

# Package ‘LilRhino’

January 20, 2025

**Type** Package

**Title** For Implementation of Feed Reduction, Learning Examples, NLP and Code Management

**Version** 1.2.2

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**Description** This is for code management functions, NLP tools, a Monty Hall simulator, and for implementing my own variable reduction technique called Feed Reduction. The Feed Reduction technique is not yet published, but is merely a tool for implementing a series of binary neural networks meant for reducing data into N dimensions, where N is the number of possible values of the response variable.

**License** GPL-2

**Encoding** UTF-8

**Suggests** textclean

**Imports** FNN, stringi, beepr, ggplot2, keras, dplyr, readr, parallel, tm, e1071, SnowballC, data.table, fastmatch, neuralnet

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2022-04-27 22:10:14 UTC

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Binary_Network	<i>Binary Decision Neural Network Wrapper</i>
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### Description

Used as a function of Feed\_Reduction, Binary\_Network uses a 3 layer neural network with an adam optimizer, leaky RELU for the first two activation functions, followed by a softmax on the last layer. The loss function is binary\_crossentropy. This is a keras wrapper, and uses tensorflow in the backend.

### Usage

```
Binary_Network(X, Y, X_test, val_split, nodes, epochs, batch_size, verbose = 0)
```

### Arguments

X	Training data.
Y	Training Labels. These must be binary.
X_test	The test Data
val_split	The validation split for keras.
nodes	The number of nodes in the hidden layers.
epochs	The number of epochs for the network
batch_size	The batch size for the network
verbose	Whether or not you want details about the run as its happening. 0 = silent, 1 = progress bar, 2 = one line per epoch.

### Details

This function is a subset for the larger function Feed\_Network. The output is the list containing the training and testing data converted into an approximation of probability space for that binary decision.

### Value

Train	The training data in approximate probability space
Test	The testing data in 'double' approximate probability space

**Author(s)**

Travis Barton

**References**

Check out [http://wbbpredictions.com/wp-content/uploads/2018/12/Redditbot\\_Paper.pdf](http://wbbpredictions.com/wp-content/uploads/2018/12/Redditbot_Paper.pdf) and Keras for details

**See Also**

Feed\_Network

**Examples**

```
## Not run:
if(8 * .Machine$sizeof.pointer == 64){
  #Feed Network Testing
  library(keras)
  library(dplyr)
  install_keras()
  dat <- keras::dataset_mnist()
  X_train = array_reshape(dat$train$x/255, c(nrow(dat$train$x/255), 784))
  y_train = to_categorical(dat$train$y, 10)
  X_test = array_reshape(dat$test$x/255, c(nrow(dat$test$x/255), 784))
  y_test =to_categorical(dat$test$y, 10)

  index_train = which(dat$train$y == 6 | dat$train$y == 5)
  index_train = sample(index_train, length(index_train))
  index_test = which(dat$test$y == 6 | dat$test$y == 5)
  index_test = sample(index_test, length(index_test))

  temp = Binary_Network(X_train[index_train,],
    y_train[index_train,c(7, 6)], X_test[index_test,], .3, 350, 30, 50)
}

## End(Not run)
```

---

Bootstrap\_Data\_Frame *A function for bootstrapping textual data so that all levels have the same number of entries.*

---

**Description**

This function takes a corpus and a set of labels and uses Bootstrap\_Vocab to increase the size of each label until they are all the same length. Stop words are not bootstrapped.

**Usage**

```
Bootstrap_Data_Frame(text, tags, stopwords, min_length = 7, max_length = 15)
```

**Arguments**

text	text is the collection of textual data to bootstrap up.
tags	tags are the collection of tags that will be used to bootstrap. There should be one for every entry in 'text'. They do not have to be unique.
stopwords	stopwords to make sure are not apart of the bootstrapping process. It is advised to eliminate the most common words. See Stop_Word_Maker()
min_length	The shortest length allowable for bootstrapped words
max_length	The longest length allowable for bootstrapped words

**Details**

Most of the bootstrapped words will be nonseneical. The intention of this package is not to create new sentences, but to instead trick your model into thinking it has equal lengthed levels. This method is meant for bag of words style models.

