

Package ‘darch’

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Type Package

Title Package for Deep Architectures and Restricted Boltzmann Machines

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Description The darch package is built on the basis of the code from G. E. Hinton and R. R. Salakhutdinov (available under Matlab Code for deep belief nets). This package is for generating neural networks with many layers (deep architectures) and train them with the method introduced by the publications “A fast learning algorithm for deep belief nets” (G. E. Hinton, S. Osindero, Y. W. Teh (2006) <DOI:10.1162/neco.2006.18.7.1527>) and “Reducing the dimensionality of data with neural networks” (G. E. Hinton, R. R. Salakhutdinov (2006) <DOI:10.1126/science.1127647>). This method includes a pre training with the contrastive divergence method published by G.E Hinton (2002) <DOI:10.1162/089976602760128018> and a fine tuning with common known training algorithms like backpropagation or conjugate gradients. Additionally, supervised fine-tuning can be enhanced with maxout and dropout, two recently developed techniques to improve fine-tuning for deep learning.

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URL <https://github.com/maddin79/darch>

BugReports <https://github.com/maddin79/darch/issues>

Depends R (>= 3.0.0)

Imports stats, methods, ggplot2, reshape2, futile.logger (>= 1.4.1),
caret, Rcpp (>= 0.12.3)

LinkingTo Rcpp

Suggests foreach, doRNG, NeuralNetTools, gputools, testthat, plyr (>= 1.8.3.9000)

Collate 'ReppExports.R' 'autosave.R' 'net.Class.R' 'darch.Class.R' 'backpropagation.R' 'benchmark.R' 'bootstrap.R' 'caret.R' 'compat.R' 'config.R' 'darch.Add.R' 'darch.Getter.R' 'dataset.R' 'darch.Learn.R' 'darch.R' 'darch.Setter.R' 'darchUnitFunctions.R' 'dropout.R' 'errorFunctions.R' 'rbm.Class.R' 'generateRBMs.R' 'generateWeightsFunctions.R' 'loadDArch.R' 'log.R' 'makeStartEndPoints.R' 'minimize.R' 'minimizeAutoencoder.R' 'minimizeClassifier.R' 'mnist.R' 'momentum.R' 'net.Getter.R' 'newDArch.R' 'params.R' 'plot.R' 'predict.R' 'print.R' 'rbm.Learn.R' 'rbm.Reset.R' 'rbmUnitFunctions.R' 'rbmUpdate.R' 'rpropagation.R' 'runDArch.R' 'saveDArch.R' 'test.R' 'util.R' 'weightUpdateFunctions.R'

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backpropagation	<i>Backpropagation learning function</i>
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Description

This function provides the backpropagation algorithm for deep architectures.

Usage

```
backpropagation(darch, trainData, targetData,
  bp.learnRate = getParameter(".bp.learnRate", rep(1, times =
    length(darch@layers))),
  bp.learnRateScale = getParameter(".bp.learnRateScale"),
  nesterovMomentum = getParameter(".darch.nesterovMomentum"),
  dropout = getParameter(".darch.dropout", rep(0, times = length(darch@layers)
    + 1), darch), dropConnect = getParameter(".darch.dropout.dropConnect"),
  matMult = getParameter(".matMult"), debugMode = getParameter(".debug", F),
  ...)
```

Arguments

darch	An instance of the class DArch .
trainData	The training data (inputs).
targetData	The target data (outputs).
bp.learnRate	Learning rates for backpropagation, length is either one or the same as the number of weight matrices when using different learning rates for each layer.
bp.learnRateScale	The learn rate is multiplied by this value after each epoch.
nesterovMomentum	See darch.nesterovMomentum parameter of darch .
dropout	See darch.dropout parameter of darch .
dropConnect	See darch.dropout.dropConnect parameter of darch .
matMult	Matrix multiplication function, internal parameter.
debugMode	Whether debug mode is enabled, internal parameter.
...	Further parameters.

Details

The only backpropagation-specific, user-relevant parameters are `bp.learnRate` and `bp.learnRateScale`; they can be passed to the `darch` function when enabling backpropagation as the fine-tuning function. `bp.learnRate` defines the backpropagation learning rate and can either be specified as a single scalar or as a vector with one entry for each weight matrix, allowing for per-layer learning rates. `bp.learnRateScale` is a single scalar which contains a scaling factor for the learning rate(s) which will be applied after each epoch.

Backpropagation supports dropout and uses the weight update function as defined via the `darch.weightUpdateFunction` parameter of `darch`.

Value

The trained deep architecture

References

Rumelhart, D., G. E. Hinton, R. J. Williams, Learning representations by backpropagating errors, Nature 323, S. 533-536, DOI: 10.1038/323533a0, 1986.

See Also

[darch](#)

Other fine-tuning functions: [minimizeAutoencoder](#), [minimizeClassifier](#), [rpropagation](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, darch.fineTuneFunction = "backpropagation")

## End(Not run)
```

crossEntropyError *Cross entropy error function*

Description

The function calculates the cross entropy error from the original and estimate parameters.

Usage

```
crossEntropyError(original, estimate)
```

Arguments

<code>original</code>	The original data matrix.
<code>estimate</code>	The calculated data matrix.

Details

This function can be used for the `darch.errorFunction` parameter of the `darch` function, but is only a valid error function if used in combination with the `softmaxUnit` activation function! It is not a valid value for the parameter `rbm.errorFunction`.

Value

A list with the name of the error function in the first entry and the error value in the second entry.

References

Rubinstein, R.Y., Kroese, D.P. (2004). The Cross-Entropy Method: A Unified Approach to Combinatorial Optimization, Monte-Carlo Simulation, and Machine Learning, Springer-Verlag, New York.

See Also

Other error functions: `mseError`, `rmseError`

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, darch.errorFunction = "crossEntropyError")

## End(Not run)
```

darch

Fit a deep neural network

Description

Fit a deep neural network with optional pre-training and one of various fine-tuning algorithms.

Usage

```
darch(x, ...)
```

```
## Default S3 method:
darch(x, y, layers = 10, ..., autosave = F,
      autosave.epochs = round(darch.numEpochs/20),
      autosave.dir = "./darch.autosave", autosave.trim = F, bp.learnRate = 1,
      bp.learnRateScale = 1, bootstrap = F, bootstrap.unique = T,
      bootstrap.num = 0, cg.length = 2, cg.switchLayers = 1, darch = NULL,
      darch.batchSize = 1, darch.dither = F, darch.dropout = 0,
      darch.dropout.dropConnect = F, darch.dropout.momentMatching = 0,
      darch.dropout.oneMaskPerEpoch = F, darch.elu.alpha = 1,
      darch.errorFunction = if (darch.isClass) crossEntropyError else mseError,
```

```

darch.finalMomentum = 0.9, darch.fineTuneFunction = backpropagation,
darch.initialMomentum = 0.5, darch.isClass = T,
darch.maxout.poolSize = 2, darch.maxout.unitFunction = linearUnit,
darch.momentumRampLength = 1, darch.nesterovMomentum = T,
darch.numEpochs = 100, darch.returnBestModel = T,
darch.returnBestModel.validationErrorFactor = 1 - exp(-1),
darch.stopClassErr = -Inf, darch.stopErr = -Inf,
darch.stopValidClassErr = -Inf, darch.stopValidErr = -Inf,
darch.trainLayers = T, darch.unitFunction = sigmoidUnit,
darch.weightDecay = 0,
darch.weightUpdateFunction = weightDecayWeightUpdate, dataSet = NULL,
dataSetValid = NULL,
generateWeightsFunction = generateWeightsGlorotUniform, gputools = F,
gputools.deviceId = 0, logLevel = NULL, normalizeWeights = F,
normalizeWeightsBound = 15, paramsList = list(),
preProc.factorToNumeric = F, preProc.factorToNumeric.targets = F,
preProc.fullRank = T, preProc.fullRank.targets = F,
preProc.orderedToFactor.targets = T, preProc.params = F,
preProc.targets = F, rbm.allData = F, rbm.batchSize = 1,
rbm.consecutive = T, rbm.errorFunction = mseError,
rbm.finalMomentum = 0.9, rbm.initialMomentum = 0.5, rbm.lastLayer = 0,
rbm.learnRate = 1, rbm.learnRateScale = 1, rbm.momentumRampLength = 1,
rbm.numCD = 1, rbm.numEpochs = 0, rbm.unitFunction = sigmoidUnitRbm,
rbm.updateFunction = rbmUpdate, rbm.weightDecay = 2e-04, retainData = F,
rprop.decFact = 0.5, rprop.incFact = 1.2, rprop.initDelta = 1/80,
rprop.maxDelta = 50, rprop.method = "iRprop+", rprop.minDelta = 1e-06,
seed = NULL, shuffleTrainData = T, weights.max = 0.1,
weights.mean = 0, weights.min = -0.1, weights.sd = 0.01,
xValid = NULL, yValid = NULL)

## S3 method for class 'formula'
darch(x, data, layers, ..., xValid = NULL, dataSet = NULL,
      dataSetValid = NULL, logLevel = NULL, paramsList = list(),
      darch = NULL)

## S3 method for class 'DataSet'
darch(x, ...)

