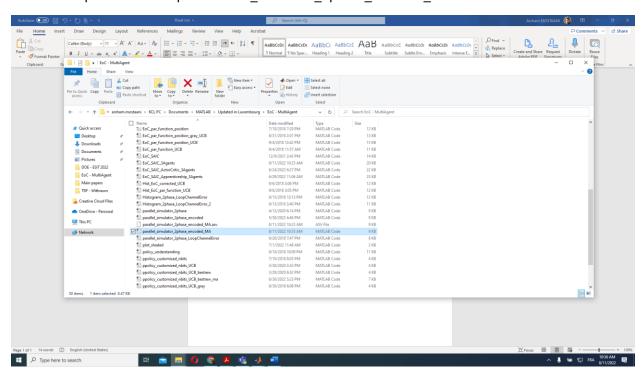
## How to adjust the setting for multiagent simulations

By Arsham Mostaani – Affiliated with SnT - Supervised by Bjorn Ottersten – Date: 11/08/2022

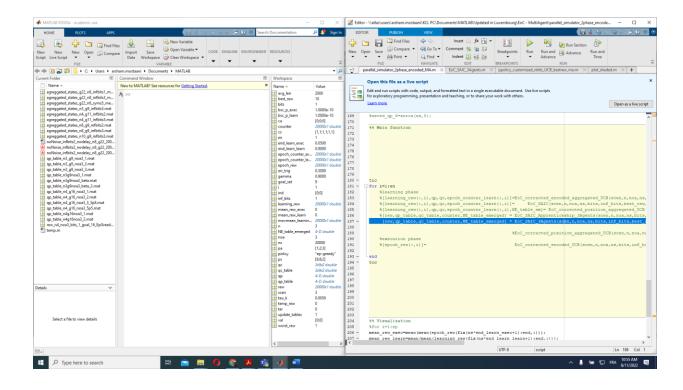
## ESAIC - SAIC

## Finding the right functions and scripts

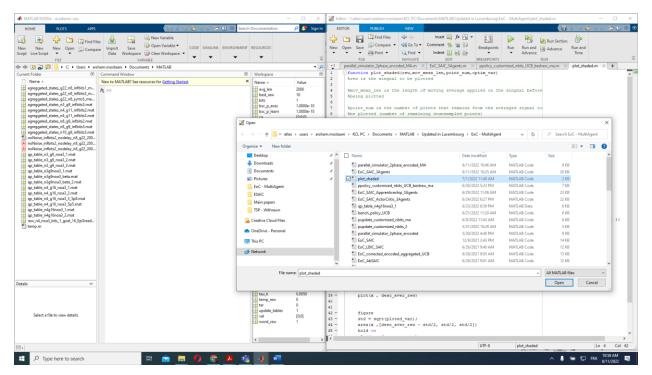
1- The parameter setup is "Parallel\_simulator\_2phase\_encoded\_MA"



- The parameter setup is control panel that gives the user all sorts of control facilities to adjust the simulations to their needs.
- 2- "EoC\_SAIC\_3Agents" is a function wrapper that encapsulates all functions that interact with each other to completely simulate one episode of training.



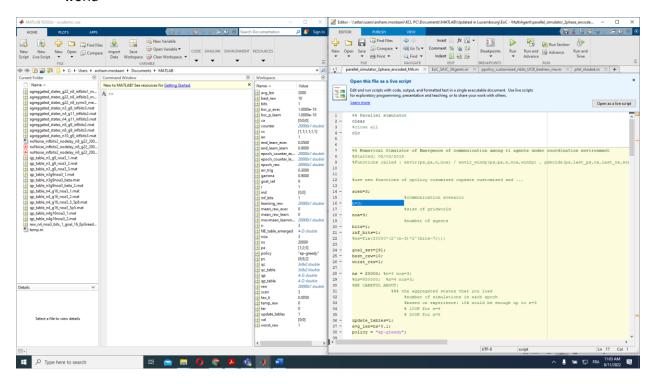
3- To have nice looking simulation results you can also use the "plot\_shaded" function, which plots the reward signal together with a shaded area of its variance. This is done such that the reward variance through a time window of the existing epoch is depicted in the form of the width of the shades – find out more in the code details and comments.



4- All these scripts and functions are available in the current repository:

## Adjustment of the parameter setup

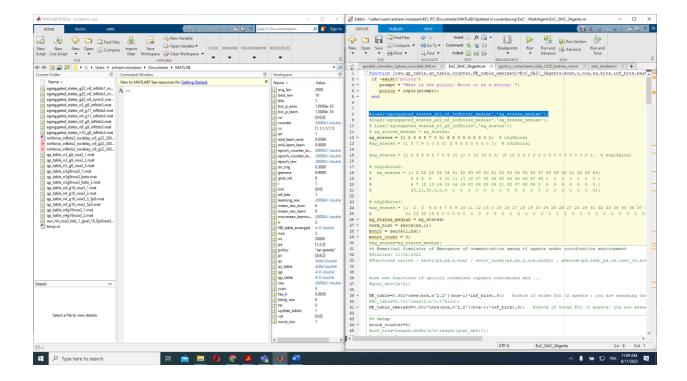
1- Select the number of grids in each direction e.g., n=3 is selected when you want a 3x3 grid world



- 2- Select the number of agents with the parameter noa (standing for the Number of Agents) noa=3;
- 3- Select the number of communication bits in the case of task-oriented quantization bits and inf\_bits should have similar values

```
bits=1;
inf_bits=1;
```

4- According to the size of the grid-world and the number of quantized bits, load the right quantizer inside the wrapper, or obtain your quantizer by running the centralized training phase



5- Select the right goal set/location

6- Update tables = 1 – This parameter should only be zero when using Apprenticeship learning wrapper.

```
update_tables=1;
```

7- Choose the right policy – we have obtained very good results with decaying epsilon greedy. Decaying epsilon greedy or Linearly anealing epsilon, has been detailed also in David Silver's seminal work "Mnih, Volodymyr, et al. "Human-level control through deep reinforcement learning." *nature* 518.7540 (2015): 529-533.":

```
policy = "ep-greedy"
```

8- Check your rewarding scheme – the default setting is adjusted according to our paper: "Task-Oriented Data Compression for Multi-Agent Communications Over Bit-Budgeted Channels".

Accordingly, the common reward signal increases exponentially as the number of agents arrived at the goal point are increased.