Using Huggingface language models

NLP master course 2022-2023

Mariano Rico (mariano.rico@upm.es)

Document created on 2023-01-05

Table of contents

1	Using HuggingFace models from R				
	Checking the python environment				
	1.2 Installing the Hugging Face module in Python	2			
	1.3 Testing NLP tasks	3			
2	Text classification 2.1 Now in Spanish	3			
3	Part of speech	7			
4	Name Entity recognition (NER)	8			

1 Using HuggingFace models from R

The key is the reticulate package. This package allows you to use any python package from R. You have the documentation here.

1.1 Checking the python environment

If you have installed the keras R package (the installation is not obvious) you will have a *Python virtual environment* r-reticulate. You can check it with this:

```
library(reticulate)
virtualenv_list() #List all available virtualenvs
```

[1] "r-reticulate"

```
#conda_list() #List all available (mini)conda envs. In this case, none.
```

To know all the details of the Python that is being used, we will do:

```
py_config() #What python(s) is/are installed and where is the ejecutable
```

```
python: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate/bin/python
```

libpython: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate/lib/libpython3.8.so

pythonhome: /nome/rstudio/.local/share/r-miniconda/envs/r-reticulate/lib/libpythons.s.so

pythonhome: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate:/home/rstudio/.local/share/r-miniconda/envs/r-reticulate:/home/rstudio/.local/share/r-miniconda/envs/r-reticulate:/home/rstudio/.local/share/r-miniconda/envs/r-reticulate:/home/rstudio/.local/share/r-miniconda/envs/r-reticulate:/home/rstudio/.local/share/r-miniconda/envs/r-reticulate/lib/libpythons.s.so

version: 3.8.13 | packaged by conda-forge | (default, Mar 25 2022, 06:04:18) [GCC 10.3.0]

numpy: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate/lib/python3.8/site-packages/numpy:

numpy_version: 1.23.4

If you are using (mini)conda you have to use a different function:

```
py_discover_config() #Use this for (mini)conda
```

python: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate/bin/python

libpython: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate/lib/libpython3.8.so

pythonhome: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate:/home/rstudio/.local/share/r-m

version: 3.8.13 | packaged by conda-forge | (default, Mar 25 2022, 06:04:18) [GCC 10.3.0]

numpy: /home/rstudio/.local/share/r-miniconda/envs/r-reticulate/lib/python3.8/site-packages/numpy:

numpy_version: 1.23.4

1.2 Installing the HuggingFace module in Python

Using the reticulate package we can install the module/library/package Python transformers with:

```
use_virtualenv("r-reticulate") #Use a given virtualenv
packs <- py_list_packages(type = "virtualenv") #Returns a df
if(!"transformers" %in% packs$package){ #If not in the list of instaled Python libs
    py_install("transformers", pip = TRUE) #Install it
}</pre>
```

1.3 Testing NLP tasks

We have to load the transformers Python module. The R package reticulate can use the methods of the module:

```
transformers <- reticulate::import("transformers")</pre>
```

The transformers R object has a function pipeline that we can use with transformers\$pipeline(). The pipeline function has and argument task to specify the kind of NLP task to do.

2 Text classification

We can call the "text-classification" task without specifying a model. In this case, a default model (in English) will be used, which occupies 268MB on disk:

```
classifier <- transformers$pipeline(task = "text-classification")</pre>
```

As the warning messages indicate, it is recommended to indicate the model that you want to use. We can use the model we just obtained:

```
classifier("I have a serious problem")

[[1]]
[[1]]$label
[1] "NEGATIVE"

[[1]]$score
[1] 0.9993677

classifier("I am filling good")

[[1]]
[[1]]$label
[1] "POSITIVE"

[[1]]$score
[1] 0.999871
```

As you can see, this model (at least at the time of execution of this source code) classifies text as positive or negative (it's a binary classifier).

