

Untitled

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R Markdown

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When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
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
```

```
iris <- iris %>%mutate_if(is.character, as.factor)
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
## 7	4.6	3.4	1.4	0.3	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa

## 10	4.9	3.1	1.5	0.1	setosa
## 11	5.4	3.7	1.5	0.2	setosa
## 12	4.8	3.4	1.6	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 15	5.8	4.0	1.2	0.2	setosa
## 16	5.7	4.4	1.5	0.4	setosa
## 17	5.4	3.9	1.3	0.4	setosa
## 18	5.1	3.5	1.4	0.3	setosa
## 19	5.7	3.8	1.7	0.3	setosa
## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa
## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 55	6.5	2.8	4.6	1.5	versicolor
## 56	5.7	2.8	4.5	1.3	versicolor
## 57	6.3	3.3	4.7	1.6	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 60	5.2	2.7	3.9	1.4	versicolor
## 61	5.0	2.0	3.5	1.0	versicolor
## 62	5.9	3.0	4.2	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor

## 64	6.1	2.9	4.7	1.4 versicolor
## 65	5.6	2.9	3.6	1.3 versicolor
## 66	6.7	3.1	4.4	1.4 versicolor
## 67	5.6	3.0	4.5	1.5 versicolor
## 68	5.8	2.7	4.1	1.0 versicolor
## 69	6.2	2.2	4.5	1.5 versicolor
## 70	5.6	2.5	3.9	1.1 versicolor
## 71	5.9	3.2	4.8	1.8 versicolor
## 72	6.1	2.8	4.0	1.3 versicolor
## 73	6.3	2.5	4.9	1.5 versicolor
## 74	6.1	2.8	4.7	1.2 versicolor
## 75	6.4	2.9	4.3	1.3 versicolor
## 76	6.6	3.0	4.4	1.4 versicolor
## 77	6.8	2.8	4.8	1.4 versicolor
## 78	6.7	3.0	5.0	1.7 versicolor
## 79	6.0	2.9	4.5	1.5 versicolor
## 80	5.7	2.6	3.5	1.0 versicolor
## 81	5.5	2.4	3.8	1.1 versicolor
## 82	5.5	2.4	3.7	1.0 versicolor
## 83	5.8	2.7	3.9	1.2 versicolor
## 84	6.0	2.7	5.1	1.6 versicolor
## 85	5.4	3.0	4.5	1.5 versicolor
## 86	6.0	3.4	4.5	1.6 versicolor
## 87	6.7	3.1	4.7	1.5 versicolor
## 88	6.3	2.3	4.4	1.3 versicolor
## 89	5.6	3.0	4.1	1.3 versicolor
## 90	5.5	2.5	4.0	1.3 versicolor
## 91	5.5	2.6	4.4	1.2 versicolor
## 92	6.1	3.0	4.6	1.4 versicolor
## 93	5.8	2.6	4.0	1.2 versicolor
## 94	5.0	2.3	3.3	1.0 versicolor
## 95	5.6	2.7	4.2	1.3 versicolor
## 96	5.7	3.0	4.2	1.2 versicolor
## 97	5.7	2.9	4.2	1.3 versicolor
## 98	6.2	2.9	4.3	1.3 versicolor
## 99	5.1	2.5	3.0	1.1 versicolor
## 100	5.7	2.8	4.1	1.3 versicolor
## 101	6.3	3.3	6.0	2.5 virginica
## 102	5.8	2.7	5.1	1.9 virginica
## 103	7.1	3.0	5.9	2.1 virginica
## 104	6.3	2.9	5.6	1.8 virginica
## 105	6.5	3.0	5.8	2.2 virginica
## 106	7.6	3.0	6.6	2.1 virginica
## 107	4.9	2.5	4.5	1.7 virginica
## 108	7.3	2.9	6.3	1.8 virginica
## 109	6.7	2.5	5.8	1.8 virginica
## 110	7.2	3.6	6.1	2.5 virginica
## 111	6.5	3.2	5.1	2.0 virginica
## 112	6.4	2.7	5.3	1.9 virginica
## 113	6.8	3.0	5.5	2.1 virginica
## 114	5.7	2.5	5.0	2.0 virginica
## 115	5.8	2.8	5.1	2.4 virginica
## 116	6.4	3.2	5.3	2.3 virginica
## 117	6.5	3.0	5.5	1.8 virginica

```
