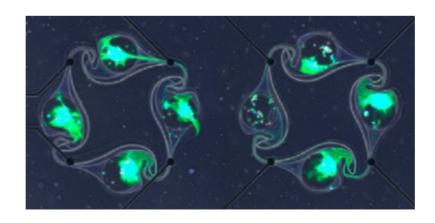
Reinforcement learning and biological networks

Stephan J. Ihle

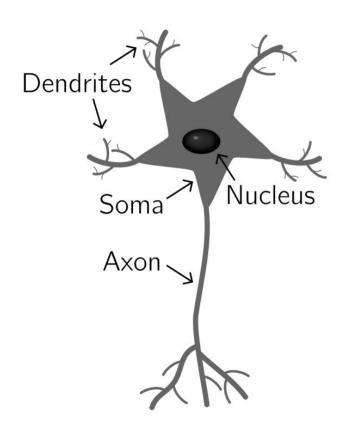
Eric and Wendy Schmidt AI-Postdoctoral Fellowship
Biological Science Department
Physical Science Department
Data Science Institute University of Chicago





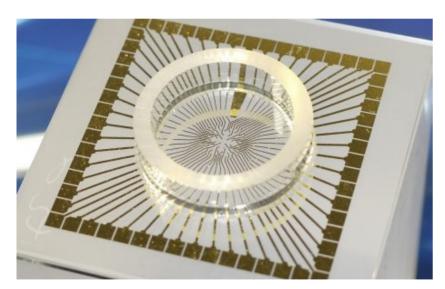


How neurons work



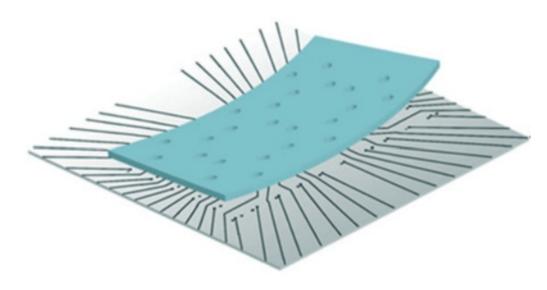
Culturing neurons in a dish

Microelectrode array for neuronal recording



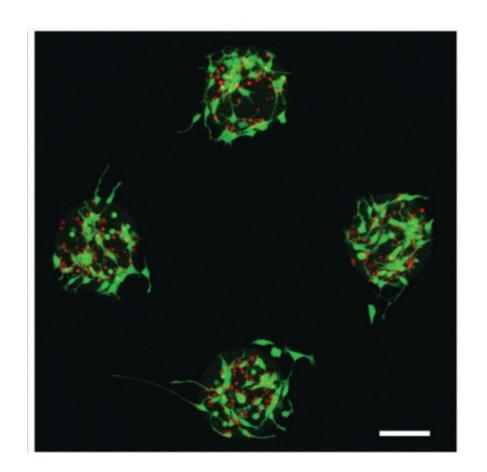
multichannelsystems.com

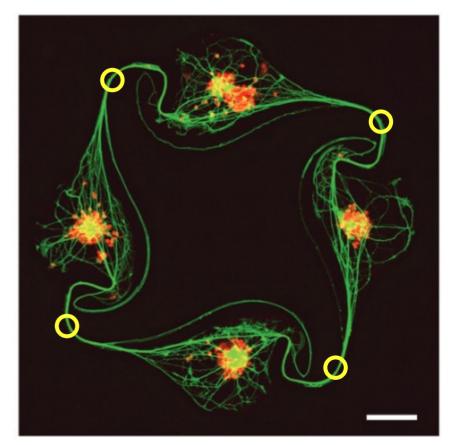
Microstructure to constrain growth



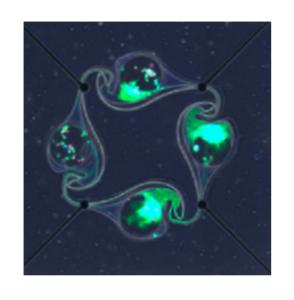
Forró et al., Biosensors and Bioelectronics, 2018

