

Package: tensorsem (via r-universe)

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Type Package

Title Estimate structural equation models using computation graphs

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Description Use lavaan code to create structural equation models, use torch to estimate them. This package provides the interface between lavaan and torch.

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Repository <https://vankesteren.r-universe.dev>

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df_to_tensor	<i>Prepare data for tensorsem model</i>
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Description

This function prepares a dataframe for a tensorsem model. It first converts the variables to a design matrix, then centers it, and lastly converts it to a torch_tensor

Usage

```
df_to_tensor(df, dtype = NULL, device = NULL)
```

Arguments

df	data frame
dtype	data type of the resulting tensor
device	device to store the resulting tensor on

Value

Torch tensor of scaled and processed data

See Also

[torch::torch_tensor\(\)](#), [stats::model.matrix\(\)](#)

lav_mod_to_torch_opts	<i>Create a torch options list from a lavaan Model class.</i>
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Description

Create a torch options list from a lavaan Model class.

Usage

```
lav_mod_to_torch_opts(lav_mod)
```

Arguments

lav_mod	lavaan Model class object
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Value

list of tensorsem options

mvn_negloglik	<i>Multivariate normal negative log-likelihood loss function for tensorsem nn module.</i>
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Description

Multivariate normal negative log-likelihood loss function for tensorsem nn module.

Usage

```
mvn_negloglik(dat, Sigma)
```

Arguments

dat	The centered dataset as a tensor
Sigma	The model() implied covariance matrix

Value

torch_tensor: scalar negative log likelihood

partable_from_torch	<i>Create a lavaan parameter table from torch free_params output</i>
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Description

See examples in tensorsem for how to save the output.

Usage

```
partable_from_torch(pars, syntax)
```

Arguments

pars	data frame of parameter estimates (est) and their standard errors (se)
syntax	syntax of the original model

`sem_fitfun`*SEM fitting function*

Description

SEM fitting function

Usage`sem_fitfun(S, Sigma)`**Arguments**

<code>S</code>	The observed covariance matrix
<code>Sigma</code>	The model implied covariance matrix

Value

torch_tensor: scalar loss function

`syntax_to_torch_opts` *Create a torch options list from lavaan syntax.*

Description

Create a torch options list from lavaan syntax.

Usage`syntax_to_torch_opts(syntax)`**Arguments**

<code>syntax</code>	lavaan syntax
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Value

list of tensorsem options

torch_jacobian	<i>Compute jacobian of output wrt input tensor</i>
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Description

Compute jacobian of output wrt input tensor

Usage

```
torch_jacobian(output, input)
```

Arguments

output	Tensor vector of size Po
input	Tensor vector of size Pi

Value

jacobian: Tensor of size Pi, Po

torch_sem	<i>Structural equation model with a Torch backend</i>
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Description

Function for creating a structural equation model

Usage

```
torch_sem(syntax, dtype = torch_float32(), device = torch_device("cpu"))
```

Arguments

syntax	lavaan syntax for the SEM model
dtype	(optional) torch dtype for the model (default torch_float32())
device	(optional) device type to put the model on. see torch::torch_device()

Details

This function instantiates a torch object for computing the model-implied covariance matrix based on a structural equation model. Through torch, gradients of this forward model can then be computed using backpropagation, and the parameters can be optimized using gradient-based optimization routines from the torch package.

Because of this, it is easy to add additional penalties to the standard objective function, or to write a new objective function altogether.

Value

A torch_sem object, which is an nn_module (torch object)

Fields

free_params Vector of free parameters

Methods

`$initialize():`

The initialize method. Don't use this, just use `torch_sem()`

Arguments:

- syntax lavaan syntax for the SEM model
- dtype (optional) torch dtype for the model (default torch_float32())
- device (optional) device type to put the model on. see `torch::torch_device()`

Value:

A torch_sem object, which is an nn_module (torch object)

`$forward():`

Compute the model-implied covariance matrix. Don't use this; nn_modules are callable, so access this method by calling the object itself as a function, e.g., `my_torch_sem()`. In the forward pass, we apply constraints to the parameter vector, and we create matrix views from it to compute the model-implied covariance matrix.

Value:

A torch_tensor of the model-implied covariance matrix

`$inverse_Hessian(loss):`

Compute and return the asymptotic covariance matrix of the parameters with respect to the loss function

Arguments:

- loss torch_tensor of freshly computed loss function (needed by torch for backwards pass)

Value:

A torch_tensor, representing the ACOV of the free parameters

`$standard_errors(loss):`

Compute and return observed information standard errors of the parameters, assuming the loss function is the likelihood and the current estimates are ML estimates.

Arguments:

- loss torch_tensor of freshly computed loss function (needed by torch for backwards pass)

Value:

A numeric vector of standard errors of the free parameters

`$partable(loss):`

Create a lavaan-like parameter table from the current parameter estimates in the torch_sem object.

Arguments:

- `loss` (optional) torch_tensor of freshly computed loss function (needed by torch for backwards pass)

Value:

lavaan partable object

`$fit(dat, lrate, maxit, verbose, tol):`

Fit a torch_sem model using the default maximum likelihood objective. This function uses the Adam optimizer to estimate the parameters of a torch_sem

Arguments:

- `dat` dataset (centered!) as a torch_tensor
- `lrate` learning rate of the Adam optimizer.
- `maxit` maximum number of epochs to train the model
- `verbose` whether to print progress to the console
- `tol` parameter change tolerance for stopping training

Value:

Self, i.e., the torch_sem object with updated parameters

`$loglik(dat):`

Multivariate normal log-likelihood of the data.

Arguments:

- `dat` dataset (centered!) as a torch_tensor

Value:

Log-likelihood value (torch scalar)

See Also

[df_to_tensor\(\)](#)

torch_vech

Half-vectorization of square matrices

Description

Half-vectorization of square matrices

Usage

`torch_vech(x)`

Arguments

`x` square (symmetric) matrix tensor

Value

column vector of stacked lower-diagonal elements

vech_dup_idx	<i>Constructs index vector for transforming a vech vector into a vec vector to create an $n \times n$ symmetric matrix from the vech vector. <code>tensor\$index_select(0, idx)\$view(3,3)</code></i>
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Description

Constructs index vector for transforming a vech vector into a vec vector to create an $n \times n$ symmetric matrix from the vech vector. `tensor$index_select(0, idx)$view(3,3)`

Usage

```
vech_dup_idx(n)
```

Arguments

n size of the resulting square matrix

Value

array containing the indices

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