## 118      7.7      3.8      6.7      2.2 virginica
## 119      7.7      2.6      6.9      2.3 virginica
## 120      6.0      2.2      5.0      1.5 virginica
## 121      6.9      3.2      5.7      2.3 virginica
## 122      5.6      2.8      4.9      2.0 virginica
## 123      7.7      2.8      6.7      2.0 virginica
## 124      6.3      2.7      4.9      1.8 virginica
## 125      6.7      3.3      5.7      2.1 virginica
## 126      7.2      3.2      6.0      1.8 virginica
## 127      6.2      2.8      4.8      1.8 virginica
## 128      6.1      3.0      4.9      1.8 virginica
## 129      6.4      2.8      5.6      2.1 virginica
## 130      7.2      3.0      5.8      1.6 virginica
## 131      7.4      2.8      6.1      1.9 virginica
## 132      7.9      3.8      6.4      2.0 virginica
## 133      6.4      2.8      5.6      2.2 virginica
## 134      6.3      2.8      5.1      1.5 virginica
## 135      6.1      2.6      5.6      1.4 virginica
## 136      7.7      3.0      6.1      2.3 virginica
## 137      6.3      3.4      5.6      2.4 virginica
## 138      6.4      3.1      5.5      1.8 virginica
## 139      6.0      3.0      4.8      1.8 virginica
## 140      6.9      3.1      5.4      2.1 virginica
## 141      6.7      3.1      5.6      2.4 virginica
## 142      6.9      3.1      5.1      2.3 virginica
## 143      5.8      2.7      5.1      1.9 virginica
## 144      6.8      3.2      5.9      2.3 virginica
## 145      6.7      3.3      5.7      2.5 virginica
## 146      6.7      3.0      5.2      2.3 virginica
## 147      6.3      2.5      5.0      1.9 virginica
## 148      6.5      3.0      5.2      2.0 virginica
## 149      6.2      3.4      5.4      2.3 virginica
## 150      5.9      3.0      5.1      1.8 virginica
```

```
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
##
##
##
```

```
set.seed(254)
data_rows <- floor(0.80 * nrow(iris))
data_rows
```

```
## [1] 120
```

```
train_indices <- sample(c(1:nrow(iris)), data_rows)
train_indices
```

```
## [1] 55 37 146 70 45 124 20 76 144 3 88 10 136 126 102 125 64 111
## [19] 122 32 147 123 95 101 149 143 94 150 11 83 54 57 61 48 29 69
## [37] 130 115 145 17 50 96 35 93 49 12 14 60 18 97 109 134 62 113
## [55] 75 119 41 27 25 89 100 91 19 137 46 103 85 6 44 86 71 36
## [73] 104 42 139 118 106 9 43 84 66 39 7 72 117 108 4 38 138 65
## [91] 5 2 87 82 40 77 128 67 92 131 74 56 59 120 23 13 33 107
## [109] 127 24 116 34 68 58 73 80 8 99 121 133
```

```
train_data <- iris[-train_indices,]
train_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
## 28	5.2	3.5	1.5	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 78	6.7	3.0	5.0	1.7	versicolor
## 79	6.0	2.9	4.5	1.5	versicolor
## 81	5.5	2.4	3.8	1.1	versicolor
## 90	5.5	2.5	4.0	1.3	versicolor
## 98	6.2	2.9	4.3	1.3	versicolor
## 105	6.5	3.0	5.8	2.2	virginica
## 110	7.2	3.6	6.1	2.5	virginica
## 112	6.4	2.7	5.3	1.9	virginica
## 114	5.7	2.5	5.0	2.0	virginica
## 129	6.4	2.8	5.6	2.1	virginica
## 132	7.9	3.8	6.4	2.0	virginica
## 135	6.1	2.6	5.6	1.4	virginica
## 140	6.9	3.1	5.4	2.1	virginica
## 141	6.7	3.1	5.6	2.4	virginica
## 142	6.9	3.1	5.1	2.3	virginica
## 148	6.5	3.0	5.2	2.0	virginica