**Value**

A data frame of your original documents along with the bootstrapped ones (column 1) along with their tags (column 2).

**Author(s)**

Travis Barton

**Examples**

```
test_set = c('I like cats', 'I like dogs', 'we love animals', 'I am a vet',
            'US politics bore me', 'I dont like to vote',
            'The rainbow looked nice today dont you think tommy')
test_tags = c('animals', 'animals', 'animals', 'animals',
            'politics', 'politics',
            'misc')
```

```
Bootstrap_Data_Frame(test_set, test_tags, c("I", "we"), min_length = 3, max_length = 8)
```

---

Bootstrap\_Vocab      *An internal function for Bootstrap\_Data\_Frame.*

---

**Description**

This function takes a selection of documents and bootstraps words from said sentences until there are N total sentences (both sudo and original).

**Usage**

```
Bootstrap_Vocab(vocab, N, stopwds, min_length = 7, max_length = 15)
```

**Arguments**

vocab	The collection of documents to bootstrap.
N	The total amount of sentences to end up with
stopwds	A list of stopwords to not include in the bootstrapping process
min_length	The shortest allowable bootstrapped document
max_length	The longest allowable bootstrapped document

**Details**

The min and max length arguments do not guarantee that a sentence will reach that length. These sentences will be nonsensical.

**Value**

A vector of bootstrapped sentences.

**Author(s)**

Travis Barton

**Examples**

```
testing_set = c(paste('this is test', as.character(seq(1, 10, 1))))  
Bootstrap_Vocab(testing_set, 20, c('this'))
```

---

Codes_done	<i>For announcing when code is done.</i>
------------	--

---

**Description**

for alerting you when your code is done.

**Usage**

```
Codes_done(title, msg, sound = FALSE, effect = 1)
```

**Arguments**

title	The title of the notification
msg	The message to be sent
sound	Optional sound to blurt as well
effect	If sound it blurted, what should it be? (check beep package for sound options)

**Details**

Only for Linux (as far as I know)

**Author(s)**

smacdonald (stack overflow) with modifcaion by Travis Barton

**References**

<https://stackoverflow.com/questions/3365657/is-there-a-way-to-make-r-beep-play-a-sound-at-the-end-of-a-script>

**Examples**

```
Codes_done("done", "check it", sound = TRUE, effect = 1)
```

---

Cross_val_maker	<i>For Creating a test and train set from a whole set</i>
-----------------	---

---

**Description**

for making one dataset into two (test and train)

**Usage**

```
Cross_val_maker(data, alpha)
```

**Arguments**

data	matrix of data you want to split
alpha	the percent of data to split

**Value**

returns a list with accessible with the '\$' sign. Test and Train are labeled as such.

**Author(s)**

Travis Barton

**Examples**

```
dat <- Cross_val_maker(iris, .1)
train <- dat$Train
test <- dat$Test
```

---

Feed_Reduction	<i>A Function for converting data into approximations of probability space.</i>
----------------	---

---

**Description**

It takes the number of unique labels in the training data and tries to predict a one vs all binary neural network for each unique label. The output is an approximation of the probability that each individual input does not match the label. Travis Barton (2018) [http://wbbpredictions.com/wp-content/uploads/2018/12/Redditbot\\_Paper.pdf](http://wbbpredictions.com/wp-content/uploads/2018/12/Redditbot_Paper.pdf)

**Usage**

```
Feed_Reduction(X, Y, X_test, val_split = .1,
              nodes = NULL, epochs = 15,
              batch_size = 30, verbose = 0)
```

**Arguments**

X	Training data
Y	Training labels
X_test	Testing data
val_split	The validation split for the keras, binary, neural networks
nodes	The number nodes for the hidden layers, default is 1/4 of the length of the training data.
epochs	The number of epochs for the fitting of the networks
batch_size	The batch size for the networks
verbose	Weither or not you want details about the run as its happening. 0 = silent, 1 = progress bar, 2 = one line per epoch.

