# **SEMINAR**

# CONFIGURATION AND RESULTS MANAGEMENT WITH MLFLOW, HYDRA ET POUTYNE

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### **OBJECTIVES OF THE PRESENTATION**

- Introduce configuration and results management tools.
- Developing good practices.
- Improve your productivity.

# **VOTRE CONFÉRENCIER**



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- Introduced to reproducible research in 2016 (R Markdown et git)
- Participation in REPROLANG of the LREC conference [Garneau et al., 2020]
- Active member in the development of a library to facilitate reproducibility (Poutyne ♥\*)

# ON THE MENU



Configuration management



Results management

The Management of a Project

```
001
        @experiment.config
        def config():
002
          seed = 42
003
004
          num runs = 10
          iteration = 0
005
          source_language = "en"
006
007
          target language = "de"
008
          src_input = "path" # The input source embeddings
          trg input = "2e path" # The input target embeddings
009
010
          other_input = "3e path" # Commentaire pas clair
```

*n*<sup>th</sup> parameters

395

Which one does that again?

Which one does that again?

Which ones necessarily go together?

Which one does that again?

Which ones necessarily go together?

Which ones are really essential?

Which one does that again?

Which ones necessarily go together?

Which ones are really essential?

How to organize them?

```
res_1.txt
res_2.txt
res_3.txt
res_4.txt
res_5_good.txt
res_5.txt
res_6_fix_a.txt
n<sup>th</sup> results file
```

Which configuration (already) used?

Which configuration (already) used?

Success or failure?

Which configuration (already) used?

 ${\tt Success\ or\ failure?}$ 

Which one is the best?

Which configuration (already) used?

Success or failure?

Which one is the best?

How to organize them?



# ON THE MENU



Configuration management



Results management

# **CONFIGURATION MANAGEMENT**



Simple and efficient

# **CONFIGURATION MANAGEMENT**



Simple and efficient



Facilitates experimentation

# **CONFIGURATION MANAGEMENT**





Facilitates experimentation



Scalable

# **HYDRA ☑**\*



Open source and MIT license



YAML structured configuration files



Hierarchical configuration files



Configurations sweeper

# STRUCTURED CONFIGURATION

conf
config.yaml
dataset
canadian.yaml
netherlands.yaml
embeddings
fast_text.yaml
model
bi_lstm_bidirectionnal.yaml
bi_lstm.yaml
lstm_bidirectionnal.yaml
lstm.yaml
optimizer
adam.yaml
SGD.yaml
•

# HIERARCHICAL CONFIGURATION

```
data_loader:
  batch_size: 2048
setting:
  seed:42
  device: "cuda:0"
defaults:
  - optimizer: SGD
  - model: bi lstm
  - dataset : canadian
```

- embeddings : fast\_text trainer : num\_epochs : 1 patience : 30

### HIERARCHICAL CONFIGURATION

optimizer: SGD

optimizer:

Ir: 0.1

type:sgd

# **EXAMPLE**

```
@hydra.main(config_path='conf/config.yaml')
def main(cfg):
    Ir = cfg.optimizer.Ir #0.1
```

### **CONFIGURATIONS SWEEPER**

python main.py -multirun task=1,2,3,4,5

python main.py -m 'main.x=int(interval(-5, 5))' 'main.y=interval(-5, 10)'

### **BONUS!**

- Automatic and customizable logging
- · Parametric instanciation

```
model:
_target_: models.LSTMNetwork
hidden_state_dim:300
num_hidden_layer:2
dropout:0.4
```

# **EXAMPLE**

```
log = logging.getLogger(__name__)
@hydra.main(config_path='conf/config.yaml')
def main(cfg):
    log.info("Init of the trainning")
:
    network = instantiate(cfg.model)
```

# **NEGATIVE POINT**

hydra.utils.get\_original\_cwd()

# ON THE MENU



Configuration management



Results management



Simple to use



Simple to use



**Experimental logging** 



Simple to use



**Experimental logging** 



Quick visualization of experiments

# **MLFLOW TRACKING ☑**\*



*Open source* and Apache 2.0 license



Automatic logging



Simple visualization



Integration with Poutyne

### **AUTOMATIC LOGGING**

- Code version (git)\*
- Training timestamp
- Training success/failure
- Computer configuration
- User

### SIMPLE VISUALIZATION

mlflow server -p 5000 -h 127.0.0.1 -backend-store-uri file:///absolute/path

#### SIMPLE VISUALIZATION

nlf/ow									
Listing	g Pri	ice Pre	diction						
Experiment ID: 0 Artifact Location: /Users/matei/milflow/demo/miruns/0									
Search Runs:		metrics.R2 > 0.24							Search
Filter Params:		alpha, Ir			Filter Metrics: rmse, r2				Clear
4 matching runs		Compare Selected Download CSV &							
					P	arameters		Metrics	
Tin	ne	User	Source	Version	alpha	I1_ratio	MAE	R2	RMSE
_ 17	:37	matei	linear.py	3a1995	0.5	0.2	84.27	0.277	158.1
□ 17	:37	matei	linear.py	3a1995	0.2	0.5	84.08	0.264	159.6
_ 17	:37	matei	linear.py	3a1995	0.5	0.5	84.12	0.272	158.6
_ 17	:37	matei	linear.py	3a1995	0	0	84.49	0.249	161.2

Figure 1 - Introducing MLflow: an Open Source Machine Learning Platform ♂\*

#### SIMPLE VISUALIZATION

- Sorting on experiments
- · Research of experiments
- · Queries on results
- · Export of results
- · Visualization of metrics

### INTEGRATION WITH POUTYNE \*\*

The "basic" version involves manual logging

- configuration parameters,
- metrics at each step and iteration,
- the version of the code.

## INTÉGRATION AVEC POUTYNE \*\*

The solution, MLFlowWriter, a callback allowing to journalize

- semi-automatically the configuration parameters,
- automatically the metrics at each step and iteration,
- automatically the version of the code,
- · manually a model,
- automatically test metrics during a test phase.

#### **EXAMPLE**

```
@hydra.main(config_path='conf/config.yaml')
def main(cfg):
    :
    mIflow_logger = MLFlowLogger(experiment_name="experiment")
    mIflow_logger.log_config_params(config_params=cfg)
    :
    mIflow_logger.log_model()
```

## **NEGATIVE POINT**

The documentation is not always easy to navigate.



## PRÉSENTATION DES RÉSULTATS



Automatic generation of tables



Dynamic report



Iterations of experiments

## TO GO FURTHER (IN ORDER)

- Training status notification Notif ✓\*
- Continuous Machine Learning (CML) ♂\*

## **QUESTIONS**



## SEMINAR

# THANK YOU FOR LISTENING!

#### REFERENCES i



Garneau, N., Godbout, M., Beauchemin, D., Durand, A., and Lamontagne, L. (2020).

A Robust Self-Learning Method for Fully Unsupervised Cross-Lingual Mappings of Word Embeddings: Making the Method Robustly Reproducible as Well.