```

Arguments

x	Input data matrix or data.frame (darch.default) or formula (darch.formula) or DataSet (darch.DataSet).
...	Additional parameters.
y	Target data matrix or data.frame , if x is an input data matrix or data.frame .
layers	Vector containing one integer for the number of neurons of each layer. Defaults to c(a, 10, b), where a is the number of columns in the training data and b the number of columns in the targets. If this has length 1, it is used as the number of neurons in the hidden layer, not as the number of layers!

<code>autosave</code>	Logical indicating whether to activate automatically saving the <code>DArch</code> instance to a file during fine-tuning.
<code>autosave.epochs</code>	After how many epochs should auto-saving happen, by default after every 5 1, the network will only be saved once when thee fine-tuning is done.
<code>autosave.dir</code>	Directory for the autosave files, the file names will be e.g. <code>autosave_010.net</code> for the <code>DArch</code> instance after 10 epochs
<code>autosave.trim</code>	Whether to trim the network before saving it. This will remove the dataset and the layer weights, resulting in a network that is no longer usable for predictions or training. Useful when only statistics and settings need to be stored.
<code>bp.learnRate</code>	Learning rates for backpropagation, length is either one or the same as the number of weight matrices when using different learning rates for each layer.
<code>bp.learnRateScale</code>	The learn rate is multiplied by this value after each epoch.
<code>bootstrap</code>	Logical indicating whether to use bootstrapping to create a training and validation data set from the given training data.
<code>bootstrap.unique</code>	Logical indicating whether to take only unique samples for the training (TRUE, default) or take all drawn samples (FALSE), which will results in a bigger training set with duplicates. Note: This is ignored if <code>bootstrap.num</code> is greater than 0.
<code>bootstrap.num</code>	If this is greater than 0, bootstrapping will draw this number of training samples without replacement.
<code>cg.length</code>	Numbers of line search
<code>cg.switchLayers</code>	Indicates when to train the full network instead of only the upper two layers
<code>darch</code>	Existing <code>DArch</code> instance for which training is to be resumed. Note: When enabling pre-training, previous training results we be lost, see explanation for parameter <code>rbm.numEpochs</code> .
<code>darch.batchSize</code>	Batch size, i.e. the number of training samples that are presented to the network before weight updates are performed, for fine-tuning.
<code>darch.dither</code>	Whether to apply dither to numeric columns in the training input data.
<code>darch.dropout</code>	Dropout rates. If this is a vector it will be treated as the dropout rates for each individual layer. If one element is missing, the input dropout will be set to 0. When enabling <code>darch.dropout.dropConnect</code> , this vector needs an additional element (one element per weight matrix between two layers as opposed to one element per layer excluding the last layer).
<code>darch.dropout.dropConnect</code>	Whether to use <code>DropConnect</code> instead of dropout for the hidden layers. Will use <code>darch.dropout</code> as the dropout rates.
<code>darch.dropout.momentMatching</code>	How many iterations to perform during moment matching for dropout inference, 0 to disable moment matching.

- `darch.dropout.oneMaskPerEpoch`
Whether to generate a new mask for each batch (FALSE, default) or for each epoch (TRUE).
- `darch.elu.alpha`
Alpha parameter for the exponential linear unit function. See [exponentialLinearUnit](#).
- `darch.errorFunction`
Error function during fine-tuning. Possible error functions include [mseError](#), [rmseError](#), and [crossEntropyError](#).
- `darch.finalMomentum`
Final momentum during fine-tuning.
- `darch.fineTuneFunction`
Fine-tuning function. Possible values include [backpropagation](#) (default), [rpropagation](#), [minimizeClassifier](#) and [minimizeAutoencoder](#) (unsupervised).
- `darch.initialMomentum`
Initial momentum during fine-tuning.
- `darch.isClass` Whether output should be treated as class labels during fine-tuning and classification rates should be printed.
- `darch.maxout.poolSize`
Pool size for maxout units, when using the maxout activation function. See [maxoutUnit](#).
- `darch.maxout.unitFunction`
Inner unit function used by maxout. See `darch.unitFunction` for possible unit functions.
- `darch.momentumRampLength`
After how many epochs, relative to the **overall** number of epochs trained, should the momentum reach `darch.finalMomentum`? A value of 1 indicates that the `darch.finalMomentum` should be reached in the final epoch, a value of 0.5 indicates that `darch.finalMomentum` should be reached after half of the training is complete. Note that this will lead to bumps in the momentum ramp if training is resumed with the same parameters for `darch.initialMomentum` and `darch.finalMomentum`. Set `darch.momentumRampLength` to 0 to avoid this problem when resuming training.
- `darch.nesterovMomentum`
Whether to use **Nesterov Accelerated Momentum**. (NAG) for gradient descent based fine-tuning algorithms.
- `darch.numEpochs`
Number of epochs of fine-tuning.
- `darch.returnBestModel`
Logical indicating whether to return the best model at the end of training, instead of the last.
- `darch.returnBestModel.validationErrorFactor`
When evaluating models with validation data, how high should the validation error be valued, compared to the training error? This is a value between 0 and 1. By default, this value is $1 - \exp(-1)$. The training error factor and the validation error factor will always add to 1, so if you pass 1 here, the training error will be ignored, and if you pass 0 here, the validation error will be ignored.