**Details**

This is a new technique for dimensionality reduction of my own creation. Data is converted to the same number of dimensions as there are unique labels. Each dimension is an approximation of the probability that the data point is inside the a unique label. The return value is a list the training and test data with their dimensionality reduced.

**Value**

Train	The training data in the new probability space
Test	The testing data in the new probability space

**Author(s)**

Travis Barton.

**References**

Check out [http://wbbpredictions.com/wp-content/uploads/2018/12/Redditbot\\_Paper.pdf](http://wbbpredictions.com/wp-content/uploads/2018/12/Redditbot_Paper.pdf) for details on the process

**See Also**

Binary\_Network

**Examples**

```
## Not run:
if(8 * .Machine$sizeof.pointer == 64){
#Feed Network Testing
library(keras)

install_keras()
dat <- keras::dataset_mnist()
X_train = array_reshape(dat$train$/255, c(nrow(dat$train$/255), 784))
```



```

y_train = dat$train$y
X_test = array_reshape(dat$test$x/255, c(nrow(dat$test$x/255), 784))
y_test = dat$test$y

Reduced_Data2 = Feed_Reduction(X_train, y_train, X_test,
                              val_split = .3, nodes = 350,
                              30, 50, verbose = 1)

library(e1071)
names(Reduced_Data2$test) = names(Reduced_Data2$train)
newdat = as.data.frame(cbind(rbind(Reduced_Data2$train, Reduced_Data2$test), c(y_train, y_test)))
colnames(newdat) = c(paste("V", c(1:11), sep = ""))
mod = svm(V11~., data = newdat, subset = c(1:60000),
          kernel = 'linear', cost = 1, type = 'C-classification')
preds = predict(mod, newdat[60001:70000,-11])
sum(preds == y_test)/10000

}

## End(Not run)

```

---

Load\_Glove\_Embeddings *Function for loading in pre-trained or personal word embedding softwares.*

---

## Description

Loads in GloVe's pretrained 42 billion token embeddings, trained on the common crawl.

## Usage

```
Load_Glove_Embeddings(path = 'glove.42B.300d.txt', d = 300)
```

## Arguments

path	The path to the embeddings file.
d	The dimension of the embeddings file.

## Details

The embeddings file should be the word, followed by numeric values, ending with a carriage return.

## Value

The embeddings matrix.

## Author(s)

Travis Barton

**Examples**

```
#This code only works if you have the 5g file found here: <https://nlp.stanford.edu/projects/glove/>

## Not run: emb = Load_Glove_Embeddings()
```

---

Monty\_Hall

*Monty Hall Simulator*

---

**Description**

A simulator for the famous Monty Hall Problem

**Usage**

```
Monty_Hall(Games = 10, Choice = "Stay")
```

**Arguments**

Games	The number of games to run on the simulation
Choice	Wether you would like the simulation to either 'Stay' with the first chosen door, 'Switch' to the other door, or 'Random' where you randomly decide to either stay or switch.

**Details**

This is just a toy example of the famous Monty Hall problem. It returns a ggplot bar chart showing the counts for wins or loses in the simulation.

**Value**

A ggplot graph is produced. There is no return value.

**Author(s)**

Travis Barton

**Examples**

```
Monty_Hall(100, 'Stay')
```

---

Nearest_Centroid	<i>For performing the nearest centroid problem (with modifications) on MNST data specifically (general to come)</i>
------------------	---

---

**Description**

For Chen's homework, I'll change this when I generalize it.

