

Tut6_memo

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2023-05-08

Write a code for the following problems:

Q1. Use the package `neuralnet` and `NeuralNetTools`. Install it if not installed in your computer. Now consider the `mtcars` dataset. Create a neural network model to predict the miles per gallon (`mpg`) of cars in the `mtcars` dataset based on the displacement (`disp`) and horsepower (`hp`) of the cars. Follow the steps below: 1. Call the package. 2. Load the dataset. 3. Scale the dataset. 4. Create a neural network model using the `neuralnet()` function and save the output in a variable. 5. Print the output using the `print()` function. 6. Predict the `mpg` using the `compute()` function. 7. Then create a neural network plot using the `plotnet()` function.

Then calculate the mean square error (MSE) for the above model. To calculate the Mean Squared Error (MSE) for the above neural network model, you can use the `predict` function to generate predictions and then compare these to the actual values.

```
library(neuralnet)
library(NeuralNetTools)
data(mtcars)

# Scale the data
maxs <- apply(mtcars, 2, max)
mins <- apply(mtcars, 2, min)
scaled_mtcars <- as.data.frame(scale(mtcars, center = mins, scale = maxs -
mins))

# Fit the neural network model
set.seed(12345)
nn <- neuralnet(mpg ~ disp + hp, data = scaled_mtcars, hidden = 2)

# Print summary and visualize
print(nn)

## $call
## neuralnet(formula = mpg ~ disp + hp, data = scaled_mtcars, hidden = 2)
##
## $response
##                mpg
## Mazda RX4      0.4510638
## Mazda RX4 Wag  0.4510638
## Datsun 710     0.5276596
```

```

## Hornet 4 Drive      0.4680851
## Hornet Sportabout  0.3531915
## Valiant             0.3276596
## Duster 360         0.1659574
## Merc 240D          0.5957447
## Merc 230           0.5276596
## Merc 280           0.3744681
## Merc 280C          0.3148936
## Merc 450SE         0.2553191
## Merc 450SL         0.2936170
## Merc 450SLC        0.2042553
## Cadillac Fleetwood 0.0000000
## Lincoln Continental 0.0000000
## Chrysler Imperial  0.1829787
## Fiat 128           0.9361702
## Honda Civic         0.8510638
## Toyota Corolla     1.0000000
## Toyota Corona      0.4723404
## Dodge Challenger    0.2170213
## AMC Javelin         0.2042553
## Camaro Z28         0.1234043
## Pontiac Firebird    0.3744681
## Fiat X1-9           0.7191489
## Porsche 914-2       0.6638298
## Lotus Europa        0.8510638
## Ford Pantera L      0.2297872
## Ferrari Dino        0.3957447
## Maserati Bora       0.1957447
## Volvo 142E         0.4680851
##
## $covariate
##              disp      hp
## Mazda RX4      0.22175106 0.20494700
## Mazda RX4 Wag  0.22175106 0.20494700
## Datsun 710     0.09204290 0.14487633
## Hornet 4 Drive 0.46620105 0.20494700
## Hornet Sportabout 0.72062859 0.43462898
## Valiant        0.38388626 0.18727915
## Duster 360     0.72062859 0.68197880
## Merc 240D      0.18857570 0.03533569
## Merc 230       0.17385882 0.15194346
## Merc 280       0.24070841 0.25088339
## Merc 280C      0.24070841 0.25088339
## Merc 450SE     0.51060115 0.45229682
## Merc 450SL     0.51060115 0.45229682
## Merc 450SLC    0.51060115 0.45229682
## Cadillac Fleetwood 1.00000000 0.54063604
## Lincoln Continental 0.97006735 0.57597173
## Chrysler Imperial 0.92017960 0.62897527
## Fiat 128       0.01895735 0.04946996

```

```

## Honda Civic          0.01147418 0.00000000
## Toyota Corolla       0.00000000 0.04593640
## Toyota Corona        0.12222499 0.15901060
## Dodge Challenger     0.61586431 0.34628975
## AMC Javelin          0.58094288 0.34628975
## Camaro Z28           0.69568471 0.68197880
## Pontiac Firebird     0.82040409 0.43462898
## Fiat X1-9            0.01970566 0.04946996
## Porsche 914-2        0.12272387 0.13780919
## Lotus Europa          0.05986530 0.21554770
## Ford Pantera L       0.69817910 0.74911661
## Ferrari Dino          0.18433525 0.43462898
## Maserati Bora         0.57345972 1.00000000
## Volvo 142E           0.12446994 0.20141343
##
## $model.list
## $model.list$response
## [1] "mpg"
##
## $model.list$variables
## [1] "disp" "hp"
##
##
## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x000000001fc6c340>
## <environment: 0x000000001fc6ac80>
## attr("type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))
## }
## <bytecode: 0x000000001fc678a8>
## <environment: 0x000000001fc66fb0>
## attr("type")
## [1] "logistic"
##
## $linear.output
## [1] TRUE
##
## $data
##              mpg cyl      disp        hp      drat
wt
## Mazda RX4          0.4510638 0.5 0.22175106 0.20494700 0.52534562

```

0.28304781	
## Mazda RX4 Wag	0.4510638 0.5 0.22175106 0.20494700 0.52534562
0.34824853	
## Datsun 710	0.5276596 0.0 0.09204290 0.14487633 0.50230415
0.20634109	
## Hornet 4 Drive	0.4680851 0.5 0.46620105 0.20494700 0.14746544
0.43518282	
## Hornet Sportabout	0.3531915 1.0 0.72062859 0.43462898 0.17972350
0.49271286	
## Valiant	0.3276596 0.5 0.38388626 0.18727915 0.00000000
0.49782664	
## Duster 360	0.1659574 1.0 0.72062859 0.68197880 0.20737327
0.52595244	
## Merc 240D	0.5957447 0.0 0.18857570 0.03533569 0.42857143
0.42879059	
## Merc 230	0.5276596 0.0 0.17385882 0.15194346 0.53456221
0.41856303	
## Merc 280	0.3744681 0.5 0.24070841 0.25088339 0.53456221
0.49271286	
## Merc 280C	0.3148936 0.5 0.24070841 0.25088339 0.53456221
0.49271286	
## Merc 450SE	0.2553191 1.0 0.51060115 0.45229682 0.14285714
0.65379698	
## Merc 450SL	0.2936170 1.0 0.51060115 0.45229682 0.14285714
0.56686269	
## Merc 450SLC	0.2042553 1.0 0.51060115 0.45229682 0.14285714
0.57964715	
## Cadillac Fleetwood	0.0000000 1.0 1.00000000 0.54063604 0.07834101
0.95551010	
## Lincoln Continental	0.0000000 1.0 0.97006735 0.57597173 0.11059908
1.00000000	
## Chrysler Imperial	0.1829787 1.0 0.92017960 0.62897527 0.21658986
0.97980056	
## Fiat 128	0.9361702 0.0 0.01895735 0.04946996 0.60829493
0.17565840	
## Honda Civic	0.8510638 0.0 0.01147418 0.00000000 1.00000000
0.02608029	
## Toyota Corolla	1.0000000 0.0 0.00000000 0.04593640 0.67281106
0.08233188	
## Toyota Corona	0.4723404 0.0 0.12222499 0.15901060 0.43317972
0.24341601	
## Dodge Challenger	0.2170213 1.0 0.61586431 0.34628975 0.00000000
0.51316799	
## AMC Javelin	0.2042553 1.0 0.58094288 0.34628975 0.17972350
0.49143442	
## Camaro Z28	0.1234043 1.0 0.69568471 0.68197880 0.44700461
0.59498849	
## Pontiac Firebird	0.3744681 1.0 0.82040409 0.43462898 0.14746544
0.59626694	
## Fiat X1-9	0.7191489 0.0 0.01970566 0.04946996 0.60829493

```

0.10790079
## Porsche 914-2      0.6638298 0.0 0.12272387 0.13780919 0.76958525
0.16031705
## Lotus Europa      0.8510638 0.0 0.05986530 0.21554770 0.46543779
0.00000000
## Ford Pantera L    0.2297872 1.0 0.69817910 0.74911661 0.67281106
0.42367681
## Ferrari Dino      0.3957447 0.5 0.18433525 0.43462898 0.39631336
0.32140118
## Maserati Bora     0.1957447 1.0 0.57345972 1.00000000 0.35944700
0.52595244
## Volvo 142E        0.4680851 0.0 0.12446994 0.20141343 0.62211982
0.32395807
##                  qsec vs am gear      carb
## Mazda RX4        0.23333333 0 1 0.5 0.4285714
## Mazda RX4 Wag    0.30000000 0 1 0.5 0.4285714
## Datsun 710        0.48928571 1 1 0.5 0.00000000
## Hornet 4 Drive    0.58809524 1 0 0.0 0.00000000
## Hornet Sportabout 0.30000000 0 0 0.0 0.1428571
## Valiant           0.68095238 1 0 0.0 0.00000000
## Duster 360        0.15952381 0 0 0.0 0.4285714
## Merc 240D         0.65476190 1 0 0.5 0.1428571
## Merc 230          1.00000000 1 0 0.5 0.1428571
## Merc 280          0.45238095 1 0 0.5 0.4285714
## Merc 280C         0.52380952 1 0 0.5 0.4285714
## Merc 450SE        0.34523810 0 0 0.0 0.2857143
## Merc 450SL        0.36904762 0 0 0.0 0.2857143
## Merc 450SLC       0.41666667 0 0 0.0 0.2857143
## Cadillac Fleetwood 0.41428571 0 0 0.0 0.4285714
## Lincoln Continental 0.39523810 0 0 0.0 0.4285714
## Chrysler Imperial 0.34761905 0 0 0.0 0.4285714
## Fiat 128          0.59166667 1 1 0.5 0.00000000
## Honda Civic       0.47857143 1 1 0.5 0.1428571
## Toyota Corolla    0.64285714 1 1 0.5 0.00000000
## Toyota Corona     0.65595238 1 0 0.0 0.00000000
## Dodge Challenger  0.28214286 0 0 0.0 0.1428571
## AMC Javelin       0.33333333 0 0 0.0 0.1428571
## Camaro Z28        0.10833333 0 0 0.0 0.4285714
## Pontiac Firebird  0.30357143 0 0 0.0 0.1428571
## Fiat X1-9         0.52380952 1 1 0.5 0.00000000
## Porsche 914-2     0.26190476 0 1 1.0 0.1428571
## Lotus Europa      0.28571429 1 1 1.0 0.1428571
## Ford Pantera L    0.00000000 0 1 1.0 0.4285714
## Ferrari Dino      0.11904762 0 1 1.0 0.7142857
## Maserati Bora     0.01190476 0 1 1.0 1.0000000
## Volvo 142E        0.48809524 1 1 0.5 0.1428571
##
## $exclude
## NULL
##