<code>darch.stopClassErr</code>	When the classification error is lower than or equal to this value, training is stopped (0..100).
<code>darch.stopErr</code>	When the value of the error function is lower than or equal to this value, training is stopped.
<code>darch.stopValidClassErr</code>	When the classification error on the validation data is lower than or equal to this value, training is stopped (0..100).
<code>darch.stopValidErr</code>	When the value of the error function on the validation data is lower than or equal to this value, training is stopped.
<code>darch.trainLayers</code>	Either TRUE to train all layers or a mask containing TRUE for all layers which should be trained and FALSE for all layers that should not be trained (no entry for the input layer).
<code>darch.unitFunction</code>	Layer function or vector of layer functions of length number of layers - 1. Note that the first entry signifies the layer function between layers 1 and 2, i.e. the output of layer 2. Layer 1 does not have a layer function, since the input values are used directly. Possible unit functions include linearUnit , sigmoidUnit , tanhUnit , rectifiedLinearUnit , softplusUnit , softmaxUnit , and maxoutUnit .
<code>darch.weightDecay</code>	Weight decay factor, defaults to 0. All weights will be multiplied by (1 - <code>darch.weightDecay</code>) prior to each weight update.
<code>darch.weightUpdateFunction</code>	Weight update function or vector of weight update functions, very similar to <code>darch.unitFunction</code> . Possible weight update functions include weightDecayWeightUpdate and maxoutWeightUpdate Note that maxoutWeightUpdate must be used on the layer after the maxout activation function!
<code>dataSet</code>	DataSet instance, passed from <code>darch.DataSet()</code> , may be specified manually.
<code>dataSetValid</code>	DataSet instance containing validation data.
<code>generateWeightsFunction</code>	Weight generation function or vector of layer generation functions of length number of layers - 1. Possible weight generation functions include generateWeightsUniform (default), generateWeightsNormal , generateWeightsGlorotNormal , generateWeightsGlorotUniform , generateWeightsHeNormal , and generateWeightsHeUniform .
<code>gputools</code>	Logical indicating whether to use gputools for matrix multiplication, if available.
<code>gputools.deviceId</code>	Integer specifying the device to use for GPU matrix multiplication. See chooseGpu .
<code>logLevel</code>	futile.logger log level. Uses the currently set log level by default, which is <code>futile.logger::flog.info</code> if it was not changed. Other available levels include, from least to most verbose: FATAL, ERROR, WARN, DEBUG, and TRACE.
<code>normalizeWeights</code>	Logical indicating whether to normalize weights (L2 norm = <code>normalizeWeightsBound</code>).

<code>normalizeWeightsBound</code>	Upper bound on the L2 norm of incoming weight vectors. Used only if <code>normalizeWeights</code> is TRUE.
<code>paramsList</code>	List of parameters, can include and does overwrite specified parameters listed above. Primary for convenience or for use in scripts.
<code>preProc.factorToNumeric</code>	Whether all factors should be converted to numeric.
<code>preProc.factorToNumeric.targets</code>	Whether all factors should be converted to numeric in the target data.
<code>preProc.fullRank</code>	Whether to use full rank encoding. See preProcess for details.
<code>preProc.fullRank.targets</code>	Whether to use full rank encoding for target data. See preProcess for details.
<code>preProc.orderedToFactor.targets</code>	Whether ordered factors in the target data should be converted to unordered factors. Note: Ordered factors are converted to numeric by dummyVars and no longer usable for classification tasks.
<code>preProc.params</code>	List of parameters to pass to the preProcess function for the input data or FALSE to disable input data pre-processing.
<code>preProc.targets</code>	Whether target data is to be centered and scaled. Unlike <code>preProc.params</code> , this is just a logical turning pre-processing for target data on or off, since this pre-processing has to be reverted when predicting new data. Most useful for regression tasks. Note: This will skew the raw network error.
<code>rbm.allData</code>	Logical indicating whether to use training and validation data for pre-training. Note: This also applies when using bootstrapping.
<code>rbm.batchSize</code>	Pre-training batch size.
<code>rbm.consecutive</code>	Logical indicating whether to train the RBMs one at a time for <code>rbm.numEpochs</code> epochs (TRUE, default) or alternately training each RBM for one epoch at a time (FALSE).
<code>rbm.errorFunction</code>	Error function during pre-training. This is only used to estimate the RBM error and does not affect the training itself. Possible error functions include mseError and rmseError .
<code>rbm.finalMomentum</code>	Final momentum during pre-training.
<code>rbm.initialMomentum</code>	Initial momentum during pre-training.
<code>rbm.lastLayer</code>	Numeric indicating at which layer to stop the pre-training. Possible values include 0, meaning that all layers are trained; positive integers, meaning to stop training after the RBM where <code>rbm.lastLayer</code> forms the visible layer; negative integers, meaning to stop the training at <code>rbm.lastLayer</code> RBMs from the top RBM.
<code>rbm.learnRate</code>	Learning rate during pre-training.

<code>rbm.learnRateScale</code>	The learn rates will be multiplied with this value after each epoch.
<code>rbm.momentumRampLength</code>	After how many epochs, relative to <code>rbm.numEpochs</code> , should the momentum reach <code>rbm.finalMomentum</code> ? A value of 1 indicates that the <code>rbm.finalMomentum</code> should be reached in the final epoch, a value of 0.5 indicates that <code>rbm.finalMomentum</code> should be reached after half of the training is complete.
<code>rbm.numCD</code>	Number of full steps for which contrastive divergence is performed. Increasing this will slow training down considerably.
<code>rbm.numEpochs</code>	Number of pre-training epochs. Note: When passing a value other than 0 here and also passing an existing <code>DArch</code> instance via the <code>darch</code> parameter, the weights of the network will be completely reset! Pre-training is essentially a form of advanced weight initialization and it makes no sense to perform pre-training on a previously trained network.
<code>rbm.unitFunction</code>	Unit function during pre-training. Possible functions include <code>sigmoidUnitRbm</code> (default), <code>tanhUnitRbm</code> , and <code>linearUnitRbm</code> .
<code>rbm.updateFunction</code>	Update function during pre-training. Currently, <code>darch</code> only provides <code>rbmUpdate</code> .
<code>rbm.weightDecay</code>	Pre-training weight decay. Weights will be multiplied by $(1 - \text{rbm.weightDecay})$ prior to each weight update.
<code>retainData</code>	Logical indicating whether to store the training data in the <code>DArch</code> instance after training or when saving it to disk.
<code>rprop.decFact</code>	Decreasing factor for the training. Default is 0.6.
<code>rprop.incFact</code>	Increasing factor for the training. Default is 1.2.
<code>rprop.initDelta</code>	Initialisation value for the update. Default is 0.0125.
<code>rprop.maxDelta</code>	Upper bound for step size. Default is 50
<code>rprop.method</code>	The method for the training. Default is "iRprop+"
<code>rprop.minDelta</code>	Lower bound for step size. Default is 0.000001
<code>seed</code>	Allows the specification of a seed which will be set via <code>set.seed</code> . Used in the context of <code>darchBench</code> .
<code>shuffleTrainData</code>	Logical indicating whether to shuffle training data before each epoch.
<code>weights.max</code>	max parameter to the <code>runif</code> function.
<code>weights.mean</code>	mean parameter to the <code>rnorm</code> function.
<code>weights.min</code>	min parameter to the <code>runif</code> function.
<code>weights.sd</code>	sd parameter to the <code>rnorm</code> function.
<code>xValid</code>	Validation input data matrix or <code>data.frame</code> .
<code>yValid</code>	Validation target data matrix or <code>data.frame</code> , if x is a data matrix or <code>data.frame</code> .
<code>data</code>	<code>data.frame</code> containing the dataset, if x is a <code>formula</code> .

Details

The darch package implements Deep Architecture Networks and restricted Boltzmann machines.

The creation of this package is motivated by the papers from G. Hinton et. al. from 2006 (see references for details) and from the MATLAB source code developed in this context. This package provides the possibility to generate deep architecture networks (darch) like the deep belief networks from Hinton et. al.. The deep architectures can then be trained with the contrastive divergence method. After this pre-training it can be fine tuned with several learning methods like backpropagation, resilient backpropagation and conjugate gradients as well as more recent techniques like dropout and maxout.

See <https://github.com/maddin79/darch> for further information, documentation, and releases.

Package:	darch
Type:	Package
Version:	0.10.0
Date:	2015-11-12
License:	GPL-2 or later
LazyLoad:	yes

Value

Fitted `DArch` instance

Author(s)

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References

Hinton, G. E., S. Osindero, Y. W. Teh, A fast learning algorithm for deep belief nets, *Neural Computation* 18(7), S. 1527-1554, DOI: 10.1162/neco.2006.18.7.1527 2006.