```
test_data <- iris[-train_indices, ]
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
## 28      5.2      3.5      1.5      0.2      setosa
## 30      4.7      3.2      1.6      0.2      setosa
## 31      4.8      3.1      1.6      0.2      setosa
## 47      5.1      3.8      1.6      0.2      setosa
## 51      7.0      3.2      4.7      1.4 versicolor
## 52      6.4      3.2      4.5      1.5 versicolor
## 53      6.9      3.1      4.9      1.5 versicolor
## 63      6.0      2.2      4.0      1.0 versicolor
## 78      6.7      3.0      5.0      1.7 versicolor
## 79      6.0      2.9      4.5      1.5 versicolor
## 81      5.5      2.4      3.8      1.1 versicolor
## 90      5.5      2.5      4.0      1.3 versicolor
## 98      6.2      2.9      4.3      1.3 versicolor
## 105     6.5      3.0      5.8      2.2 virginica
## 110     7.2      3.6      6.1      2.5 virginica
## 112     6.4      2.7      5.3      1.9 virginica
## 114     5.7      2.5      5.0      2.0 virginica
## 129     6.4      2.8      5.6      2.1 virginica
## 132     7.9      3.8      6.4      2.0 virginica
## 135     6.1      2.6      5.6      1.4 virginica
## 140     6.9      3.1      5.4      2.1 virginica
## 141     6.7      3.1      5.6      2.4 virginica
## 142     6.9      3.1      5.1      2.3 virginica
## 148     6.5      3.0      5.2      2.0 virginica
```

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

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

```

## 19 FALSE      TRUE      FALSE
## 20 FALSE      FALSE     TRUE
## 21 FALSE      FALSE     TRUE
## 22 FALSE      FALSE     TRUE
## 23 FALSE      FALSE     TRUE
## 24 FALSE      FALSE     TRUE
## 25 FALSE      FALSE     TRUE
## 26 FALSE      FALSE     TRUE
## 27 FALSE      FALSE     TRUE
## 28 FALSE      FALSE     TRUE
## 29 FALSE      FALSE     TRUE
## 30 FALSE      FALSE     TRUE
##
## $covariate
##      Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1           5.1         3.5         1.4         0.2
## 15          5.8         4.0         1.2         0.2
## 16          5.7         4.4         1.5         0.4
## 21          5.4         3.4         1.7         0.2
## 22          5.1         3.7         1.5         0.4
## 26          5.0         3.0         1.6         0.2
## 28          5.2         3.5         1.5         0.2
## 30          4.7         3.2         1.6         0.2
## 31          4.8         3.1         1.6         0.2
## 47          5.1         3.8         1.6         0.2
## 51          7.0         3.2         4.7         1.4
## 52          6.4         3.2         4.5         1.5
## 53          6.9         3.1         4.9         1.5
## 63          6.0         2.2         4.0         1.0
## 78          6.7         3.0         5.0         1.7
## 79          6.0         2.9         4.5         1.5
## 81          5.5         2.4         3.8         1.1
## 90          5.5         2.5         4.0         1.3
## 98          6.2         2.9         4.3         1.3
## 105         6.5         3.0         5.8         2.2
## 110         7.2         3.6         6.1         2.5
## 112         6.4         2.7         5.3         1.9
## 114         5.7         2.5         5.0         2.0
## 129         6.4         2.8         5.6         2.1
## 132         7.9         3.8         6.4         2.0
## 135         6.1         2.6         5.6         1.4
## 140         6.9         3.1         5.4         2.1
## 141         6.7         3.1         5.6         2.4
## 142         6.9         3.1         5.1         2.3
## 148         6.5         3.0         5.2         2.0
##
## $model.list
## $model.list$response
## [1] "setosa"      "versicolor" "virginica"
##
## $model.list$variables
## [1] "Sepal.Length" "Sepal.Width"  "Petal.Length" "Petal.Width"
##
##

```

```

## $err.fct
## function (x, y)
## {
##     1/2 * (y - x)^2
## }
## <bytecode: 0x60c2bf0b29f8>
## <environment: 0x60c2bf0b0448>
## attr(,"type")
## [1] "sse"
##
## $act.fct
## function (x)
## {
##     1/(1 + exp(-x))
## }
## <bytecode: 0x60c2bf0bae58>
## <environment: 0x60c2bf0b7990>
## attr(,"type")
## [1] "logistic"
##
## $linear.output
## [1] FALSE
##
## $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
## 28          5.2           3.5           1.5           0.2    setosa
## 30          4.7           3.2           1.6           0.2    setosa
## 31          4.8           3.1           1.6           0.2    setosa
## 47          5.1           3.8           1.6           0.2    setosa
## 51          7.0           3.2           4.7           1.4 versicolor
## 52          6.4           3.2           4.5           1.5 versicolor
## 53          6.9           3.1           4.9           1.5 versicolor
## 63          6.0           2.2           4.0           1.0 versicolor
## 78          6.7           3.0           5.0           1.7 versicolor
## 79          6.0           2.9           4.5           1.5 versicolor
## 81          5.5           2.4           3.8           1.1 versicolor
## 90          5.5           2.5           4.0           1.3 versicolor
## 98          6.2           2.9           4.3           1.3 versicolor
## 105         6.5           3.0           5.8           2.2  virginica
## 110         7.2           3.6           6.1           2.5  virginica
## 112         6.4           2.7           5.3           1.9  virginica
## 114         5.7           2.5           5.0           2.0  virginica
## 129         6.4           2.8           5.6           2.1  virginica
## 132         7.9           3.8           6.4           2.0  virginica
## 135         6.1           2.6           5.6           1.4  virginica
## 140         6.9           3.1           5.4           2.1  virginica
## 141         6.7           3.1           5.6           2.4  virginica
## 142         6.9           3.1           5.1           2.3  virginica

```



