

WEEK 13 - LESSONS IN R

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Installing packages

```
install.packages(c("neuralnet", "keras", "tensorflow"), dependancies = T)
```

```
## Installing packages into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'  
## (as 'lib' is unspecified)
```

```
library(neuralnet)  
install.packages("tidyverse")
```

```
## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'  
## (as 'lib' is unspecified)
```

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --  
## v dplyr      1.1.4      v readr      2.1.5  
## v forcats    1.0.0      v stringr   1.5.1  
## v ggplot2    3.5.1      v tibble    3.2.1  
## v lubridate  1.9.3      v tidyr     1.3.1  
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --  
## x dplyr::compute() masks neuralnet::compute()  
## x dplyr::filter()  masks stats::filter()  
## x dplyr::lag()     masks stats::lag()  
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

Data analysis

```
iris <- iris %>%mutate_if(is.character, as.factor)  
head(iris)
```

```
##   Sepal.Length Sepal.Width Petal.Length Petal.Width Species  
## 1          5.1         3.5          1.4          0.2  setosa  
## 2          4.9         3.0          1.4          0.2  setosa  
## 3          4.7         3.2          1.3          0.2  setosa  
## 4          4.6         3.1          1.5          0.2  setosa  
## 5          5.0         3.6          1.4          0.2  setosa  
## 6          5.4         3.9          1.7          0.4  setosa
```

```
summary(iris)
```

```
##   Sepal.Length   Sepal.Width   Petal.Length   Petal.Width
```

```
## Min.      :4.300   Min.      :2.000   Min.      :1.000   Min.      :0.100
## 1st Qu.:5.100   1st Qu.:2.800   1st Qu.:1.600   1st Qu.:0.300
## Median :5.800   Median :3.000   Median :4.350   Median :1.300
## Mean    :5.843   Mean    :3.057   Mean    :3.758   Mean    :1.199
## 3rd Qu.:6.400   3rd Qu.:3.300   3rd Qu.:5.100   3rd Qu.:1.800
## Max.    :7.900   Max.    :4.400   Max.    :6.900   Max.    :2.500
##          Species
## setosa      :50
## versicolor:50
## virginica  :50
##
##
##
```

Train and test split

```
set.seed(254)
data_rows <- floor(0.80 * nrow(iris))
train_indices <- sample(c(1:nrow(iris)), data_rows)
head(train_indices)
```

```
## [1] 55 37 146 70 45 124
```

```
train_data <- iris[train_indices,]
head(train_data)
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 55           6.5         2.8         4.6         1.5 versicolor
## 37           5.5         3.5         1.3         0.2  setosa
## 146          6.7         3.0         5.2         2.3 virginica
## 70           5.6         2.5         3.9         1.1 versicolor
## 45           5.1         3.8         1.9         0.4  setosa
## 124          6.3         2.7         4.9         1.8 virginica
```

```
test_data <- iris[-train_indices, ]
head(test_data)
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1           5.1         3.5         1.4         0.2  setosa
## 15          5.8         4.0         1.2         0.2  setosa
## 16          5.7         4.4         1.5         0.4  setosa
## 21          5.4         3.4         1.7         0.2  setosa
## 22          5.1         3.7         1.5         0.4  setosa
## 26          5.0         3.0         1.6         0.2  setosa
```

Two hidden layers with 4 and 2 neurons

```
model <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length +
Petal.Width, data = train_data, hidden = c(4,2), linear.output = FALSE)
# Print the model summary
head(model)
```

```
## $call
## neuralnet(formula = Species ~ Sepal.Length + Sepal.Width + Petal.Length +
```

```

##      Petal.Width, data = train_data, hidden = c(4, 2), linear.output = FALSE)
##
## $response
##      versicolor setosa virginica
## 1      FALSE    TRUE     FALSE
## 2      TRUE     FALSE    FALSE
## 3      FALSE    FALSE     TRUE
## 4      FALSE    TRUE     FALSE
## 5      TRUE     FALSE    FALSE
## 6      FALSE    FALSE     TRUE
## 7      TRUE     FALSE    FALSE
## 8      FALSE    TRUE     FALSE
## 9      FALSE    FALSE     TRUE
## 10     TRUE     FALSE    FALSE
## 11     FALSE    TRUE     FALSE
## 12     TRUE     FALSE    FALSE
## 13     FALSE    FALSE     TRUE
## 14     FALSE    FALSE     TRUE
## 15     FALSE    FALSE     TRUE
## 16     FALSE    FALSE     TRUE
## 17     FALSE    TRUE     FALSE
## 18     FALSE    FALSE     TRUE
## 19     FALSE    FALSE     TRUE
## 20     TRUE     FALSE    FALSE
## 21     FALSE    FALSE     TRUE
## 22     FALSE    FALSE     TRUE
## 23     FALSE    TRUE     FALSE
## 24     FALSE    FALSE     TRUE
## 25     FALSE    FALSE     TRUE
## 26     FALSE    FALSE     TRUE
## 27     FALSE    TRUE     FALSE
## 28     FALSE    FALSE     TRUE
## 29     TRUE     FALSE    FALSE
## 30     FALSE    TRUE     FALSE
## 31     FALSE    TRUE     FALSE
## 32     FALSE    TRUE     FALSE
## 33     FALSE    TRUE     FALSE
## 34     TRUE     FALSE    FALSE
## 35     TRUE     FALSE    FALSE
## 36     FALSE    TRUE     FALSE
## 37     FALSE    FALSE     TRUE
## 38     FALSE    FALSE     TRUE
## 39     FALSE    FALSE     TRUE
## 40     TRUE     FALSE    FALSE
## 41     TRUE     FALSE    FALSE
## 42     FALSE    TRUE     FALSE
## 43     TRUE     FALSE    FALSE
## 44     FALSE    TRUE     FALSE
## 45     TRUE     FALSE    FALSE
## 46     TRUE     FALSE    FALSE
## 47     TRUE     FALSE    FALSE
## 48     FALSE    TRUE     FALSE
## 49     TRUE     FALSE    FALSE
## 50     FALSE    TRUE     FALSE

```

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

```
## 105      TRUE FALSE FALSE
## 106      TRUE FALSE FALSE
## 107      TRUE FALSE FALSE
## 108      FALSE FALSE TRUE
## 109      FALSE FALSE TRUE
## 110      TRUE FALSE FALSE
## 111      FALSE FALSE TRUE
## 112      TRUE FALSE FALSE
## 113      FALSE TRUE FALSE
## 114      FALSE TRUE FALSE
## 115      FALSE TRUE FALSE
## 116      FALSE TRUE FALSE
## 117      TRUE FALSE FALSE
## 118      FALSE TRUE FALSE
## 119      FALSE FALSE TRUE
## 120      FALSE FALSE TRUE
```

```
##
```

```
## $covariate
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## 55           6.5         2.8         4.6         1.5
## 37           5.5         3.5         1.3         0.2
## 146          6.7         3.0         5.2         2.3
## 70           5.6         2.5         3.9         1.1
## 45           5.1         3.8         1.9         0.4
## 124          6.3         2.7         4.9         1.8
## 20           5.1         3.8         1.5         0.3
## 76           6.6         3.0         4.4         1.4
## 144          6.8         3.2         5.9         2.3
## 3           4.7         3.2         1.3         0.2
## 88           6.3         2.3         4.4         1.3
## 10           4.9         3.1         1.5         0.1
## 136          7.7         3.0         6.1         2.3
## 126          7.2         3.2         6.0         1.8
## 102          5.8         2.7         5.1         1.9
## 125          6.7         3.3         5.7         2.1
## 64           6.1         2.9         4.7         1.4
## 111          6.5         3.2         5.1         2.0
## 122          5.6         2.8         4.9         2.0
## 32           5.4         3.4         1.5         0.4
## 147          6.3         2.5         5.0         1.9
## 123          7.7         2.8         6.7         2.0
## 95           5.6         2.7         4.2         1.3
## 101          6.3         3.3         6.0         2.5
## 149          6.2         3.4         5.4         2.3
## 143          5.8         2.7         5.1         1.9
## 94           5.0         2.3         3.3         1.0
## 150          5.9         3.0         5.1         1.8
## 11           5.4         3.7         1.5         0.2
## 83           5.8         2.7         3.9         1.2
## 54           5.5         2.3         4.0         1.3
## 57           6.3         3.3         4.7         1.6
## 61           5.0         2.0         3.5         1.0
## 48           4.6         3.2         1.4         0.2
## 29           5.2         3.4         1.4         0.2
```