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Goodfellow, Ian J. et al. (2013). "Maxout Networks". In: *Proceedings of the 30th International Conference on Machine Learning, ICML 2013, Atlanta, GA, USA, 16-21 June 2013*, pp. 1319-1327. URL: <http://jmlr.org/proceedings/papers/v28/goodfellow13.html>.

Drees, Martin (2013). "Implementierung und Analyse von tiefen Architekturen in R". German. Master's thesis. Fachhochschule Dortmund.

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See Also

Other darch interface functions: `darchBench`, `darchTest`, `plot.DArch`, `predict.DArch`, `print.DArch`

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris)
print(model)
predictions <- predict(model, newdata = iris, type = "class")
cat(paste("Incorrect classifications:", sum(predictions != iris[,5])))

trainData <- matrix(c(0,0,0,1,1,0,1,1), ncol = 2, byrow = TRUE)
trainTargets <- matrix(c(0,1,1,0), nrow = 4)
model2 <- darch(trainData, trainTargets, layers = c(2, 10, 1),
  darch.numEpochs = 500, darch.stopClassErr = 0, retainData = T)
e <- darchTest(model2)
cat(paste0("Incorrect classifications on all examples: ", e[3], " (",
  e[2], "%)\n"))

plot(model2)

## End(Not run)

#
# More examples can be found at
# https://github.com/maddin79/darch/tree/v0.12.0/examples
```

darchBench

Benchmarking wrapper for darch

Description

Simple benchmarking function which wraps around the [darch](#) function for users who can't or don't want to use the caret package for benchmarking. This function requires the `foreach` package to work, and will perform parallel benchmarks if an appropriate backend was registered beforehand.

Usage

```
darchBench(..., bench.times = 1, bench.save = F,
  bench.dir = "./darch.benchmark", bench.continue = T, bench.delete = F,
  bench.seeds = NULL, output.capture = bench.save, logLevel = NULL)
```

Arguments

<code>...</code>	Parameters to the darch function
<code>bench.times</code>	How many benchmark runs to perform
<code>bench.save</code>	Whether to save benchmarking results to a directory
<code>bench.dir</code>	Path (relative or absolute) including directory where benchmark results are saved if <code>bench.save</code> is true

<code>bench.continue</code>	Whether the benchmark is to be continued from an earlier run. If TRUE, existing benchmark results are looked for in the directory given in <code>bench.dir</code> and new results are appended. If both this and <code>bench.continue</code> are FALSE and the directory given in <code>bench.dir</code> does already exist, the training will be aborted with an error.
<code>bench.delete</code>	Whether to delete the contents of <code>bench.dir</code> if <code>bench.continue</code> is FALSE. Caution: This will attempt to delete ALL files in the given directory, use at your own risk!
<code>bench.seeds</code>	Vector of seeds, one for each run. Will be passed to darch .
<code>output.capture</code>	Whether to capture R output in <code>.Rout</code> files in the given directory. This is the only way of gaining access to the R output since the <code>foreach</code> loop will not print anything to the console. Will be ignored if <code>bench.save</code> is FALSE.
<code>logLevel</code>	futile.logger log level. Uses the currently set log level by default, which is <code>futile.logger::flog.info</code> if it was not changed. Other available levels include, from least to most verbose: FATAL, ERROR, WARN, DEBUG, and TRACE.

Value

List of DArch instances; the results of each call to `darch`.

See Also

Other `darch` interface functions: [darchTest](#), [darch](#), [plot.DArch](#), [predict.DArch](#), [print.DArch](#)

Examples

```
## Not run:
data(iris)
modelList <- darchBench(Species ~ ., iris, c(0, 50, 0),
  preProc.params = list(method = c("center", "scale")),
  darch.unitFunction = c("sigmoidUnit", "softmaxUnit"),
  darch.numEpochs = 30, bench.times = 10, bench.save = T)

## End(Not run)
```

<code>darchModelInfo</code>	<i>Creates a custom caret model for darch.</i>
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Description

This function creates a caret model description to enable training DArch instances with the [train](#) function. See the documentation on custom caret models for further information and examples on how to create valid params and grid values.

Usage

```
darchModelInfo(params = NULL, grid = NULL)
```

Arguments

params	data.frame of parameters or NULL to use a simple default (bp.learnRate).
grid	Function which produces a data.frame containing a grid of parameter combinations or NULL to use a simple default.

Value

A valid caret model which can be passed to [train](#).

See Also

[Caret custom models](#)

Examples

```
## Not run:
data(iris)
tc <- trainControl(method = "boot", number = 5, allowParallel = F,
  verboseIter = T)

parameters <- data.frame(parameter = c("layers", "bp.learnRate", "darch.unitFunction"),
  class = c("character", "numeric", "character"),
  label = c("Network structure", "Learning rate", "unitFunction"))

grid <- function(x, y, len = NULL, search = "grid")
{
  df <- expand.grid(layers = c("c(0,20,0)", "c(0,10,10,0)", "c(0,10,5,5,0)"),
    bp.learnRate = c(1,2,5,10))

  df[["darch.unitFunction"]] <- rep(c("c(tanhUnit, softmaxUnit)",
    "c(tanhUnit, tanhUnit, softmaxUnit)",
    "c(tanhUnit, tanhUnit, tanhUnit, softmaxUnit)"), 4)

  df
}

caretModel <- train(Species ~ ., data = iris, tuneLength = 12, trControl = tc,
  method = darchModelInfo(parameters, grid), preProc = c("center", "scale"),
  darch.numEpochs = 15, darch.batchSize = 6, testing = T, ...)

## End(Not run)
```

darchTest

Test classification network.

Description

Forward-propagate given data through the deep neural network and return classification accuracy using the given labels.

Usage

```
darchTest(darch, newdata = NULL, targets = T)
```

Arguments

darch	DArch instance.
newdata	New data to use, NULL to use training data.
targets	Labels for the data, NULL to use training labels (only possible when data is NULL as well).

Details

This is primarily a convenience function similar to [predict.DArch](#) with classification performance measurements instead of network output, and it returns a list of accuracy indicators (raw network error, percentage of incorrect classifications and absolute number of incorrect classifications).

Value

Vector containing error function output, percentage of incorrect classifications and absolute number of incorrect classifications.

See Also

Other darch interface functions: [darchBench](#), [darch](#), [plot.DArch](#), [predict.DArch](#), [print.DArch](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, retainData = T)
classificationStats <- darchTest(model)

## End(Not run)
```

exponentialLinearUnit *Exponential linear unit (ELU) function with unit derivatives.*

Description

The function calculates the activation of the units and returns a list, in which the first entry is the exponential linear activation of the units and the second entry is the derivative of the transfer function.

Usage

```
exponentialLinearUnit(input, alpha = getParameter(".darch.elu.alpha", 1, ...),
  ...)
```

Arguments

input	Input for the activation function.
alpha	ELU hyperparameter.
...	Additional parameters.

Value

A list with the ELU activation in the first entry and the derivative of the activation in the second entry.

References

Clevert, Djork-Arne, Thomas Unterthiner, and Sepp Hochreiter (2015). "Fast and Accurate Deep Network Learning by Exponential Linear Units (ELUs)". In: CoRR abs/1511.07289. URL : <http://arxiv.org/abs/1511.07289>

See Also

Other darch unit functions: [linearUnit](#), [maxoutUnit](#), [rectifiedLinearUnit](#), [sigmoidUnit](#), [softmaxUnit](#), [softplusUnit](#), [tanhUnit](#)

Examples

```
## Not run:  
data(iris)  
model <- darch(Species ~ ., iris, darch.unitFunction = "exponentialLinearUnit",  
  darch.elu.alpha = 2)  
  
## End(Not run)
```

generateWeightsGlorotNormal

Glorot normal weight initialization

Description

This function is used to generate random weights and biases using Glorot normal weight initialization as described in Glorot & Bengio, AISTATS 2010.