```

## 148          6.5          3.0          5.2          2.0  virginica
##
## $exclude
## NULL
##
## $net.result
## $net.result[[1]]
##          [,1]          [,2]          [,3]
## 1  9.991355e-01 0.001659219 2.888748e-10
## 15 9.991374e-01 0.001656208 2.881641e-10
## 16 9.991378e-01 0.001655553 2.880095e-10
## 21 9.991203e-01 0.001682876 2.944759e-10
## 22 9.991360e-01 0.001658348 2.886692e-10
## 26 9.991141e-01 0.001692545 2.967733e-10
## 28 9.991335e-01 0.001662327 2.896091e-10
## 30 9.991297e-01 0.001668198 2.909975e-10
## 31 9.991249e-01 0.001675776 2.927920e-10
## 47 9.991361e-01 0.001658264 2.886495e-10
## 51 1.193086e-03 0.991841402 1.173366e-03
## 52 1.161553e-03 0.991731584 1.201323e-03
## 53 8.680128e-04 0.990089810 1.542250e-03
## 63 1.151139e-03 0.991662037 1.210063e-03
## 78 4.176569e-05 0.936557818 2.041222e-02
## 79 6.882928e-04 0.988550031 1.881375e-03
## 81 1.186391e-03 0.991822512 1.179286e-03
## 90 9.992943e-04 0.990924100 1.366906e-03
## 98 1.180271e-03 0.991803859 1.184722e-03
## 105 2.978676e-11 0.001970871 9.997422e-01
## 110 3.726085e-11 0.002268828 9.996876e-01
## 112 8.684217e-11 0.003860900 9.993550e-01
## 114 4.650096e-11 0.002607858 9.996223e-01
## 129 3.229135e-11 0.002073528 9.997237e-01
## 132 4.964612e-10 0.011494713 9.971312e-01
## 135 6.020337e-11 0.003067363 9.995288e-01
## 140 5.744430e-10 0.012587730 9.967502e-01
## 141 3.729272e-11 0.002270048 9.996874e-01
## 142 5.238735e-09 0.048828764 9.787701e-01
## 148 1.252679e-09 0.020411292 9.936789e-01
##
##
## $weights
## $weights[[1]]
## $weights[[1]][[1]]
##          [,1]          [,2]          [,3]          [,4]
## [1,] 1.5850607 0.07276711 1.2989067 0.820164988
## [2,] -0.5460290 -1.01895101 0.2002771 -0.003823878
## [3,] 0.7748516 -0.20160588 0.3085626 1.229014294
## [4,] -0.7914394 1.24033480 -0.6046214 -0.849104820
## [5,] 0.0689953 1.00543305 -0.6077275 -0.583076099
##
## $weights[[1]][[2]]
##          [,1]          [,2]
## [1,] -7.255405 0.4441351
## [2,] 13.302418 -3.6911194

```