## 69	6.2	2.2	4.5	1.5
## 130	7.2	3.0	5.8	1.6
## 115	5.8	2.8	5.1	2.4
## 145	6.7	3.3	5.7	2.5
## 17	5.4	3.9	1.3	0.4
## 50	5.0	3.3	1.4	0.2
## 96	5.7	3.0	4.2	1.2
## 35	4.9	3.1	1.5	0.2
## 93	5.8	2.6	4.0	1.2
## 49	5.3	3.7	1.5	0.2
## 12	4.8	3.4	1.6	0.2
## 14	4.3	3.0	1.1	0.1
## 60	5.2	2.7	3.9	1.4
## 18	5.1	3.5	1.4	0.3
## 97	5.7	2.9	4.2	1.3
## 109	6.7	2.5	5.8	1.8
## 134	6.3	2.8	5.1	1.5
## 62	5.9	3.0	4.2	1.5
## 113	6.8	3.0	5.5	2.1
## 75	6.4	2.9	4.3	1.3
## 119	7.7	2.6	6.9	2.3
## 41	5.0	3.5	1.3	0.3
## 27	5.0	3.4	1.6	0.4
## 25	4.8	3.4	1.9	0.2
## 89	5.6	3.0	4.1	1.3
## 100	5.7	2.8	4.1	1.3
## 91	5.5	2.6	4.4	1.2
## 19	5.7	3.8	1.7	0.3
## 137	6.3	3.4	5.6	2.4
## 46	4.8	3.0	1.4	0.3
## 103	7.1	3.0	5.9	2.1
## 85	5.4	3.0	4.5	1.5
## 6	5.4	3.9	1.7	0.4
## 44	5.0	3.5	1.6	0.6
## 86	6.0	3.4	4.5	1.6
## 71	5.9	3.2	4.8	1.8
## 36	5.0	3.2	1.2	0.2
## 104	6.3	2.9	5.6	1.8
## 42	4.5	2.3	1.3	0.3
## 139	6.0	3.0	4.8	1.8
## 118	7.7	3.8	6.7	2.2
## 106	7.6	3.0	6.6	2.1
## 9	4.4	2.9	1.4	0.2
## 43	4.4	3.2	1.3	0.2
## 84	6.0	2.7	5.1	1.6
## 66	6.7	3.1	4.4	1.4
## 39	4.4	3.0	1.3	0.2
## 7	4.6	3.4	1.4	0.3
## 72	6.1	2.8	4.0	1.3
## 117	6.5	3.0	5.5	1.8
## 108	7.3	2.9	6.3	1.8
## 4	4.6	3.1	1.5	0.2
## 38	4.9	3.6	1.4	0.1
## 138	6.4	3.1	5.5	1.8

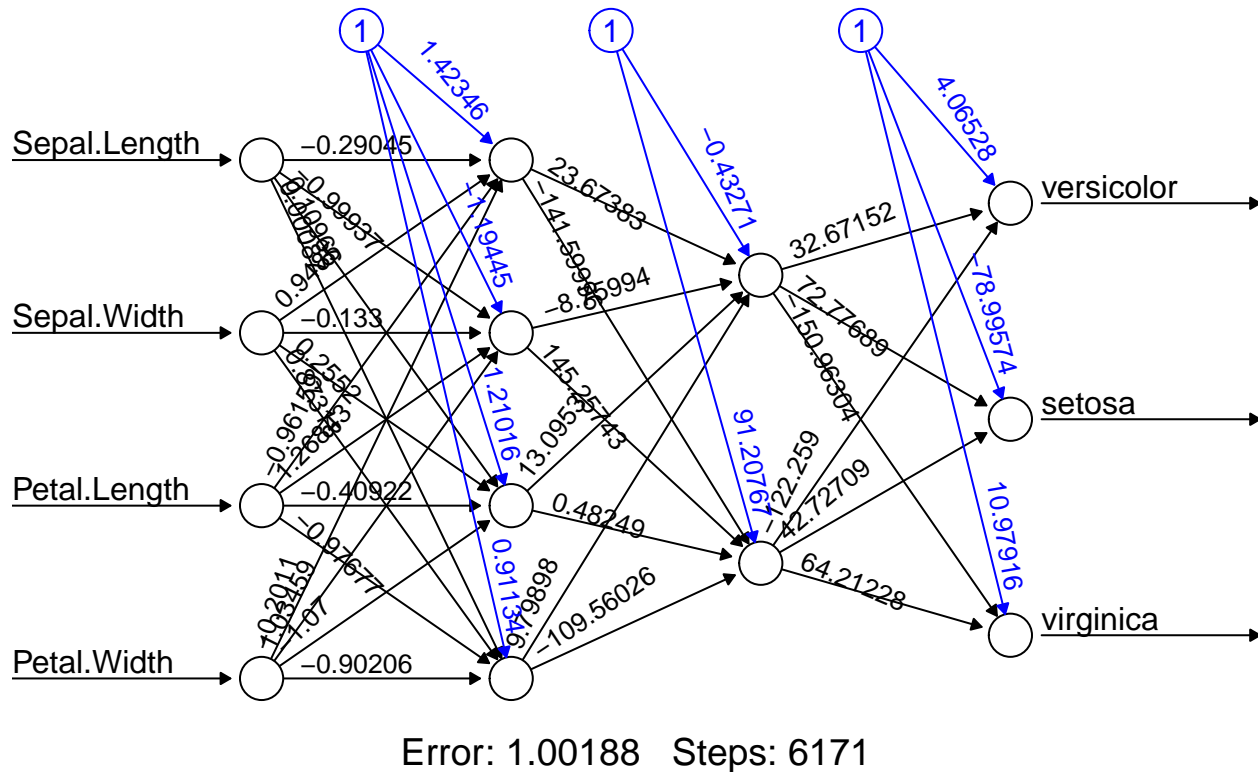
```

## 65      5.6      2.9      3.6      1.3
## 5       5.0      3.6      1.4      0.2
## 2       4.9      3.0      1.4      0.2
## 87      6.7      3.1      4.7      1.5
## 82      5.5      2.4      3.7      1.0
## 40      5.1      3.4      1.5      0.2
## 77      6.8      2.8      4.8      1.4
## 128     6.1      3.0      4.9      1.8
## 67      5.6      3.0      4.5      1.5
## 92      6.1      3.0      4.6      1.4
## 131     7.4      2.8      6.1      1.9
## 74      6.1      2.8      4.7      1.2
## 56      5.7      2.8      4.5      1.3
## 59      6.6      2.9      4.6      1.3
## 120     6.0      2.2      5.0      1.5
## 23      4.6      3.6      1.0      0.2
## 13      4.8      3.0      1.4      0.1
## 33      5.2      4.1      1.5      0.1
## 107     4.9      2.5      4.5      1.7
## 127     6.2      2.8      4.8      1.8
## 24      5.1      3.3      1.7      0.5
## 116     6.4      3.2      5.3      2.3
## 34      5.5      4.2      1.4      0.2
## 68      5.8      2.7      4.1      1.0
## 58      4.9      2.4      3.3      1.0
## 73      6.3      2.5      4.9      1.5
## 80      5.7      2.6      3.5      1.0
## 8       5.0      3.4      1.5      0.2
## 99      5.1      2.5      3.0      1.1
## 121     6.9      3.2      5.7      2.3
## 133     6.4      2.8      5.6      2.2
##
## $model.list
## $model.list$response
## [1] "versicolor" "setosa"      "virginica"
##
## $model.list$variables
## [1] "Sepal.Length" "Sepal.Width"  "Petal.Length" "Petal.Width"
##
##
## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x5bfe53991878>
## <environment: 0x5bfe53992fa8>
## attr(,"type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))

```

```
## }
## <bytecode: 0x5bfe53997cd8>
## <environment: 0x5bfe539985d0>
## attr(,"type")
## [1] "logistic"
```

```
plot(model, rep = 'best')
```



```
pred <- predict(model, test_data)
pred
```

```
##           [,1]      [,2]      [,3]
## 1  1.000000e+00 1.987582e-03 1.606099e-61
## 15 1.000000e+00 1.987582e-03 1.606099e-61
## 16 1.000000e+00 1.987582e-03 1.606099e-61
## 21 1.000000e+00 1.987582e-03 1.606099e-61
## 22 1.000000e+00 1.987582e-03 1.606099e-61
## 26 1.000000e+00 1.987582e-03 1.606099e-61
## 28 1.000000e+00 1.987582e-03 1.606099e-61
## 30 1.000000e+00 1.987582e-03 1.606099e-61
## 31 1.000000e+00 1.987582e-03 1.606099e-61
## 47 1.000000e+00 1.987582e-03 1.606099e-61
## 51 5.976903e-38 1.000000e+00 2.953469e-33
## 52 5.723452e-38 1.000000e+00 3.608146e-33
## 53 1.384220e-38 1.000000e+00 2.544987e-30
## 63 6.966252e-38 1.000000e+00 1.455306e-33
## 78 5.834333e-43 9.999693e-01 4.187287e-10
## 79 1.736209e-38 1.000000e+00 8.933657e-31
## 81 7.119429e-38 1.000000e+00 1.316157e-33
## 90 6.249596e-38 1.000000e+00 2.403280e-33
```



```
## 98 6.688873e-38 1.000000e+00 1.755865e-33
## 105 5.423696e-52 2.476923e-16 1.000000e+00
## 110 5.316714e-52 2.369408e-16 1.000000e+00
## 112 1.893062e-51 4.010254e-15 1.000000e+00
## 114 9.329015e-52 8.290613e-16 1.000000e+00
## 129 6.037474e-52 3.145041e-16 1.000000e+00
## 132 1.404842e-51 2.063591e-15 1.000000e+00
## 135 2.891381e-51 1.030162e-14 1.000000e+00
## 140 3.342740e-51 1.423096e-14 1.000000e+00
## 141 5.820653e-52 2.898980e-16 1.000000e+00
## 142 1.001202e-50 1.638601e-13 1.000000e+00
## 148 7.647401e-51 8.991549e-14 1.000000e+00
```