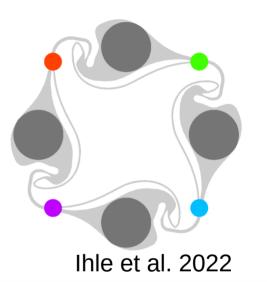
Culturing neurons in a dish

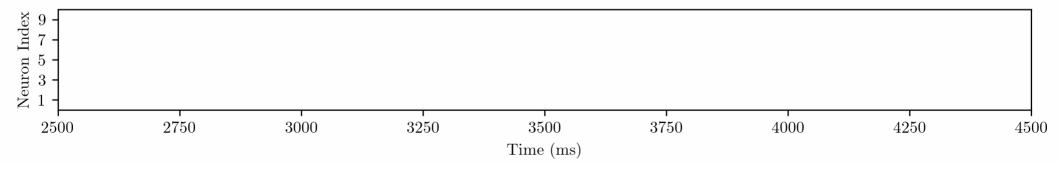




Measuring neuronal activity

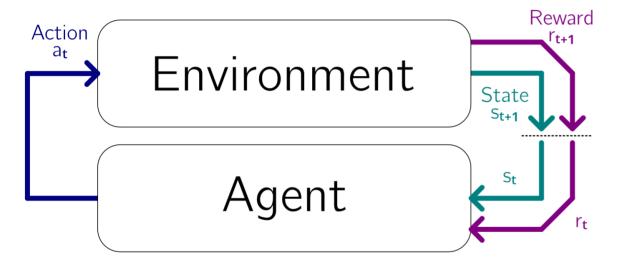






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Reinforcement learning

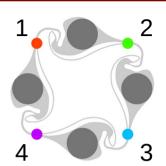


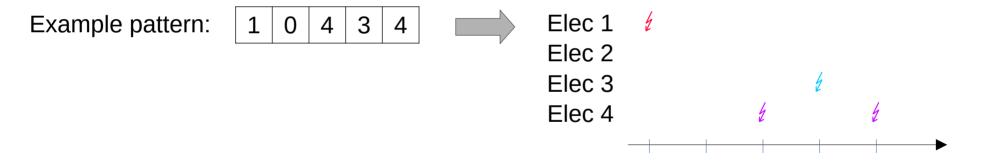
The network is stimulated every 250 ms

What actions should you use?

You have up to 5 slots per action.

Each slot can have value 0 (no stimulus), or numbers 1,2,3,4





Advice: Especially at the beginning, do not use all 5 slots, but only the last 1,2, or 3

What states should you use?

There are different state functions.

Reward functions can be static (easier) or dynamic (harder). Dynamic functions need to be trained on the data but contain in general more useful information.

There exists 2 jupyter notebook examples with these two types of functions.

Attention: Your state representation may not be complete

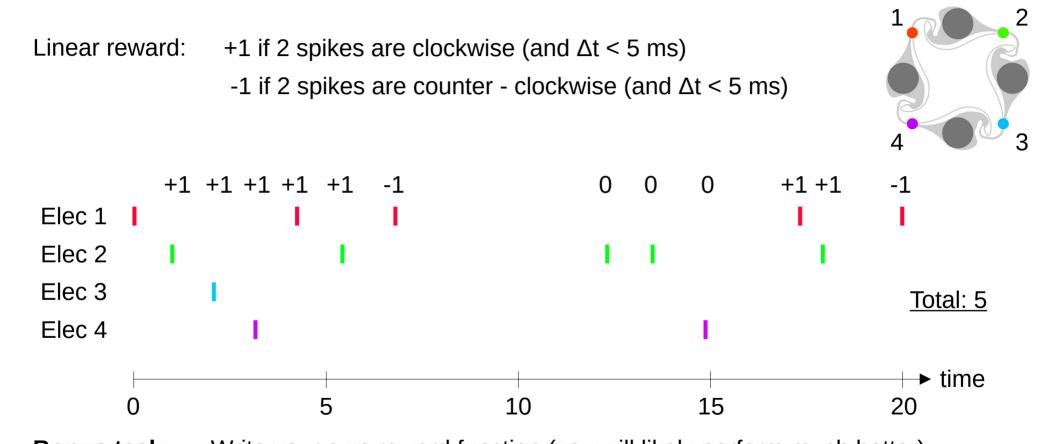
Your state is not necessarily Markovian

Bonus task: Write your own state function (you will likely perform much better)