**Usage**

```
Nearest_Centroid(X_train, X_test, Y_train)
```

**Arguments**

X_train	Training data
X_test	data to be tested
Y_train	training labels

**Note**

Based on homework from Guangling Chen's M251 class at SJSU

**Author(s)**

Travis Barton

---

Num_Al_Sep	<i>Number/alpha numeric separator for strings.</i>
------------	--

---

**Description**

A Function for the separating of numbers from letters. 'b4' for example would be converted to 'b4'.

**Usage**

```
Num_Al_Sep(vec)
```

**Arguments**

vec	The string vector in which you wish to separate the numbers from the letters.
-----	---

**Value**

output	The separated vector.
--------	-----------------------

**Note**

This is a really simple function really used inside other functions.

**Author(s)**

Travis Barton

**Examples**

```
test_vec = 'The most iconic American weapon has to be the AR15'  
res = Num_Al_Sep(test_vec)  
print(res)
```

---

Percent

*Percent of confusion matrix*

---

**Description**

For finding the accuracy of confusion matrices with true/pred values

**Usage**

```
Percent(true, test)
```

**Arguments**

true	The true values
test	the test values

**Details**

Make sure your strings have the right values and create a square matrix.

**Value**

the percent acc.

**Author(s)**

Travis Barton

**Examples**

```
true <- rep(1:10, 10)  
test <- rep(1:10, 10)  
test[c(2, 22, 33, 89)] = 1  
Percent(true, test)  
#or  
#percent(table(true, test))
```

## Description

This function goes through a number of pretreatment steps in preparation for vectorization. These steps are designed to help the data become more standard so that there are fewer outliers when training during NLP. The following effects are applied: 1. Non-alpha/numerics are removed. 2. Numbers are separated from letters. 3. Numbers are replaced with their word equivalents. 4. Words are stemmed (optional). 5. Words are lowercased (optional).

## Usage

```
Pretreatment(title_vec, stem = TRUE, lower = TRUE, parallel = FALSE)
```

## Arguments

<code>title_vec</code>	Vector of documents to be pre-treated.
<code>stem</code>	Boolean variable to decide whether to stem or not.
<code>lower</code>	Boolean variable to decide whether to lowercase words or not.
<code>parallel</code>	Boolean variable to decide whether to run this function in parallel or not.

## Details

This function returns a list. It should be able to accept any format that the function lapply would accept. The parallelization is done with the function `Mcapply` from the package 'parallel' and will only work on systems that allow forking (Sorry windows users). Future updates will allow for socketing.

## Value

<code>output</code>	The list of character strings post-pretreatment
---------------------	---

## Author(s)

Travis Barton

## Examples

```
## Not run: # for some reason it takes longer than 5 seconds on CRAN's computers
test_vec = c('This is a test', 'Ahoy!', 'my battle-ship is on... b6!')
res = Pretreatment(test_vec)
print(res)

## End(Not run)
```

**Description**

Creates a random forest style collection of neural networks for classification

**Usage**

```
Random_Brains(data, y, x_test,
              variables = ceiling(ncol(data)/10),
              brains = floor(sqrt(ncol(data))),
              hiddens = c(3, 4))
```

**Arguments**

data	The data that holds the predictors ONLY.
y	The response variable
x_test	The testing predictors
variables	The number of predictors to select for each brain in 'data'. The default is one tenth of the number of columns in 'data'.
brains	The number of neural networks to create. The default is the square root of the number of columns in 'data'.
hiddens	This is a vector with length equal to the desired number of hidden layers. Each entry in the vector corresponds to the number of nodes in that layer. The default is c(3, 4) which is a two layer network with 3 and 4 nodes in the layers respectively.

**Details**

This function is meant to mirror the classic random forest function exactly. The only difference being that it uses shallow neural networks to build the forest instead of decision trees.

**Value**

predictions	The predictions for x_test.
num_brains	The number of neural networks used to decide the predictions.
predictors_per_brain	The number of variables used for the neural networks used to decide the predictions.
hidden_layers	The vector describing the number of layers, as well as how many there were.
preds_per_brain	This matrix describes which columns were selected by each brain. Each row is a new brain. Each column describes the index of the column used.
raw_results	The matrix of raw predictions from the brains. Each row is the cumulative predictions of all the brains. Which prediction won by majority vote can be seen in 'predictions'

**Note**

The neural networks are created using the neuralnet package!