```

## [3,] -13.922219  8.0373052
## [4,]  2.365087 -7.6101736
## [5,]  5.607081 -12.5279189
##
## $weights[[1]][[3]]
##      [,1]      [,2]      [,3]
## [1,] -6.703558  4.822063 -6.768551
## [2,] 13.759986 -11.225064 -15.200882
## [3,] -17.809196 -11.223174 15.267968
##
##
##
## $generalized.weights
## $generalized.weights[[1]]
##      [,1]      [,2]      [,3]      [,4]      [,5]
## 1  -0.005685301  0.010861959 -0.012227302 -9.228596e-04 0.004637927
## 15 -0.002720428  0.004389956 -0.004682075  1.120169e-05 0.002219256
## 16 -0.001771065  0.003065011 -0.003386790 -1.150891e-04 0.001444789
## 21 -0.031429623  0.062039050 -0.069646516 -6.193651e-03 0.025639519
## 22 -0.004347101  0.008996646 -0.010458119 -1.102484e-03 0.003546257
## 26 -0.038074132  0.086018039 -0.099084295 -1.342101e-02 0.031059950
## 28 -0.009172052  0.017580883 -0.019765557 -1.519396e-03 0.007482333
## 30 -0.012717459  0.031402218 -0.037776711 -5.988049e-03 0.010374594
## 31 -0.020357092  0.048693528 -0.057624622 -8.692756e-03 0.016606821
## 47 -0.004402584  0.008760877 -0.010094442 -9.164908e-04 0.003591518
## 51  0.077848382  0.155948674 -0.206713989 -1.606611e-01 0.033188241
## 52  0.128711976  0.272924195 -0.355484391 -2.747692e-01 0.076236632
## 53  0.799175435  1.604466118 -2.116109619 -1.659425e+00 0.501925805
## 63  0.152099317  0.315226384 -0.414650787 -3.218353e-01 0.082583721
## 78  6.010937870 11.919084762 -15.843896607 -1.244767e+01 3.787893504
## 79  1.232104935  2.707606028 -3.478557945 -2.691667e+00 0.775960339
## 81  0.085752552  0.182049599 -0.240240241 -1.827583e-01 0.042596309
## 90  0.456457070  1.001007571 -1.297041674 -9.956993e-01 0.285934196
## 98  0.095778373  0.200424563 -0.262964139 -2.027670e-01 0.051520438
## 105 0.250031258  0.693430548 -0.825212107 -6.518235e-01 0.157566885
## 110 0.492433921  1.455023585 -1.668518562 -1.304675e+00 0.310326380
## 112 1.851636707  3.829323668 -5.066983590 -4.021626e+00 1.166880213
## 114 0.763987727  2.045091743 -2.489599190 -1.949609e+00 0.481456558
## 129 0.370016152  0.901346244 -1.125636307 -8.934018e-01 0.233179980
## 132 4.394138000  9.177165589 -11.842156515 -9.365144e+00 2.769134591
## 135 1.250325387  2.705246180 -3.526617029 -2.814858e+00 0.787941021
## 140 4.505154698  9.402278037 -12.271166859 -9.680679e+00 2.839095870
## 141 0.531576805  1.390076916 -1.679261011 -1.320174e+00 0.334993713
## 142 6.920754508 14.386460597 -18.803122405 -1.476195e+01 4.361374603
## 148 5.086318176 11.589299969 -14.655322154 -1.149146e+01 3.205339328
##      [,6]      [,7]      [,8]      [,9]      [,10]
## 1  -0.008860897  0.009974705  7.528333e-04  0.006280644 -0.011999379
## 15 -0.003581208  0.003819510 -9.141895e-06  0.003005301 -0.004849654
## 16 -0.002500351  0.002762848  9.388204e-05  0.001956524 -0.003385966
## 21 -0.050609897  0.056815874  5.052600e-03  0.034720818 -0.068535544
## 22 -0.007339218  0.008531445  8.993639e-04  0.004802313 -0.009938737
## 26 -0.070171333  0.080830452  1.094848e-02  0.042061116 -0.095025518
## 28 -0.014342026  0.016124224  1.239470e-03  0.010132515 -0.019421883
## 30 -0.025617102  0.030817243  4.884864e-03  0.014049185 -0.034690532

```