Three hidden layers with 8, 4 and 2 neurons

```
model2 <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length + Petal.Width, data = train_data,
# Print the model summary
head(model2)
```

```
## $call
## neuralnet(formula = Species ~ Sepal.Length + Sepal.Width + Petal.Length +
##     Petal.Width, data = train_data, hidden = c(8, 4, 2), linear.output = FALSE)
##
## $response
##      versicolor setosa virginica
## 1      FALSE    TRUE     FALSE
## 2       TRUE   FALSE     FALSE
## 3      FALSE   FALSE      TRUE
## 4      FALSE    TRUE     FALSE
## 5       TRUE   FALSE     FALSE
## 6      FALSE   FALSE      TRUE
## 7       TRUE   FALSE     FALSE
## 8      FALSE    TRUE     FALSE
## 9      FALSE   FALSE      TRUE
## 10      TRUE   FALSE     FALSE
## 11      FALSE    TRUE     FALSE
## 12      TRUE   FALSE     FALSE
## 13      FALSE   FALSE      TRUE
## 14      FALSE   FALSE      TRUE
## 15      FALSE   FALSE      TRUE
## 16      FALSE   FALSE      TRUE
## 17      FALSE    TRUE     FALSE
## 18      FALSE   FALSE      TRUE
## 19      FALSE   FALSE      TRUE
## 20      TRUE   FALSE     FALSE
## 21      FALSE   FALSE      TRUE
## 22      FALSE   FALSE      TRUE
## 23      FALSE    TRUE     FALSE
## 24      FALSE   FALSE      TRUE
## 25      FALSE   FALSE      TRUE
## 26      FALSE   FALSE      TRUE
## 27      FALSE    TRUE     FALSE
## 28      FALSE   FALSE      TRUE
```

## 29	TRUE	FALSE	FALSE
## 30	FALSE	TRUE	FALSE
## 31	FALSE	TRUE	FALSE
## 32	FALSE	TRUE	FALSE
## 33	FALSE	TRUE	FALSE
## 34	TRUE	FALSE	FALSE
## 35	TRUE	FALSE	FALSE
## 36	FALSE	TRUE	FALSE
## 37	FALSE	FALSE	TRUE
## 38	FALSE	FALSE	TRUE
## 39	FALSE	FALSE	TRUE
## 40	TRUE	FALSE	FALSE
## 41	TRUE	FALSE	FALSE
## 42	FALSE	TRUE	FALSE
## 43	TRUE	FALSE	FALSE
## 44	FALSE	TRUE	FALSE
## 45	TRUE	FALSE	FALSE
## 46	TRUE	FALSE	FALSE
## 47	TRUE	FALSE	FALSE
## 48	FALSE	TRUE	FALSE
## 49	TRUE	FALSE	FALSE
## 50	FALSE	TRUE	FALSE
## 51	FALSE	FALSE	TRUE
## 52	FALSE	FALSE	TRUE
## 53	FALSE	TRUE	FALSE
## 54	FALSE	FALSE	TRUE
## 55	FALSE	TRUE	FALSE
## 56	FALSE	FALSE	TRUE
## 57	TRUE	FALSE	FALSE
## 58	TRUE	FALSE	FALSE
## 59	TRUE	FALSE	FALSE
## 60	FALSE	TRUE	FALSE
## 61	FALSE	TRUE	FALSE
## 62	FALSE	TRUE	FALSE
## 63	TRUE	FALSE	FALSE
## 64	FALSE	FALSE	TRUE
## 65	TRUE	FALSE	FALSE
## 66	FALSE	FALSE	TRUE
## 67	FALSE	TRUE	FALSE
## 68	TRUE	FALSE	FALSE
## 69	TRUE	FALSE	FALSE
## 70	FALSE	TRUE	FALSE
## 71	FALSE	TRUE	FALSE
## 72	TRUE	FALSE	FALSE
## 73	FALSE	FALSE	TRUE
## 74	TRUE	FALSE	FALSE
## 75	FALSE	FALSE	TRUE
## 76	FALSE	FALSE	TRUE
## 77	FALSE	FALSE	TRUE
## 78	TRUE	FALSE	FALSE
## 79	TRUE	FALSE	FALSE
## 80	FALSE	TRUE	FALSE
## 81	FALSE	TRUE	FALSE
## 82	TRUE	FALSE	FALSE

```

## 83      TRUE FALSE FALSE
## 84      FALSE TRUE  FALSE
## 85      FALSE FALSE  TRUE
## 86      FALSE FALSE  TRUE
## 87      TRUE  FALSE FALSE
## 88      TRUE  FALSE FALSE
## 89      FALSE FALSE  TRUE
## 90      FALSE TRUE  FALSE
## 91      TRUE  FALSE FALSE
## 92      TRUE  FALSE FALSE
## 93      FALSE TRUE  FALSE
## 94      FALSE TRUE  FALSE
## 95      TRUE  FALSE FALSE
## 96      FALSE TRUE  FALSE
## 97      FALSE FALSE  TRUE
## 98      FALSE TRUE  FALSE
## 99      FALSE TRUE  FALSE
## 100     FALSE FALSE  TRUE
## 101     FALSE TRUE  FALSE
## 102     FALSE TRUE  FALSE
## 103     FALSE TRUE  FALSE
## 104     FALSE FALSE  TRUE
## 105     TRUE  FALSE FALSE
## 106     TRUE  FALSE FALSE
## 107     TRUE  FALSE FALSE
## 108     FALSE FALSE  TRUE
## 109     FALSE FALSE  TRUE
## 110     TRUE  FALSE FALSE
## 111     FALSE FALSE  TRUE
## 112     TRUE  FALSE FALSE
## 113     FALSE TRUE  FALSE
## 114     FALSE TRUE  FALSE
## 115     FALSE TRUE  FALSE
## 116     FALSE TRUE  FALSE
## 117     TRUE  FALSE FALSE
## 118     FALSE TRUE  FALSE
## 119     FALSE FALSE  TRUE
## 120     FALSE FALSE  TRUE
##
## $covariate
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## 55             6.5         2.8         4.6         1.5
## 37             5.5         3.5         1.3         0.2
## 146            6.7         3.0         5.2         2.3
## 70             5.6         2.5         3.9         1.1
## 45             5.1         3.8         1.9         0.4
## 124            6.3         2.7         4.9         1.8
## 20             5.1         3.8         1.5         0.3
## 76             6.6         3.0         4.4         1.4
## 144            6.8         3.2         5.9         2.3
## 3             4.7         3.2         1.3         0.2
## 88             6.3         2.3         4.4         1.3
## 10            4.9         3.1         1.5         0.1
## 136            7.7         3.0         6.1         2.3

```

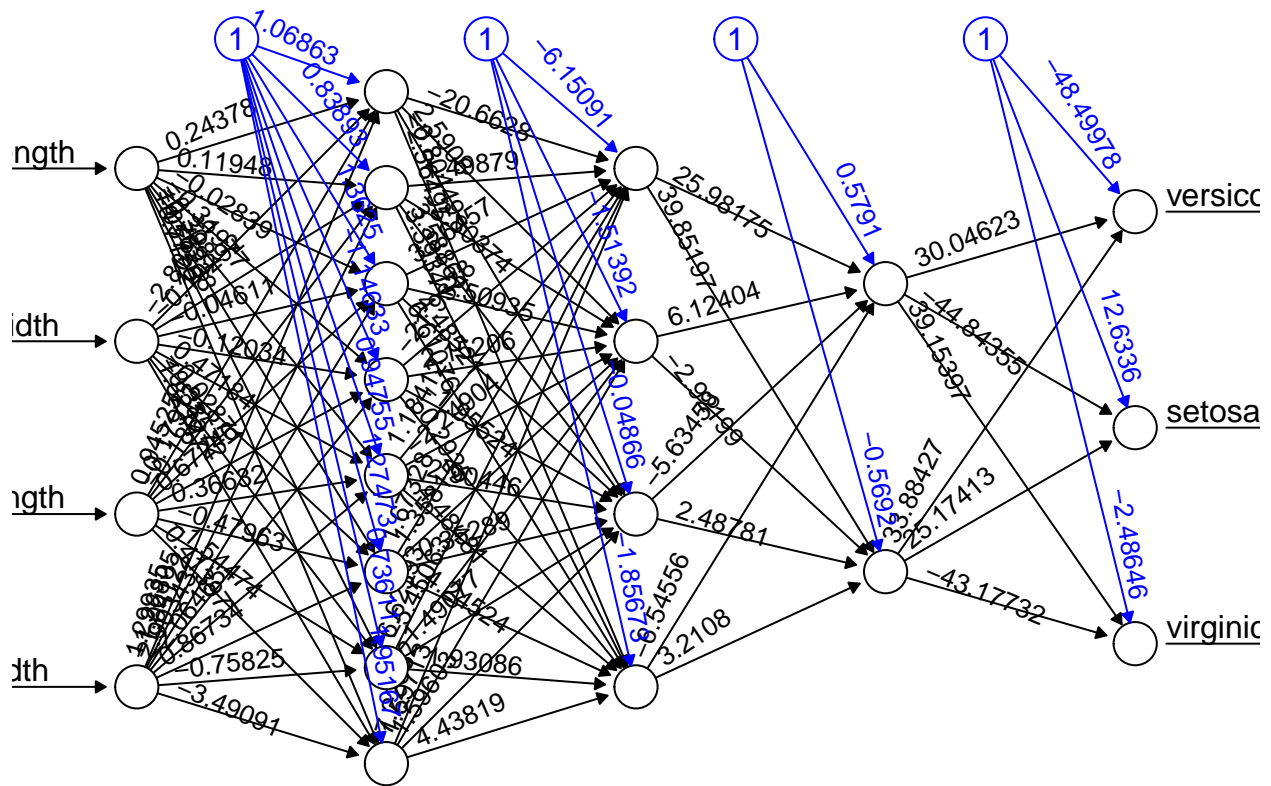
## 126	7.2	3.2	6.0	1.8
## 102	5.8	2.7	5.1	1.9
## 125	6.7	3.3	5.7	2.1
## 64	6.1	2.9	4.7	1.4
## 111	6.5	3.2	5.1	2.0
## 122	5.6	2.8	4.9	2.0
## 32	5.4	3.4	1.5	0.4
## 147	6.3	2.5	5.0	1.9
## 123	7.7	2.8	6.7	2.0
## 95	5.6	2.7	4.2	1.3
## 101	6.3	3.3	6.0	2.5
## 149	6.2	3.4	5.4	2.3
## 143	5.8	2.7	5.1	1.9
## 94	5.0	2.3	3.3	1.0
## 150	5.9	3.0	5.1	1.8
## 11	5.4	3.7	1.5	0.2
## 83	5.8	2.7	3.9	1.2
## 54	5.5	2.3	4.0	1.3
## 57	6.3	3.3	4.7	1.6
## 61	5.0	2.0	3.5	1.0
## 48	4.6	3.2	1.4	0.2
## 29	5.2	3.4	1.4	0.2
## 69	6.2	2.2	4.5	1.5
## 130	7.2	3.0	5.8	1.6
## 115	5.8	2.8	5.1	2.4
## 145	6.7	3.3	5.7	2.5
## 17	5.4	3.9	1.3	0.4
## 50	5.0	3.3	1.4	0.2
## 96	5.7	3.0	4.2	1.2
## 35	4.9	3.1	1.5	0.2
## 93	5.8	2.6	4.0	1.2
## 49	5.3	3.7	1.5	0.2
## 12	4.8	3.4	1.6	0.2
## 14	4.3	3.0	1.1	0.1
## 60	5.2	2.7	3.9	1.4
## 18	5.1	3.5	1.4	0.3
## 97	5.7	2.9	4.2	1.3
## 109	6.7	2.5	5.8	1.8
## 134	6.3	2.8	5.1	1.5
## 62	5.9	3.0	4.2	1.5
## 113	6.8	3.0	5.5	2.1
## 75	6.4	2.9	4.3	1.3
## 119	7.7	2.6	6.9	2.3
## 41	5.0	3.5	1.3	0.3
## 27	5.0	3.4	1.6	0.4
## 25	4.8	3.4	1.9	0.2
## 89	5.6	3.0	4.1	1.3
## 100	5.7	2.8	4.1	1.3
## 91	5.5	2.6	4.4	1.2
## 19	5.7	3.8	1.7	0.3
## 137	6.3	3.4	5.6	2.4
## 46	4.8	3.0	1.4	0.3
## 103	7.1	3.0	5.9	2.1
## 85	5.4	3.0	4.5	1.5

