

Artificial Neural Networks

2024-08-01

```
# Install the necessary packages for neural networks and data manipulation
install.packages(c('neuralnet', 'keras', 'tensorflow'), dependencies = TRUE)
```

```
##
## The downloaded binary packages are in
## /var/folders/bb/9352ds8s1g5cscthpcw8v4t40000gn/T//RtmpxiLVxs/downloaded_packages
```

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

```
##
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```

```
# Load the required libraries
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()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
# Prepare the iris dataset by converting any character columns to factors
iris <- iris %>% mutate_if(is.character, as.factor)
summary(iris) # Display summary statistics of the dataset
```

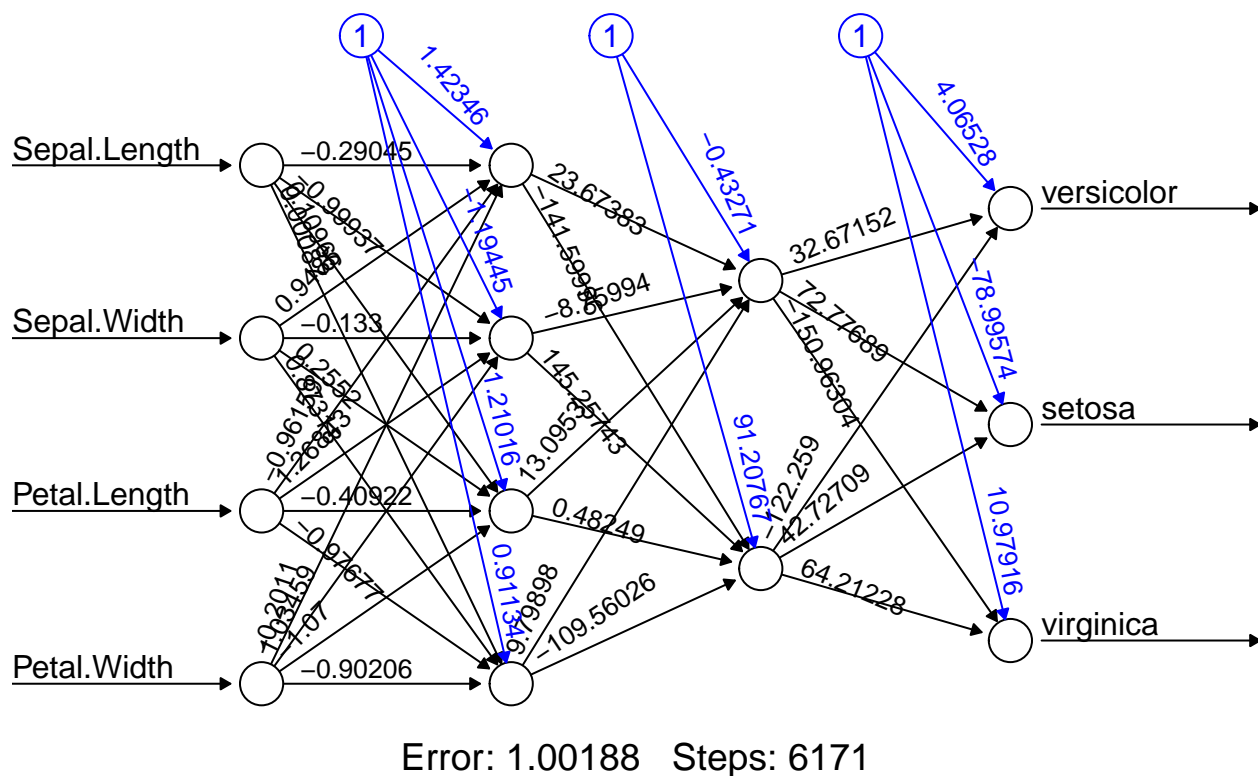
```
##   Sepal.Length   Sepal.Width   Petal.Length   Petal.Width
##   Min.    :4.300   Min.    :2.000   Min.    :1.000   Min.    :0.100
##   1st Qu.:5.100   1st Qu.:2.800   1st Qu.:1.600   1st Qu.:0.300
##   Median :5.800   Median :3.000   Median :4.350   Median :1.300
##   Mean   :5.843   Mean   :3.057   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
##
##
##

# Split the data into training and test sets
set.seed(254) # Set seed for reproducibility
data_rows <- floor(0.80 * nrow(iris)) # Calculate the number of rows for the training set
train_indices <- sample(c(1:nrow(iris)), data_rows) # Randomly sample indices for the training set
train_data <- iris[train_indices, ] # Create the training dataset
test_data <- iris[-train_indices, ] # Create the test dataset

# Train a neural network model
# Define the formula and structure of the neural network
model <- neuralnet(Species ~ Sepal.Length + Sepal.Width + Petal.Length + Petal.Width,
                    data = train_data, hidden = c(4, 2), linear.output = FALSE)

# Visualize the trained neural network model
plot(model, rep = 'best')
```



```
# Predict the species for the test dataset
pred <- predict(model, test_data)
```

```

# Map the predictions to species labels
labels <- c("setosa", "versicolor", "virginica")
prediction_label <- data.frame(max.col(pred)) %>%
  mutate(pred = labels[max.col.pred.]) %>%
  select(2) %>%
  unlist()

```

```

# Generate a confusion matrix to compare predicted and actual species
table(test_data$Species, prediction_label)

```

```

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

```

```

# Calculate the accuracy of the model
check <- as.numeric(test_data$Species) == max.col(pred) # Compare predictions to actual species
accuracy <- (sum(check) / nrow(test_data)) * 100 # Calculate the accuracy percentage
accuracy # Print the accuracy

```

```

## [1] 100

```