```

## 31 -0.039722920 0.047008672 7.091291e-03 0.022488813 -0.053792523
## 47 -0.007146888 0.008234772 7.476384e-04 0.004863606 -0.009678280
## 51 0.084101471 -0.101036782 -7.940404e-02 -0.069455622 -0.136121581
## 52 0.167376828 -0.214569995 -1.663083e-01 -0.111180179 -0.234770153
## 53 1.009735580 -1.330548698 -1.043481e+00 -0.685431582 -1.375758445
## 63 0.186359836 -0.236019522 -1.842950e-01 -0.132666122 -0.272349444
## 78 7.511174111 -9.984418502 -7.844213e+00 -5.153249671 -10.218347579
## 79 1.705819917 -2.191169897 -1.695552e+00 -1.056379088 -2.321335831
## 81 0.102508712 -0.127264288 -9.835483e-02 -0.075474470 -0.158162978
## 90 0.629035691 -0.813813812 -6.249753e-01 -0.391618730 -0.858478062
## 98 0.117829957 -0.148460364 -1.153100e-01 -0.083623803 -0.173275800
## 105 0.436992552 -0.520039700 -4.107721e-01 -0.214353880 -0.594483712
## 110 0.916940502 -1.051482452 -8.221919e-01 -0.422167691 -1.247403665
## 112 2.413198728 -3.193157112 -2.534385e+00 -1.587423640 -3.282910726
## 114 1.288795073 -1.568918410 -1.228623e+00 -0.654973026 -1.753273968
## 129 0.568018776 -0.709363680 -5.630121e-01 -0.317217935 -0.772731554
## 132 5.783351298 -7.462795578 -5.901810e+00 -3.767131403 -7.867659674
## 135 1.704817428 -2.222435729 -1.773893e+00 -1.071914363 -2.319229787
## 140 5.925214224 -7.733152361 -6.100656e+00 -3.862306977 -8.060650520
## 141 0.876011684 -1.058252016 -8.319595e-01 -0.455725223 -1.191724372
## 142 9.066188703 -11.849510211 -9.302810e+00 -5.933221683 -12.333631963
## 148 7.303452613 -9.235622283 -7.241793e+00 -4.360543134 -9.935602335
##      [,11]      [,12]
## 1 0.013507695 1.019496e-03
## 15 0.005172362 -1.237534e-05
## 16 0.003741440 1.271400e-04
## 21 0.076939635 6.842223e-03
## 22 0.011553249 1.217929e-03
## 26 0.109460022 1.482640e-02
## 28 0.021835327 1.678499e-03
## 30 0.041732537 6.615089e-03
## 31 0.063658846 9.603022e-03
## 47 0.011151490 1.012460e-03
## 51 0.182219298 1.414734e-01
## 52 0.306377035 2.367336e-01
## 53 1.814671670 1.423027e+00
## 63 0.359810537 2.790812e-01
## 78 13.583143342 1.067152e+01
## 79 2.982364068 2.307710e+00
## 81 0.210089073 1.595577e-01
## 90 1.112574997 8.540496e-01
## 98 0.228394006 1.759678e-01
## 105 0.707461159 5.588137e-01
## 110 1.430434718 1.118508e+00
## 112 4.343967053 3.447774e+00
## 114 2.134353949 1.671416e+00
## 129 0.965017323 7.659208e-01
## 132 10.152379107 8.028816e+00
## 135 3.023397919 2.413201e+00
## 140 10.520173370 8.299326e+00
## 141 1.439644345 1.131796e+00
## 142 16.120074469 1.265554e+01
## 148 12.564129206 9.851726e+00
##

```

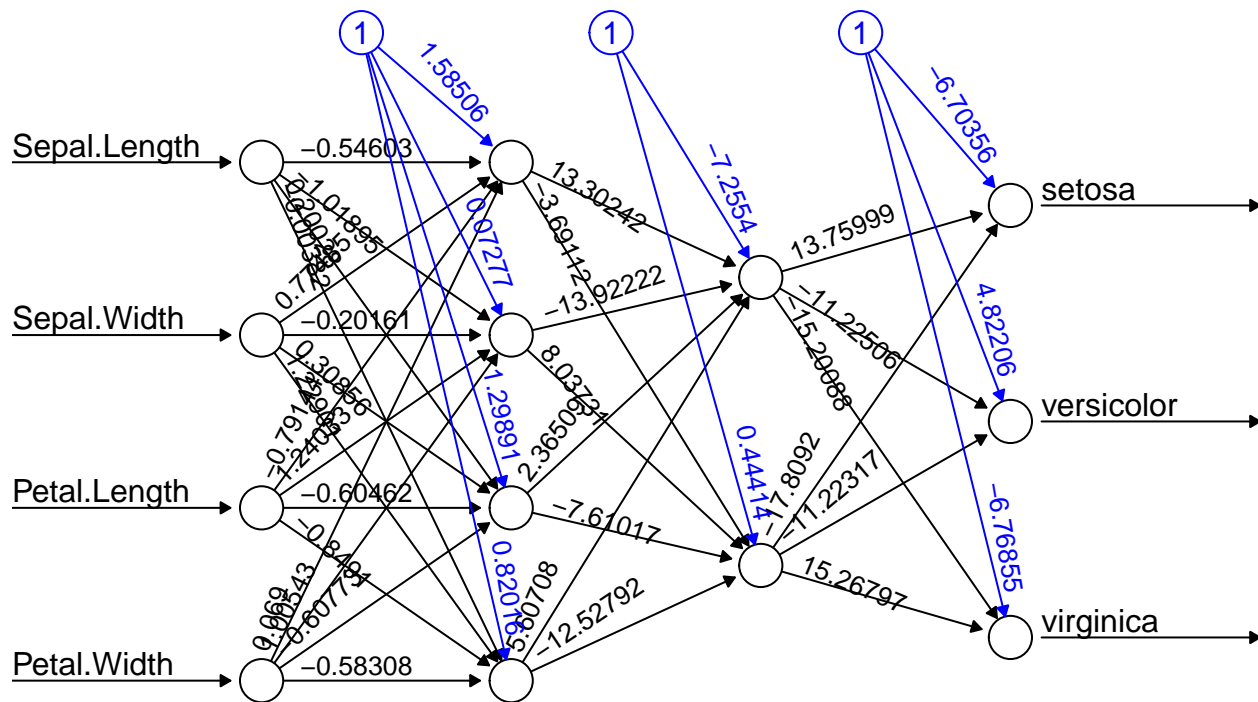
```

