# NEURAL SYSTEM SIMULATION – IMPLEMENTATION PLAN

**AGENT SYSTEMS** 

### GENERAL IDEA

- Representation of a group of interconnected neurons
- Neurons modeled as agents
- Connections between neurons are formed dynamically
- Additional agent types include Sensors and ReceptoryFields
- Technology: MASON

### REMAINDER: MASON UNDERSTANDING OF AGENTS

#### In MASON:

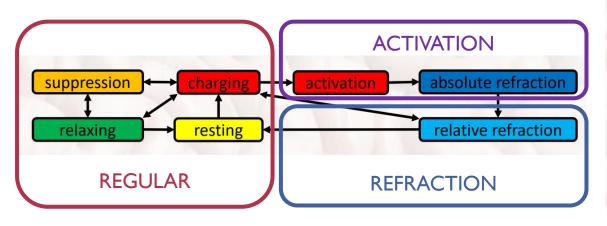
- An agent is an object that contains the step(SimState) method and is (can be) scheduled for execution
- MASON holds a global schedule of upcoming agents' activations
- At each step MASON fires the agent that's sheduled to execute at the earliest time (multiple agents if ties occure)
- Upon execution, agent's step(SimState) method is called; an agent has access to global simulation state and consequently to all other agents

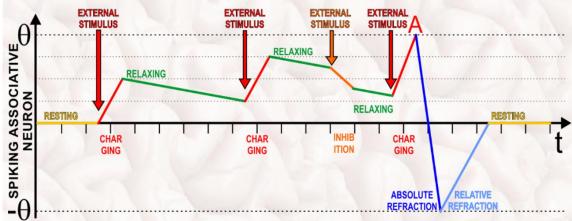
# MAIN AGENT TYPE

ASSOCIATIVE SPIKING NEURON

### ASSOCIATIVE SPIKING NEURON

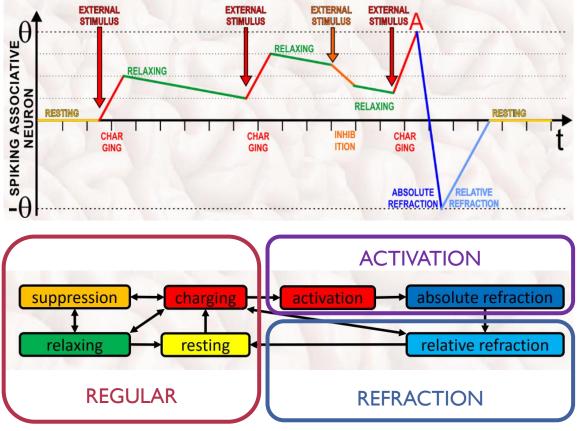
Neuron maintains internal excitation level, influenced by external stimulation and passage of time; it can be described as a state machine (proposed model involves a simplified verison of states space, with states grouped into three distinct groups, marked on figure below).





### ASSOCIATIVE SPIKING NEURON

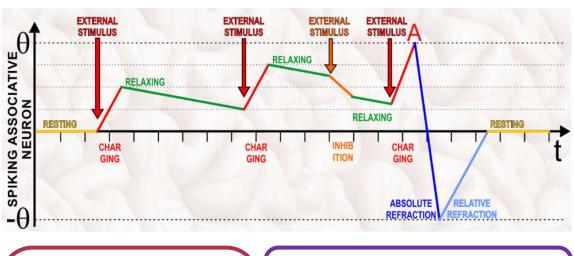
- Neuron starts simulation in REGULAR state with no stimuli on inputs – it's predicted activation time equals infinity
- An external stimuli (from other agents neurons and sensors – or simulation scenario) may cause neuron to expect activation
- When neuron's input state changes one of it's neighbours is either activated or it's activation ends – several things happen:
  - Neuron updates it's excitation level based on past input state and time passed
  - Neuron's predicted activation time is recomputed (with the assumption that the current input state will be constant). If it turns out to be finite, neuron is scheduled for activation after computed time.
  - If neuron was already scheduled for activation (based on past, outdated input state) this old event is descheduled.