```

```

## $net.result
## $net.result[[1]]
##                                [,1]
## Mazda RX4                    0.42184184
## Mazda RX4 Wag                0.42184184
## Datsun 710                    0.64380473
## Hornet 4 Drive               0.33444979
## Hornet Sportabout           0.22461018
## Valiant                      0.35629309
## Duster 360                  0.18061349
## Merc 240D                   0.48147148
## Merc 230                    0.48026018
## Merc 280                    0.40296401
## Merc 280C                   0.40296401
## Merc 450SE                  0.29154665
## Merc 450SL                  0.29154665
## Merc 450SLC                 0.29154665
## Cadillac Fleetwood          0.09933257
## Lincoln Continental          0.10423581
## Chrysler Imperial           0.11353670
## Fiat 128                     0.86649834
## Honda Civic                  0.89709715
## Toyota Corolla               0.91077284
## Toyota Corona                0.56879390
## Dodge Challenger             0.27457060
## AMC Javelin                  0.28539310
## Camaro Z28                   0.19014274
## Pontiac Firebird             0.18741204
## Fiat X1-9                    0.86468489
## Porsche 914-2                0.57322128
## Lotus Europa                 0.70740127
## Ford Pantera L               0.17698460
## Ferrari Dino                 0.42076475
## Maserati Bora                0.17912652
## Volvo 142E                   0.55330749
##
##
## $weights
## $weights[[1]]
## $weights[[1]][[1]]
##                                [,1]    [,2]
## [1,] -1.085499 -3.082977
## [2,] 15.493586  2.767810
## [3,]  1.591692  1.315673
##
## $weights[[1]][[2]]
##                                [,1]
## [1,]  1.1195829
## [2,] -0.6861368
## [3,] -0.5589443

```

```

##
##
##
## $generalized.weights
## $generalized.weights[[1]]
##           [,1]      [,2]
## Mazda RX4      -3.195287 -0.5405825
## Mazda RX4 Wag  -3.195287 -0.5405825
## Datsun 710      -11.113565 -1.2983304
## Hornet 4 Drive  -1.095704 -0.4931870
## Hornet Sportabout -2.080202 -0.9883427
## Valiant         -1.099463 -0.4246247
## Duster 360      -2.590689 -1.2310997
## Merc 240D       -5.275380 -0.7018338
## Merc 230        -5.472189 -0.7380027
## Merc 280        -2.550942 -0.4973866
## Merc 280C       -2.550942 -0.4973866
## Merc 450SE      -1.448330 -0.6783357
## Merc 450SL      -1.448330 -0.6783357
## Merc 450SLC     -1.448330 -0.6783357
## Cadillac Fleetwood -4.157807 -1.9763950
## Lincoln Continental -4.011069 -1.9066386
## Chrysler Imperial -3.762422 -1.7884321
## Fiat 128        -20.908481 -2.3802481
## Honda Civic     -24.310542 -2.7667736
## Toyota Corolla  -26.414790 -3.0276333
## Toyota Corona   -8.706789 -1.0543361
## Dodge Challenger -1.587232 -0.7520523
## AMC Javelin     -1.488953 -0.7036861
## Camaro Z28      -2.471040 -1.1740678
## Pontiac Firebird -2.503631 -1.1899795
## Fiat X1-9       -20.751803 -2.3619493
## Porsche 914-2   -8.824009 -1.0631969
## Lotus Europa    -13.197839 -1.5299421
## Ford Pantera L  -2.638560 -1.2537440
## Ferrari Dino    -3.822807 -0.6408806
## Maserati Bora   -2.615145 -1.2408561
## Volvo 142E      -8.249625 -1.0147210
##
##
## $startweights
## $startweights[[1]]
## $startweights[[1]][[1]]
##           [,1]      [,2]
## [1,]  0.5855288 -0.4534972
## [2,]  0.7094660  0.6058875
## [3,] -0.1093033 -1.8179560
##
## $startweights[[1]][[2]]
##           [,1]

```

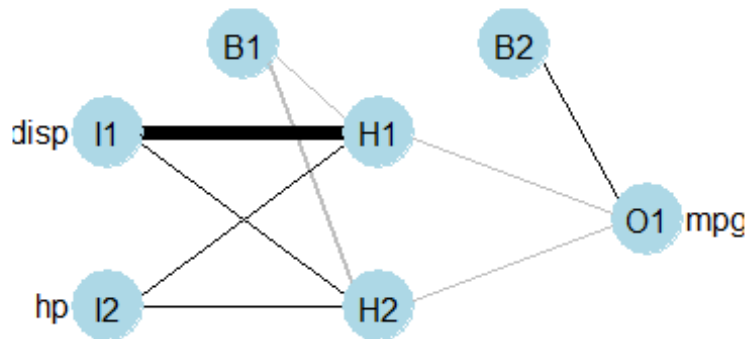
```
## [1,] 0.6300986
## [2,] -0.2761841
## [3,] -0.2841597
##
##
## $result.matrix
##                                [,1]
## error                        0.121773974
## reached.threshold            0.008093771
## steps                       260.000000000
## Intercept.to.1layhid1      -1.085498971
## disp.to.1layhid1           15.493586009
## hp.to.1layhid1              1.591692173
## Intercept.to.1layhid2     -3.082977358
## disp.to.1layhid2           2.767810494
## hp.to.1layhid2             1.315673187
## Intercept.to.mpg           1.119582852
## 1layhid1.to.mpg            -0.686136820
## 1layhid2.to.mpg            -0.558944306
##
## attr(,"class")
## [1] "nn"
```

Now we are going to predict mpg values using the neural net model we have constructed.

```
# Predict the mpg value
predictions <- compute(nn, scaled_mtcars[,c('disp', 'hp')])

# Train the neural network
set.seed(12345)
nn <- neuralnet(mpg ~ disp + hp, data = scaled_mtcars, hidden = 2)

# Plot the neural network with NeuralNetTools
plotnet(nn)
```

The `plotnet` function from the `NeuralNetTools` package produces a visualization of a neural network model trained using the `neuralnet` package. This diagram provides a schematic representation of the architecture and weights of the model. Here's how to interpret the output:

Layers: The diagram consists of three types of layers: an input layer on the left, hidden layers in the middle, and an output layer on the right. Each circle (also known as a node or a neuron) represents a variable (in the input layer) or a computation (in the hidden and output layers).

