

# Package ‘nnetpredint’

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**Title** Prediction Intervals of Multi-Layer Neural Networks

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**Imports** stats, RSNNS

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**Description** Computing prediction intervals of neural network models (e.g.backpropagation) at certain confidence level. It can take the output from models trained by other packages like 'nnet', 'neuralnet', 'RSNNS', etc.

**License** GPL (>= 2)

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activate                      *Neuron Activation Function*

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### Description

activation function for each neuron node, we provide the popular activation functions including 'sigmoid', 'tanh', etc.

### Usage

```
activate(x, funName)
```

### Arguments

x	input value to the activation function
funName	This package provides the most popular activation functions which can be used by setting the funName parameter to the following strings: 'sigmoid', 'tanh'. 'sigmoid' sigmoid function $1/(1 + \exp(-x))$ . 'tanh' tanh is the hyperbolic tangent function equal to $(\exp(x) - \exp(-x)) / (\exp(x) + \exp(-x))$ .

### Examples

```
x <- c(-5:5)
y1 <- activate(x, funName = 'sigmoid')
y2 <- activate(x, funName = 'tanh')
```

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jacobian                      *Jacobian Matrix of Gradient Function for Training Datasets*

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### Description

Calculate the Jacobian matrix of gradient function for the training dataset. It takes input from neural network models and the gradient at each weight parameters. The matrix has dimension of  $R [nObs * nPara]$ , nObs denotes the number of training observations and nPara denotes the number of weights parameters.

### Usage

```
jacobian(object, ...)

## S3 method for class 'nnet'
jacobian(object, xTrain, funName = 'sigmoid',...)
## S3 method for class 'nn'
jacobian(object, xTrain, funName = 'sigmoid',...)
## S3 method for class 'rsnns'
jacobian(object, xTrain, funName = 'sigmoid',...)
```

**Arguments**

object	object of class: nnet as returned by 'nnet' package, nn as returned by 'neuralnet' package, rsnnns as returned by 'RSNNS' package.
xTrain	matrix or data frame of input values for the training dataset.
funName	activation function name of neuron, e.g. 'sigmoid', 'tanh', etc. In default, it is set to 'sigmoid'.
...	additional arguments passed to the method.

**Details**

Jacobian matrix with gradient function, in which  $J[ij]$  element denotes the gradient function at the  $j$ th weight parameters for the  $i$ th training observation. The dimension is equal to  $nObs * nPara$ .

**Value**

matrix which denotes the Jacobian matrix for training datasets.

**Author(s)**

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**See Also**

[nnetPredInt](#)

**Examples**

```
library(nnet)
xTrain <- rbind(cbind(runif(150,min = 0, max = 0.5),runif(150,min = 0, max = 0.5)) ,
cbind(runif(150,min = 0.5, max = 1),runif(150,min = 0.5, max = 1))
)
nObs <- dim(xTrain)[1]
yTrain <- 0.5 + 0.4 * sin(2* pi * xTrain %*% c(0.4,0.6)) +rnorm(nObs,mean = 0, sd = 0.05)
# Training nnet models
net <- nnet(yTrain ~ xTrain,size = 3, rang = 0.1,decay = 5e-4, maxit = 500)

# Calculating Jacobian Matrix of the training samples
library(nnetpredint)
jacobMat = jacobian(net,xTrain)
dim(jacobMat)
```

**Description**

Get the prediction intervals of new dataset at certain confidence level based on the training datasets and the gradient at weight parameters of the neural network model.

**Usage**

```
nnetPredInt(object, ...)

## Default S3 method:
nnetPredInt(object = NULL, xTrain, yTrain, yFit, node, wts, newData,
            alpha = 0.05, lambda = 0.5, funName = 'sigmoid', ...)
## S3 method for class 'nnet'
nnetPredInt(object, xTrain, yTrain, newData, alpha = 0.05, lambda = 0.5,
            funName = 'sigmoid', ...)
## S3 method for class 'nn'
nnetPredInt(object, xTrain, yTrain, newData, alpha = 0.05, lambda = 0.5,
            funName = 'sigmoid', ...)
## S3 method for class 'rsnns'
nnetPredInt(object, xTrain, yTrain, newData, alpha = 0.05, lambda = 0.5,
            funName = 'sigmoid', ...)
```

