Tensorflow and Keras installation steps for Deep Learning applications in Rstudio Ricardo Dalagnol¹ and Fabien Wagner

1. Install OSGEO (<u>https://trac.osgeo.org/osgeo4w/</u>) to have GDAL utilities and QGIS. Do not modify the installation path.

- 2. Install ImageMagick (<u>https://imagemagick.org/index.php</u>) for image visualization.
- 3. Install the latest Miniconda (<u>https://docs.conda.io/en/latest/miniconda.html</u>)
- 4. Install Visual Studioa. Install the 2019 or 2017 free version. The free version is called 'Community'.
 - b. During installation, select the box to include C++ development
- 5. Install the latest Nvidia driver for your GPU card from the official Nvidia site
- 6. Install CUDA NVIDIA
 - a. I installed CUDA Toolkit 10.1 update2, to check if the installed or available NVIDIA driver is compatible with the CUDA version (check this in the cuDNN page <u>https://docs.nvidia.com/deeplearning/sdk/cudnn-support-</u><u>matrix/index.html</u>)
 - b. Downloaded from NVIDIA website, searched for CUDA NVIDIA download in google: <u>https://developer.nvidia.com/cuda-10.1-download-archive-update2</u> (list of all archives here <u>https://developer.nvidia.com/cuda-toolkit-archive</u>)

7. Install cuDNN (deep neural network library) – I have installed cudnn-10.1windows10-x64-v7.6.5.32 for CUDA 10.1, the last version https://docs.nvidia.com/deeplearning/sdk/cudnn-install/#install-windows

- a. To download you have to create/login an account with NVIDIA Developer
- b. Before installing, check the NVIDIA driver version if it is compatible
- c. To install on windows, follow the instructions at section 3.3 <u>https://docs.nvidia.com/deeplearning/sdk/cudnn-install/#install-windows</u>

d. To check for env vars Win+R "control sysdm.cpl"

- 8. Install R v4.X (<u>https://cran.r-project.org/bin/windows/base/</u>)
- 9. Install Rstudio (<u>http://www.rstudio.com/products/rstudio/download/</u>)

10. Install Rtools (<u>https://cran.r-project.org/bin/windows/Rtools/</u>) and add in the path. This is explained in the link above how to do this from R. It is just a line of code.

11. Create and setup a virtual environment with conda prompt and install python 3.7, tensorflow-gpu and keras. To do this open Anaconda Prompt in Start menu, then type in the commands in the Table 1, one by one. If a prompt appears asking y or n, type y.

Table 1 – Commands to create and setup the virtual environment.

Command	Reason
conda create -n r-tensorflow python=3.7 anaconda	Create an environment named r-
	tensorflow inside conda with the
	python v3.7
activate r-tensorflow	Activate the conda environment
Obs: depending on the conda version you may need	so we can start installing stuff
to type conda activate r-tensorflow	

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Send an email for collaboration or if you have suggestions for the improvement of this manual.

pip install tensorflow-gpu==2.2.0-rc4	Install the tensorflow v2.2					
conda install pillow	Need this to work with .tif					
conda install libtiff=4.0.10	Need this to work with .tif					
pip install tensorflow-addons	Install tensorflow addons					
python	Start python so we can quickly					
	test if both python and tensorflow					
	are working					
import tensorflow as tf	Try to import tensorflow. If no					
	error, you are fine; Else, start the					
	installation again and/or try to					
	find the source of error					
import numpy	Try to import numpy; another					
	important library. If no error, you					
	are fine; Else, start the					
	installation again and/or try to					
	find the source of error					
exit()	Exit python					
deactivate	Deactivate the conda					
Obs: depending on the conda version you may need	environment					
to type conda deactivate						

If all worked, the environment should be ready to use in Rstudio.

12. Now install the packages in R. Open Rstudio and run the commands below. If it asks if you want to install from source, click yes.

install.packages("devtools")
library(devtools)
devtools::install_github("rstudio/keras")
install.packages("tensorflow")
install.packages("reticulate")
library(keras)
library(tensorflow)
library(reticulate)

- 13. If everything installed correctly, it is time for testing training a model.
 - a. Download the UNET model and data from the paper of Wagner et al. (2020) and run the code (<u>https://zenodo.org/record/3926822#.Xv4c1ChKiUl</u>).
 - b. If everything is working properly, you should see the information regarding UNET training being outputted in the console (Figure 1) and the variation in training and validation loss and accuracy metric in the viewer (Figure 2). In Figure 2, see that the training (blue) goes up quite fast at the beginning and then slowly increases until the end of the training. Meanwhile, the validation (green) only increases after 25 epochs. Your outputs will not be exactly identical to this, but this is expected.
- 14. The end. Have fun with deep learning!

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12/12	[=========]	- 5s	438ms/step	- loss:	0.2100	- custom:	0.8216 -	- val_loss:	1.0540 -	val_custom:	0.1548
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12/12	[=========			- 5s	386ms/step	- loss:	0.1917	- custom:	0.8370 -	- val_loss:	0.7169 -	val_custom:	0.4108
Epoch	28/30		-		, 1								
12/12	[======]	- 6s	487ms/step	- loss:	0.1791	- custom:	0.8468 -	- val_loss:	0.5691 -	val_custom:	0.5317
Epoch	29/30												
12/12	[========]	- 6s	483ms/step	- loss:	0.1750	- custom:	0.8507 -	- val_loss:	0.2822 -	val_custom:	0.7711
Epoch	30/30												
12/12	[=======			- 6s	4/6ms/step	- IOSS:	0.1710	- custom:	0.8534 -	- vai_loss:	0.2561 -	val_custom:	0./918
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Figure 1 – Information output in the console during training.



Figure 2 – Graph of training and validation loss (up) and accuracy metric (bottom).