## 6	5.4	3.9	1.7	0.4
## 44	5.0	3.5	1.6	0.6
## 86	6.0	3.4	4.5	1.6
## 71	5.9	3.2	4.8	1.8
## 36	5.0	3.2	1.2	0.2
## 104	6.3	2.9	5.6	1.8
## 42	4.5	2.3	1.3	0.3
## 139	6.0	3.0	4.8	1.8
## 118	7.7	3.8	6.7	2.2
## 106	7.6	3.0	6.6	2.1
## 9	4.4	2.9	1.4	0.2
## 43	4.4	3.2	1.3	0.2
## 84	6.0	2.7	5.1	1.6
## 66	6.7	3.1	4.4	1.4
## 39	4.4	3.0	1.3	0.2
## 7	4.6	3.4	1.4	0.3
## 72	6.1	2.8	4.0	1.3
## 117	6.5	3.0	5.5	1.8
## 108	7.3	2.9	6.3	1.8
## 4	4.6	3.1	1.5	0.2
## 38	4.9	3.6	1.4	0.1
## 138	6.4	3.1	5.5	1.8
## 65	5.6	2.9	3.6	1.3
## 5	5.0	3.6	1.4	0.2
## 2	4.9	3.0	1.4	0.2
## 87	6.7	3.1	4.7	1.5
## 82	5.5	2.4	3.7	1.0
## 40	5.1	3.4	1.5	0.2
## 77	6.8	2.8	4.8	1.4
## 128	6.1	3.0	4.9	1.8
## 67	5.6	3.0	4.5	1.5
## 92	6.1	3.0	4.6	1.4
## 131	7.4	2.8	6.1	1.9
## 74	6.1	2.8	4.7	1.2
## 56	5.7	2.8	4.5	1.3
## 59	6.6	2.9	4.6	1.3
## 120	6.0	2.2	5.0	1.5
## 23	4.6	3.6	1.0	0.2
## 13	4.8	3.0	1.4	0.1
## 33	5.2	4.1	1.5	0.1
## 107	4.9	2.5	4.5	1.7
## 127	6.2	2.8	4.8	1.8
## 24	5.1	3.3	1.7	0.5
## 116	6.4	3.2	5.3	2.3
## 34	5.5	4.2	1.4	0.2
## 68	5.8	2.7	4.1	1.0
## 58	4.9	2.4	3.3	1.0
## 73	6.3	2.5	4.9	1.5
## 80	5.7	2.6	3.5	1.0
## 8	5.0	3.4	1.5	0.2
## 99	5.1	2.5	3.0	1.1
## 121	6.9	3.2	5.7	2.3
## 133	6.4	2.8	5.6	2.2
##				

```

## $model.list
## $model.list$response
## [1] "versicolor" "setosa"      "virginica"
##
## $model.list$variables
## [1] "Sepal.Length" "Sepal.Width"  "Petal.Length" "Petal.Width"
##
##
## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x5bfe53991878>
## <environment: 0x5bfe54f978b0>
## attr("type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))
## }
## <bytecode: 0x5bfe53997cd8>
## <environment: 0x5bfe54f97d48>
## attr("type")
## [1] "logistic"
plot(model2, rep = 'best')

```



Error: 1.00079 Steps: 666

```
pred2 <- predict(model2, test_data)
pred2
```

```
##           [,1]      [,2]      [,3]
## 1  9.999998e-01  8.790491e-04  1.486519e-03
## 15 9.999998e-01  8.790488e-04  1.486520e-03
## 16 9.999998e-01  8.790500e-04  1.486518e-03
## 21 9.999998e-01  8.790493e-04  1.486519e-03
## 22 9.999998e-01  8.790539e-04  1.486512e-03
## 26 9.999998e-01  8.790497e-04  1.486519e-03
## 28 9.999998e-01  8.790491e-04  1.486519e-03
## 30 9.999998e-01  8.790500e-04  1.486518e-03
## 31 9.999998e-01  8.790499e-04  1.486518e-03
## 47 9.999998e-01  8.790492e-04  1.486519e-03
## 51 9.798366e-08  1.000000e+00  1.090960e-19
## 52 4.944054e-08  1.000000e+00  2.834624e-19
## 53 3.322452e-09  1.000000e+00  2.235976e-17
## 63 2.720924e-07  1.000000e+00  2.807449e-20
## 78 1.129563e-08  2.223755e-02  9.787111e-01
## 79 4.837272e-09  1.000000e+00  1.120450e-17
## 81 2.408606e-07  1.000000e+00  3.289534e-20
## 90 7.333305e-08  1.000000e+00  1.623310e-19
## 98 1.440887e-07  1.000000e+00  6.478605e-20
## 105 7.265736e-08  7.646931e-14  1.000000e+00
## 110 5.827280e-08  5.665371e-14  1.000000e+00
## 112 1.789452e-07  4.093547e-13  1.000000e+00
## 114 1.200711e-07  1.731450e-13  1.000000e+00
```

```
## 129 8.424443e-08 9.518050e-14 1.000000e+00
## 132 4.144426e-07 1.785009e-11 1.000000e+00
## 135 2.595521e-10 1.000000e+00 1.357644e-14
## 140 1.142640e-07 1.569868e-13 1.000000e+00
## 141 5.678086e-08 5.488341e-14 1.000000e+00
## 142 8.017150e-08 8.785982e-14 1.000000e+00
## 148 1.854539e-07 4.400379e-13 1.000000e+00
```

10 hidden layers with 8, 20, 6, 5, 4, 2, 1, 4, 5, and 30 neurons

```
model3 <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length + Petal.Width, data = train_data,
# Print the model summary
head(model3)
```

```
## $call
## neuralnet(formula = Species ~ Sepal.Length + Sepal.Width + Petal.Length +
##      Petal.Width, data = train_data, hidden = c(8, 20, 6, 5, 4,
##      2, 1, 4, 5, 30), linear.output = FALSE)
##
## $response
##      versicolor setosa virginica
## 1      FALSE    TRUE     FALSE
## 2      TRUE    FALSE     FALSE
## 3      FALSE    FALSE     TRUE
## 4      FALSE    TRUE     FALSE
## 5      TRUE    FALSE     FALSE
## 6      FALSE    FALSE     TRUE
## 7      TRUE    FALSE     FALSE
## 8      FALSE    TRUE     FALSE
## 9      FALSE    FALSE     TRUE
## 10     TRUE    FALSE     FALSE
## 11     FALSE    TRUE     FALSE
## 12     TRUE    FALSE     FALSE
## 13     FALSE    FALSE     TRUE
## 14     FALSE    FALSE     TRUE
## 15     FALSE    FALSE     TRUE
## 16     FALSE    FALSE     TRUE
## 17     FALSE    TRUE     FALSE
## 18     FALSE    FALSE     TRUE
## 19     FALSE    FALSE     TRUE
## 20     TRUE    FALSE     FALSE
## 21     FALSE    FALSE     TRUE
## 22     FALSE    FALSE     TRUE
## 23     FALSE    TRUE     FALSE
## 24     FALSE    FALSE     TRUE
## 25     FALSE    FALSE     TRUE
## 26     FALSE    FALSE     TRUE
## 27     FALSE    TRUE     FALSE
## 28     FALSE    FALSE     TRUE
## 29     TRUE    FALSE     FALSE
## 30     FALSE    TRUE     FALSE
## 31     FALSE    TRUE     FALSE
## 32     FALSE    TRUE     FALSE
```