**Arguments**

object	object of class: nnet as returned by 'nnet' package, nn as returned by 'neuralnet' package, rsnns as returned by 'RSNNS' package. Object set as NULL will use the default method which takes the weight parameters as the input from user.
xTrain	matrix or data frame of input values for the training dataset.
yTrain	vector of target values for the training dataset.
newData	matrix or data frame of the prediction dataset.
yFit	vector of the fitted values, as the output produced by the training model, e.g. nnet\$fitted.values ('nnet'), nn\$net.result[[1]] ('neuralnet') and rsnns\$fitted.values ('RSNNS')
node	a vector of integers specifying the number of hidden nodes in each layer. Multi-layer network has the structure (s0, s1, ..., sm), in which s0 denotes the dimension for input layer and sm denotes the dimension of the output layer. sm is usually set as 1.
wts	a numeric vector of optimal weight parameters as the output of the neural network training model. The order of wts parameter is as follows: For any node i in layer k: c(bias ik, wi1k, wi2k, ..., wijk).  nnet object, returned by 'nnet' package. We can directly set the wts as: wts = nnet\$wts

nn object, returned by 'neuralnet' package. We need to use [transWeightListToVect](#) function to transform the list of weights to a single vector first: `wts = transWeightListToVect(wtsList, m)`.

rsnns object, returned by 'RSNNS' package. We need to transform and combine the weight and bias parameters to a single vector: `weightMatrix(object)` and `extractNetInfo(object)$unitDefinitions$unitBias`.

alpha	confidence level. The confidence level is set to (1-alpha). In default, alpha = 0.05.
lambda	decay parameter of weights when the Jacobian matrix of training dataset is singular. In default, lamda is set to 0.5 .
funName	activation function name of neuron, e.g. 'sigmoid', 'tanh', etc. In default, it is set to 'sigmoid'.
...	additional arguments passed to the method.

### Value

data frame of the prediction intervals, including prediction value, lower and upper bounds of the interval.

yPredValue	the column of prediction value in the data frame.
lowerBound	the column of prediction lower bounds in the data frame.
upperBound	the column of prediction upper bounds in the data frame.

### Author(s)

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### References

De Veaux R. D., Schumi J., Schweinsberg J., Ungar L. H., 1998, "Prediction intervals for neural networks via nonlinear regression", *Technometrics* 40(4): 273-282.

Chryssolouris G., Lee M., Ramsey A., "Confidence interval prediction for neural networks models", *IEEE Trans. Neural Networks*, 7 (1), 1996, pp. 229-232.

'neuralnet' package by Stefan Fritsch, Frauke Guenther.

'nnet' package by Brian Ripley, William Venables.

'RSNNS' package by Christoph Bergmeir, Jose M. Benitez.

### See Also

[transWeightListToVect](#) [jacobian](#)

## Examples

```

# Example 1: Using the nn object trained by neuralnet package
set.seed(500)
library(MASS)
data <- Boston
maxs <- apply(data, 2, max)
mins <- apply(data, 2, min)
scaled <- as.data.frame(scale(data, center = mins, scale = maxs - mins)) # normalization
index <- sample(1:nrow(data),round(0.75*nrow(data)))
train_ <- scaled[index,]
test_ <- scaled[-index,]

library(neuralnet) # Training
n <- names(train_)
f <- as.formula(paste("medv ~", paste(n[!n %in% "medv"], collapse = " + ")))
nn <- neuralnet(f,data = train_,hidden = c(5,3),linear.output = FALSE)
plot(nn)

library(nnetpredint) # Getting prediction confidence interval
x <- train_[,-14]
y <- train_[,14]
newData <- test_[,-14]

# S3 generic method: Object of nn
yPredInt <- nnetPredInt(nn, x, y, newData)
print(yPredInt[1:20,])

# S3 default method for user defined weights input, without model object trained:
yFit <- c(nn$net.result[[1]])
nodeNum <- c(13,5,3,1)
m <- 3
wtsList <- nn$weights[[1]]
wts <- transWeightListToVect(wtsList,m)
yPredInt2 <- nnetPredInt(object = NULL, x, y, yFit, nodeNum, wts, newData, alpha = 0.05)
print(yPredInt2[1:20,])

# Compare to the predict values from the neuralnet Compute method
predValue <- compute(nn,newData)
print(matrix(predValue$net.result[1:20]))

# Example 2: Using the nnet object trained by nnet package
library(nnet)
xTrain <- rbind(cbind(runif(150,min = 0, max = 0.5),runif(150,min = 0, max = 0.5)) ,
cbind(runif(150,min = 0.5, max = 1),runif(150,min = 0.5, max = 1))
)
nObs <- dim(xTrain)[1]
yTrain <- 0.5 + 0.4 * sin(2* pi * xTrain %>% c(0.4,0.6)) + rnorm(nObs,mean = 0, sd = 0.05)
plot(xTrain %>% c(0.4,0.6),yTrain)

# Training nnet models