2.1 Now in Spanish

If you try to do it for Spanish using something like this:

You will get an error message like this:

```
Downloading: 100% | 613/613 [00:00<00:00, 574kB/s]

Error in py_call_impl(callable, dots$args, dots$keywords):

ValueError: Could not load model PlanTL-GOB-ES/roberta-base-bne with any of the following classes: (

<class 'transformers.models.auto.modeling_tf_auto.TFAutoModelForSequenceClassification'>,

<class 'transformers.models.roberta.modeling tf roberta.TFRobertaForMaskedLM'>.
```

This error is because some models are written in PyTorch, so we need to have PyTorch (and some other classes) installed. The easiest is to execute the following:

```
reticulate::py_install("pytorch-pretrained-bert", pip = TRUE)#torch (887MB) and others
```

After the installation of the python packages, the R session must be restarted so that the values of some python environment variables are updated.

To restart the R session, click on the RStudio menu: Session->Restart R.

If R is not restarted you will get the following error (we saw it before):

```
Downloading: 100% | 613/613 [00:00<00:00, 407kB/s]

Error in py_call_impl(callable, dots$args, dots$keywords):

ValueError: Could not load model PlanTL-GOB-ES/roberta-base-bne with any of the following classes: (

<class 'transformers.models.auto.modeling_tf_auto.TFAutoModelForSequenceClassification'>,

<class 'transformers.models.roberta.modeling_tf_roberta.TFRobertaForMaskedLM'>
).
```

IMPORTANT: Even if R is restarted, the downloaded models are NOT lost, so they could grow up quickly and require a lot of disk space. The same goes for installed python packages. Therefore, check from time to time the size occupied by:

- 1) HuggingFace models. On linux they are stored in /home/rstudio/.cache/huggingface/hub (with rstudio being the RStudio user).
- 2) the python packages. On linux they are stored in /home/rstudio/.virtualenvs/r-reticulate/lib/python3.8/site-packages (with rstudio being the RStudio user), with directories transformers, pytorch_pretrained_bert, etc., each with multiple .py scripts.

Now we can execute the following code (it takes a while to download the 600MB model):

Despite the intimidanting warning, we continue. This is the warning:

```
| 499M/499M [00:49<00:00, 10.1MB/s]
Downloading: 100%
Some weights of the model checkpoint at PlanTL-GOB-ES/roberta-base-bne were not used when
initializing RobertaForSequenceClassification: ['lm_head.decoder.weight', 'lm_head.bias',
'lm_head.dense.bias', 'lm_head.dense.weight', 'lm_head.decoder.bias',
'lm_head.layer_norm.weight', 'lm_head.layer_norm.bias']
- This IS expected if you are initializing RobertaForSequenceClassification from the
checkpoint of a model trained on another task or with another architecture (e.g. initializing
a BertForSequenceClassification model from a BertForPreTraining model).
- This IS NOT expected if you are initializing RobertaForSequenceClassification from the
checkpoint of a model that you expect to be exactly identical (initializing a
BertForSequenceClassification model from a BertForSequenceClassification model).
Some weights of RobertaForSequenceClassification were not initialized from the model
checkpoint at PlanTL-GOB-ES/roberta-base-bne and are newly initialized: ['classifier.out_proj.bias', 'c
'classifier.dense.weight']
You should probably TRAIN this model on a down-stream task to be able to use it for
predictions and inference.
Downloading: 100%|
                        | 1.39k/1.39k [00:00<00:00, 1.23MB/s]
Downloading: 100%|
                        | 851k/851k [00:00<00:00, 1.59MB/s]
Downloading: 100%| | 509k/509k [00:00<00:00, 1.03MB/s]
Downloading: 100%| | 2.21M/2.21M [00:00<00:00, 2.64MB/s]
Downloading: 100%| | 957/957 [00:00<00:00, 695kB/s]
We can use the model:
classifier("Tengo un problema serio")
[[1]]
[[1]]$label
[1] "LABEL_1"
[[1]]$score
[1] 0.53885
classifier("Me siento fenomenal")
[[1]]
[[1]]$label
[1] "LABEL_1"
[[1]]$score
[1] 0.5383498
Clearly, the model is not working properly.
This is because this model is prepared for another task: "fill-mask" (you can see the model documentation
here).
unmasker <- transformers$pipeline(task = "fill-mask",</pre>
                                    model="PlanTL-GOB-ES/roberta-base-bne")
unmasker("Gracias a los datos de la BNE se ha podido <mask> este modelo del lenguaje.")
```

[[1]]