##
## $startweights
## $startweights[[1]]
## $startweights[[1]][[1]]
##      [,1]      [,2]      [,3]      [,4]
## [1,] 0.1784364 0.03180517 1.02430425 0.86798143
## [2,] -1.5608613 -1.07548182 -0.07104261 0.04399257
## [3,] -0.7420118 -0.25677290 -0.69559296 1.28046684
## [4,] -1.5417923 1.18186111 -0.55477281 -0.67644125
## [5,] 0.6627263 0.89462278 -0.48677958 -0.19119685
##
## $startweights[[1]][[2]]
##      [,1]      [,2]
## [1,] 1.38259854 0.5498319
## [2,] 0.74860902 0.3447832
## [3,] 2.34618058 1.5688060
## [4,] 0.21910371 1.1177046
## [5,] -0.02222762 -2.2190741
##
## $startweights[[1]][[3]]
##      [,1]      [,2]      [,3]
## [1,] -0.6102503 -0.3899154 -0.2255378
## [2,] -0.2183231 -0.1562910 -0.2408516
## [3,] -1.0091955 -0.5116348 1.4387121
##
##
##
## $result.matrix
##      [,1]
## error      0.004401359
## reached.threshold 0.009244609
## steps      169.000000000
## Intercept.to.1layhid1 1.585060688
## Sepal.Length.to.1layhid1 -0.546029023
## Sepal.Width.to.1layhid1 0.774851604
## Petal.Length.to.1layhid1 -0.791439362
## Petal.Width.to.1layhid1 0.068995300
## Intercept.to.1layhid2 0.072767109
## Sepal.Length.to.1layhid2 -1.018951007
## Sepal.Width.to.1layhid2 -0.201605879
## Petal.Length.to.1layhid2 1.240334805
## Petal.Width.to.1layhid2 1.005433051
## Intercept.to.1layhid3 1.298906684
## Sepal.Length.to.1layhid3 0.200277124
## Sepal.Width.to.1layhid3 0.308562585
## Petal.Length.to.1layhid3 -0.604621396
## Petal.Width.to.1layhid3 -0.607727533
## Intercept.to.1layhid4 0.820164988
## Sepal.Length.to.1layhid4 -0.003823878
## Sepal.Width.to.1layhid4 1.229014294
## Petal.Length.to.1layhid4 -0.849104820
## Petal.Width.to.1layhid4 -0.583076099
## Intercept.to.2layhid1 -7.255404980
## 1layhid1.to.2layhid1 13.302418227

```

```
## 1layhid2.to.2layhid1      -13.922219420
## 1layhid3.to.2layhid1       2.365086708
## 1layhid4.to.2layhid1       5.607080502
## Intercept.to.2layhid2      0.444135101
## 1layhid1.to.2layhid2      -3.691119418
## 1layhid2.to.2layhid2       8.037305200
## 1layhid3.to.2layhid2      -7.610173637
## 1layhid4.to.2layhid2     -12.527918866
## Intercept.to.setosa       -6.703558292
## 2layhid1.to.setosa        13.759986212
## 2layhid2.to.setosa       -17.809195535
## Intercept.to.versicolor    4.822063224
## 2layhid1.to.versicolor   -11.225064330
## 2layhid2.to.versicolor   -11.223173824
## Intercept.to.virginica     -6.768551046
## 2layhid1.to.virginica    -15.200882084
## 2layhid2.to.virginica     15.267968382
##
## attr(,"class")
## [1] "nn"
```

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



Error: 0.004401 Steps: 169

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

```
##           [,1]      [,2]      [,3]
## 1  9.991355e-01 0.001659219 2.888748e-10
## 15 9.991374e-01 0.001656208 2.881641e-10
## 16 9.991378e-01 0.001655553 2.880095e-10
```

