
## b->h8 i1->h8 i2->h8 i3->h8 i4->h8 i5->h8 i6->h8 i7->h8 i8->h8 i9->h8
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## i10->h8 i11->h8 i12->h8 i13->h8 i14->h8 i15->h8 i16->h8 i17->h8
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## b->h9 i1->h9 i2->h9 i3->h9 i4->h9 i5->h9 i6->h9 i7->h9 i8->h9 i9->h9
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## i10->h9 i11->h9 i12->h9 i13->h9 i14->h9 i15->h9 i16->h9 i17->h9
## 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## b->h10 i1->h10 i2->h10 i3->h10 i4->h10 i5->h10 i6->h10 i7->h10
## 0.01 0.08 -0.22 -0.16 0.00 0.02 0.19 -0.10
## i8->h10 i9->h10 i10->h10 i11->h10 i12->h10 i13->h10 i14->h10 i15->h10
## 0.04 0.02 -0.41 0.05 0.28 0.23 0.10 -0.03
## i16->h10 i17->h10
## -0.09 0.36
## b->o h1->o h2->o h3->o h4->o h5->o h6->o h7->o h8->o h9->o h10->o
## 4.18 3.05 -4.09 -0.03 -8.11 4.18 0.00 0.35 0.00 0.00 -3.24
```

You could also use wts to get the best weights found and fitted.values to get the fitted values on training data

```
nn$wts
```

```
## [1] -8.346814e-03 -3.180840e-02 1.383018e-01 4.092997e-01 -3.184893e-01
## [6] -3.070548e-01 -1.693836e-01 3.159437e-02 3.780833e-02 -1.987079e-02
## [11] 4.510113e-02 -3.709286e-02 1.461482e-01 -1.564430e-01 -3.425698e-01
## [16] -7.996408e-02 -1.764706e-03 5.723435e-03 -3.809041e-04 -4.158933e-02
## [21] 8.783067e-02 8.942135e-04 -3.537822e-02 -4.959560e-02 3.248600e-02
## [26] 7.563627e-02 -9.572936e-02 9.037993e-02 2.284504e-01 1.140116e-02
## [31] 2.915567e-02 1.233496e-02 1.016949e-02 6.662271e-03 4.209715e-04
## [36] 1.021499e-02 7.504003e-04 -9.645047e-03 -1.110022e-02 1.439678e-04
## [41] 1.267905e-03 -5.782217e-06 8.102928e-03 4.318705e-03 1.361195e-03
## [46] 1.744511e-02 -1.383612e-03 2.214859e-03 1.813942e-03 -1.276769e-03
## [51] -2.855351e-04 -3.203392e-04 -9.915250e-03 -5.918951e-04 2.125271e-03
## [56] 1.021530e-01 -4.062111e-04 3.147085e-02 2.264027e-02 -2.814349e-03
## [61] -4.003394e-02 1.496051e-02 -3.245579e-02 4.834716e-03 -4.022348e-02
## [66] 2.774757e-02 1.215735e-01 6.872600e-02 2.032940e-02 -1.382640e-02
## [71] 2.704714e-02 -1.350968e-02 9.291463e-04 8.177963e-04 2.559609e-03
## [76] 2.047389e-03 -1.374688e-03 2.312299e-04 1.307436e-03 -1.461727e-03
## [81] 5.645802e-03 4.444709e-03 2.380481e-03 -1.189458e-03 -3.715773e-04
## [86] -1.970768e-03 -3.566358e-04 -1.228527e-03 7.355922e-03 -1.113603e-03
## [91] 2.064103e-03 -9.792936e-04 1.359719e-04 1.912850e-03 1.734161e-03
## [96] 1.922780e-03 -8.638021e-04 1.132063e-03 -1.341269e-03 -1.731161e-03
```



