

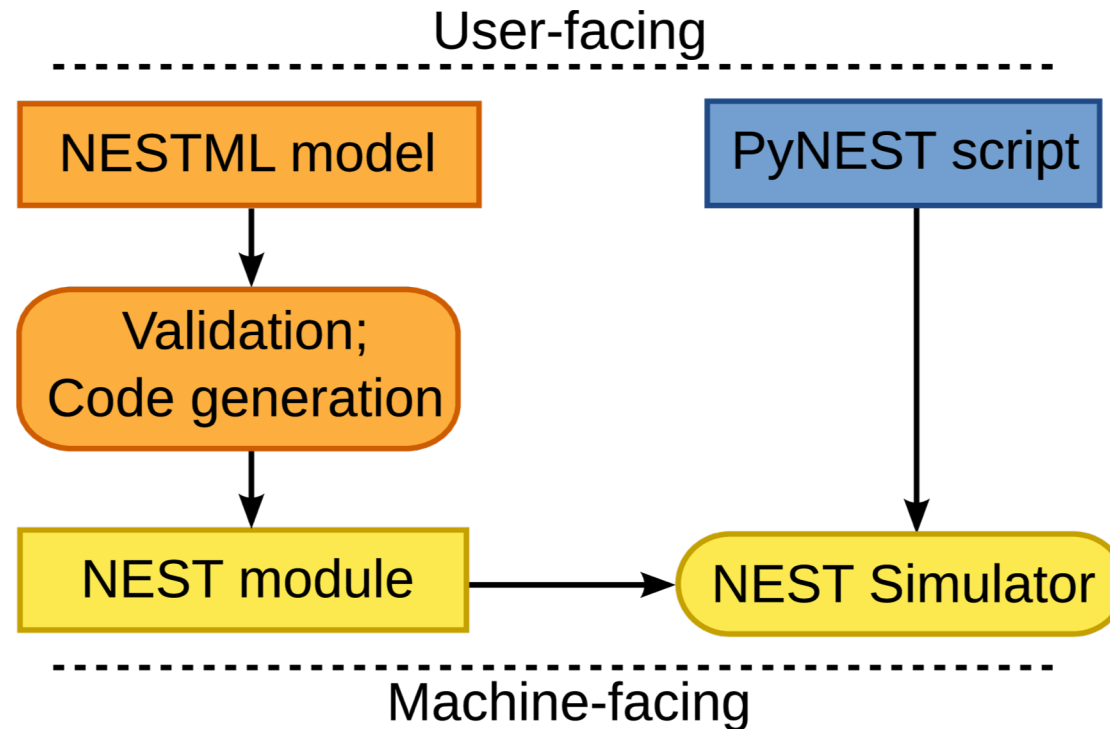
NESTML Tutorial

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Introducing NESTML

NESTML is a domain-specific language for neuron and synapse models.



Using PyNEST, you instantiate and connect the models that you define in NESTML.

Introducing NESTML



- Concise; low on boilerplate
- Speak in the vernacular of the neuroscientist (keywords such as `neuron`, `synapse`)
- Easy (dynamical) equation handling coupled with imperative-style programming (`if V_m >= V_threshold: ...`)

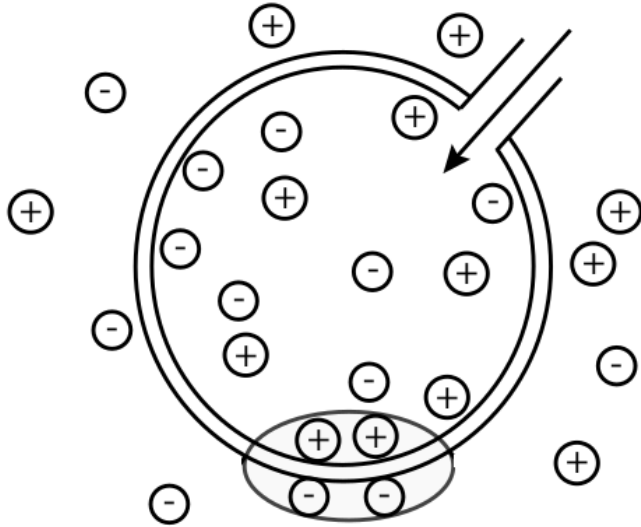
NESTML comes with a **code generation** toolbox.