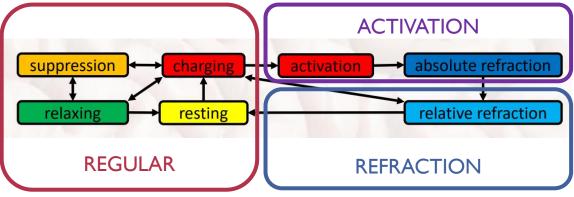


Images source: <a href="https://home.agh.edu.pl/~horzyk/lectures/ci/Cl-AssociativeNeuralGraphs.pdf">https://home.agh.edu.pl/~horzyk/lectures/ci/Cl-AssociativeNeuralGraphs.pdf</a>, slide 73

### ASSOCIATIVE SPIKING NEURON

- After neuron reaches activation, it stimulates all it's neighbours and schedule an end of activation event (it enters the ACTIVATION state)
- When end of activation event occures, neuron notifies all it's neighbours (they are forced to recompute their estimated activation time) and schedules itself for end of refraction (it enters the REFRACTION state)
- When end of refraction event occures, neuron resets it's internal excitation level and enters the REGULAR state
- In ACTIVATION and REFRACTION states
   neuron mostly ignores stimuli changes it does
   update it's input state, but doesn't update excitement
   nor schedule new activations





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```
enum NeuronState:
                  # encompasses RESTING, CHARGING, RELAXING and SUPPRESSION
    REGULAR,
    ACTIVATED,
    REFRACTED
base agent Neuron:
    neighbours = []
                       # neurons that are OUTPUTS of current neuron (that react to it's activation)
    current input excitations = {}
                                     # { neighbour: (weight, start time) }
    state = REGULAR
    excitation = 0.0
    last_update_time = None
    nearest activation = None
     ... # Neuron's methods are presented on the following slides
```

```
# base agent Neuron:
       step():
             # step() is called in three cases: when neuron is activated, when it activation ends (refraction starts) and when refraction period ends
             if state == REGULAR:
                                       # neuron was in regular state and has just been activated
                   for neigh in neighbours:
                          neigh.start input excitation(self)
                   state = ACTIVATED
                   self.schedule(ACTIVATION TIME)
                                                           # activation time is constant, it may be one of experiment parameters
             elif state == ACTIVATED: # activation has just ended and neuron should start refraction
                   for neigh in neighbours:
                          neigh.stop_input_excitation(self)
                   state = REFRACTED
                   self.schedule(REFRACTION TIME) # refraction time is constant, it may be one of experiment parameters
             else: # state == REFRACTED; refraction has ended and neuron has just returned to regular state
                   excitation = 0.0
                   for input in current input excitations:
                          weight, = current input excitations[input]
                          current input excitations [input] = (weight, curr time())
                   state = REGULAR
                   reschedule activation()
```

```
# base agent Neuron:
      start_input_excitation(neighbour):
     # called by neighbour neuron when it's activation starts
            current_input_excitations.put(neighbour, (WEIGHTS[neigh, self], curr_time())
                  #WEIGHTS are read from global simulation state (it's a property of connection between neurons)
            if state == REGULAR:
                  excitation = new excitation()
                  reschedule activation()
      stop_input_excitation(neighbour):
            # called by neigbour neuron when it's activation ends
            current_input_excitations.remove(neighbour)
            if state == REGULAR:
                  excitation = new excitation()
                  reschedule_activation()
```

