## Tamburo ANN

## 2024-08-01

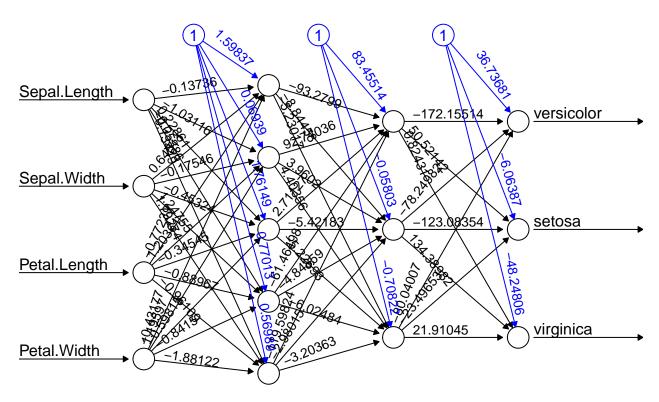
```
# Install and load required packages
install.packages(c('neuralnet', 'keras', 'tensorflow'), dependencies = TRUE)
##
## The downloaded binary packages are in
  /var/folders/bb/9352ds8s1g5cscthpcw8v4t40000gn/T//RtmpT7MLQq/downloaded_packages
install.packages("tidyverse")
##
## The downloaded binary packages are in
   /var/folders/bb/9352ds8s1g5cscthpcw8v4t40000gn/T//RtmpT7MLQq/downloaded_packages
install.packages("cowplot")
##
## The downloaded binary packages are in
  /var/folders/bb/9352ds8s1g5cscthpcw8v4t40000gn/T//RtmpT7MLQq/downloaded_packages
library(neuralnet)
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()
                     masks stats::lag()
## x dplyr::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(cowplot)
##
## Attaching package: 'cowplot'
## The following object is masked from 'package:lubridate':
##
```

##

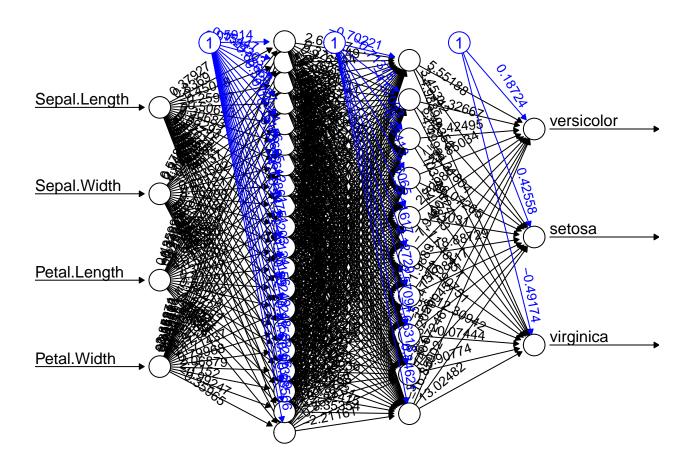
stamp

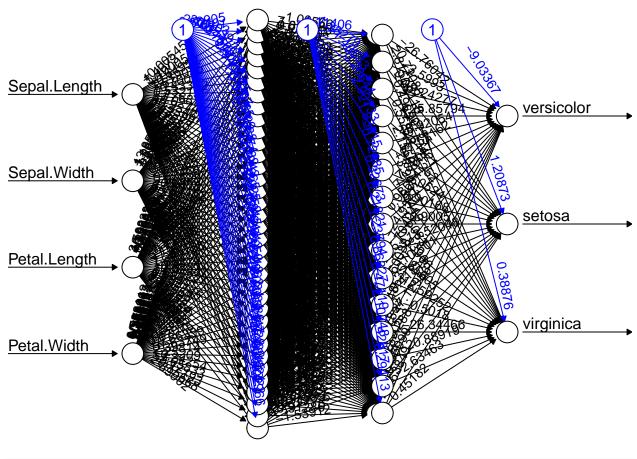
```
# Prepare the iris dataset
iris <- iris %>% mutate_if(is.character, as.factor)
summary(iris)
##
    Sepal.Length
                     Sepal.Width
                                     Petal.Length
                                                     Petal.Width
## Min.
          :4.300 Min.
                           :2.000
                                          :1.000
                                                           :0.100
                                    Min.
                                                     Min.
## 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 the seed and split the data into training and test sets
set.seed(254)
data_rows <- floor(0.80 * nrow(iris))</pre>
train indices <- sample(c(1:nrow(iris)), data rows)</pre>
train_data <- iris[train_indices, ]</pre>
test_data <- iris[-train_indices, ]</pre>
# Define a function to train and evaluate the neural network model
train_and_evaluate <- function(hidden_layers) {</pre>
  model <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length + Petal.Width,
                     data = train_data, hidden = hidden_layers, linear.output = FALSE)
  pred <- predict(model, test_data)</pre>
  labels <- c("setosa", "versicolor", "virginica")</pre>
  prediction_label <- data.frame(max.col(pred)) %>%
   mutate(pred = labels[max.col(pred)]) %>%
    select(pred) %>%
   unlist()
  check <- as.numeric(test_data$Species) == max.col(pred)</pre>
  accuracy <- (sum(check) / nrow(test_data)) * 100</pre>
  list(model = model, accuracy = accuracy)
}
# Train and evaluate models with different hidden layer configurations
hidden_layers_list <- list(c(5, 3), c(20, 10), c(35, 15))
results <- data.frame(Hidden_Layers = character(), Accuracy = numeric(), stringsAsFactors = FALSE)
models <- list()</pre>
for (hidden_layers in hidden_layers_list) {
  result <- train_and_evaluate(hidden_layers)</pre>
  models <- c(models, list(result$model))</pre>
  results <- rbind(results, data.frame(Hidden_Layers = paste(hidden_layers, collapse = ", "), Accuracy
}
# Create plots for the models
```

```
model_plots <- lapply(1:length(models), function(i) {
   plot(models[[i]], rep = 'best', main = paste("Model:", paste(hidden_layers_list[[i]], collapse = ", "
})</pre>
```



Error: 1.002153 Steps: 3195





```
# Plot the accuracy bar graph
accuracy_plot <- ggplot(results, aes(x = Hidden_Layers, y = Accuracy)) +
    geom_bar(stat = "identity") +
    labs(title = "Accuracy of Neural Network Models", x = "Hidden Layers", y = "Accuracy (%)") +
    theme_minimal()
# Print the accuracy of the models
print(results)</pre>
```

```
## Hidden_Layers Accuracy
## 1 5, 3 100
## 2 20, 10 100
## 3 35, 15 100
```

```
# Combine the model plots and accuracy plot using cowplot
top_row <- plot_grid(plotlist = model_plots, nrow = 1)
bottom_row <- plot_grid(accuracy_plot)
final_plot <- plot_grid(top_row, bottom_row, ncol = 1, rel_heights = c(2, 1))
# Display the final combined plot
print(final_plot)</pre>
```