Nodes: Each node in the input layer represents a feature that the model uses to make its prediction. Each node in the hidden layer represents a transformation of the input features, and the node in the output layer represents the predicted outcome.

Edges: The lines connecting the nodes (also known as edges) represent the weights of the model. These weights determine the importance of the corresponding input or transformation in predicting the outcome. In some visualizations, the thickness and color of the line might indicate the size and sign of the weight (a thicker line for a larger absolute value, a red line for a negative weight, and a blue line for a positive weight).

Calculation of MSE:

```
# Generate predictions using the neural network model
predictions <- compute(nn, scaled_mtcars[,c('disp', 'hp')])

# The predictions are on the same scale as the training data. So we need to
```

```

convert them back.
actuals <- mtcars$mpg
predicted <- predictions$net.result * (maxs["mpg"] - mins["mpg"]) +
mins["mpg"]

# Calculate the mean squared error
mse <- mean((predicted - actuals)^2)
print(mse)

## [1] 4.203105

```

Q2. Predict the species of iris flowers based on four input variables: Sepal.Length, Sepal.Width, Petal.Length, and Petal.Width. Also create a neural network plot using the plotnet() function.

```

library(neuralnet)
library(NeuralNetTools)

# Load the iris dataset
data(iris)

# Convert the Species to numerical values
iris$Species <- as.numeric(iris$Species)

# Scale the dataset
maxs <- apply(iris, 2, max)
mins <- apply(iris, 2, min)
scaled_iris <- as.data.frame(scale(iris, center = mins, scale = maxs - mins))

# Create a neural network model
set.seed(12345)
nn <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length +
Petal.Width,
                data = scaled_iris, hidden = 5)

# Print the output
print(nn)

## $call
## neuralnet(formula = Species ~ Sepal.Length + Sepal.Width + Petal.Length +
##   Petal.Width, data = scaled_iris, hidden = 5)
##
## $response
##   Species
## 1      0.0
## 2      0.0
## 3      0.0
## 4      0.0
## 5      0.0
## 6      0.0

```

## 7	0.0
## 8	0.0
## 9	0.0
## 10	0.0
## 11	0.0
## 12	0.0
## 13	0.0
## 14	0.0
## 15	0.0
## 16	0.0
## 17	0.0
## 18	0.0
## 19	0.0
## 20	0.0
## 21	0.0
## 22	0.0
## 23	0.0
## 24	0.0
## 25	0.0
## 26	0.0
## 27	0.0
## 28	0.0
## 29	0.0
## 30	0.0
## 31	0.0
## 32	0.0
## 33	0.0
## 34	0.0
## 35	0.0
## 36	0.0
## 37	0.0
## 38	0.0
## 39	0.0
## 40	0.0
## 41	0.0
## 42	0.0
## 43	0.0
## 44	0.0
## 45	0.0
## 46	0.0
## 47	0.0
## 48	0.0
## 49	0.0
## 50	0.0
## 51	0.5
## 52	0.5
## 53	0.5
## 54	0.5
## 55	0.5
## 56	0.5

## 57	0.5
## 58	0.5
## 59	0.5
## 60	0.5
## 61	0.5
## 62	0.5
## 63	0.5
## 64	0.5
## 65	0.5
## 66	0.5
## 67	0.5
## 68	0.5
## 69	0.5
## 70	0.5
## 71	0.5
## 72	0.5
## 73	0.5
## 74	0.5
## 75	0.5
## 76	0.5
## 77	0.5
## 78	0.5
## 79	0.5
## 80	0.5
## 81	0.5
## 82	0.5
## 83	0.5
## 84	0.5
## 85	0.5
## 86	0.5
## 87	0.5
## 88	0.5
## 89	0.5
## 90	0.5
## 91	0.5
## 92	0.5
## 93	0.5
## 94	0.5
## 95	0.5
## 96	0.5
## 97	0.5
## 98	0.5
## 99	0.5
## 100	0.5
## 101	1.0
## 102	1.0
## 103	1.0
## 104	1.0
## 105	1.0
## 106	1.0

```

## 107      1.0
## 108      1.0
## 109      1.0
## 110      1.0
## 111      1.0
## 112      1.0
## 113      1.0
## 114      1.0
## 115      1.0
## 116      1.0
## 117      1.0
## 118      1.0
## 119      1.0
## 120      1.0
## 121      1.0
## 122      1.0
## 123      1.0
## 124      1.0
## 125      1.0
## 126      1.0
## 127      1.0
## 128      1.0
## 129      1.0
## 130      1.0
## 131      1.0
## 132      1.0
## 133      1.0
## 134      1.0
## 135      1.0
## 136      1.0
## 137      1.0
## 138      1.0
## 139      1.0
## 140      1.0
## 141      1.0
## 142      1.0
## 143      1.0
## 144      1.0
## 145      1.0
## 146      1.0
## 147      1.0
## 148      1.0
## 149      1.0
## 150      1.0
##
## $covariate
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## [1,] 0.22222222 0.62500000 0.06779661 0.04166667
## [2,] 0.16666667 0.41666667 0.06779661 0.04166667
## [3,] 0.11111111 0.50000000 0.05084746 0.04166667

```

##	[4,]	0.08333333	0.45833333	0.08474576	0.04166667
##	[5,]	0.19444444	0.66666667	0.06779661	0.04166667
##	[6,]	0.30555556	0.79166667	0.11864407	0.12500000
##	[7,]	0.08333333	0.58333333	0.06779661	0.08333333
##	[8,]	0.19444444	0.58333333	0.08474576	0.04166667
##	[9,]	0.02777778	0.37500000	0.06779661	0.04166667
##	[10,]	0.16666667	0.45833333	0.08474576	0.00000000
##	[11,]	0.30555556	0.70833333	0.08474576	0.04166667
##	[12,]	0.13888889	0.58333333	0.10169492	0.04166667
##	[13,]	0.13888889	0.41666667	0.06779661	0.00000000
##	[14,]	0.00000000	0.41666667	0.01694915	0.00000000
##	[15,]	0.41666667	0.83333333	0.03389831	0.04166667
##	[16,]	0.38888889	1.00000000	0.08474576	0.12500000
##	[17,]	0.30555556	0.79166667	0.05084746	0.12500000
##	[18,]	0.22222222	0.62500000	0.06779661	0.08333333
##	[19,]	0.38888889	0.75000000	0.11864407	0.08333333
##	[20,]	0.22222222	0.75000000	0.08474576	0.08333333
##	[21,]	0.30555556	0.58333333	0.11864407	0.04166667
##	[22,]	0.22222222	0.70833333	0.08474576	0.12500000
##	[23,]	0.08333333	0.66666667	0.00000000	0.04166667
##	[24,]	0.22222222	0.54166667	0.11864407	0.16666667
##	[25,]	0.13888889	0.58333333	0.15254237	0.04166667
##	[26,]	0.19444444	0.41666667	0.10169492	0.04166667
##	[27,]	0.19444444	0.58333333	0.10169492	0.12500000
##	[28,]	0.25000000	0.62500000	0.08474576	0.04166667
##	[29,]	0.25000000	0.58333333	0.06779661	0.04166667
##	[30,]	0.11111111	0.50000000	0.10169492	0.04166667
##	[31,]	0.13888889	0.45833333	0.10169492	0.04166667
##	[32,]	0.30555556	0.58333333	0.08474576	0.12500000
##	[33,]	0.25000000	0.87500000	0.08474576	0.00000000
##	[34,]	0.33333333	0.91666667	0.06779661	0.04166667
##	[35,]	0.16666667	0.45833333	0.08474576	0.04166667
##	[36,]	0.19444444	0.50000000	0.03389831	0.04166667
##	[37,]	0.33333333	0.62500000	0.05084746	0.04166667
##	[38,]	0.16666667	0.66666667	0.06779661	0.00000000
##	[39,]	0.02777778	0.41666667	0.05084746	0.04166667
##	[40,]	0.22222222	0.58333333	0.08474576	0.04166667
##	[41,]	0.19444444	0.62500000	0.05084746	0.08333333
##	[42,]	0.05555556	0.12500000	0.05084746	0.08333333
##	[43,]	0.02777778	0.50000000	0.05084746	0.04166667
##	[44,]	0.19444444	0.62500000	0.10169492	0.20833333
##	[45,]	0.22222222	0.75000000	0.15254237	0.12500000
##	[46,]	0.13888889	0.41666667	0.06779661	0.08333333
##	[47,]	0.22222222	0.75000000	0.10169492	0.04166667
##	[48,]	0.08333333	0.50000000	0.06779661	0.04166667
##	[49,]	0.27777778	0.70833333	0.08474576	0.04166667
##	[50,]	0.19444444	0.54166667	0.06779661	0.04166667
##	[51,]	0.75000000	0.50000000	0.62711864	0.54166667
##	[52,]	0.58333333	0.50000000	0.59322034	0.58333333
##	[53,]	0.72222222	0.45833333	0.66101695	0.58333333