Can you figure out an approach to use the spiking activity directly as a state?

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What reward function should you use?



Bonus task: Write your own reward function (you will likely perform much better)

Your final score:

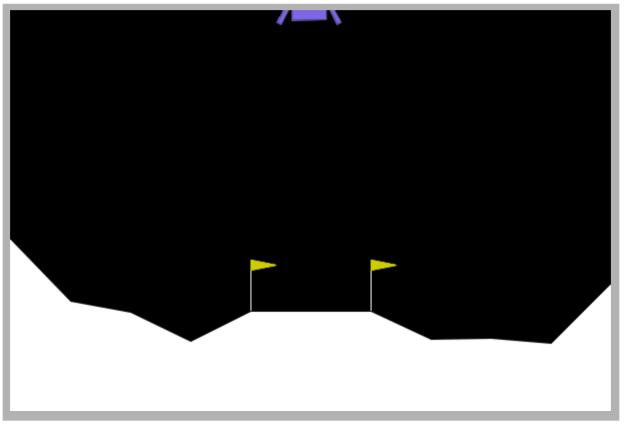
Send me your code by 5pm on XX/XX: ihles@uchicago.edu

We will run your code directly on our setup. It will run for exactly 2 hours on a network (28800 actions). The final score will be your average reward over the last 30 min of this time period (7200 actions).

For the reward function, we are using: "linear_reward()"

Important consideration: Use RL gymnasium

https://gymnasium.farama.org/index.html



Important consideration: Use RL gymnasium

```
import gymnasium as gym
# Initialise the environment
env = gym.make("LunarLander-v3", render mode="human")
# Reset the environment to generate the first observation
observation, info = env.reset(seed=42)
for in range(1000):
    # this is where you would insert your policy
    action = env.action space.sample()
    # step (transition) through the environment with the action
    # receiving the next observation, reward and if the episode has terminated or truncated
    observation, reward, terminated, truncated, info = env.step(action)
    # If the episode has ended then we can reset to start a new episode
    if terminated or truncated:
        observation, info = env.reset()
env.close()
```

Important consideration: We are data limited

At the end, you will have **28800 stimulation events** (actions) in total (of which your performance is judged on the last 7200).

Action space size: $5^5 = 3125$

Action space is huge for RL

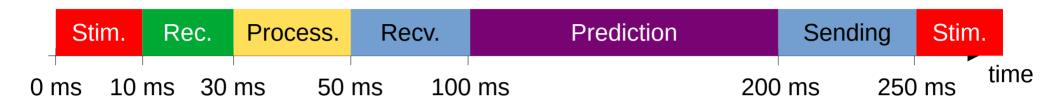
Rule of thumb: Use small networks!

Number of weights in network should be sqrt(samples) → 170

Use experience replay: https://deeplizard.com/learn/video/Bcuj2fTH4_4

https://arxiv.org/abs/2007.06700

Important consideration: We are time critical



Stimulation: Requires 10 ms

Recording: We look at 20 ms of neuronal activity

Processing: Spike detection / clustering takes 20 ms

Receiving data: From laboratory computer to RCC (50 ms)

Prediction: You have roughly 100 ms to decide on your next action

Sending data: From RCC to laboratory computer (50 ms)

Use 1 network to decide on your next action and (at least) one network for training!

Why/how to use multiple networks: https://pytorch.org/tutorials/intermediate/reinforcement_q_learning.html

Happy Hacking!

Questions?