## 33	FALSE	TRUE	FALSE
## 34	TRUE	FALSE	FALSE
## 35	TRUE	FALSE	FALSE
## 36	FALSE	TRUE	FALSE
## 37	FALSE	FALSE	TRUE
## 38	FALSE	FALSE	TRUE
## 39	FALSE	FALSE	TRUE
## 40	TRUE	FALSE	FALSE
## 41	TRUE	FALSE	FALSE
## 42	FALSE	TRUE	FALSE
## 43	TRUE	FALSE	FALSE
## 44	FALSE	TRUE	FALSE
## 45	TRUE	FALSE	FALSE
## 46	TRUE	FALSE	FALSE
## 47	TRUE	FALSE	FALSE
## 48	FALSE	TRUE	FALSE
## 49	TRUE	FALSE	FALSE
## 50	FALSE	TRUE	FALSE
## 51	FALSE	FALSE	TRUE
## 52	FALSE	FALSE	TRUE
## 53	FALSE	TRUE	FALSE
## 54	FALSE	FALSE	TRUE
## 55	FALSE	TRUE	FALSE
## 56	FALSE	FALSE	TRUE
## 57	TRUE	FALSE	FALSE
## 58	TRUE	FALSE	FALSE
## 59	TRUE	FALSE	FALSE
## 60	FALSE	TRUE	FALSE
## 61	FALSE	TRUE	FALSE
## 62	FALSE	TRUE	FALSE
## 63	TRUE	FALSE	FALSE
## 64	FALSE	FALSE	TRUE
## 65	TRUE	FALSE	FALSE
## 66	FALSE	FALSE	TRUE
## 67	FALSE	TRUE	FALSE
## 68	TRUE	FALSE	FALSE
## 69	TRUE	FALSE	FALSE
## 70	FALSE	TRUE	FALSE
## 71	FALSE	TRUE	FALSE
## 72	TRUE	FALSE	FALSE
## 73	FALSE	FALSE	TRUE
## 74	TRUE	FALSE	FALSE
## 75	FALSE	FALSE	TRUE
## 76	FALSE	FALSE	TRUE
## 77	FALSE	FALSE	TRUE
## 78	TRUE	FALSE	FALSE
## 79	TRUE	FALSE	FALSE
## 80	FALSE	TRUE	FALSE
## 81	FALSE	TRUE	FALSE
## 82	TRUE	FALSE	FALSE
## 83	TRUE	FALSE	FALSE
## 84	FALSE	TRUE	FALSE
## 85	FALSE	FALSE	TRUE
## 86	FALSE	FALSE	TRUE

```

## 87      TRUE FALSE FALSE
## 88      TRUE FALSE FALSE
## 89     FALSE FALSE  TRUE
## 90     FALSE  TRUE FALSE
## 91      TRUE FALSE FALSE
## 92      TRUE FALSE FALSE
## 93     FALSE  TRUE FALSE
## 94     FALSE  TRUE FALSE
## 95      TRUE FALSE FALSE
## 96     FALSE  TRUE FALSE
## 97     FALSE FALSE  TRUE
## 98     FALSE  TRUE FALSE
## 99     FALSE  TRUE FALSE
## 100     FALSE FALSE  TRUE
## 101     FALSE  TRUE FALSE
## 102     FALSE  TRUE FALSE
## 103     FALSE  TRUE FALSE
## 104     FALSE FALSE  TRUE
## 105      TRUE FALSE FALSE
## 106      TRUE FALSE FALSE
## 107      TRUE FALSE FALSE
## 108     FALSE FALSE  TRUE
## 109     FALSE FALSE  TRUE
## 110      TRUE FALSE FALSE
## 111     FALSE FALSE  TRUE
## 112      TRUE FALSE FALSE
## 113     FALSE  TRUE FALSE
## 114     FALSE  TRUE FALSE
## 115     FALSE  TRUE FALSE
## 116     FALSE  TRUE FALSE
## 117      TRUE FALSE FALSE
## 118     FALSE  TRUE FALSE
## 119     FALSE FALSE  TRUE
## 120     FALSE FALSE  TRUE

```

```
##
```

```
## $covariate
```

```

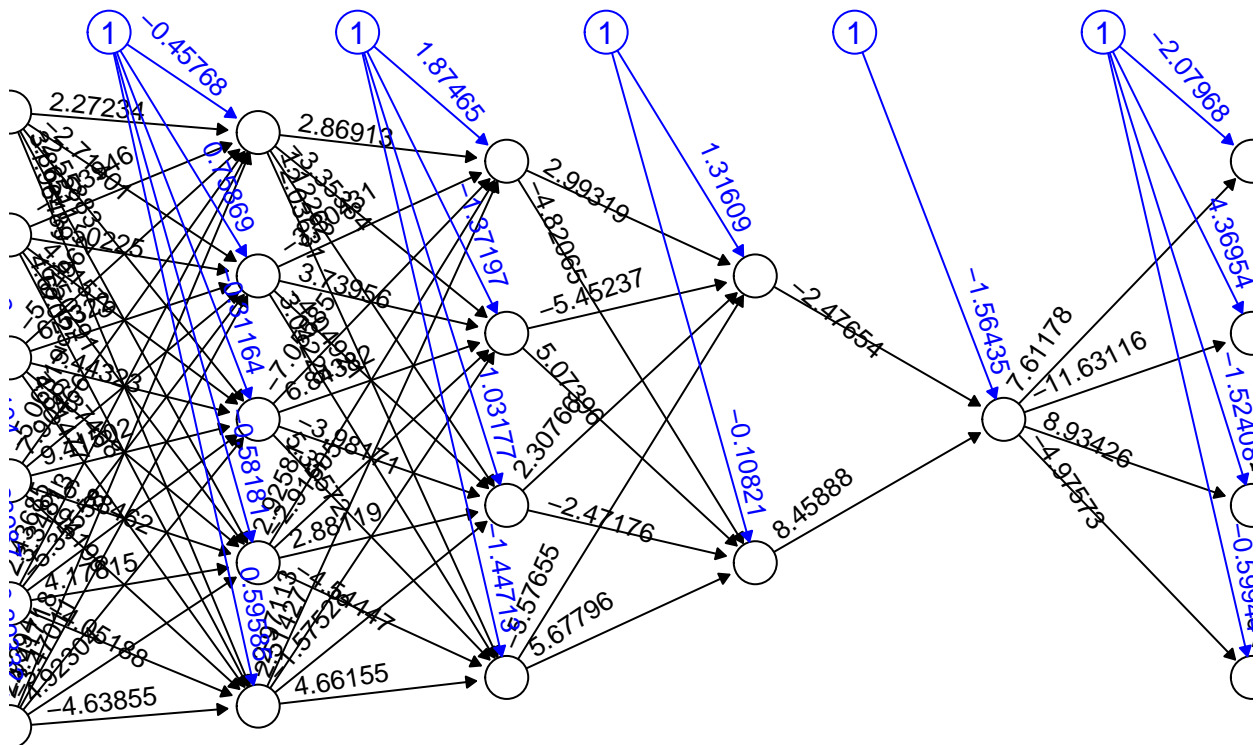
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## 55             6.5         2.8         4.6         1.5
## 37             5.5         3.5         1.3         0.2
## 146            6.7         3.0         5.2         2.3
## 70             5.6         2.5         3.9         1.1
## 45             5.1         3.8         1.9         0.4
## 124            6.3         2.7         4.9         1.8
## 20             5.1         3.8         1.5         0.3
## 76             6.6         3.0         4.4         1.4
## 144            6.8         3.2         5.9         2.3
## 3             4.7         3.2         1.3         0.2
## 88             6.3         2.3         4.4         1.3
## 10            4.9         3.1         1.5         0.1
## 136            7.7         3.0         6.1         2.3
## 126            7.2         3.2         6.0         1.8
## 102            5.8         2.7         5.1         1.9
## 125            6.7         3.3         5.7         2.1
## 64             6.1         2.9         4.7         1.4