Usage

```
generateWeightsGlorotNormal(numUnits1, numUnits2,  
  weights.mean = getParameter(".weights.mean", 0, ...), ...)
```

Arguments

<code>numUnits1</code>	Number of units in the lower layer.
<code>numUnits2</code>	Number of units in the upper layer.
<code>weights.mean</code>	mean parameter to the mnorm function.
<code>...</code>	Additional parameters, used for parameter resolution and passed to generateWeightsNormal .

Value

Weight matrix.

References

Glorot, Xavier and Yoshua Bengio (2010). "Understanding the difficulty of training deep feed-forward neural networks". In: International conference on artificial intelligence and statistics, pp. 249-256

See Also

Other weight generation functions: [generateWeightsGlorotUniform](#), [generateWeightsHeNormal](#), [generateWeightsHeUniform](#), [generateWeightsNormal](#), [generateWeightsUniform](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, generateWeightsFunction = "generateWeightsGlorotNormal",
  weights.mean = .1)

## End(Not run)
```

`generateWeightsGlorotUniform`

Glorot uniform weight initialization

Description

This function is used to generate random weights and biases using Glorot uniform weight initialization as described in Glorot & Bengio, AISTATS 2010.

Usage

```
generateWeightsGlorotUniform(numUnits1, numUnits2, ...)
```

Arguments

<code>numUnits1</code>	Number of units in the lower layer.
<code>numUnits2</code>	Number of units in the upper layer.
<code>...</code>	Additional parameters, used for parameter resolution and passed to generateWeightsUniform .

Value

Weight matrix.

References

Glorot, Xavier and Yoshua Bengio (2010). "Understanding the difficulty of training deep feed-forward neural networks". In: International conference on artificial intelligence and statistics, pp. 249-256

See Also

Other weight generation functions: [generateWeightsGlorotNormal](#), [generateWeightsHeNormal](#), [generateWeightsHeUniform](#), [generateWeightsNormal](#), [generateWeightsUniform](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, generateWeightsFunction = "generateWeightsGlorotUniform")

## End(Not run)
```

generateWeightsHeNormal

He normal weight initialization

Description

This function is used to generate random weights and biases using He normal weight initialization as described in He et al., <http://arxiv.org/abs/1502.01852>.

Usage

```
generateWeightsHeNormal(numUnits1, numUnits2,
  weights.mean = getParameter(".weights.mean", 0, ...), ...)
```

Arguments

numUnits1	Number of units in the lower layer.
numUnits2	Number of units in the upper layer.
weights.mean	mean parameter to the rnorm function.
...	Additional parameters, used for parameter resolution and passed to generateWeightsNormal .

Value

Weight matrix.

References

He, Kaiming, Xiangyu Zhang, Shaoqing Ren, and Jian Sun (2015). "Delving Deep into Rectifiers: Surpassing Human-Level Performance on ImageNet Classification". In: CoRR abs/1502.01852. URL: <http://arxiv.org/abs/1502.01852>

See Also

Other weight generation functions: [generateWeightsGlorotNormal](#), [generateWeightsGlorotUniform](#), [generateWeightsHeUniform](#), [generateWeightsNormal](#), [generateWeightsUniform](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, generateWeightsFunction = "generateWeightsHeNormal",
  weights.mean = .1)

## End(Not run)
```

generateWeightsHeUniform

He uniform weight initialization

Description

This function is used to generate random weights and biases using He uniform weight initialization as described in He et al., <http://arxiv.org/abs/1502.01852>.

Usage

```
generateWeightsHeUniform(numUnits1, numUnits2, ...)
```

Arguments

numUnits1	Number of units in the lower layer.
numUnits2	Number of units in the upper layer.
...	Additional parameters, used for parameter resolution and passed to generateWeightsUniform .

Value

Weight matrix.

References

He, Kaiming, Xiangyu Zhang, Shaoqing Ren, and Jian Sun (2015). "Delving Deep into Rectifiers: Surpassing Human-Level Performance on ImageNet Classification". In: CoRR abs/1502.01852. URL: <http://arxiv.org/abs/1502.01852>

See Also

Other weight generation functions: [generateWeightsGlorotNormal](#), [generateWeightsGlorotUniform](#), [generateWeightsHeNormal](#), [generateWeightsNormal](#), [generateWeightsUniform](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, generateWeightsFunction = "generateWeightsHeUniform")

## End(Not run)
```

`generateWeightsNormal` *Generates a weight matrix using [rnorm](#).*

Description

This function is the standard method for generating weights for instances of [Net](#). It uses [rnorm](#) to do so.

Usage

```
generateWeightsNormal(numUnits1, numUnits2,
  weights.mean = getParameter(".weights.mean", 0, ...),
  weights.sd = getParameter(".weights.sd", 0.01, ...), ...)
```

Arguments

<code>numUnits1</code>	Number of units in the lower layer.
<code>numUnits2</code>	Number of units in the upper layer.
<code>weights.mean</code>	mean parameter to the rnorm function.
<code>weights.sd</code>	sd parameter to the rnorm function.
<code>...</code>	Additional parameters, used for parameter resolution.

Value

Weight matrix.

See Also

Other weight generation functions: [generateWeightsGlorotNormal](#), [generateWeightsGlorotUniform](#), [generateWeightsHeNormal](#), [generateWeightsHeUniform](#), [generateWeightsUniform](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, generateWeightsFunction = "generateWeightsNormal",
  weights.mean = .1, weights.sd = .05)

## End(Not run)
```

generateWeightsUniform

Generates a weight matrix using [runif](#)

Description

This function is used to generate random weights and biases using [runif](#).

Usage

```
generateWeightsUniform(numUnits1, numUnits2,
  weights.min = getParameter(".weights.min", -0.1, ...),
  weights.max = getParameter(".weights.max", 0.1, ...), ...)
```

Arguments

numUnits1	Number of units in the lower layer.
numUnits2	Number of units in the upper layer.
weights.min	min parameter to the runif function.
weights.max	max parameter to the runif function.
...	Additional parameters, used for parameter resolution.

Value

Weight matrix.

See Also

Other weight generation functions: [generateWeightsGlorotNormal](#), [generateWeightsGlorotUniform](#), [generateWeightsHeNormal](#), [generateWeightsHeUniform](#), [generateWeightsNormal](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, generateWeightsFunction = "generateWeightsUniform",
  weights.min = -.1, weights.max = .5)

## End(Not run)
```

linearUnit	<i>Linear unit function with unit derivatives.</i>
------------	--

Description

The function calculates the activation of the units and returns a list, in which the first entry is the linear activation of the units and the second entry is the derivative of the transfer function.

Usage

```
linearUnit(input, ...)
```

Arguments

input	Input for the activation function.
...	Additional parameters, not used.

Value

A list with the linear activation in the first entry and the derivative of the activation in the second entry.

See Also

Other darch unit functions: [exponentialLinearUnit](#), [maxoutUnit](#), [rectifiedLinearUnit](#), [sigmoidUnit](#), [softmaxUnit](#), [softplusUnit](#), [tanhUnit](#)

Examples

```
## Not run:  
data(iris)  
model <- darch(Species ~ ., iris, darch.unitFunction = "linearUnit")  
  
## End(Not run)
```

maxoutUnit	<i>Maxout / LWTA unit function</i>
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Description

The function calculates the activation of the units and returns a list, in which the first entry is the result through the maxout transfer function and the second entry is the derivative of the transfer function.

Usage

```
maxoutUnit(input, ..., poolSize = getParameter(".darch.maxout.poolSize", 2,
  ...), unitFunc = getParameter(".darch.maxout.unitFunction", linearUnit,
  ...), dropoutMask = vector())
```

Arguments

input	Input for the activation function.
...	Additional parameters, passed on to inner unit function.
poolSize	The size of each maxout pool.
unitFunc	Inner unit function for maxout.
dropoutMask	Vector containing the dropout mask.

