rnn: Recurrent Neural Network architectures in native R

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Abstract

The R package rnn implements several Recurrent Neural Network (RNN) architectures in the R language. The native R implementations of these architectures allow scientists familiar with the R language, to develop an intuitive understanding of these architectures, something which is not possible with production frameworks, such as TensorFlow, PyTorch or CNTK.

1 About package rnn in R

The ${\tt rnn}$ package is available on CRAN at https://cran.r-project.org/package=rnn and can be installed using:

```
install.packages('rnn')
```

After installation, the package can be loaded using:

```
library(rnn)
```

A list of all the exported functions can be printed using:

A list of all the functions - including non-exported ones - can be printed using:

```
ls(getNamespace('rnn'), all.names=TRUE)
    [1] ".__NAMESPACE__."
                                 ".__S3MethodsTable__." ".packageName"
##
##
    [4] "b2i"
                                 "backprop_gru"
                                                         "backprop_lstm"
                                                         "bin2int"
##
    [7] "backprop_r"
                                 "backprop_rnn"
  [10] "clean_lstm"
                                 "clean_r"
                                                         "clean_rnn"
  Г137
        "epoch_annealing"
                                 "epoch_print"
                                                         "i2b"
        "init_gru"
                                 "init_lstm"
                                                         "init_r"
##
   Г167
        "init_rnn"
                                 "int2bin"
                                                         "loss_L1"
        "predict_gru"
                                 "predict_lstm"
                                                         "predict_rnn"
  [22]
##
                                                         "run.rnn_demo"
   [25]
        "predictr"
                                 "run.finance_demo"
   [28] "trainr"
                                 "update_adagrad"
                                                         "update_r"
  [31] "update_sgd"
```

The rnn has one dependency, the sigmoid package (Quast 2016), which is on CRAN at https://cran.r-project.org/package=sigmoid. The sigmoid package provides a collection of sigmoid functions such as the Rectified Linear Unit (ReLU()), Gompertz(), etc. Until version 0.8.0 of the rnn package, the sigmoid functions were included in the package, after which they were released as a separate package for more general use.

In addition to this, the rnn package includes a Shiny app demonstrating a Recurrent Neural Network analysis of a time series (Foreign Exchange rates). In order to run the app locally, the Shiny package needs to be installed.

2 trainr

The workhorse of the rnn package is the trainr() function, it trains a model based on input and output data, given the specified hyperparameters.

The documentation of the trainr() function can be called up using:

```
help('trainr')
```

Recurrent Neural Networks have the ability to learn bit-by-bit binary addition (including carrying over) with as little as 3 hidden nodes, whereas feed-forward neural networks would need many more.

First training data is generated, the training data is between 0-127, or an 8-bit binary.

```
set.seed(123) # for reproducible random numbers
X1 = sample(0:127, 7000, replace=TRUE)
X2 = sample(0:127, 7000, replace=TRUE)
```

The training data is used to generate the output data or labels.

```
Y <- X1 + X2
```

Following this, both the input data and the output data are converted into binary format, using the built-in int2bin() function.

```
X1 <- int2bin(X1, length=8)
X2 <- int2bin(X2, length=8)
Y <- int2bin(Y, length=8)</pre>
```

Finally, the two input variables are stored in a single 3 dimensional tensor. Where the first dimension contains observations, the second dimension time, and the third dimension variables.

```
X <- array( c(X1,X2), dim=c(dim(X1),2) )</pre>
```

The objects X and Y can now be fed to the trainr() function, which will output a trained model (stored here in the object called m1).

3 predictr

Using a trained model to make predictions is done using the predictr() function.

Take the first two observation of the training data, the first input variable.

```
X[1:2,,1]
## [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,] 0 0 1 0 0 1 0 0
## [2,] 0 0 1 0 0 1 1 0
```

Observation 1 & 2; variable 1 in decimal format.

```
bin2int( X[1:2,,1] )
## [1] 36 100
```

Take the first two observation of the training data, the second input variable.

```
X[1:2,,2]
## [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,] 1 1 1 0 1 1 1 0
## [2,] 1 1 0 1 0 0 1 0
```

Observation 1 & 2; variable 2 in decimal format.

```
bin2int( X[1:2,,2] )
## [1] 119 75
```

Summing observation 1 of variable 1: 36, with observation 1 of variable 2: 119, gives 155.

Summing observation 2 of variable 1: 100, with observation 2 of variable 2: 75, gives 175.

Make predictions using the predictr() function.

Compared to the ground truth values.

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References

Quast, Bastiaan (2016). sigmoid: Sigmoid Functions for Machine Learning. R package version 0.3.0. URL: https://cran.r-project.org/package=sigmoid.