##	[54,]	0.33333333	0.12500000	0.50847458	0.50000000
##	[55,]	0.61111111	0.33333333	0.61016949	0.58333333
##	[56,]	0.38888889	0.33333333	0.59322034	0.50000000
##	[57,]	0.55555556	0.54166667	0.62711864	0.62500000
##	[58,]	0.16666667	0.16666667	0.38983051	0.37500000
##	[59,]	0.63888889	0.37500000	0.61016949	0.50000000
##	[60,]	0.25000000	0.29166667	0.49152542	0.54166667
##	[61,]	0.19444444	0.00000000	0.42372881	0.37500000
##	[62,]	0.44444444	0.41666667	0.54237288	0.58333333
##	[63,]	0.47222222	0.08333333	0.50847458	0.37500000
##	[64,]	0.50000000	0.37500000	0.62711864	0.54166667
##	[65,]	0.36111111	0.37500000	0.44067797	0.50000000
##	[66,]	0.66666667	0.45833333	0.57627119	0.54166667
##	[67,]	0.36111111	0.41666667	0.59322034	0.58333333
##	[68,]	0.41666667	0.29166667	0.52542373	0.37500000
##	[69,]	0.52777778	0.08333333	0.59322034	0.58333333
##	[70,]	0.36111111	0.20833333	0.49152542	0.41666667
##	[71,]	0.44444444	0.50000000	0.64406780	0.70833333
##	[72,]	0.50000000	0.33333333	0.50847458	0.50000000
##	[73,]	0.55555556	0.20833333	0.66101695	0.58333333
##	[74,]	0.50000000	0.33333333	0.62711864	0.45833333
##	[75,]	0.58333333	0.37500000	0.55932203	0.50000000
##	[76,]	0.63888889	0.41666667	0.57627119	0.54166667
##	[77,]	0.69444444	0.33333333	0.64406780	0.54166667
##	[78,]	0.66666667	0.41666667	0.67796610	0.66666667
##	[79,]	0.47222222	0.37500000	0.59322034	0.58333333
##	[80,]	0.38888889	0.25000000	0.42372881	0.37500000
##	[81,]	0.33333333	0.16666667	0.47457627	0.41666667
##	[82,]	0.33333333	0.16666667	0.45762712	0.37500000
##	[83,]	0.41666667	0.29166667	0.49152542	0.45833333
##	[84,]	0.47222222	0.29166667	0.69491525	0.62500000
##	[85,]	0.30555556	0.41666667	0.59322034	0.58333333
##	[86,]	0.47222222	0.58333333	0.59322034	0.62500000
##	[87,]	0.66666667	0.45833333	0.62711864	0.58333333
##	[88,]	0.55555556	0.12500000	0.57627119	0.50000000
##	[89,]	0.36111111	0.41666667	0.52542373	0.50000000
##	[90,]	0.33333333	0.20833333	0.50847458	0.50000000
##	[91,]	0.33333333	0.25000000	0.57627119	0.45833333
##	[92,]	0.50000000	0.41666667	0.61016949	0.54166667
##	[93,]	0.41666667	0.25000000	0.50847458	0.45833333
##	[94,]	0.19444444	0.12500000	0.38983051	0.37500000
##	[95,]	0.36111111	0.29166667	0.54237288	0.50000000
##	[96,]	0.38888889	0.41666667	0.54237288	0.45833333
##	[97,]	0.38888889	0.37500000	0.54237288	0.50000000
##	[98,]	0.52777778	0.37500000	0.55932203	0.50000000
##	[99,]	0.22222222	0.20833333	0.33898305	0.41666667
##	[100,]	0.38888889	0.33333333	0.52542373	0.50000000
##	[101,]	0.55555556	0.54166667	0.84745763	1.00000000
##	[102,]	0.41666667	0.29166667	0.69491525	0.75000000
##	[103,]	0.77777778	0.41666667	0.83050847	0.83333333

```

## [104,] 0.55555556 0.37500000 0.77966102 0.70833333
## [105,] 0.61111111 0.41666667 0.81355932 0.87500000
## [106,] 0.91666667 0.41666667 0.94915254 0.83333333
## [107,] 0.16666667 0.20833333 0.59322034 0.66666667
## [108,] 0.83333333 0.37500000 0.89830508 0.70833333
## [109,] 0.66666667 0.20833333 0.81355932 0.70833333
## [110,] 0.80555556 0.66666667 0.86440678 1.00000000
## [111,] 0.61111111 0.50000000 0.69491525 0.79166667
## [112,] 0.58333333 0.29166667 0.72881356 0.75000000
## [113,] 0.69444444 0.41666667 0.76271186 0.83333333
## [114,] 0.38888889 0.20833333 0.67796610 0.79166667
## [115,] 0.41666667 0.33333333 0.69491525 0.95833333
## [116,] 0.58333333 0.50000000 0.72881356 0.91666667
## [117,] 0.61111111 0.41666667 0.76271186 0.70833333
## [118,] 0.94444444 0.75000000 0.96610169 0.87500000
## [119,] 0.94444444 0.25000000 1.00000000 0.91666667
## [120,] 0.47222222 0.08333333 0.67796610 0.58333333
## [121,] 0.72222222 0.50000000 0.79661017 0.91666667
## [122,] 0.36111111 0.33333333 0.66101695 0.79166667
## [123,] 0.94444444 0.33333333 0.96610169 0.79166667
## [124,] 0.55555556 0.29166667 0.66101695 0.70833333
## [125,] 0.66666667 0.54166667 0.79661017 0.83333333
## [126,] 0.80555556 0.50000000 0.84745763 0.70833333
## [127,] 0.52777778 0.33333333 0.64406780 0.70833333
## [128,] 0.50000000 0.41666667 0.66101695 0.70833333
## [129,] 0.58333333 0.33333333 0.77966102 0.83333333
## [130,] 0.80555556 0.41666667 0.81355932 0.62500000
## [131,] 0.86111111 0.33333333 0.86440678 0.75000000
## [132,] 1.00000000 0.75000000 0.91525424 0.79166667
## [133,] 0.58333333 0.33333333 0.77966102 0.87500000
## [134,] 0.55555556 0.33333333 0.69491525 0.58333333
## [135,] 0.50000000 0.25000000 0.77966102 0.54166667
## [136,] 0.94444444 0.41666667 0.86440678 0.91666667
## [137,] 0.55555556 0.58333333 0.77966102 0.95833333
## [138,] 0.58333333 0.45833333 0.76271186 0.70833333
## [139,] 0.47222222 0.41666667 0.64406780 0.70833333
## [140,] 0.72222222 0.45833333 0.74576271 0.83333333
## [141,] 0.66666667 0.45833333 0.77966102 0.95833333
## [142,] 0.72222222 0.45833333 0.69491525 0.91666667
## [143,] 0.41666667 0.29166667 0.69491525 0.75000000
## [144,] 0.69444444 0.50000000 0.83050847 0.91666667
## [145,] 0.66666667 0.54166667 0.79661017 1.00000000
## [146,] 0.66666667 0.41666667 0.71186441 0.91666667
## [147,] 0.55555556 0.20833333 0.67796610 0.75000000
## [148,] 0.61111111 0.41666667 0.71186441 0.79166667
## [149,] 0.52777778 0.58333333 0.74576271 0.91666667
## [150,] 0.44444444 0.41666667 0.69491525 0.70833333
##
## $model.list
## $model.list$response