Details

Maxout sets the activations of all neurons but the one with the highest activation within a pool to 0. If this is used without [maxoutWeightUpdate](#), it becomes the local-winner-takes-all algorithm, as the only difference between the two is that outgoing weights are shared for maxout.

Value

A list with the maxout activation in the first entry and the derivative of the transfer function in the second entry.

References

Srivastava, Rupesh Kumar, Jonathan Masci, Sohrab Kazerounian, Faustino Gomez, and Juergen Schmidhuber (2013). "Compete to Compute". In: *Advances in Neural Information Processing Systems 26*. Ed. by C.J.C. Burges, L. Bottou, M. Welling, Z. Ghahramani, and K.Q. Weinberger. Curran Associates, Inc., pp. 2310-2318. URL: <http://papers.nips.cc/paper/5059-competeto-compute.pdf>

Goodfellow, Ian J., David Warde-Farley, Mehdi Mirza, Aaron C. Courville, and Yoshua Bengio (2013). "Maxout Networks". In: *Proceedings of the 30th International Conference on Machine Learning, ICML 2013, Atlanta, GA, USA, 16-21 June 2013*, pp. 1319-1327. URL: <http://jmlr.org/proceedings/papers/v28/goodfellow13a.pdf>

See Also

Other darch unit functions: [exponentialLinearUnit](#), [linearUnit](#), [rectifiedLinearUnit](#), [sigmoidUnit](#), [softmaxUnit](#), [softplusUnit](#), [tanhUnit](#)

Examples

```
## Not run:
data(iris)
# LWTA:
model <- darch(Species ~ ., iris, c(0, 50, 0),
  darch.unitFunction = c("maxoutUnit", "softmaxUnit"),
  darch.maxout.poolSize = 5, darch.maxout.unitFunction = "sigmoidUnit")
```

```
# Maxout:
model <- darch(Species ~ ., iris, c(0, 50, 0),
  darch.unitFunction = c("maxoutUnit", "softmaxUnit"),
  darch.maxout.poolSize = 5, darch.maxout.unitFunction = "sigmoidUnit",
  darch.weightUpdateFunction = c("weightDecayWeightUpdate", "maxoutWeightUpdate"))

## End(Not run)
```

maxoutWeightUpdate *Updates the weight on maxout layers*

Description

On maxout layers, only the weights of active units are altered, additionally all weights within a pool must be the same.

Usage

```
maxoutWeightUpdate(darch, layerIndex, weightsInc, biasesInc, ...,
  weightDecay = getParameter(".darch.weightDecay", 0, darch),
  poolSize = getParameter(".darch.maxout.poolSize", 2, darch))
```

Arguments

darch	DArch instance.
layerIndex	Layer index within the network.
weightsInc	Matrix containing scheduled weight updates from the fine-tuning algorithm.
biasesInc	Bias weight updates.
...	Additional parameters, not used.
weightDecay	Weights are multiplied by (1 - weightDecay) before each update. Corresponds to the <code>darch.weightDecay</code> parameter of darch.default .
poolSize	Size of maxout pools, see parameter <code>darch.maxout.poolSize</code> of darch .

Value

The updated weights.

References

Goodfellow, Ian J., David Warde-Farley, Mehdi Mirza, Aaron C. Courville, and Yoshua Bengio (2013). "Maxout Networks". In: Proceedings of the 30th International Conference on Machine Learning, ICML 2013, Atlanta, GA, USA, 16-21 June 2013, pp. 1319-1327. URL: <http://jmlr.org/proceedings/papers/v28/goodfellow13a.pdf>

See Also

Other weight update functions: [weightDecayWeightUpdate](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, c(0, 50, 0),
  darch.unitFunction = c("maxoutUnit", "softmaxUnit"),
  darch.maxout.poolSize = 5, darch.maxout.unitFunction = "sigmoidUnit",
  darch.weightUpdateFunction = c("weightDecayWeightUpdate", "maxoutWeightUpdate"))

## End(Not run)
```

minimizeAutoencoder *Conjugate gradient for a autoencoder network*

Description

This function trains a [DArch](#) autoencoder network with the conjugate gradient method.

Usage

```
minimizeAutoencoder(darch, trainData, targetData,
  cg.length = getParameter(".cg.length"),
  dropout = getParameter(".darch.dropout"),
  dropConnect = getParameter(".darch.dropout.dropConnect"),
  matMult = getParameter(".matMult"), debugMode = getParameter(".debug"),
  ...)
```

Arguments

darch	A instance of the class DArch .
trainData	The training data matrix.
targetData	The labels for the training data.
cg.length	Numbers of line search
dropout	See darch.dropout parameter of darch .
dropConnect	See darch.dropout.dropConnect parameter of darch .
matMult	Matrix multiplication function, internal parameter.
debugMode	Whether debug mode is enabled, internal parameter.
...	Further parameters.

Details

This function is built on the basis of the code from G. Hinton et. al. (<http://www.cs.toronto.edu/~hinton/MatlabForSciencePaper> - last visit 2016-04-30) for the fine tuning of deep belief nets. The original code is located in the files 'backpropclassify.m', 'CG_MNIST.m' and 'CG_CLASSIFY_INIT.m'. It implements the fine tuning for a classification net with backpropagation using a direct translation of the [minimize](#) function from C. Rasmussen (available at <http://www.gatsby.ucl.ac.uk/~edward/code/minimize/> - last visit 2016-04-30) to R.

minimizeAutoencoder supports dropout but does not use the weight update function as defined via the `darch.weightUpdateFunction` parameter of `darch`, so that weight decay, momentum etc. are not supported.

Value

The trained `DArch` object.

See Also

`darch`, `fineTuneDArch`

Other fine-tuning functions: `backpropagation`, `minimizeClassifier`, `rpropagation`

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, c(6,10,2,10,6), darch.isClass = F,
  preProc.params = list(method=c("center", "scale")),
  darch.numEpochs = 20, darch.batchSize = 6, darch.unitFunction = tanhUnit
  darch.fineTuneFunction = "minimizeAutoencoder")

## End(Not run)
```

minimizeClassifier *Conjugate gradient for a classification network*

Description

This function trains a `DArch` classifier network with the conjugate gradient method.

Usage

```
minimizeClassifier(darch, trainData, targetData,
  cg.length = getParameter(".cg.length"),
  cg.switchLayers = getParameter(".cg.length"),
  dropout = getParameter(".darch.dropout"),
  dropConnect = getParameter(".darch.dropout.dropConnect"),
  matMult = getParameter(".matMult"), debugMode = getParameter(".debug"),
  ...)
```

Arguments

<code>darch</code>	A instance of the class <code>DArch</code> .
<code>trainData</code>	The training data matrix.
<code>targetData</code>	The labels for the training data.
<code>cg.length</code>	Numbers of line search

<code>cg.switchLayers</code>	Indicates when to train the full network instead of only the upper two layers
<code>dropout</code>	See <code>darch.dropout</code> parameter of darch .
<code>dropConnect</code>	See <code>darch.dropout.dropConnect</code> parameter of darch .
<code>matMult</code>	Matrix multiplication function, internal parameter.
<code>debugMode</code>	Whether debug mode is enabled, internal parameter.
<code>...</code>	Further parameters.

Details

This function is build on the basis of the code from G. Hinton et. al. (<http://www.cs.toronto.edu/~hinton/MatlabForSciencePa> - last visit 2016-04-30) for the fine tuning of deep belief nets. The original code is located in the files 'backpropclassify.m', 'CG_MNIST.m' and 'CG_CLASSIFY_INIT.m'. It implements the fine tuning for a classification net with backpropagation using a direct translation of the [minimize](#) function from C. Rasmussen (available at <http://www.gatsby.ucl.ac.uk/~edward/code/minimize/> - last visit 2016-04-30) to R. The parameter `cg.switchLayers` is for the switch between two training type. Like in the original code, the top two layers can be trained alone until epoch is equal to `epochSwitch`. Afterwards the entire network will be trained.