```
## [101] -1.777638e-03  3.921842e-05  3.322880e-04  7.994634e-04 -1.064268e-03
## [106] -2.337382e-04  1.325078e-04  2.913920e-04  1.839477e-03  3.618418e-02
## [111]  1.868118e-02  2.398438e-02  3.675391e-04  4.058815e-03 -1.549374e-02
## [116]  2.376740e-02 -2.782790e-02 -2.597028e-02  1.534999e-02  6.567838e-03
## [121]  1.441848e-03  6.705053e-04  1.281724e-03  2.313750e-03  1.439441e-02
## [126]  5.681919e-03 -2.082583e-04  1.628096e-03 -2.149866e-03 -7.148944e-04
## [131] -1.048437e-03  1.176948e-03 -1.005163e-03 -1.055594e-03 -2.203453e-03
## [136] -2.592809e-03 -1.287332e-03 -1.827870e-03 -1.810755e-03 -2.666722e-04
## [141]  2.063867e-03  1.982976e-03 -4.849844e-04  1.700788e-04 -1.894135e-04
## [146] -1.238095e-03 -1.154004e-03  3.898819e-04 -1.035475e-03  1.102433e-03
## [151]  2.200860e-04 -1.723675e-03 -1.498730e-03 -1.518853e-03  2.100760e-03
## [156] -2.030408e-03  4.571084e-04 -1.704186e-03  4.035629e-05 -1.338792e-03
## [161]  2.769769e-05 -4.918236e-04  8.152693e-03  7.554874e-02 -2.219631e-01
## [166] -1.586807e-01  1.505371e-03  2.233091e-02  1.947682e-01 -1.010053e-01
## [171]  3.711506e-02  2.019794e-02 -4.143202e-01  4.890080e-02  2.843147e-01
## [176]  2.348469e-01  9.914329e-02 -3.337568e-02 -9.274277e-02  3.641261e-01
## [181]  4.180649e+00  3.051322e+00 -4.089120e+00 -2.562073e-02 -8.110837e+00
## [186]  4.181220e+00  1.249743e-03  3.483872e-01 -5.223906e-04  6.956624e-04
## [191] -3.235796e+00
```

```
##nn$fitted.values
```

```
nn.preds = predict(nn, TestData, type = "class")
nn.preds
```

```
## [1] "Yes" "Yes" "Yes" "No" "No" "Yes" "Yes" "Yes" "No" "Yes" "Yes" "Yes"
## [13] "No" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes"
## [25] "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "No"
## [37] "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "No"
## [49] "Yes" "Yes" "Yes" "No" "Yes" "Yes" "Yes" "No" "Yes" "Yes" "Yes" "Yes"
## [61] "Yes" "Yes" "Yes" "Yes" "No" "Yes" "Yes" "Yes" "Yes" "No" "Yes" "Yes"
## [73] "Yes" "No" "Yes" "No" "No" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes"
## [85] "No" "Yes" "Yes" "Yes" "No" "Yes" "No" "Yes" "Yes" "Yes" "Yes" "No"
## [97] "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "No" "No" "No"
## [109] "Yes" "Yes" "No" "Yes" "No" "Yes" "Yes" "Yes" "Yes" "Yes" "No" "No"
## [121] "No" "No" "Yes" "No" "No" "Yes" "Yes" "Yes" "Yes" "Yes" "No" "Yes"
## [133] "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "No" "Yes" "Yes" "Yes" "Yes"
## [145] "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes" "Yes"
## [157] "Yes" "No" "Yes" "Yes" "No" "Yes" "Yes" "No" "Yes" "Yes" "Yes" "Yes"
## [169] "Yes" "Yes" "Yes" "Yes" "Yes" "No" "No" "No" "No" "No" "Yes" "Yes"
## [181] "Yes" "Yes" "Yes" "No" "No" "No" "No" "Yes" "No" "No" "Yes" "Yes"
## [193] "Yes" "No" "No" "No" "Yes" "No" "No" "No" "No" "No" "No" "No"
## [205] "Yes" "Yes" "No" "No" "No" "No" "No" "No" "No" "Yes" "No" "No"
## [217] "Yes" "No" "No" "Yes" "Yes" "Yes" "Yes" "No" "Yes" "No" "Yes" "No"
## [229] "No" "Yes" "Yes" "Yes" "Yes" "No" "Yes" "Yes" "Yes" "No"
```

```
##CONFUSION MATRIX
```

```
table(TestData$Private, nn.preds)
```

```
##      nn.preds
##      No Yes
## No    55  10
## Yes   16 157
```

```
##Accuracy
```

```
(55+157)/238
```

```
## [1] 0.8907563
```