```



```

## [1] "Species"
##
## $model.list$variables
## [1] "Sepal.Length" "Sepal.Width" "Petal.Length" "Petal.Width"
##
##
## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x000000001fc6c340>
## <environment: 0x00000000158b6698>
## attr(,"type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))
## }
## <bytecode: 0x000000001fc678a8>
## <environment: 0x00000000158b6200>
## attr(,"type")
## [1] "logistic"
##
## $linear.output
## [1] TRUE
##
## $data
##      Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1      0.22222222  0.62500000  0.06779661  0.04166667      0.0
## 2      0.16666667  0.41666667  0.06779661  0.04166667      0.0
## 3      0.11111111  0.50000000  0.05084746  0.04166667      0.0
## 4      0.08333333  0.45833333  0.08474576  0.04166667      0.0
## 5      0.19444444  0.66666667  0.06779661  0.04166667      0.0
## 6      0.30555556  0.79166667  0.11864407  0.12500000      0.0
## 7      0.08333333  0.58333333  0.06779661  0.08333333      0.0
## 8      0.19444444  0.58333333  0.08474576  0.04166667      0.0
## 9      0.02777778  0.37500000  0.06779661  0.04166667      0.0
## 10     0.16666667  0.45833333  0.08474576  0.00000000      0.0
## 11     0.30555556  0.70833333  0.08474576  0.04166667      0.0
## 12     0.13888889  0.58333333  0.10169492  0.04166667      0.0
## 13     0.13888889  0.41666667  0.06779661  0.00000000      0.0
## 14     0.00000000  0.41666667  0.01694915  0.00000000      0.0
## 15     0.41666667  0.83333333  0.03389831  0.04166667      0.0
## 16     0.38888889  1.00000000  0.08474576  0.12500000      0.0
## 17     0.30555556  0.79166667  0.05084746  0.12500000      0.0
## 18     0.22222222  0.62500000  0.06779661  0.08333333      0.0
## 19     0.38888889  0.75000000  0.11864407  0.08333333      0.0

```

## 20	0.22222222	0.75000000	0.08474576	0.08333333	0.0
## 21	0.30555556	0.58333333	0.11864407	0.04166667	0.0
## 22	0.22222222	0.70833333	0.08474576	0.12500000	0.0
## 23	0.08333333	0.66666667	0.00000000	0.04166667	0.0
## 24	0.22222222	0.54166667	0.11864407	0.16666667	0.0
## 25	0.13888889	0.58333333	0.15254237	0.04166667	0.0
## 26	0.19444444	0.41666667	0.10169492	0.04166667	0.0
## 27	0.19444444	0.58333333	0.10169492	0.12500000	0.0
## 28	0.25000000	0.62500000	0.08474576	0.04166667	0.0
## 29	0.25000000	0.58333333	0.06779661	0.04166667	0.0
## 30	0.11111111	0.50000000	0.10169492	0.04166667	0.0
## 31	0.13888889	0.45833333	0.10169492	0.04166667	0.0
## 32	0.30555556	0.58333333	0.08474576	0.12500000	0.0
## 33	0.25000000	0.87500000	0.08474576	0.00000000	0.0
## 34	0.33333333	0.91666667	0.06779661	0.04166667	0.0
## 35	0.16666667	0.45833333	0.08474576	0.04166667	0.0
## 36	0.19444444	0.50000000	0.03389831	0.04166667	0.0
## 37	0.33333333	0.62500000	0.05084746	0.04166667	0.0
## 38	0.16666667	0.66666667	0.06779661	0.00000000	0.0
## 39	0.02777778	0.41666667	0.05084746	0.04166667	0.0
## 40	0.22222222	0.58333333	0.08474576	0.04166667	0.0
## 41	0.19444444	0.62500000	0.05084746	0.08333333	0.0
## 42	0.05555556	0.12500000	0.05084746	0.08333333	0.0
## 43	0.02777778	0.50000000	0.05084746	0.04166667	0.0
## 44	0.19444444	0.62500000	0.10169492	0.20833333	0.0
## 45	0.22222222	0.75000000	0.15254237	0.12500000	0.0
## 46	0.13888889	0.41666667	0.06779661	0.08333333	0.0
## 47	0.22222222	0.75000000	0.10169492	0.04166667	0.0
## 48	0.08333333	0.50000000	0.06779661	0.04166667	0.0
## 49	0.27777778	0.70833333	0.08474576	0.04166667	0.0
## 50	0.19444444	0.54166667	0.06779661	0.04166667	0.0
## 51	0.75000000	0.50000000	0.62711864	0.54166667	0.5
## 52	0.58333333	0.50000000	0.59322034	0.58333333	0.5
## 53	0.72222222	0.45833333	0.66101695	0.58333333	0.5
## 54	0.33333333	0.12500000	0.50847458	0.50000000	0.5
## 55	0.61111111	0.33333333	0.61016949	0.58333333	0.5
## 56	0.38888889	0.33333333	0.59322034	0.50000000	0.5
## 57	0.55555556	0.54166667	0.62711864	0.62500000	0.5
## 58	0.16666667	0.16666667	0.38983051	0.37500000	0.5
## 59	0.63888889	0.37500000	0.61016949	0.50000000	0.5
## 60	0.25000000	0.29166667	0.49152542	0.54166667	0.5
## 61	0.19444444	0.00000000	0.42372881	0.37500000	0.5
## 62	0.44444444	0.41666667	0.54237288	0.58333333	0.5
## 63	0.47222222	0.08333333	0.50847458	0.37500000	0.5
## 64	0.50000000	0.37500000	0.62711864	0.54166667	0.5
## 65	0.36111111	0.37500000	0.44067797	0.50000000	0.5
## 66	0.66666667	0.45833333	0.57627119	0.54166667	0.5
## 67	0.36111111	0.41666667	0.59322034	0.58333333	0.5
## 68	0.41666667	0.29166667	0.52542373	0.37500000	0.5
## 69	0.52777778	0.08333333	0.59322034	0.58333333	0.5