```
# base agent Neuron:
```

#### new\_excitation():

# returns current excitation of neuron based on last known excitation value, current inputs and time passed

- # inputs must not change between subsequent calls to new excitation()
- # function sets last update time to curr time each time it is called
- # the exact formula for return value is the following:

$$excitation_{curr} = \begin{cases} excitation_{last} + \sum w_i * (t_{curr} - t_{last}), & curr_{inputs} \neq \emptyset \\ excitation_{last} - signum(excitation_{last}) * relaxation(excitation_{last}, t_{curr} - t_{last}), & w.p.p \end{cases}$$

#### compute\_activation\_time():

# returns time at which activation would occure if all the inputs were steady

# the exact formula for return value is the following:

$$t_{rel} = \frac{threshold - excitation_{curr}}{\sum w_i}$$

#### reschedule\_activation():

if nearest\_activation is not None:

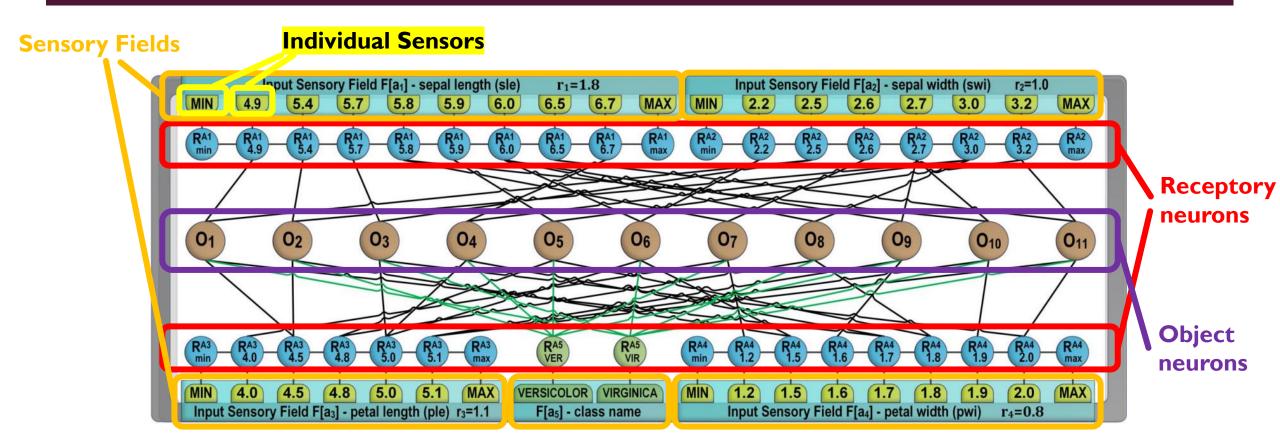
nearest activation.deschedule()

activation\_time = compute\_activation\_time()
self.schedule(activation\_time)

# **NEURONS ORGANIZATION**

**ACTIVE ASSOCIATIVE NEURAL GRAPH** 

### ACTIVE ASSOCIATIVE NEURAL GRAPH



# SPECIFIC NEURON TYPES

AGENTS DERIVED FROM BASE NEURON AGENT

### RECEPTORY NEURON

- Connects to exactly one Sensor (in), two neighbouring Receptory Neurons (in/out) and one or more Object
   Neuron (in/out)
- Adheres to the Plasticity Rule (more details can be found further into the presentation)
- Weights of connections to other Receptory Neurons are computed after the formula:

$$w_{ij} = 1 - \frac{|v_i - v_j|}{r}$$

Weights on connections to Object Neurons are computed after the following formula:

$$w_i^a = \frac{count(rn_i)}{count(a)}$$

# OBJECT NEURON

- Connects to exactly  $n_{attr}$  Receptory Neurons (in/out)
- Weights of connections to Receptory Neurons are constantly equal to 1.0

# **OTHER AGENTS**

AGENTS **NOT** DERIVED FROM BASE NEURON AGENT

### RECEPTORY FIELD

- Represents a single receptor/sense
- Modelled as agent for convenience it's scheduled activations form the Simulation Scenario
- It's job is to stimulate all it's associated Sensors at given times with given values

### SENSOR

- Represents specific value/range received from Receptory Field
- Sensor's excitement level when stimulated by sensory field is determined by difference between value represented by Sensor and value "observed" by Receptory Field:

$$x_{s_i} = |v_{rf} - v_i|$$

- Each Sensor is connected to a single Receptory Neuron
- Every sensor stimulates it's Receptory Neuron immediately (it does not have an activation threshold)

## DYNAMIC STRUCTURE CHANGES

CREATING OF NEURONS AND CONNECTIONS

### **RULES**

- **New patterns:** every time the network is presented with an unknown pattern (on it's **Receptory Fields**) that does not activate any **Object Neuron** within given time (simulation parameter), a new **Object Neuron** is created, connected to the first activated **Receptory Neuron** associated with each **Receptory Field**.
- Plasticity Rule: every time a Receptory Field "observes" value that is not represented by any existing Sensor, a new Sensor and Receptory Neuron are created to represent it. Then, when an existing Receptory Neuron detects that it is stimulated stronger by it's own Sensor than by it's neighbouring Receptory Neuron, it disconnects with it and connects with the newly created Receptory Neuron.