# Tutorial: Setting up a Reinforcement Learning pipeline for a Telco Core Network (part 2)

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# Installing the Gym environment and pipeline

Operating System	Linux	macOS	Windows		
Repository access	Git command or download zip from GitHub				
Package installer and dependency management	pip , Miniconda or Anaconda				
Virtual environment	venv or Poetry				
Interpreter	CPython 3.9+				

A possible alternative is to use Google Colab, but you would miss the fun animation of the examples.





# Installing the Gym environment and pipeline

Requires around 10GB to install and compile all dependencies

#### Installation, follow instructions from README.md

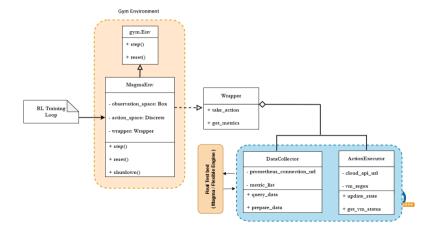
- git clone https://github.com/gfraysse/icin2024\_tutorial.git
- python -m venv <your virtual env>
- source <your virtual env>/bin/activate
- pip install -r requirements.txt





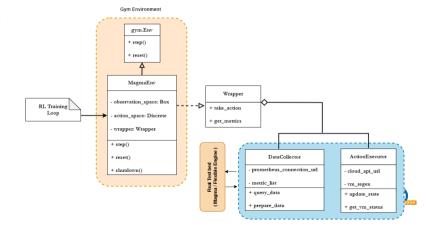
**Role**: provides standard interface for RL algorithms to interact with actual environment

- Gymnasium: Open-source python library for developing & comparing reinforcement learning algorithms
  - Provides standard API for learning algorithms and environment to communicate
  - Set of environments. E.g. Classic Control, Atari, MuJoCo etc.
  - Drop-in replacement of OpenAl Gym since 2021



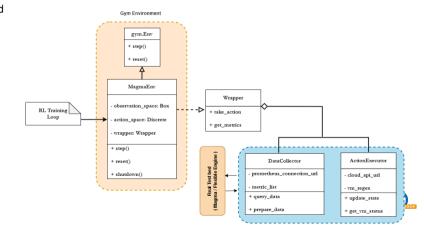


- MagmaEnv: Environment definition complying with Gym
  - a way to manage the environment while adhering to MDP
  - helps the learning algorithm to control the environment
  - implements the crucial methods like step() and reset()

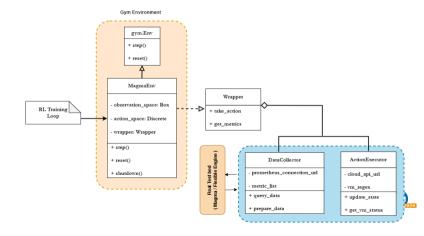




- step(): accepts an 'action' and transitions the environment to a new state. It then returns:
  - a new state: in which agent finds itself after executing the action
  - reward: feedback for taking the action
  - terminated: indicates whether environment reached terminal state.
  - info: general info about environment
- reset(): Moves the environment to an initial state.



 TelcoCoreScalingEnv: simulation environment mimicking Magma





## Finilizing the installation the Gym environment and pipeline

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## Running an experiment

#### Adjust the configuration

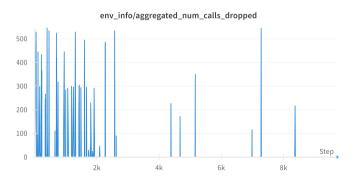
- Configuration of the pipeline config\_per.yaml
- Configuration of the environment telco\_core\_scaling/envs/config/config\_env.yaml

#### Run an experiment

python pipeline-exp\_d3qn\_per\_sim.py







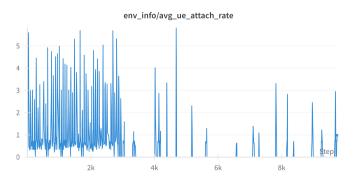
### Number of calls dropped:

- Computed by the pipeline
- Measures the number of sessions that could not be initiated





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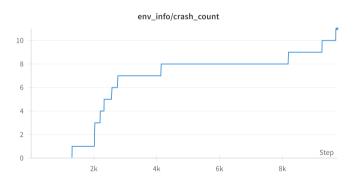


#### UE attach rate:

- Metric collected from Magma NMS
- Attach rate of the User Equipments (UEs) connecting to the Access Gateways (AGWs)







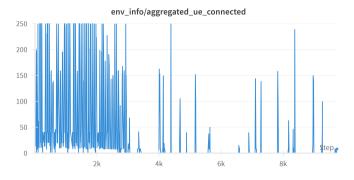
#### Crash count:

- Measured by the pipeline
- Detect when something is wrong with the environment during the training (usually a crash, or an issue when starting a new instance)
- Initiate a reset





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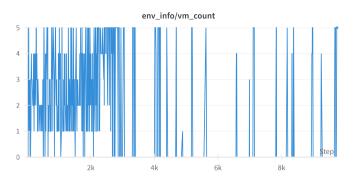
#### UEs connected:

- Metric collected from Magma NMS
- Number of UEs connected on all the AGWs currently running
- Different from Normalized UEs connected metric





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#### VM count:

- Metric collected from Magma NMS
- Measure the number of active instances of AGW(s)





### Reward function

Metric	Definition				
U	Number of UEs connected				
M	Memory usage, in MB				
D	Number of dropped sessions				

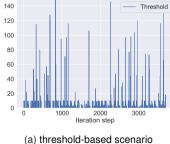
- Maximum reward value is 1
- Encourages the Network Functions (NFs) to use resources optimally around 70%
- Resource usage above 80% or crashes are penalized with lowest reward value -1.

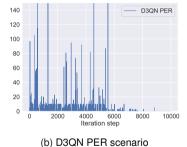
$$r = \begin{cases} 1 - (0.7 - \max(M, U) - D) & \text{if M, U} \in [0, 80] \\ \max(-\max(M, U) - 10 * D, -1) \text{otherwise.} \end{cases}$$

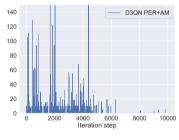




## Number of sessions dropped







cenano (b) D3QN PER scenan

(c) D3QN PER+AM scenario

Evolution of the number D of dropped sessions, during the three experiments





## Average values of metrics during the experiments

Experiment	Steps	DUR	Metric (average)							
		(h)	U	M	P	C	D			
buffer filling + $\epsilon$ decay + pure exploitation										
D3QN PER	10k	154	345	78	5.76	3.25	0.60			
D3QN PER+AM	10k	187	368	59	4.81	3.96	0.49			
$\epsilon$ decay + pure exploitation										
D3QN PER	6k	63	382	61	4.48	3.57	0.25			
D3QN PER+AM	6k	76	408	41	3.11	4.63	0.12			
pure exploitation										
D3QN PER	4k	29	407	51	3.65	3.80	0.03			
D3QN PER+AM	4k	34	420	33	2.60	4.84	0.01			
Threshold-based	4k	145	247	119	9.72	2.16	1.73			





### Lessons learned

Reinforcement Learning (RL) is complex:

- Not your typical network or software engineer skill
- Very active area of research







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#### Automation of Network Core scaling:

- Development is required
- Probably better to use Infrastructure as Code framework to be laaS-independent





### Lessons learned

#### RL is complex:

- Not your typical network or software engineer skill
- Very active area of research

#### Automation of Network Core scaling:

- Development is required
- Probably better to use Infrastructure as Code framework to be laaS-independent

#### Load generation is always tricky:

- Commercial products exist
- Lack of open source tools to generate traffic in a consistent way for a long time
- Traffic needs to be balanced across all instances





Q & A

• Any questions left ?