## 70	0.36111111	0.20833333	0.49152542	0.41666667	0.5
## 71	0.44444444	0.50000000	0.64406780	0.70833333	0.5
## 72	0.50000000	0.33333333	0.50847458	0.50000000	0.5
## 73	0.55555556	0.20833333	0.66101695	0.58333333	0.5
## 74	0.50000000	0.33333333	0.62711864	0.45833333	0.5
## 75	0.58333333	0.37500000	0.55932203	0.50000000	0.5
## 76	0.63888889	0.41666667	0.57627119	0.54166667	0.5
## 77	0.69444444	0.33333333	0.64406780	0.54166667	0.5
## 78	0.66666667	0.41666667	0.67796610	0.66666667	0.5
## 79	0.47222222	0.37500000	0.59322034	0.58333333	0.5
## 80	0.38888889	0.25000000	0.42372881	0.37500000	0.5
## 81	0.33333333	0.16666667	0.47457627	0.41666667	0.5
## 82	0.33333333	0.16666667	0.45762712	0.37500000	0.5
## 83	0.41666667	0.29166667	0.49152542	0.45833333	0.5
## 84	0.47222222	0.29166667	0.69491525	0.62500000	0.5
## 85	0.30555556	0.41666667	0.59322034	0.58333333	0.5
## 86	0.47222222	0.58333333	0.59322034	0.62500000	0.5
## 87	0.66666667	0.45833333	0.62711864	0.58333333	0.5
## 88	0.55555556	0.12500000	0.57627119	0.50000000	0.5
## 89	0.36111111	0.41666667	0.52542373	0.50000000	0.5
## 90	0.33333333	0.20833333	0.50847458	0.50000000	0.5
## 91	0.33333333	0.25000000	0.57627119	0.45833333	0.5
## 92	0.50000000	0.41666667	0.61016949	0.54166667	0.5
## 93	0.41666667	0.25000000	0.50847458	0.45833333	0.5
## 94	0.19444444	0.12500000	0.38983051	0.37500000	0.5
## 95	0.36111111	0.29166667	0.54237288	0.50000000	0.5
## 96	0.38888889	0.41666667	0.54237288	0.45833333	0.5
## 97	0.38888889	0.37500000	0.54237288	0.50000000	0.5
## 98	0.52777778	0.37500000	0.55932203	0.50000000	0.5
## 99	0.22222222	0.20833333	0.33898305	0.41666667	0.5
## 100	0.38888889	0.33333333	0.52542373	0.50000000	0.5
## 101	0.55555556	0.54166667	0.84745763	1.00000000	1.0
## 102	0.41666667	0.29166667	0.69491525	0.75000000	1.0
## 103	0.77777778	0.41666667	0.83050847	0.83333333	1.0
## 104	0.55555556	0.37500000	0.77966102	0.70833333	1.0
## 105	0.61111111	0.41666667	0.81355932	0.87500000	1.0
## 106	0.91666667	0.41666667	0.94915254	0.83333333	1.0
## 107	0.16666667	0.20833333	0.59322034	0.66666667	1.0
## 108	0.83333333	0.37500000	0.89830508	0.70833333	1.0
## 109	0.66666667	0.20833333	0.81355932	0.70833333	1.0
## 110	0.80555556	0.66666667	0.86440678	1.00000000	1.0
## 111	0.61111111	0.50000000	0.69491525	0.79166667	1.0
## 112	0.58333333	0.29166667	0.72881356	0.75000000	1.0
## 113	0.69444444	0.41666667	0.76271186	0.83333333	1.0
## 114	0.38888889	0.20833333	0.67796610	0.79166667	1.0
## 115	0.41666667	0.33333333	0.69491525	0.95833333	1.0
## 116	0.58333333	0.50000000	0.72881356	0.91666667	1.0
## 117	0.61111111	0.41666667	0.76271186	0.70833333	1.0
## 118	0.94444444	0.75000000	0.96610169	0.87500000	1.0
## 119	0.94444444	0.25000000	1.00000000	0.91666667	1.0

```

## 120 0.47222222 0.08333333 0.67796610 0.58333333 1.0
## 121 0.72222222 0.50000000 0.79661017 0.91666667 1.0
## 122 0.36111111 0.33333333 0.66101695 0.79166667 1.0
## 123 0.94444444 0.33333333 0.96610169 0.79166667 1.0
## 124 0.55555556 0.29166667 0.66101695 0.70833333 1.0
## 125 0.66666667 0.54166667 0.79661017 0.83333333 1.0
## 126 0.80555556 0.50000000 0.84745763 0.70833333 1.0
## 127 0.52777778 0.33333333 0.64406780 0.70833333 1.0
## 128 0.50000000 0.41666667 0.66101695 0.70833333 1.0
## 129 0.58333333 0.33333333 0.77966102 0.83333333 1.0
## 130 0.80555556 0.41666667 0.81355932 0.62500000 1.0
## 131 0.86111111 0.33333333 0.86440678 0.75000000 1.0
## 132 1.00000000 0.75000000 0.91525424 0.79166667 1.0
## 133 0.58333333 0.33333333 0.77966102 0.87500000 1.0
## 134 0.55555556 0.33333333 0.69491525 0.58333333 1.0
## 135 0.50000000 0.25000000 0.77966102 0.54166667 1.0
## 136 0.94444444 0.41666667 0.86440678 0.91666667 1.0
## 137 0.55555556 0.58333333 0.77966102 0.95833333 1.0
## 138 0.58333333 0.45833333 0.76271186 0.70833333 1.0
## 139 0.47222222 0.41666667 0.64406780 0.70833333 1.0
## 140 0.72222222 0.45833333 0.74576271 0.83333333 1.0
## 141 0.66666667 0.45833333 0.77966102 0.95833333 1.0
## 142 0.72222222 0.45833333 0.69491525 0.91666667 1.0
## 143 0.41666667 0.29166667 0.69491525 0.75000000 1.0
## 144 0.69444444 0.50000000 0.83050847 0.91666667 1.0
## 145 0.66666667 0.54166667 0.79661017 1.00000000 1.0
## 146 0.66666667 0.41666667 0.71186441 0.91666667 1.0
## 147 0.55555556 0.20833333 0.67796610 0.75000000 1.0
## 148 0.61111111 0.41666667 0.71186441 0.79166667 1.0
## 149 0.52777778 0.58333333 0.74576271 0.91666667 1.0
## 150 0.44444444 0.41666667 0.69491525 0.70833333 1.0
##
## $exclude
## NULL
##
## $net.result
## $net.result[[1]]
##           [,1]
## [1,] -0.0086914078
## [2,] -0.0030266962
## [3,] 0.0009198357
## [4,] 0.0192867055
## [5,] -0.0057320049
## [6,] 0.0150618527
## [7,] 0.0192488469
## [8,] 0.0018135313
## [9,] 0.0216968943
## [10,] -0.0049541760
## [11,] -0.0135699572
## [12,] 0.0157314226

```

```
## [13,] -0.0084953308
## [14,] -0.0059459207
## [15,] -0.0439319947
## [16,] -0.0093306517
## [17,] -0.0085438099
## [18,]  0.0008031175
## [19,] -0.0028171096
## [20,]  0.0043618200
## [21,] -0.0012892842
## [22,]  0.0146495593
## [23,] -0.0169625505
## [24,]  0.0429608332
## [25,]  0.0359237417
## [26,]  0.0060841244
## [27,]  0.0283388358
## [28,] -0.0060162978
## [29,] -0.0126687461
## [30,]  0.0215056271
## [31,]  0.0176136317
## [32,]  0.0053641369
## [33,] -0.0182387092
## [34,] -0.0242343767
## [35,]  0.0056662284
## [36,] -0.0188616423
## [37,] -0.0308721149
## [38,] -0.0115646434
## [39,]  0.0141469753
## [40,] -0.0020606610
## [41,] -0.0019750056
## [42,] -0.0057591566
## [43,]  0.0124288409
## [44,]  0.0469661835
## [45,]  0.0379741515
## [46,]  0.0134646114
## [47,]  0.0014432943
## [48,]  0.0116593663
## [49,] -0.0101677715
## [50,] -0.0047178757
## [51,]  0.4897162979
## [52,]  0.5173818139
## [53,]  0.6086299664
## [54,]  0.6322536966
## [55,]  0.6458168152
## [56,]  0.6051544704
## [57,]  0.5641748953
## [58,]  0.3920117803
## [59,]  0.5325658585
## [60,]  0.5588264786
## [61,]  0.4949951215
## [62,]  0.5352291990
```

```
## [63,] 0.4076146555
## [64,] 0.6367323472
## [65,] 0.3913019513
## [66,] 0.4706256701
## [67,] 0.6101539926
## [68,] 0.4243325035
## [69,] 0.6861636758
## [70,] 0.4763251527
## [71,] 0.7016530511
## [72,] 0.4635469430
## [73,] 0.7773112214
## [74,] 0.5883203679
## [75,] 0.4823305166
## [76,] 0.5024579632
## [77,] 0.6128640923
## [78,] 0.7535017356
## [79,] 0.6328210048
## [80,] 0.3253857444
## [81,] 0.4829836276
## [82,] 0.4193928699
## [83,] 0.4526431151
## [84,] 0.8674095152
## [85,] 0.6137158132
## [86,] 0.4955283500
## [87,] 0.5765688300
## [88,] 0.5748542913
## [89,] 0.4599129653
## [90,] 0.5891611210
## [91,] 0.6177232122
## [92,] 0.5808848406
## [93,] 0.4975668070
## [94,] 0.4027550956
## [95,] 0.5722672006
## [96,] 0.4452615340
## [97,] 0.5058928700
## [98,] 0.4991416317
## [99,] 0.3317874736
## [100,] 0.5138604316
## [101,] 1.1224092578
## [102,] 0.9763355882
## [103,] 1.0228754671
## [104,] 0.9681402514
## [105,] 1.0797064766
## [106,] 1.0565964964
## [107,] 0.8873992850
## [108,] 0.9765856535
## [109,] 0.9534591008
## [110,] 1.0421926562
## [111,] 0.8321799858
## [112,] 0.9629750362
```

```

## [113,] 0.9833684920
## [114,] 1.0258278447
## [115,] 1.0859237684
## [116,] 0.9745592677
## [117,] 0.9128068125
## [118,] 0.9998359289
## [119,] 0.9957785080
## [120,] 0.8304066995
## [121,] 1.0303098453
## [122,] 0.9500359579
## [123,] 0.9778422132
## [124,] 0.8664418936
## [125,] 0.9542964304
## [126,] 0.9209820537
## [127,] 0.8338357824
## [128,] 0.7996252388
## [129,] 1.0572072760
## [130,] 0.8404113291
## [131,] 0.9373669373
## [132,] 0.8681247296
## [133,] 1.0807356832
## [134,] 0.7838279914
## [135,] 0.9088925910
## [136,] 1.0169016562
## [137,] 0.9965445594
## [138,] 0.8889332602
## [139,] 0.7796489070
## [140,] 0.9392606939
## [141,] 1.0681259369
## [142,] 0.9422437029
## [143,] 0.9763355882
## [144,] 1.0686288399
## [145,] 1.0656209361
## [146,] 0.9931629412
## [147,] 0.9280548815
## [148,] 0.9160829209
## [149,] 0.9254856336
## [150,] 0.8506341337
##
##
## $weights
## $weights[[1]]
## $weights[[1]][[1]]
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 1.2118446 -1.7329851 -0.6269448 3.6976994 0.5834111
## [2,] 0.3571057 4.6067043 0.5741632 -0.7837544 0.9161600
## [3,] 1.0413444 -6.2172964 -1.0809956 5.5128476 -1.2208272
## [4,] -1.4887963 -0.6877557 1.2307358 -5.2359423 -0.7973437
## [5,] -1.0778383 -1.2384032 -0.4820558 -3.6042089 -1.2517228
##