```

```

net <- nnet(yTrain ~ xTrain,size = 3, rang = 0.1,decay = 5e-4, maxit = 500)
yFit <- c(net$fitted.values)
nodeNum <- c(2,3,1)
wts <- net$wts

# New data for prediction intervals
library(nnetpredint)
newData <- cbind(seq(0,1,0.05),seq(0,1,0.05))
yTest <- 0.5 + 0.4 * sin(2* pi * newData %*% c(0.4,0.6))+rnorm(dim(newData)[1],
  mean = 0, sd = 0.05)

# S3 generic method: Object of nnet
yPredInt <- nnetPredInt(net, xTrain, yTrain, newData)
print(yPredInt[1:20,])

# S3 default method: xTrain,yTrain,yFit,...
yPredInt2 <- nnetPredInt(object = NULL, xTrain, yTrain, yFit, node = nodeNum, wts = wts,
  newData, alpha = 0.05, funName = 'sigmoid')

plot(newData %*% c(0.4,0.6),yTest,type = 'b')
lines(newData %*% c(0.4,0.6),yPredInt$yPredValue,type = 'b',col='blue')
lines(newData %*% c(0.4,0.6),yPredInt$lowerBound,type = 'b',col='red') # lower bound
lines(newData %*% c(0.4,0.6),yPredInt$upperBound,type = 'b',col='red') # upper bound

# Example 3: Using the rsnn object trained by RSNNS package
library(RSNNS)
data(iris)
#shuffle the vector
iris <- iris[sample(1:nrow(iris),length(1:nrow(iris))),1:ncol(iris)]
irisValues <- iris[,1:4]
irisTargets <- decodeClassLabels(iris[,5]),['setosa']

iris <- splitForTrainingAndTest(irisValues, irisTargets, ratio=0.15)
iris <- normTrainingAndTestSet(iris)
model <- mlp(iris$inputsTrain, iris$targetsTrain, size=5, learnFuncParams=c(0.1),
maxit=50, inputsTest=iris$inputsTest, targetsTest=iris$targetsTest)
predictions <- predict(model,iris$inputsTest)

# Generating prediction intervals
library(nnetpredint)
# S3 Method for rsnn class prediction intervals
xTrain <- iris$inputsTrain
yTrain <- iris$targetsTrain
newData <- iris$inputsTest
yPredInt <- nnetPredInt(model, xTrain, yTrain, newData)
print(yPredInt[1:20,])

```

**Description**

This function transforms the list of weight parameters, typically the output of neuralnet package (`nn$weights`) to a single numeric vector in specific order, which will be used as the input to the prediction interval function. The order of weight parameters is the same as the output of nnet package (`nnet$wts`).

**Usage**

```
transWeightListToVect(wtsList, m = 2)
```

**Arguments**

<code>wtsList</code>	the list of weights found by neuralnet package ( <code>nn\$weights</code> ). For nnet weights output ( <code>nnet\$wts</code> ), there is no need for transformation and it can be used directly by the prediction interval method.
<code>m</code>	the number of layers of the neural networks, which is the number of hidden layer plus + 1. <code>m</code> is default to 2 for single hidden layer networks.

**Value**

the numeric vector of optimal weights found by neural networks, dimension  $1 * nPara$ .

**Author(s)**

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**References**

neuralnet package by Stefan Fritsch, Frauke Guenther

**See Also**

[nnetPredInt](#)

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