`minimizeClassifier` supports dropout but does not use the weight update function as defined via the `darch.weightUpdateFunction` parameter of [darch](#), so that weight decay, momentum etc. are not supported.

Value

The trained [DArch](#) object.

See Also

[darch](#), [fineTuneDArch](#)

Other fine-tuning functions: [backpropagation](#), [minimizeAutoencoder](#), [rpropagation](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris,
  preProc.params = list(method = c("center", "scale")),
  darch.unitFunction = c("sigmoidUnit", "softmaxUnit"),
  darch.fineTuneFunction = "minimizeClassifier",
  cg.length = 3, cg.switchLayers = 5)

## End(Not run)
```

mseError	<i>Mean squared error function</i>
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Description

The function calculates the mean squared error (MSE) from the original and estimate parameters.

Usage

```
mseError(original, estimate)
```

Arguments

original	The original data matrix.
estimate	The calculated data matrix.

Details

This function is a valid value for both [darch](#) parameters `rbm.errorFunction` and `darch.errorFunction`.

Value

A list with the name of the error function in the first entry and the error value in the second entry.

See Also

Other error functions: [crossEntropyError](#), [rmseError](#)

Examples

```
## Not run:  
data(iris)  
model <- darch(Species ~ ., iris, rbm.errorFunction = "mseError",  
  darch.errorFunction = "mseError")  
  
## End(Not run)
```

plot.DArch *Plot DArch statistics or structure.*

Description

This function provides different plots depending on the type parameter:

Usage

```
## S3 method for class 'DArch'  
plot(x, y = "raw", ..., type = y)
```

Arguments

x	DArch instance.
y	See type.
...	Additional parameters, passed to plotting functions.
type	Which type of plot to create, one of raw, class, time, momentum, and net.

Details

- raw. Plots the raw network error (e.g. MSE), this is the default
- class. Plots the classification error
- time. Plots the times needed for each epoch
- momentum. Plots the momentum ramp
- net. Calls [plotnet](#) to plot the network

Value

The plotted graph.

See Also

Other darch interface functions: [darchBench](#), [darchTest](#), [darch](#), [predict.DArch](#), [print.DArch](#)

Examples

```
## Not run:  
data(iris)  
model <- darch(Species ~ ., iris)  
plot(model)  
plot(model, "class")  
plot(model, "time")  
plot(model, "momentum")  
plot(model, "net")  
  
## End(Not run)
```

predict.DArch	<i>Forward-propagate data.</i>
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Description

Forward-propagate given data through the deep neural network.

Usage

```
## S3 method for class 'DArch'
predict(object, ..., newdata = NULL, type = "raw",
        inputLayer = 1, outputLayer = 0)
```

Arguments

object	DArch instance
...	Further parameters, if newdata is NULL, the first unnamed parameter will be used for newdata instead.
newdata	New data to predict, NULL to return latest network output
type	Output type, one of: raw, bin, class, or character. raw returns the layer output, bin returns 1 for every layer output > 0.5, 0 otherwise, and class returns 1 for the output unit with the highest activation, otherwise 0. Additionally, when using class, class labels are returned when available. character is the same as class, except using character vectors instead of factors.
inputLayer	Layer number (> 0). The data given in newdata will be fed into this layer. Note that absolute numbers count from the input layer, i.e. for a network with three layers, 1 would indicate the input layer.
outputLayer	Layer number (if > 0) or offset (if <= 0) relative to the last layer. The output of the given layer is returned. Note that absolute numbers count from the input layer, i.e. for a network with three layers, 1 would indicate the input layer.

Value

Vector or matrix of networks outputs, output type depending on the type parameter.

See Also

Other darch interface functions: [darchBench](#), [darchTest](#), [darch](#), [plot.DArch](#), [print.DArch](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, retainData = T)
predict(model)

## End(Not run)
```

`print.DArch` *Print DArch details.*

Description

Print verbose information about a [DArch](#) instance.

Usage

```
## S3 method for class 'DArch'  
print(x, ...)
```

Arguments

`x` [DArch](#) instance
`...` Further parameters, not used.

Details

Information printed include [darch](#) parameters and a summary of training statistics.

See Also

Other darch interface functions: [darchBench](#), [darchTest](#), [darch](#), [plot.DArch](#), [predict.DArch](#)

Examples

```
## Not run:  
data(iris)  
model <- darch(Species ~ ., iris)  
print(model)  
  
## End(Not run)
```

`provideMNIST` *Provides MNIST data set in the given folder.*

Description

This function will, if necessary and allowed, download the compressed MNIST data set and save it to .RData files using [readMNIST](#). If the compressed MNIST archives are available, it will convert them into RData files loadable from within R. If the RData files are already available, nothing will be done.

Usage

```
provideMNIST(folder = "data/", download = F)
```

Arguments

folder	Folder name, including a trailing slash.
download	Logical indicating whether download is allowed.

Value

Boolean value indicating success or failure.

Examples

```
## Not run:  
provideMNIST("mnist/", download = T)  
  
## End(Not run)
```

rectifiedLinearUnit *Rectified linear unit function with unit derivatives.*

Description

The function calculates the activation of the units and returns a list, in which the first entry is the rectified linear activation of the units and the second entry is the derivative of the transfer function.

Usage

```
rectifiedLinearUnit(input, ...)
```

Arguments

input	Input for the activation function.
...	Additional parameters, not used.

Value

A list with the rectified linear activation in the first entry and the derivative of the activation in the second entry.

References

Glorot, Xavier, Antoine Bordes, and Yoshua Bengio (2011). "Deep Sparse Rectifier Neural Networks". In: Proceedings of the Fourteenth International Conference on Artificial Intelligence and Statistics (AISTATS-11). Ed. by Geoffrey J. Gordon and David B. Dunson. Vol. 15. Journal of Machine Learning Research - Workshop and Conference Proceedings, pp. 315-323. URL : <http://www.jmlr.org/proceedings/papers/v15/glorot11a/glorot11a.pdf>

See Also

Other darch unit functions: [exponentialLinearUnit](#), [linearUnit](#), [maxoutUnit](#), [sigmoidUnit](#), [softmaxUnit](#), [softplusUnit](#), [tanhUnit](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, darch.unitFunction = "rectifiedLinearUnit")

## End(Not run)
```

rmseError	<i>Root-mean-square error function</i>
-----------	--

Description

The function calculates the root-mean-square error (RMSE) from the original and estimate parameters.

Usage

```
rmseError(original, estimate)
```

Arguments

original	The original data matrix.
estimate	The calculated data matrix.

Details

This function is a valid value for both [darch](#) parameters `rbm.errorFunction` and `darch.errorFunction`.

Value

A list with the name of the error function in the first entry and the error value in the second entry.

See Also

Other error functions: [crossEntropyError](#), [mseError](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, rbm.errorFunction = "rmseError",
  darch.errorFunction = "rmseError")

## End(Not run)
```

rpropagation	<i>Resilient backpropagation training for deep architectures.</i>
--------------	---

Description

The function trains a deep architecture with the resilient backpropagation algorithm. It is able to use four different types of training (see details). For details of the resilient backpropagation algorithm see the references.