```

## 111	6.5	3.2	5.1	2.0
## 122	5.6	2.8	4.9	2.0
## 32	5.4	3.4	1.5	0.4
## 147	6.3	2.5	5.0	1.9
## 123	7.7	2.8	6.7	2.0
## 95	5.6	2.7	4.2	1.3
## 101	6.3	3.3	6.0	2.5
## 149	6.2	3.4	5.4	2.3
## 143	5.8	2.7	5.1	1.9
## 94	5.0	2.3	3.3	1.0
## 150	5.9	3.0	5.1	1.8
## 11	5.4	3.7	1.5	0.2
## 83	5.8	2.7	3.9	1.2
## 54	5.5	2.3	4.0	1.3
## 57	6.3	3.3	4.7	1.6
## 61	5.0	2.0	3.5	1.0
## 48	4.6	3.2	1.4	0.2
## 29	5.2	3.4	1.4	0.2
## 69	6.2	2.2	4.5	1.5
## 130	7.2	3.0	5.8	1.6
## 115	5.8	2.8	5.1	2.4
## 145	6.7	3.3	5.7	2.5
## 17	5.4	3.9	1.3	0.4
## 50	5.0	3.3	1.4	0.2
## 96	5.7	3.0	4.2	1.2
## 35	4.9	3.1	1.5	0.2
## 93	5.8	2.6	4.0	1.2
## 49	5.3	3.7	1.5	0.2
## 12	4.8	3.4	1.6	0.2
## 14	4.3	3.0	1.1	0.1
## 60	5.2	2.7	3.9	1.4
## 18	5.1	3.5	1.4	0.3
## 97	5.7	2.9	4.2	1.3
## 109	6.7	2.5	5.8	1.8
## 134	6.3	2.8	5.1	1.5
## 62	5.9	3.0	4.2	1.5
## 113	6.8	3.0	5.5	2.1
## 75	6.4	2.9	4.3	1.3
## 119	7.7	2.6	6.9	2.3
## 41	5.0	3.5	1.3	0.3
## 27	5.0	3.4	1.6	0.4
## 25	4.8	3.4	1.9	0.2
## 89	5.6	3.0	4.1	1.3
## 100	5.7	2.8	4.1	1.3
## 91	5.5	2.6	4.4	1.2
## 19	5.7	3.8	1.7	0.3
## 137	6.3	3.4	5.6	2.4
## 46	4.8	3.0	1.4	0.3
## 103	7.1	3.0	5.9	2.1
## 85	5.4	3.0	4.5	1.5
## 6	5.4	3.9	1.7	0.4
## 44	5.0	3.5	1.6	0.6
## 86	6.0	3.4	4.5	1.6
## 71	5.9	3.2	4.8	1.8

## 36	5.0	3.2	1.2	0.2
## 104	6.3	2.9	5.6	1.8
## 42	4.5	2.3	1.3	0.3
## 139	6.0	3.0	4.8	1.8
## 118	7.7	3.8	6.7	2.2
## 106	7.6	3.0	6.6	2.1
## 9	4.4	2.9	1.4	0.2
## 43	4.4	3.2	1.3	0.2
## 84	6.0	2.7	5.1	1.6
## 66	6.7	3.1	4.4	1.4
## 39	4.4	3.0	1.3	0.2
## 7	4.6	3.4	1.4	0.3
## 72	6.1	2.8	4.0	1.3
## 117	6.5	3.0	5.5	1.8
## 108	7.3	2.9	6.3	1.8
## 4	4.6	3.1	1.5	0.2
## 38	4.9	3.6	1.4	0.1
## 138	6.4	3.1	5.5	1.8
## 65	5.6	2.9	3.6	1.3
## 5	5.0	3.6	1.4	0.2
## 2	4.9	3.0	1.4	0.2
## 87	6.7	3.1	4.7	1.5
## 82	5.5	2.4	3.7	1.0
## 40	5.1	3.4	1.5	0.2
## 77	6.8	2.8	4.8	1.4
## 128	6.1	3.0	4.9	1.8
## 67	5.6	3.0	4.5	1.5
## 92	6.1	3.0	4.6	1.4
## 131	7.4	2.8	6.1	1.9
## 74	6.1	2.8	4.7	1.2
## 56	5.7	2.8	4.5	1.3
## 59	6.6	2.9	4.6	1.3
## 120	6.0	2.2	5.0	1.5
## 23	4.6	3.6	1.0	0.2
## 13	4.8	3.0	1.4	0.1
## 33	5.2	4.1	1.5	0.1
## 107	4.9	2.5	4.5	1.7
## 127	6.2	2.8	4.8	1.8
## 24	5.1	3.3	1.7	0.5
## 116	6.4	3.2	5.3	2.3
## 34	5.5	4.2	1.4	0.2
## 68	5.8	2.7	4.1	1.0
## 58	4.9	2.4	3.3	1.0
## 73	6.3	2.5	4.9	1.5
## 80	5.7	2.6	3.5	1.0
## 8	5.0	3.4	1.5	0.2
## 99	5.1	2.5	3.0	1.1
## 121	6.9	3.2	5.7	2.3
## 133	6.4	2.8	5.6	2.2
##				
## \$model.list				
## \$model.list\$response				
## [1] "versicolor" "setosa" "virginica"				
##				

```
## $model.list$variables
## [1] "Sepal.Length" "Sepal.Width" "Petal.Length" "Petal.Width"
##
##
## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x5bfe53991878>
## <environment: 0x5bfe56f96240>
## attr("type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))
## }
## <bytecode: 0x5bfe53997cd8>
## <environment: 0x5bfe56f966d8>
## attr("type")
## [1] "logistic"
plot(model3, rep = 'best')
```



```
pred3 <- predict(model3, test_data)
pred3
```

```
##           [,1]           [,2]           [,3]
## 1  9.999036e-01  1.421452e-13  3.158454e-04
```

```
## 15 9.999036e-01 1.421452e-13 3.158454e-04
## 16 9.999036e-01 1.421452e-13 3.158454e-04
## 21 9.999036e-01 1.421452e-13 3.158454e-04
## 22 9.999036e-01 1.421452e-13 3.158454e-04
## 26 9.999036e-01 1.421452e-13 3.158454e-04
## 28 9.999036e-01 1.421452e-13 3.158454e-04
## 30 9.999036e-01 1.421452e-13 3.158454e-04
## 31 9.999036e-01 1.421452e-13 3.158454e-04
## 47 9.999036e-01 1.421452e-13 3.158454e-04
## 51 8.700491e-109 1.000000e+00 1.336467e-05
## 52 1.297176e-108 1.000000e+00 1.408254e-05
## 53 2.071300e-108 1.000000e+00 1.515804e-05
## 63 7.315655e-109 1.000000e+00 1.311504e-05
## 78 7.205867e-102 9.969511e-01 6.257672e-02
## 79 4.758556e-108 1.000000e+00 1.773884e-05
## 81 7.420057e-109 1.000000e+00 1.313390e-05
## 90 9.760581e-109 1.000000e+00 1.355172e-05
## 98 8.269050e-109 1.000000e+00 1.328727e-05
## 105 3.961383e-04 2.714381e-15 9.990498e-01
## 110 4.266206e-06 1.163679e-15 9.999579e-01
## 112 9.453037e-08 9.148913e-16 9.999839e-01
## 114 1.761865e-02 6.660749e-15 9.761323e-01
## 129 7.258686e-04 3.114619e-15 9.984396e-01
## 132 1.015953e-10 9.434979e-16 9.999723e-01
## 135 3.846955e-06 1.148504e-15 9.999600e-01
## 140 2.117849e-14 1.020414e-15 9.998810e-01
## 141 2.085738e-05 1.486541e-15 9.998942e-01
## 142 2.308972e-15 1.040769e-15 9.998185e-01
## 148 4.906116e-20 1.127430e-15 9.985372e-01
```

17 hidden layers with 10, 20, 30, 40, 40, 20, 10, 40, 50, 10, 20, 30, 40, 40, 20, 10, and 40 neurons

```
model4 <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length + Petal.Width, data = train_data,
# Print the model summary
head(model4))
```

```
## $call
## neuralnet(formula = Species ~ Sepal.Length + Sepal.Width + Petal.Length +
##     Petal.Width, data = train_data, hidden = c(10, 20, 30, 40,
##     40, 20, 10, 40, 50, 10, 20, 30, 40, 40, 20, 10, 40), linear.output = FALSE)
##
## $response
##      versicolor setosa virginica
## 1      FALSE      TRUE      FALSE
## 2       TRUE     FALSE      FALSE
## 3      FALSE     FALSE       TRUE
## 4      FALSE     TRUE      FALSE
## 5       TRUE     FALSE      FALSE
## 6      FALSE     FALSE       TRUE
## 7       TRUE     FALSE      FALSE
## 8      FALSE     TRUE      FALSE
## 9      FALSE     FALSE       TRUE
```

## 10	TRUE	FALSE	FALSE
## 11	FALSE	TRUE	FALSE
## 12	TRUE	FALSE	FALSE
## 13	FALSE	FALSE	TRUE
## 14	FALSE	FALSE	TRUE
## 15	FALSE	FALSE	TRUE
## 16	FALSE	FALSE	TRUE
## 17	FALSE	TRUE	FALSE
## 18	FALSE	FALSE	TRUE
## 19	FALSE	FALSE	TRUE
## 20	TRUE	FALSE	FALSE
## 21	FALSE	FALSE	TRUE
## 22	FALSE	FALSE	TRUE
## 23	FALSE	TRUE	FALSE
## 24	FALSE	FALSE	TRUE
## 25	FALSE	FALSE	TRUE
## 26	FALSE	FALSE	TRUE
## 27	FALSE	TRUE	FALSE
## 28	FALSE	FALSE	TRUE
## 29	TRUE	FALSE	FALSE
## 30	FALSE	TRUE	FALSE
## 31	FALSE	TRUE	FALSE
## 32	FALSE	TRUE	FALSE
## 33	FALSE	TRUE	FALSE
## 34	TRUE	FALSE	FALSE
## 35	TRUE	FALSE	FALSE
## 36	FALSE	TRUE	FALSE
## 37	FALSE	FALSE	TRUE
## 38	FALSE	FALSE	TRUE
## 39	FALSE	FALSE	TRUE
## 40	TRUE	FALSE	FALSE
## 41	TRUE	FALSE	FALSE
## 42	FALSE	TRUE	FALSE
## 43	TRUE	FALSE	FALSE
## 44	FALSE	TRUE	FALSE
## 45	TRUE	FALSE	FALSE
## 46	TRUE	FALSE	FALSE
## 47	TRUE	FALSE	FALSE
## 48	FALSE	TRUE	FALSE
## 49	TRUE	FALSE	FALSE
## 50	FALSE	TRUE	FALSE
## 51	FALSE	FALSE	TRUE
## 52	FALSE	FALSE	TRUE
## 53	FALSE	TRUE	FALSE
## 54	FALSE	FALSE	TRUE
## 55	FALSE	TRUE	FALSE
## 56	FALSE	FALSE	TRUE
## 57	TRUE	FALSE	FALSE
## 58	TRUE	FALSE	FALSE
## 59	TRUE	FALSE	FALSE
## 60	FALSE	TRUE	FALSE
## 61	FALSE	TRUE	FALSE
## 62	FALSE	TRUE	FALSE
## 63	TRUE	FALSE	FALSE