```

```

## $weights[[1]][[2]]
##      [,1]
## [1,]  1.4211563
## [2,] -0.7918154
## [3,] -0.9813023
## [4,]  0.5249991
## [5,] -0.6354708
## [6,] -0.4949521
##
##
##
## $generalized.weights
## $generalized.weights[[1]]
##      [,1]      [,2]      [,3]      [,4]
## [1,]  14.9887916  7.009057e-01 -42.658671 -25.794512
## [2,]  66.1921664 -1.652502e+01 -142.096815 -87.552197
## [3,] -158.8399216 -6.309122e+00  432.860547 263.357040
## [4,]  -8.0381218 -7.238172e-01   22.112080  13.373229
## [5,]  20.6590134  3.008434e+00  -63.021904 -38.190239
## [6,]  -7.3141933 -2.040435e+00   23.993065  14.471409
## [7,]  -6.4125730 -1.880334e+00   20.327694  12.386860
## [8,] -75.5708758 -5.256969e+00  214.213885 129.013661
## [9,]  -8.0296692 -1.319226e-01   20.586576  12.568563
## [10,] 35.9681161 -6.243744e+00  -83.813740 -50.750008
## [11,]  9.0408419  3.446807e-01  -26.253238 -15.796455
## [12,] -8.2202841 -1.768203e+00   25.209196  15.122423
## [13,] 22.4437831 -5.185643e+00  -49.489334 -30.215714
## [14,] 25.2262479  5.041341e-01  -67.277401 -41.073085
## [15,]  2.5246335 -2.439236e-01   -7.118580  -4.367308
## [16,]  9.9939191  1.593381e+00  -32.567083 -20.175718
## [17,] 12.8282389  1.741867e+00  -39.521064 -24.395666
## [18,] -162.7993734 -1.538414e+01  470.647076 286.105218
## [19,]  44.3925201  2.304813e+00 -129.323286 -77.385556
## [20,] -25.0582932 -6.145331e+00   81.122049  49.350887
## [21,] 125.3653957 -1.532037e+01 -311.015520 -186.727470
## [22,]  -7.9020463 -2.115205e+00   25.284579  15.404293
## [23,]  6.4553538  1.043452e+00  -19.979531 -12.456239
## [24,] -3.6359131 -6.239101e-01   10.330510   6.281178
## [25,] -3.6277164 -1.177796e+00   11.664895   6.911045
## [26,] -34.9949022  9.308500e+00  73.458636  45.186750
## [27,] -4.8699331 -1.028817e+00   14.556142   8.815740
## [28,] 22.4071066  5.193535e-01  -62.647533 -37.706205
## [29,] 11.6288874 -9.462225e-01  -30.053160 -18.201283
## [30,] -6.8407637 -9.558498e-01   19.580200  11.772937
## [31,] -9.6123433  1.568078e-01   24.621157  14.891122
## [32,] -29.8553907  2.292501e+00  75.184203  45.940908
## [33,]  5.2951777  8.293882e-01  -17.335092 -10.575897
## [34,]  3.9881442  3.607865e-01  -12.675015  -7.804814
## [35,] -31.3077076  3.750025e+00  74.986525  45.636682
## [36,]  8.7761640 -1.592862e+00  -20.654425 -12.671440

```


##	[37,]	4.8982165	-1.037090e+00	-11.756769	-7.159338
##	[38,]	9.9523353	1.441336e+00	-30.712810	-18.534581
##	[39,]	-11.0919871	-8.047751e-01	29.994283	18.293449
##	[40,]	68.8935512	7.009541e-01	-188.298754	-113.453041
##	[41,]	64.2441375	7.251976e+00	-188.233783	-115.030763
##	[42,]	74.2293261	-4.590795e+01	-99.015388	-67.350426
##	[43,]	-10.6902828	-2.267238e+00	32.171358	19.576615
##	[44,]	-2.7502841	-1.065503e+00	8.884213	5.419576
##	[45,]	-2.9180411	-1.310937e+00	10.167251	6.075756
##	[46,]	-13.9961417	1.683623e+00	32.764450	20.194154
##	[47,]	-75.2932345	-1.773744e+01	245.136442	147.638725
##	[48,]	-12.1297537	-1.617414e+00	34.834061	21.099183
##	[49,]	11.7503214	1.009934e+00	-35.071926	-21.119557
##	[50,]	31.7290418	-1.184081e+00	-83.200666	-50.421081
##	[51,]	-1.2223146	-2.223776e+00	5.330691	3.558080
##	[52,]	-0.5840944	-2.963387e+00	5.059166	3.263004
##	[53,]	-1.3005300	-2.615524e+00	5.952949	3.984498
##	[54,]	-2.0677119	-1.635227e+00	6.157799	4.259499
##	[55,]	-1.7818603	-2.126839e+00	6.242673	4.255081
##	[56,]	-0.7347850	-3.305557e+00	5.770500	3.771794
##	[57,]	-0.3383166	-3.434529e+00	5.173723	3.289325
##	[58,]	-1.1402809	-2.064347e+00	5.017546	3.333696
##	[59,]	-1.6024082	-1.966079e+00	5.700983	3.883556
##	[60,]	-0.5163653	-3.302399e+00	5.345315	3.465304
##	[61,]	-2.4737626	-8.477047e-01	5.903247	4.181658
##	[62,]	-0.5324707	-3.134173e+00	5.167448	3.337701
##	[63,]	-3.7978906	8.375717e-01	6.247555	4.622147
##	[64,]	-0.8908903	-3.303460e+00	6.060501	3.982862
##	[65,]	-0.7257238	-2.282412e+00	4.522255	2.916777
##	[66,]	-1.1501090	-2.246349e+00	5.228615	3.478742
##	[67,]	-0.3159280	-3.783101e+00	5.568894	3.556193
##	[68,]	-1.3515596	-1.917672e+00	5.216762	3.500749
##	[69,]	-4.3662283	2.292881e+00	5.555976	4.162523
##	[70,]	-1.5555424	-1.909045e+00	5.551371	3.781896
##	[71,]	-0.1672610	-4.747407e+00	6.489547	4.106686
##	[72,]	-1.2725569	-2.144807e+00	5.326763	3.575740
##	[73,]	-3.6143484	-8.989569e-02	7.215733	5.088460
##	[74,]	-1.2205622	-2.631771e+00	5.832416	3.900189
##	[75,]	-1.3511792	-2.104647e+00	5.419127	3.649306
##	[76,]	-1.2620457	-2.258706e+00	5.440222	3.647759
##	[77,]	-2.3800563	-1.115511e+00	6.078818	4.254380
##	[78,]	-1.5290478	-3.372783e+00	7.387477	4.899363
##	[79,]	-0.7554176	-3.452676e+00	5.987087	3.914177
##	[80,]	-1.8459982	-1.254384e+00	5.359081	3.647750
##	[81,]	-1.7363634	-1.734736e+00	5.661978	3.890947
##	[82,]	-1.8989815	-1.447397e+00	5.623520	3.881179
##	[83,]	-1.2255043	-2.158788e+00	5.263908	3.524002
##	[84,]	-2.4539679	-4.610774e+00	10.902395	7.171985
##	[85,]	-0.2274406	-3.887405e+00	5.529194	3.511774
##	[86,]	-0.2219831	-3.038590e+00	4.462218	2.782214