[[1]]\$score [1] 0.08422067
[[1]]\$token [1] 3832
<pre>[[1]]\$token_str [1] " desarrollar"</pre>
<pre>[[1]]\$sequence [1] "Gracias a los datos de la BNE se ha podido desarrollar este modelo del lenguaje."</pre>
[[2]] [[2]]\$score [1] 0.06348325
[[2]]\$token [1] 3078
[[2]]\$token_str [1] " crear"
[[2]]\$sequence [1] "Gracias a los datos de la BNE se ha podido crear este modelo del lenguaje."
[[3]] [[3]]\$score [1] 0.0614842
[[3]]\$token [1] 2171
[[3]]\$token_str [1] " realizar"
[[3]]\$sequence [1] "Gracias a los datos de la BNE se ha podido realizar este modelo del lenguaje."
[[4]] [[4]]\$score [1] 0.05621832
[[4]]\$token [1] 10880
[[4]]\$token_str [1] " elaborar"
[[4]]\$sequence [1] "Gracias a los datos de la BNE se ha podido elaborar este modelo del lenguaje."

```
[[5]] $score
[1] 0.05133353

[[5]] $token
[1] 31915

[[5]] $token_str
[1] " validar"

[[5]] $sequence
[1] "Gracias a los datos de la BNE se ha podido validar este modelo del lenguaje."
```

3 Part of speech

You have a Part Of Speech neural model for Spanish: Warning! The model uses 1.4GB disc space.

entity	score	index	word	start	end
NOUN	0.9996961	1	ĠFestival	0	8
ADP	0.9996126	2	$\dot{\mathrm{G}}\mathrm{d}\mathrm{e}$	9	11
PROPN	0.9993528	3	$\dot{\mathrm{G}}\mathrm{San}$	12	15
PROPN	0.9884313	4	ĠSebastiÃjn	16	25
PUNCT	0.9493771	5	:	25	26
PROPN	0.9998086	6	$\dot{\mathrm{G}}\mathrm{Johnny}$	27	33
PROPN	0.9978847	7	$\dot{\mathrm{G}}\mathrm{De}$	34	36
PROPN	0.8299615	8	pp	36	38
VERB	0.9997770	9	Ġrecibirá	39	47
DET	0.9998410	10	Ġel	48	50
NOUN	0.9998456	11	$\dot{\mathrm{G}}\mathrm{premio}$	51	57
PROPN	0.9996608	12	\dot{G} Donostia	58	66
ADP	0.9998453	13	$\dot{\mathrm{G}}\mathrm{en}$	67	69
ADJ	0.8772469	14	$\dot{\mathrm{G}}\mathrm{pleno}$	70	75
NOUN	0.9996101	15	Ġrif	76	79
X	0.7009529	16	ir	79	81
X	0.6158468	17	ra	81	83
NOUN	0.9970248	18	fe	83	85

entity	score	index	word	start	end
ADJ	0.9998363	19	Ġjudicial	86	94
ADP	0.9998705	20	$\dot{\mathrm{G}}\mathrm{con}$	95	98
PROPN	0.9998962	21	$\dot{\mathrm{G}}\mathrm{Amber}$	99	104
PROPN	0.7110988	22	ĠНе	105	107
X	0.9438123	23	ard	107	110

4 Name Entity recognition (NER)

We will use the ner-plus model from PlanTL, that takes 500MB disc space. The ner model is heavier (1.5GB).

entity	score	index	word	start	end
B-OTH	0.9795215	1	ĠFestival	0	8
I-OTH	0.9771883	2	$\dot{\mathrm{G}}\mathrm{d}\mathrm{e}$	9	11
I-OTH	0.9894046	3	$\dot{G}San$	12	15
E-OTH	0.9754279	4	ĠSebastián	16	25
B-PER	0.9998310	6	ĠJohnny	27	33
E-PER	0.9994376	7	ĠDe	34	36
E-PER	0.9986013	8	pp	36	38
S-OTH	0.9964671	12	ĠDonostia	58	66
B-PER	0.9998432	21	$\dot{\mathrm{G}}\mathrm{Amber}$	99	104
E-PER	0.9997476	22	$\dot{\mathrm{G}}\mathrm{He}$	105	107
E-PER	0.9975891	23	ard	107	110