## 64	FALSE	FALSE	TRUE
## 65	TRUE	FALSE	FALSE
## 66	FALSE	FALSE	TRUE
## 67	FALSE	TRUE	FALSE
## 68	TRUE	FALSE	FALSE
## 69	TRUE	FALSE	FALSE
## 70	FALSE	TRUE	FALSE
## 71	FALSE	TRUE	FALSE
## 72	TRUE	FALSE	FALSE
## 73	FALSE	FALSE	TRUE
## 74	TRUE	FALSE	FALSE
## 75	FALSE	FALSE	TRUE
## 76	FALSE	FALSE	TRUE
## 77	FALSE	FALSE	TRUE
## 78	TRUE	FALSE	FALSE
## 79	TRUE	FALSE	FALSE
## 80	FALSE	TRUE	FALSE
## 81	FALSE	TRUE	FALSE
## 82	TRUE	FALSE	FALSE
## 83	TRUE	FALSE	FALSE
## 84	FALSE	TRUE	FALSE
## 85	FALSE	FALSE	TRUE
## 86	FALSE	FALSE	TRUE
## 87	TRUE	FALSE	FALSE
## 88	TRUE	FALSE	FALSE
## 89	FALSE	FALSE	TRUE
## 90	FALSE	TRUE	FALSE
## 91	TRUE	FALSE	FALSE
## 92	TRUE	FALSE	FALSE
## 93	FALSE	TRUE	FALSE
## 94	FALSE	TRUE	FALSE
## 95	TRUE	FALSE	FALSE
## 96	FALSE	TRUE	FALSE
## 97	FALSE	FALSE	TRUE
## 98	FALSE	TRUE	FALSE
## 99	FALSE	TRUE	FALSE
## 100	FALSE	FALSE	TRUE
## 101	FALSE	TRUE	FALSE
## 102	FALSE	TRUE	FALSE
## 103	FALSE	TRUE	FALSE
## 104	FALSE	FALSE	TRUE
## 105	TRUE	FALSE	FALSE
## 106	TRUE	FALSE	FALSE
## 107	TRUE	FALSE	FALSE
## 108	FALSE	FALSE	TRUE
## 109	FALSE	FALSE	TRUE
## 110	TRUE	FALSE	FALSE
## 111	FALSE	FALSE	TRUE
## 112	TRUE	FALSE	FALSE
## 113	FALSE	TRUE	FALSE
## 114	FALSE	TRUE	FALSE
## 115	FALSE	TRUE	FALSE
## 116	FALSE	TRUE	FALSE
## 117	TRUE	FALSE	FALSE


```
## 118      FALSE  TRUE      FALSE
## 119      FALSE FALSE      TRUE
## 120      FALSE FALSE      TRUE
```

```
##
```

```
## $covariate
```

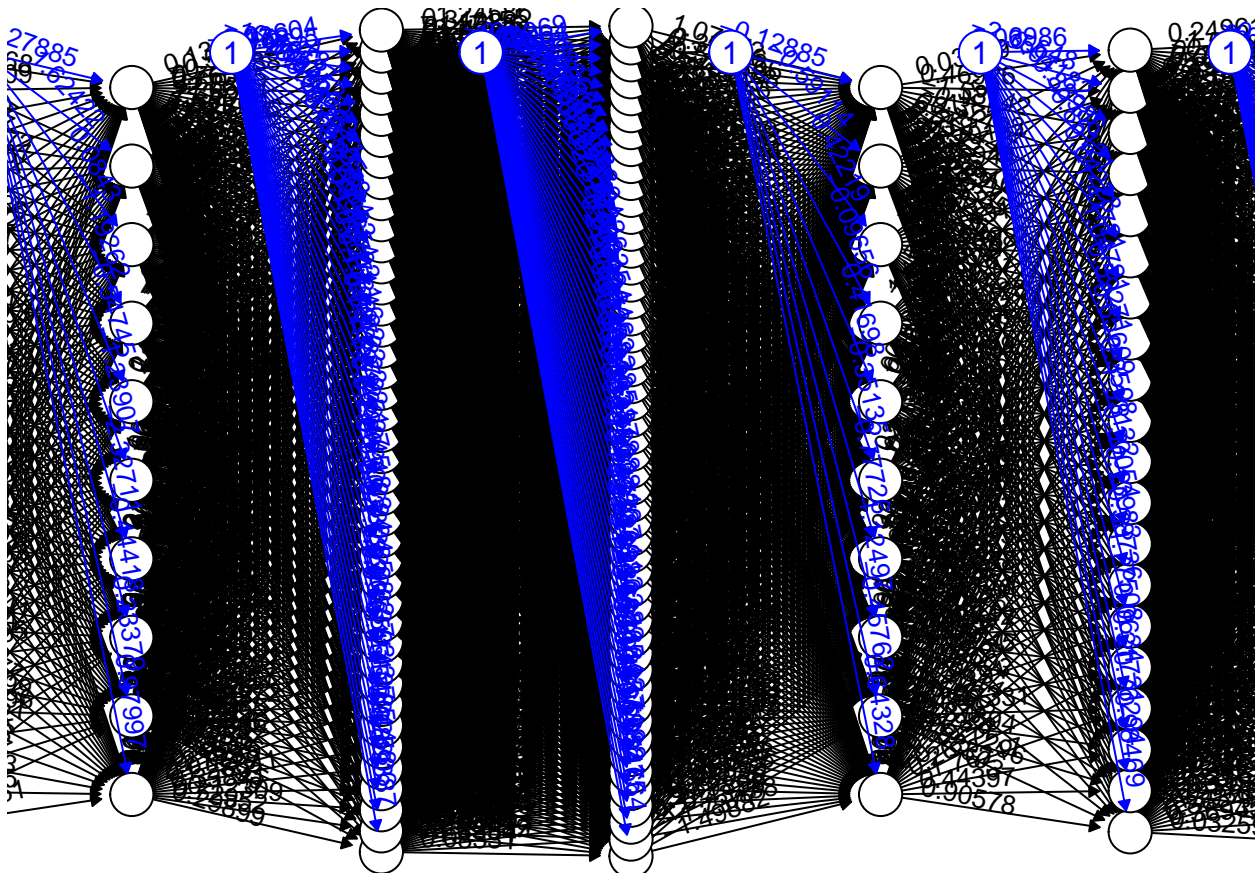
```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## 55           6.5         2.8         4.6         1.5
## 37           5.5         3.5         1.3         0.2
## 146          6.7         3.0         5.2         2.3
## 70           5.6         2.5         3.9         1.1
## 45           5.1         3.8         1.9         0.4
## 124          6.3         2.7         4.9         1.8
## 20           5.1         3.8         1.5         0.3
## 76           6.6         3.0         4.4         1.4
## 144          6.8         3.2         5.9         2.3
## 3           4.7         3.2         1.3         0.2
## 88           6.3         2.3         4.4         1.3
## 10           4.9         3.1         1.5         0.1
## 136          7.7         3.0         6.1         2.3
## 126          7.2         3.2         6.0         1.8
## 102          5.8         2.7         5.1         1.9
## 125          6.7         3.3         5.7         2.1
## 64           6.1         2.9         4.7         1.4
## 111          6.5         3.2         5.1         2.0
## 122          5.6         2.8         4.9         2.0
## 32           5.4         3.4         1.5         0.4
## 147          6.3         2.5         5.0         1.9
## 123          7.7         2.8         6.7         2.0
## 95           5.6         2.7         4.2         1.3
## 101          6.3         3.3         6.0         2.5
## 149          6.2         3.4         5.4         2.3
## 143          5.8         2.7         5.1         1.9
## 94           5.0         2.3         3.3         1.0
## 150          5.9         3.0         5.1         1.8
## 11           5.4         3.7         1.5         0.2
## 83           5.8         2.7         3.9         1.2
## 54           5.5         2.3         4.0         1.3
## 57           6.3         3.3         4.7         1.6
## 61           5.0         2.0         3.5         1.0
## 48           4.6         3.2         1.4         0.2
## 29           5.2         3.4         1.4         0.2
## 69           6.2         2.2         4.5         1.5
## 130          7.2         3.0         5.8         1.6
## 115          5.8         2.8         5.1         2.4
## 145          6.7         3.3         5.7         2.5
## 17           5.4         3.9         1.3         0.4
## 50           5.0         3.3         1.4         0.2
## 96           5.7         3.0         4.2         1.2
## 35           4.9         3.1         1.5         0.2
## 93           5.8         2.6         4.0         1.2
## 49           5.3         3.7         1.5         0.2
## 12           4.8         3.4         1.6         0.2
## 14           4.3         3.0         1.1         0.1
## 60           5.2         2.7         3.9         1.4
```