```
## 21 9.991203e-01 0.001682876 2.944759e-10
## 22 9.991360e-01 0.001658348 2.886692e-10
## 26 9.991141e-01 0.001692545 2.967733e-10
## 28 9.991335e-01 0.001662327 2.896091e-10
## 30 9.991297e-01 0.001668198 2.909975e-10
## 31 9.991249e-01 0.001675776 2.927920e-10
## 47 9.991361e-01 0.001658264 2.886495e-10
## 51 1.193086e-03 0.991841402 1.173366e-03
## 52 1.161553e-03 0.991731584 1.201323e-03
## 53 8.680128e-04 0.990089810 1.542250e-03
## 63 1.151139e-03 0.991662037 1.210063e-03
## 78 4.176569e-05 0.936557818 2.041222e-02
## 79 6.882928e-04 0.988550031 1.881375e-03
## 81 1.186391e-03 0.991822512 1.179286e-03
## 90 9.992943e-04 0.990924100 1.366906e-03
## 98 1.180271e-03 0.991803859 1.184722e-03
## 105 2.978676e-11 0.001970871 9.997422e-01
## 110 3.726085e-11 0.002268828 9.996876e-01
## 112 8.684217e-11 0.003860900 9.993550e-01
## 114 4.650096e-11 0.002607858 9.996223e-01
## 129 3.229135e-11 0.002073528 9.997237e-01
## 132 4.964612e-10 0.011494713 9.971312e-01
## 135 6.020337e-11 0.003067363 9.995288e-01
## 140 5.744430e-10 0.012587730 9.967502e-01
## 141 3.729272e-11 0.002270048 9.996874e-01
## 142 5.238735e-09 0.048828764 9.787701e-01
## 148 1.252679e-09 0.020411292 9.936789e-01

labels <- c("setosa", "versicolor", "virginica")
prediction_label <- data.frame(max.col(pred)) %>%
mutate(pred=labels[max.col.pred.]) %>%
select(2) %>%
unlist()
table(test_data$Species, prediction_label)

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

check = as.numeric(test_data$Species) == max.col(pred)
check

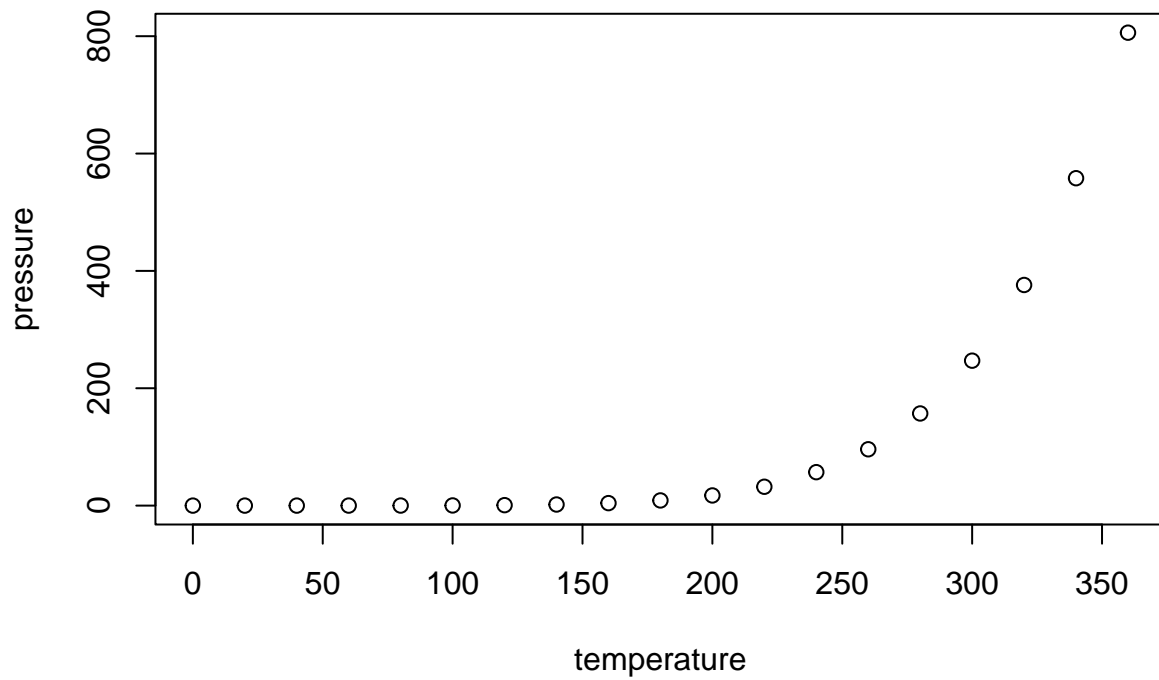
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [16] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE

accuracy <- (sum(check)/nrow(test_data))*100
print(accuracy)

## [1] 100
```

Including Plots

You can also embed plots, for example:



Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.