**Author(s)**

Travis Barton

**Examples**

```
dat = Cross_val_maker(iris, .2)

train = dat$Train
test = dat$Test

Final_Test = Random_Brains(train[,-5],
  train$Species, as.matrix(test[,-5]),
  variables = 3, brains = 2)
table(Final_Test$predictions, as.numeric(test$Species))
```

---

Sentence_Vector	<i>Function for extracting the sentence vector from an embeddings matrix.</i>
-----------------	---

---

**Description**

Function for extracting the sentence vector from an embeddings matrix in a fast and convenient manner.

**Usage**

```
Sentence_Vector(Sentence, emb_matrix, dimension, stopwords)
```

**Arguments**

Sentence	The sentence to find the vector of.
emb_matrix	The embeddings matrix to search.
dimension	The dimension of the vector to return.
stopwords	Words that should not be included in the averaging process.

**Details**

The function splits the sentence into words, eliminates all stopwords, finds the vectors of each word, then averages the word vectors into a sentence vector.

**Value**

The sentence vector from an embeddings matrix.

**Author(s)**

Travis Barton

**Examples**

```
emb = data.frame(matrix(c(1, 2, 3, 4, 5, 5,
4, 3, 2, 1, 1, 5, 3, 2, 4), nrow = 3),
row.names = c('sentence', 'in', 'question'))
```

```
Sentence_Vector(c('this is the sentence in question'), emb, 5, c('this', 'is', 'the'))
```

---

Stopword\_Maker

*For the finding of the  $N$  most populous words in a corpus.*

---

**Description**

This function finds the  $N$  most used words in a corpus. This is done to identify stop words to better prune data sets before training.

**Usage**

```
Stopword_Maker(titles, cutoff = 20)
```

**Arguments**

titles	The documents in which the most populous words are sought.
cutoff	The number of $N$ top most used words to keep as stop words.

**Value**

output	A vector of the $N$ most populous words.
--------	--

**Author(s)**

Travis Barton

**Examples**

```
test_set = c('this is a testset', 'I am searching for a list of words',
'I like turtles',
'A rocket would be a fast way of getting to work, but I do not think it is very practical')
res = Stopword_Maker(test_set, 4)
print(res)
```



---

Table_percent	<i>Table Percent</i>
---------------	----------------------

---

**Description**

Finds the acc of square tables.

**Usage**

```
Table_percent(in_table)
```

**Arguments**

in\_table      a confusion matrix

**Details**

The table must be square

**Note**

make sure its square.

**Author(s)**

Travis Barton

**Examples**

```
true <- rep(1:10, 10)
test <- rep(1:10, 10)
test[c(2, 22, 33, 89)] = 1
Table_percent(table(true, test))
```

---

Vector_Puller	<i>Function for extracting word vectors from embeddings.</i>
---------------	--

---

**Description**

Function for extracting word vectors from embeddings. This function is an internal function for 'Sentence\_Puller'. It averages the word vectors and returns the average of these vectors.

**Usage**

```
Vector_Puller(words, emb_matrix, dimension)
```

**Arguments**

words	The word to be extracted.
emb_matrix	The embeddings matrix. It must be a data frame.
dimension	The Dimension of the embeddings to extract. They do not have to match that of the matrix, but they cannot exceed its maximum column count.

**Details**

This is a simple and fast internal function.

**Value**

The vector that corresponds to the average of the word vectors.

**Author(s)**

Travis Barton

**Examples**

```
# This is an example emb_matrix  
  
emb = data.frame(matrix(c(1, 2, 3, 4, 5, 5, 4, 3, 2, 1), nrow = 2), row.names = c('cow', 'moo'))  
  
Vector_Puller(c('cow', 'moo'), emb, 5)
```

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