## 18	5.1	3.5	1.4	0.3
## 97	5.7	2.9	4.2	1.3
## 109	6.7	2.5	5.8	1.8
## 134	6.3	2.8	5.1	1.5
## 62	5.9	3.0	4.2	1.5
## 113	6.8	3.0	5.5	2.1
## 75	6.4	2.9	4.3	1.3
## 119	7.7	2.6	6.9	2.3
## 41	5.0	3.5	1.3	0.3
## 27	5.0	3.4	1.6	0.4
## 25	4.8	3.4	1.9	0.2
## 89	5.6	3.0	4.1	1.3
## 100	5.7	2.8	4.1	1.3
## 91	5.5	2.6	4.4	1.2
## 19	5.7	3.8	1.7	0.3
## 137	6.3	3.4	5.6	2.4
## 46	4.8	3.0	1.4	0.3
## 103	7.1	3.0	5.9	2.1
## 85	5.4	3.0	4.5	1.5
## 6	5.4	3.9	1.7	0.4
## 44	5.0	3.5	1.6	0.6
## 86	6.0	3.4	4.5	1.6
## 71	5.9	3.2	4.8	1.8
## 36	5.0	3.2	1.2	0.2
## 104	6.3	2.9	5.6	1.8
## 42	4.5	2.3	1.3	0.3
## 139	6.0	3.0	4.8	1.8
## 118	7.7	3.8	6.7	2.2
## 106	7.6	3.0	6.6	2.1
## 9	4.4	2.9	1.4	0.2
## 43	4.4	3.2	1.3	0.2
## 84	6.0	2.7	5.1	1.6
## 66	6.7	3.1	4.4	1.4
## 39	4.4	3.0	1.3	0.2
## 7	4.6	3.4	1.4	0.3
## 72	6.1	2.8	4.0	1.3
## 117	6.5	3.0	5.5	1.8
## 108	7.3	2.9	6.3	1.8
## 4	4.6	3.1	1.5	0.2
## 38	4.9	3.6	1.4	0.1
## 138	6.4	3.1	5.5	1.8
## 65	5.6	2.9	3.6	1.3
## 5	5.0	3.6	1.4	0.2
## 2	4.9	3.0	1.4	0.2
## 87	6.7	3.1	4.7	1.5
## 82	5.5	2.4	3.7	1.0
## 40	5.1	3.4	1.5	0.2
## 77	6.8	2.8	4.8	1.4
## 128	6.1	3.0	4.9	1.8
## 67	5.6	3.0	4.5	1.5
## 92	6.1	3.0	4.6	1.4
## 131	7.4	2.8	6.1	1.9
## 74	6.1	2.8	4.7	1.2
## 56	5.7	2.8	4.5	1.3

```

## 59          6.6          2.9          4.6          1.3
## 120         6.0          2.2          5.0          1.5
## 23          4.6          3.6          1.0          0.2
## 13          4.8          3.0          1.4          0.1
## 33          5.2          4.1          1.5          0.1
## 107         4.9          2.5          4.5          1.7
## 127         6.2          2.8          4.8          1.8
## 24          5.1          3.3          1.7          0.5
## 116         6.4          3.2          5.3          2.3
## 34          5.5          4.2          1.4          0.2
## 68          5.8          2.7          4.1          1.0
## 58          4.9          2.4          3.3          1.0
## 73          6.3          2.5          4.9          1.5
## 80          5.7          2.6          3.5          1.0
## 8           5.0          3.4          1.5          0.2
## 99          5.1          2.5          3.0          1.1
## 121         6.9          3.2          5.7          2.3
## 133         6.4          2.8          5.6          2.2
##
## $model.list
## $model.list$response
## [1] "versicolor" "setosa"      "virginica"
##
## $model.list$variables
## [1] "Sepal.Length" "Sepal.Width"  "Petal.Length" "Petal.Width"
##
##
## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x5bfe53991878>
## <environment: 0x5bfe557ee810>
## attr("type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))
## }
## <bytecode: 0x5bfe53997cd8>
## <environment: 0x5bfe557eece0>
## attr("type")
## [1] "logistic"
plot(model4, rep = 'best')

```



```
pred4 <- predict(model4, test_data)
pred4
```

```
##           [,1]      [,2]      [,3]
## 1  0.998740896 0.002362521 0.003209156
## 15 0.998749978 0.002366242 0.003214961
## 16 0.998745391 0.002365070 0.003211674
## 21 0.998736578 0.002357736 0.003207876
## 22 0.998734385 0.002353665 0.003208018
## 26 0.998730123 0.002335525 0.003213250
## 28 0.998739772 0.002361558 0.003208685
## 30 0.998730214 0.002300796 0.003230491
## 31 0.998729458 0.002317132 0.003221704
## 47 0.998736261 0.002357223 0.003207859
## 51 0.001152903 0.512633875 0.487269718
## 52 0.001152837 0.512644036 0.487260514
## 53 0.001152912 0.512632525 0.487271073
## 63 0.001152909 0.512633085 0.487270433
## 78 0.001153255 0.512580086 0.487318972
## 79 0.001153021 0.512615803 0.487286376
## 81 0.001152810 0.512648196 0.487256644
## 90 0.001152894 0.512635372 0.487268475
## 98 0.001152823 0.512646172 0.487258511
## 105 0.001153932 0.512476340 0.487413336
## 110 0.001153924 0.512477526 0.487412260
## 112 0.001153811 0.512494822 0.487396552
```

```
## 114 0.001153896 0.512481785 0.487408393
## 129 0.001153915 0.512478841 0.487411065
## 132 0.001153642 0.512520748 0.487372986
## 135 0.001153711 0.512510130 0.487382635
## 140 0.001153743 0.512505219 0.487387105
## 141 0.001153920 0.512478181 0.487411665
## 142 0.001153700 0.512511911 0.487381026
## 148 0.001153738 0.512506054 0.487386345
```

Model Evaluation

predict categories using test data

Create list of category name

prediction dataframe

create a table to display the actual and the predicted

```
evaluate_model <- function(pred, test_data) {
  labels <- c("setosa", "versicolor", "virginica")
  prediction_label <- data.frame(max.col(pred)) %>%
    mutate(pred = labels[max.col(pred)]) %>%
    select(2) %>%
    unlist()
  confusion_matrix <- table(test_data$Species, prediction_label)
  check <- as.numeric(test_data$Species) == max.col(pred)
  check
  accuracy <- (sum(check) / nrow(test_data)) * 100
  list(confusion_matrix = confusion_matrix, accuracy = accuracy)
}
```

Evaluate the model with two hidden layers

```
evaluation1 <- evaluate_model(pred, test_data)
print("Evaluation of Model 1:")
```

```
## [1] "Evaluation of Model 1:"
```

```
print(evaluation1$confusion_matrix)
```

```
##           prediction_label
##           setosa versicolor virginica
## setosa           10           0           0
## versicolor        0           9           0
## virginica          0           0          11
```

```
print(paste("Accuracy:", evaluation1$accuracy))
```

```
## [1] "Accuracy: 100"
```

Evaluate the model with three hidden layers

```
evaluation2 <- evaluate_model(pred2, test_data)
print("Evaluation of Model 2:")
```

```
## [1] "Evaluation of Model 2:"
```

```
print(evaluation2$confusion_matrix)
```

```
##           prediction_label
##           setosa versicolor virginica
##  setosa           10           0           0
##  versicolor        0           8           1
##  virginica         0           1          10
```

```
print(paste("Accuracy:", evaluation2$accuracy))
```

```
## [1] "Accuracy: 93.3333333333333"
```

Evaluate the model with 10 hidden layers

```
evaluation3 <- evaluate_model(pred3, test_data)
print("Evaluation of Model 3:")
```

```
## [1] "Evaluation of Model 3:"
```

```
print(evaluation3$confusion_matrix)
```

```
##           prediction_label
##           setosa versicolor virginica
##  setosa           10           0           0
##  versicolor        0           9           0
##  virginica         0           0          11
```

```
print(paste("Accuracy:", evaluation3$accuracy))
```

```
## [1] "Accuracy: 100"
```

Evaluate the model with 17 hidden layers

```
evaluation4 <- evaluate_model(pred4, test_data)
print("Evaluation of Model 4:")
```

```
## [1] "Evaluation of Model 4:"
```

```
print(evaluation4$confusion_matrix)
```

```
##           prediction_label
##           setosa versicolor
##  setosa           10           0
##  versicolor        0           9
##  virginica         0          11
```

```
print(paste("Accuracy:", evaluation4$accuracy))
```

```
## [1] "Accuracy: 63.3333333333333"
```

Tabular report

Number of Hidden Layers	Accuracy(%)
2	100
3	93.333333
10	100
17	63.333333

Increasing the complexity of a neural network by adding more hidden layers and neurons can enhance model performance, as evidenced by the highest accuracy achieved with a model containing 10 hidden layers. However, this improvement is not always guaranteed and heavily depends on the dataset's specific characteristics and the problem at hand. For instance, in model 4, despite increasing the hidden layers to 17, the accuracy decreased to 63.33%, illustrating that a more complex architecture does not necessarily lead to better performance and may even hinder it. Additionally, in terms of RAM consumption, the RAM gauge on the Posit website turned red, indicating that all available memory was consumed, highlighting the increased computational demands and resource constraints associated with overly complex models.