

Simulating Language: Lab 5 Worksheet

This simulation introduces the idea of *learning* a signalling system. In the previous simulations, an individual agent's signalling system was provided innately, and didn't change. Populations of agents evolved through natural selection according to the fitness function we specified, to be 'optimal' in some way for communication.

In this simulation, we're ignoring evolution, and instead allowing the weights in an agent's signalling system to change through learning, as a result of their experiences. We'll be using `learning1.py`, which you can download from the website in the usual way.

The first section of the code is similar to the code we used in our first simulation (`signalling1.py`), when we introduced the following:

- a signalling system represented as a list of lists - you can think of this as a matrix or as a neural network.
- how to choose a signal to express a meaning;
- how to decide which meaning a signal is expressing;
- communication as a measure of how well the speaker's meaning matches the hearer's meaning after being transmitted via a signal.

These should all be very familiar by now. We have made one major change to the code though. In `signalling1.py` we had separate matrices for reception and production. From now on we are going to use a model where we just have a single matrix which handles both processes. There are some small changes to the code to accomplish this.

Identify the changes required to go from a two-matrix model to a one-matrix model, and figure out why they have been made.

Learning

In learning, agents remembering the association between the meaning and signal. We need one simple function to implement learning. The function **learn** takes three parameters and just two lines of code:

- a signalling system
- a meaning
- a signal

The function finds the appropriate cell in the signalling system matrix indexed by the meaning and signal, and adds one to the value of the weight in this cell.

```
# ----- new code below -----  
  
def learn(system, meaning, signal):  
    system[meaning][signal] += 1
```

```
>>> s = [[0,0,0],[0,0,0],[0,0,0],[0,0,0]]
>>> learn(s,0,2)
>>> learn(s,1,1)
>>> learn(s,0,2)
>>> learn(s,3,0)
>>> s
[[0,0,2],[0,1,0],[0,0,0],[1,0,0]]
```

Make sure you understand how this learning function works, what the parameters mean, and how the function updates the correct cell in the matrix.

Enter the code in the box and try it out.

Create a signalling system, then modify it by learning some random meaning-signal pairs.

Make sure you understand how and why the weights in the matrix have changed.

Training

Rather than input each learning episode individually, we can give an agent a list of meaning-signal pairs, and learn them all through the single function **train**. This function goes through each item in the list, and learns each meaning-signal pair individually.

```
def train(system, word_list):
    for ms_pair in word_list:
        learn(system, ms_pair[0], ms_pair[1])
```

Create a signalling system, then provide it with a list of learning exposures and check that the system has learnt from the data you gave it.

Questions

1. How good is this model of learning? What does 'good' mean for a model of learning? How can you test it?
2. Can you write some code to test how well an agent has learnt a language?
3. Learning is implemented as a frequency count of associations. Are there other reasonable ways of updating the matrix based on observed meaning-signal pairs?