

Package ‘mafR’

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Type Package

Title Interface for Masked Autoregressive Flows

Description Interfaces the 'python' library 'zuko' implementing Masked Autoregressive Flows. See Rozet, Divo and Schnake (2023) <[doi:10.5281/zenodo.7625672](https://doi.org/10.5281/zenodo.7625672)> and Papamakarios, Pavlakou and Murray (2017) <[doi:10.48550/arXiv.1705.07057](https://doi.org/10.48550/arXiv.1705.07057)>.

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Author Jean-Michel Marin [aut, cph],
François Rousset [aut, cre, cph] (ORCID:
<<https://orcid.org/0000-0003-4670-0371>>)

Maintainer François Rousset <francois.rousset@umontpellier.fr>

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`.r_to_torch` *Utility to manage torch tensors*

Description

(Currently not used nor exported) utility converting an R object to a torch tensor.

Usage

```
.r_to_torch(x, py_handle, device)
```

Arguments

<code>x</code>	An R object suitable for use in <code>reticulate::r_to_py(x)</code> (this being as indefinite as the <code>r_to_py</code> documentation in this respect.)
<code>py_handle</code>	The return value of <code>get_py_MAF_handle</code> , or possibly more generally an environment with (at least) elements <code>torch</code> and <code>device</code> defined as in such a return value.
<code>device</code>	Character: <code>"cpu"</code> ; or a GPU backend, either <code>"cuda"</code> (or <code>"cuda:0"</code> , etc.) or <code>"mps"</code> depending on system capabilities.

Value

`r_to_torch` returns a 32-bit floating-point **torch** tensor allocated on the given device.

Examples

```
# See help("get_py_MAF_handle") for session setup, providing 'py_handle' argument in particular.
```

`control_py_env` *Python controls*

Description

Interface to control variables in a Python environment possibly used by Infusion. Currently the only implemented control is that of the **torch** random seed.

Usage

```
control_py_env(py_handle, seed = NULL)
```

Arguments

<code>py_handle</code>	An R environment that provides access to a Python evaluation environment, as produced by <code>get_py_MAF_handle</code>
<code>seed</code>	Numeric: passed (as integer value) to <code>torch.random.manual_seed</code> .

Value

Returns NULL invisibly.

See Also

[get_py_MAF_handle](#)

Examples

```
# See help("get_py_MAF_handle") for basic usage

# The following test of the first argument may be useful in broader contexts:
## Not run:
if (inherits(py_handle, "environment")) control_py_env(py_handle, seed=0L)

## End(Not run)
```

get_py_MAF_handle *Utilities to manage Python environment and torch tensors*

Description

Utility initializing a Python environment for running `zuko.flows.MAF` and retrieving it.

Usage

```
get_py_MAF_handle(envir, reset=FALSE, torch_device="cpu", GPU_mem=NULL,
                  verbose = TRUE)
```

Arguments

envir	An environment (in the R sense) initialized as shown in the Examples.
reset	Boolean: Whether to reinitialize the Python session or not.
torch_device	Character: "cpu"; or a GPU backend, either "cuda" (or "cuda:0", etc.) or "mps" depending on system capabilities.
GPU_mem	For development purposes (effect is complicated). An amount of (dedicated) GPU memory, in bytes.
verbose	Boolean. Whether to print some messages or not.

Value

If successful, `get_py_MAF_handle` returns the modified input environment. If sourcing the Python code provided by **mafR** failed (presumably from trying to use an improperly set-up Python environment), the error condition message is returned.

See Also

[init_py_env](#), [control_py_env](#)

Examples

```
## Not run: # too slow for CRAN checks
# Initialization of python virtual environment:
init_py_env(test_cuda=FALSE)

# Initialization of python session (defines functions etc.
# in python session and provide access to them from R):
#
my_env <- list2env(list(is_set=FALSE),parent = emptyenv())
py_handle <- get_py_MAF_handle(my_env, reset=FALSE, torch_device="cpu")

# => get_py_MAF_handle() has added objects in myenv:
my_env$torch # Imported Python package (result of reticulate::import("torch"))
my_env$device # the torch_device
# and functions for MAF training, such as myenv$MAF_density_estimation.

# The returned 'py_handle' can be used by other functions, e.g.:
control_py_env(py_handle, seed=0L)
# alters the seed in the python session.

## End(Not run)
```

init_py_env

Recipe for installation using py_require

Description

Initialize the required Python environment with the required modules, using procedures introduced in **reticulate** 1.42.0.