Usage

```
rpropagation(darch, trainData, targetData,
  rprop.method = getParameter(".rprop.method"),
  rprop.decFact = getParameter(".rprop.decFact"),
  rprop.incFact = getParameter(".rprop.incFact"),
  rprop.initDelta = getParameter(".rprop.initDelta"),
  rprop.minDelta = getParameter(".rprop.minDelta"),
  rprop.maxDelta = getParameter(".rprop.maxDelta"),
  nesterovMomentum = getParameter(".darch.nesterovMomentum"),
  dropout = getParameter(".darch.dropout"),
  dropConnect = getParameter(".darch.dropout.dropConnect"),
  errorFunction = getParameter(".darch.errorFunction"),
  matMult = getParameter(".matMult"), debugMode = getParameter(".debug", F,
  darch), ...)
```

Arguments

darch	The deep architecture to train
trainData	The training data
targetData	The expected output for the training data
rprop.method	The method for the training. Default is "iRprop+"
rprop.decFact	Decreasing factor for the training. Default is 0.6.
rprop.incFact	Increasing factor for the training Default is 1.2.
rprop.initDelta	Initialisation value for the update. Default is 0.0125.
rprop.minDelta	Lower bound for step size. Default is 0.000001
rprop.maxDelta	Upper bound for step size. Default is 50
nesterovMomentum	See darch.nesterovMomentum parameter of darch .
dropout	See darch.dropout parameter of darch .
dropConnect	See darch.dropout.dropConnect parameter of darch .
errorFunction	See darch.errorFunction parameter of darch .
matMult	Matrix multiplication function, internal parameter.
debugMode	Whether debug mode is enabled, internal parameter.
...	Further parameters.

Details

RPROP supports dropout and uses the weight update function as defined via the `darch.weightUpdateFunction` parameter of `darch`.

The code for the calculation of the weight change is a translation from the MATLAB code from the Rprop Optimization Toolbox implemented by R. Calandra (see References).

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The possible training methods (parameter `rprop.method`) are the following (see References for details):

Rprop+:	Rprop with Weight-Backtracking
Rprop-:	Rprop without Weight-Backtracking
iRprop+:	Improved Rprop with Weight-Backtracking
iRprop-:	Improved Rprop without Weight-Backtracking

Value

[DArch](#) - The trained deep architecture

References

M. Riedmiller, H. Braun. A direct adaptive method for faster backpropagation learning: The RPROP algorithm. In Proceedings of the IEEE International Conference on Neural Networks, pp 586-591. IEEE Press, 1993.

C. Igel, M. Huesken. Improving the Rprop Learning Algorithm, Proceedings of the Second International Symposium on Neural Computation, NC 2000, ICSC Academic Press, Canada/Switzerland, pp. 115-121., 2000.

Kohavi, R., A Study of Cross-Validation and Bootstrap for Accuracy Estimation and Model Selection, Proceedings of the 14th Int. Joint Conference on Artificial Intelligence 2, S. 1137-1143, Morgan Kaufmann, Morgan Kaufmann Publishers Inc., San Francisco, CA, USA, 1995.

See Also

[darch](#)

Other fine-tuning functions: [backpropagation](#), [minimizeAutoencoder](#), [minimizeClassifier](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, darch.fineTuneFunction = "rpropagation",
  preProc.params = list(method = c("center", "scale")),
  darch.unitFunction = c("softplusUnit", "softmaxUnit"),
  rprop.method = "iRprop+", rprop.decFact = .5, rprop.incFact = 1.2,
  rprop.initDelta = 1/100, rprop.minDelta = 1/1000000, rprop.maxDelta = 50)

## End(Not run)
```

sigmoidUnit

Sigmoid unit function with unit derivatives.

Description

The function calculates the activation and returns a list which the first entry is the result through the sigmoid transfer function and the second entry is the derivative of the transfer function.

Usage

```
sigmoidUnit(input, ...)
```

Arguments

input	Input for the activation function.
...	Additional parameters, not used.

Value

A list with the activation in the first entry and the derivative of the transfer function in the second entry.

See Also

Other darch unit functions: [exponentialLinearUnit](#), [linearUnit](#), [maxoutUnit](#), [rectifiedLinearUnit](#), [softmaxUnit](#), [softplusUnit](#), [tanhUnit](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, darch.unitFunction = "sigmoidUnit")

## End(Not run)
```

softmaxUnit	<i>Softmax unit function with unit derivatives.</i>
-------------	---

Description

The function calculates the activation of the units and returns a list, in which the first entry is the result through the softmax transfer function and the second entry is the derivative of the transfer function.

Usage

```
softmaxUnit(input, ...)
```

Arguments

input	Input for the activation function.
...	Additional parameters, not used.

Value

A list with the softmax activation in the first entry and the derivative of the transfer function in the second entry.

References

<http://www.faqs.org/faqs/ai-faq/neural-nets/part2/section-12.html>

See Also

Other darch unit functions: [exponentialLinearUnit](#), [linearUnit](#), [maxoutUnit](#), [rectifiedLinearUnit](#), [sigmoidUnit](#), [softplusUnit](#), [tanhUnit](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris,
  darch.unitFunction = c("sigmoidUnit", "softmaxUnit"))

## End(Not run)
```

softplusUnit	<i>Softplus unit function with unit derivatives.</i>
--------------	--

Description

The function calculates the activation of the units and returns a list, in which the first entry is the softmax activation of the units and the second entry is the derivative of the transfer function. Softplus is a smoothed version of the rectified linear activation function.

Usage

```
softplusUnit(input, ...)
```

Arguments

input	Input for the activation function.
...	Additional parameters, not used.

Value

A list with the softplus activation in the first entry and the derivative of the activation in the second entry.

References

Dugas, Charles, Yoshua Bengio, Francois Belisle, Claude Nadeau, and Rene Garcia (2001). "Incorporating Second-Order Functional Knowledge for Better Option Pricing". In: Advances in Neural Information Processing Systems, pp. 472-478.

See Also

Other darch unit functions: [exponentialLinearUnit](#), [linearUnit](#), [maxoutUnit](#), [rectifiedLinearUnit](#), [sigmoidUnit](#), [softmaxUnit](#), [tanhUnit](#)

Examples

```
## Not run:  
data(iris)  
model <- darch(Species ~ ., iris, darch.unitFunction = "softplusUnit")  
  
## End(Not run)
```

tanhUnit	<i>Continuous Tan-Sigmoid unit function.</i>
----------	--

Description

Calculates the unit activations and returns them in a list.

Usage

```
tanhUnit(input, ...)
```

Arguments

input	Input for the activation function.
...	Additional parameters, not used.

Value

A list with the activation in the first entry and the derivative of the transfer function in the second entry.

See Also

Other darch unit functions: [exponentialLinearUnit](#), [linearUnit](#), [maxoutUnit](#), [rectifiedLinearUnit](#), [sigmoidUnit](#), [softmaxUnit](#), [softplusUnit](#)

Examples

```
## Not run:
data(iris)
model <- darch(Species ~ ., iris, darch.unitFunction = "tanhUnit")

## End(Not run)
```

weightDecayWeightUpdate	<i>Updates the weight using weight decay.</i>
-------------------------	---

Description

Multiplies the weights by (1 - weightDecay) before applying the scheduled weight changes.

Usage

```
weightDecayWeightUpdate(darch, layerIndex, weightsInc, biasesInc, ...,
  weightDecay = getParameter(".darch.weightDecay", 0, darch),
  debug = getParameter(".debug", F, darch))
```

Arguments

darch	DArch instance.
layerIndex	Layer index within the network.
weightsInc	Matrix containing scheduled weight updates from the fine-tuning algorithm.
biasesInc	Bias weight updates.
...	Additional parameters, not used.
weightDecay	Weights are multiplied by $(1 - \text{weightDecay})$ before each update. Corresponds to the <code>darch.weightDecay</code> parameter of darch.default .
debug	Internal debugging flag.

Value

updated weights

See Also

Other weight update functions: [maxoutWeightUpdate](#)

Examples

```
## Not run:  
model <- darch(Species ~ ., iris, c(0, 50, 0),  
  darch.weightUpdateFunction = "weightDecayWeightUpdate")  
  
## End(Not run)
```

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