##	[87,]	-1.0305091	-2.801356e+00	5.681412	3.762253
##	[88,]	-3.7092638	1.194062e+00	5.648530	4.184394
##	[89,]	-0.4843928	-2.732868e+00	4.606793	2.937416
##	[90,]	-1.2723022	-2.593117e+00	5.871277	3.947121
##	[91,]	-1.0260541	-3.031442e+00	5.974114	3.961332
##	[92,]	-0.6747016	-3.224296e+00	5.552374	3.613083
##	[93,]	-1.4593965	-2.063207e+00	5.561185	3.772756
##	[94,]	-1.5067339	-1.811722e+00	5.364300	3.638524
##	[95,]	-0.8599996	-3.001194e+00	5.618458	3.702395
##	[96,]	-0.5756016	-2.579684e+00	4.600328	2.942233
##	[97,]	-0.6167264	-2.896058e+00	5.044132	3.266500
##	[98,]	-1.0792904	-2.437971e+00	5.330884	3.542653
##	[99,]	-1.2698764	-1.757451e+00	4.912817	3.263818
##	[100,]	-0.7825430	-2.803788e+00	5.233352	3.429268
##	[101,]	0.4228866	4.366131e+00	-6.671950	-3.877665
##	[102,]	-8.9454284	-2.254795e+01	47.752975	30.391801
##	[103,]	16.8636558	4.795941e+00	-40.413026	-26.401587
##	[104,]	-7.3161146	-1.661409e+01	36.561679	23.389554
##	[105,]	2.3269195	4.501352e+00	-10.767418	-6.612050
##	[106,]	10.2763614	-5.476188e+00	-13.504910	-9.329768
##	[107,]	-1.1600533	-7.925123e+00	12.679520	8.090836
##	[108,]	-27.5876390	1.074581e+01	41.011181	28.722980
##	[109,]	-17.6454204	1.172621e+01	19.859328	14.495532
##	[110,]	1.5307979	1.594868e+01	-24.000071	-14.411603
##	[111,]	-0.7161159	-6.224932e+00	9.499130	6.042821
##	[112,]	-11.3532430	-4.828017e+00	29.153588	19.352498
##	[113,]	-17.4911232	-2.233783e+01	64.557419	41.513018
##	[114,]	11.4104964	9.742431e+00	-36.208391	-23.232778
##	[115,]	1.5202919	4.984988e+00	-9.771282	-5.906001
##	[116,]	-3.3377253	-3.037202e+01	46.366027	28.667465
##	[117,]	-2.7049864	-7.553467e+00	15.276090	9.839565
##	[118,]	-460.1673550	-4.668772e+03	7024.242883	4288.821634
##	[119,]	-252.8325124	2.576452e+02	165.931578	133.321630
##	[120,]	-5.8462165	3.132644e+00	7.466149	5.491805
##	[121,]	5.7248134	1.594782e+01	-32.619992	-20.207973
##	[122,]	-2.3263614	-1.541684e+01	24.860807	15.618965
##	[123,]	-42.5504867	3.685825e+01	36.387814	27.749939
##	[124,]	-3.2354657	-3.081432e+00	10.440023	6.978629
##	[125,]	-2.2495658	-1.749275e+01	27.356054	17.035396
##	[126,]	-4.0333938	-6.315497e+00	16.283239	10.627034
##	[127,]	-1.8561965	-4.333949e+00	9.318442	6.118249
##	[128,]	-0.7451923	-5.363788e+00	8.445843	5.420314
##	[129,]	5.0099790	3.694401e+00	-15.154633	-9.619456
##	[130,]	-3.7426175	-1.365524e+00	9.151814	6.281812
##	[131,]	-13.4890661	8.655240e+00	15.486495	11.303227
##	[132,]	-0.9260738	-7.396983e+00	11.442180	7.212963
##	[133,]	3.3726874	2.297612e+00	-9.998688	-6.279434
##	[134,]	-1.8775642	-3.264375e+00	7.949296	5.301240
##	[135,]	-5.1642368	-2.757391e+00	13.904249	9.385939
##	[136,]	37.4681080	-2.013768e+01	-48.738930	-33.876126

```

## [137,] -6.1727694 -2.472809e+02 334.259792 202.785532
## [138,] -1.3762223 -7.797653e+00 12.915419 8.219947
## [139,] -0.5922691 -5.216190e+00 7.945859 5.091856
## [140,] -4.2949914 -8.799727e+00 20.091875 12.919943
## [141,] 2.5022124 5.877509e+00 -12.906484 -7.895857
## [142,] -4.2367907 -9.400476e+00 20.742161 13.282048
## [143,] -8.9454284 -2.254795e+01 47.752975 30.391801
## [144,] 2.1903158 6.551363e+00 -13.153720 -8.016785
## [145,] 1.3808694 8.565478e+00 -14.110261 -8.470737
## [146,] -35.4346108 -6.219419e+01 153.148376 97.206180
## [147,] -8.4434915 1.285844e+00 15.185836 10.482381
## [148,] -2.6256506 -7.933713e+00 15.604990 10.009472
## [149,] -0.1063037 -1.386358e+01 18.157416 11.176735
## [150,] -0.6757035 -6.958702e+00 10.400511 6.597885
##
##
## $startweights
## $startweights[[1]]
## $startweights[[1]][[1]]
##      [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 0.5855288 -1.8179560 -0.1162478 0.8168998 0.7796219
## [2,] 0.7094660 0.6300986 1.8173120 -0.8863575 1.4557851
## [3,] -0.1093033 -0.2761841 0.3706279 -0.3315776 -0.6443284
## [4,] -0.4534972 -0.2841597 0.5202165 1.1207127 -1.5531374
## [5,] 0.6058875 -0.9193220 -0.7505320 0.2987237 -1.5977095
##
## $startweights[[1]][[2]]
##      [,1]
## [1,] 1.8050975
## [2,] -0.4816474
## [3,] 0.6203798
## [4,] 0.6121235
## [5,] -0.1623110
## [6,] 0.8118732
##
##
##
## $result.matrix
##      [,1]
## error 0.570312030
## reached.threshold 0.009159972
## steps 802.000000000
## Intercept.to.1layhid1 1.211844619
## Sepal.Length.to.1layhid1 0.357105651
## Sepal.Width.to.1layhid1 1.041344392
## Petal.Length.to.1layhid1 -1.488796330
## Petal.Width.to.1layhid1 -1.077838284
## Intercept.to.1layhid2 -1.732985090
## Sepal.Length.to.1layhid2 4.606704293
## Sepal.Width.to.1layhid2 -6.217296399

```

```
## Petal.Length.to.1layhid2 -0.687755695
## Petal.Width.to.1layhid2 -1.238403234
## Intercept.to.1layhid3 -0.626944842
## Sepal.Length.to.1layhid3 0.574163190
## Sepal.Width.to.1layhid3 -1.080995645
## Petal.Length.to.1layhid3 1.230735761
## Petal.Width.to.1layhid3 -0.482055805
## Intercept.to.1layhid4 3.697699399
## Sepal.Length.to.1layhid4 -0.783754443
## Sepal.Width.to.1layhid4 5.512847618
## Petal.Length.to.1layhid4 -5.235942255
## Petal.Width.to.1layhid4 -3.604208945
## Intercept.to.1layhid5 0.583411127
## Sepal.Length.to.1layhid5 0.916160010
## Sepal.Width.to.1layhid5 -1.220827165
## Petal.Length.to.1layhid5 -0.797343731
## Petal.Width.to.1layhid5 -1.251722800
## Intercept.to.Species 1.421156257
## 1layhid1.to.Species -0.791815427
## 1layhid2.to.Species -0.981302336
## 1layhid3.to.Species 0.524999113
## 1layhid4.to.Species -0.635470818
## 1layhid5.to.Species -0.494952064
##
## attr(,"class")
## [1] "nn"

# Plot the neural network
plotnet(nn)
```