Usage

```
init_py_env(
  cuda=FALSE, test_cuda=cuda, test_dynamo=TRUE, force=FALSE,
  test_torch=test_cuda || test_dynamo,
  verbose=interactive())
```

Arguments

cuda	Boolean: Whether to install cuda device (only needed once).
test_cuda	Boolean: whether to test that cuda device can be used.
test_dynamo	Boolean: whether to test that "torch._dynamo" module can be imported. In particular, this test whether libstdc++ is accessible where this module expects it. See Details if the test fails
test_torch	Boolean: whether to test that "torch" module can be imported.
force	Boolean: Whether to force re-installation of miniconda.
verbose	Boolean: Whether to report progress of the different steps of the procedure.

Details

The attempt to import the "torch._dynamo" module may fail on linux with message `ImportError: cannot import name 'NP_SUPPORTED_MODULES' from 'torch._dynamo.utils'`. If so, seek the `libstdc++` library in the `r-reticulate` architecture (something like `/home/<MY_USER_NAME>/.local/share/r-miniconda/lib` and either create a symbolic link to it as `libstdc++.so.6` in (likely path) `/usr/lib/x86_64-linux-gnu/`, or change the `PATH` variable: `export LD_LIBRARY_PATH=/home/<MY_USER_NAME>/.local/share/r-miniconda/lib/:$LD_LIBRARY_PATH`. On the WSL, I found that the symbolic link is overwritten each time the WSL is restarted.

Value

Returns NULL invisibly.

See Also

[get_py_MAF_handle](#)

Examples

```
# See help("get_py_MAF_handle") for basic usage

# The following test may be useful in broader contexts:
## Not run: # too slow for CRAN checks
if ( ! reticulate::py_available(initialize = FALSE)) init_py_env(test_cuda=FALSE)

## End(Not run)
```

mafR

Interface for masked autoregressive flows

Description

This wraps Python procedures to train Masked Autoregressive Flows (MAFs, Paramakarios et al. 2017) using the Python package `zuko`. It has been tested with version 1.1.0, 1.2.0 and 1.4.0 of that package. Note that objects created by its version 1.2.0 cannot be read with its version 1.1.0 (i.e., when saved in and read from pickle files).

The simplest portable way to get **mafR** working may be to install it in a conda environment. Below is a complete installation recipe. Alternative installation procedures (notably, using **reticulate** functions) *may* be found in files not necessarily included in this package but available on its Git repository, <https://github.com/f-rousset/mafR>.

```
mkdir -p ~/miniconda3
wget https://repo.anaconda.com/miniconda/Miniconda3-latest-Linux-x86_64.sh -O ~/miniconda3/miniconda3.sh
bash ~/miniconda3/miniconda3.sh -b -u -p ~/miniconda3
rm ~/miniconda3/miniconda3.sh
```

```
~/miniconda3/bin/conda init bash
conda create --name maf-conda python==3.10
```

```
conda activate maf-conda

pip install zuko

conda install R
conda install conda-forge::r-gmp
conda install conda-forge::gsl
```

and, in an R session within the maf-conda environment:

```
install.packages("reticulate")
library(reticulate)
use_condaenv(condaenv="maf-conda", conda=~/.miniconda3/bin/conda)
install.packages("mafR")

# 'mafR' was first designed for use with 'Infusion':
install.packages("Infusion")
install.packages("Rmixmod") # only a Suggested dependency of Infusion, but needed.
```

References

- Papamakarios, G., D. Sterratt, and I. Murray. 2019. Sequential Neural Likelihood: Fast Likelihood-free Inference with Autoregressive Flows. Proceedings of the Twenty-Second International Conference on Artificial Intelligence and Statistics, PMLR 89:837-848, 2019. <https://doi.org/10.48550/arXiv.1705.07057> ; <https://proceedings.mlr.press/v89/papamakarios19a.html>
- Rozet, F., Divo, F., Schnake, S (2023) Zuko: Normalizing flows in PyTorch. <https://doi.org/10.5281/zenodo.7625672>

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