- Code generation (model definition but not instantiation)
- Automated ODE analysis and solver selection
- Flexible addition of targets using Jinja2 templates

Mapping biological neurons to NESTML

Model

neuron-model based on an RC-circuit



NESTML

<<rc_neuron.nestml>>

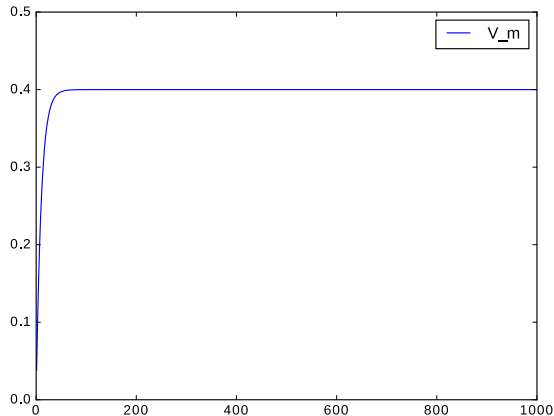
neuron rc_neuron:

end

Mapping biological neurons to NESTML

Model

$$\frac{dV_m}{dt} = -\frac{V_m}{\tau_m} + \frac{I_{syn}}{C_m}$$



NESTML

<<rc_neuron.nestml>>

neuron rc_neuron:

initial_values:

V_m mV = 0 mV

end

equations:

$V_m' = -V_m/\tau_m + I_{syn}/C_m$

end

parameters:

values taken from experiments

C_m pF = 250 pF

τ_m ms = 10 ms

I_{syn} pA = 10 pA

end

update:

integrate_odes()

end

end

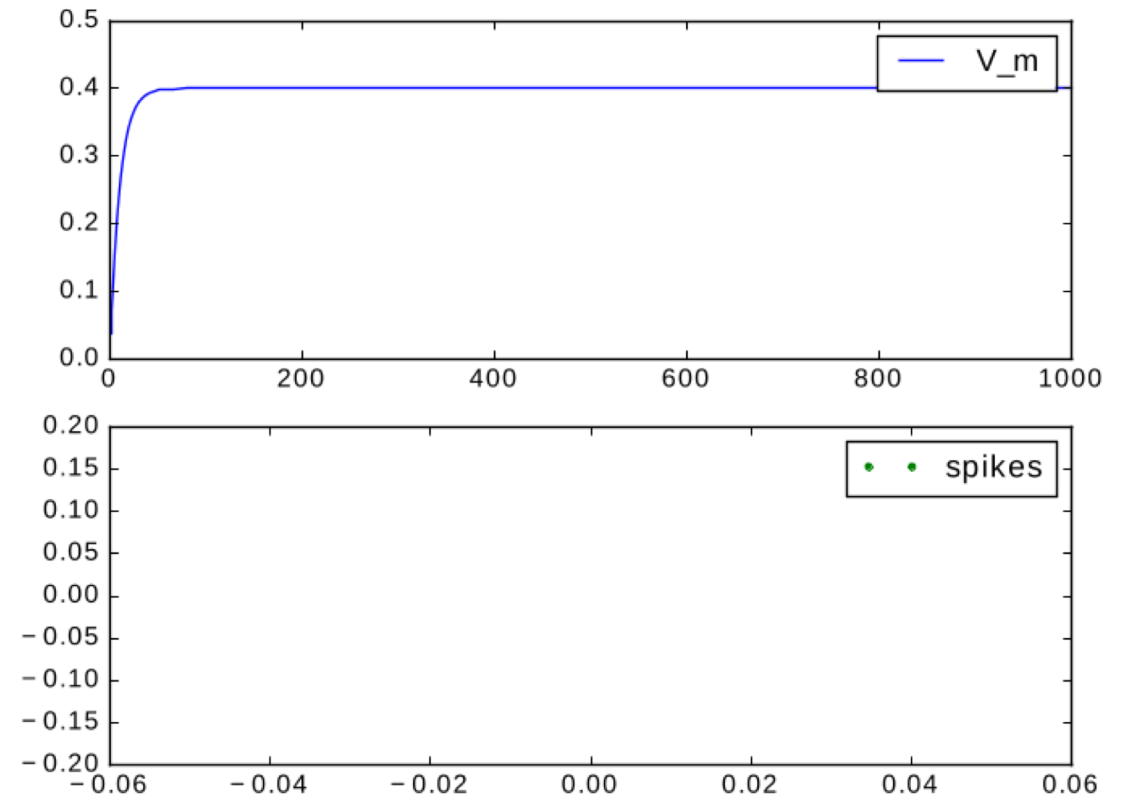
Simulating `rc_neuron`

NEST
<<Runtime>>

- Simulating `rc_neuron` for 1000 ms with constant input current of 10 pA

→ Strictly positive membrane potential

→ No spikes



Adding the resting potential E_L

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<<Runtime>>

- Shift V_m by E_L :

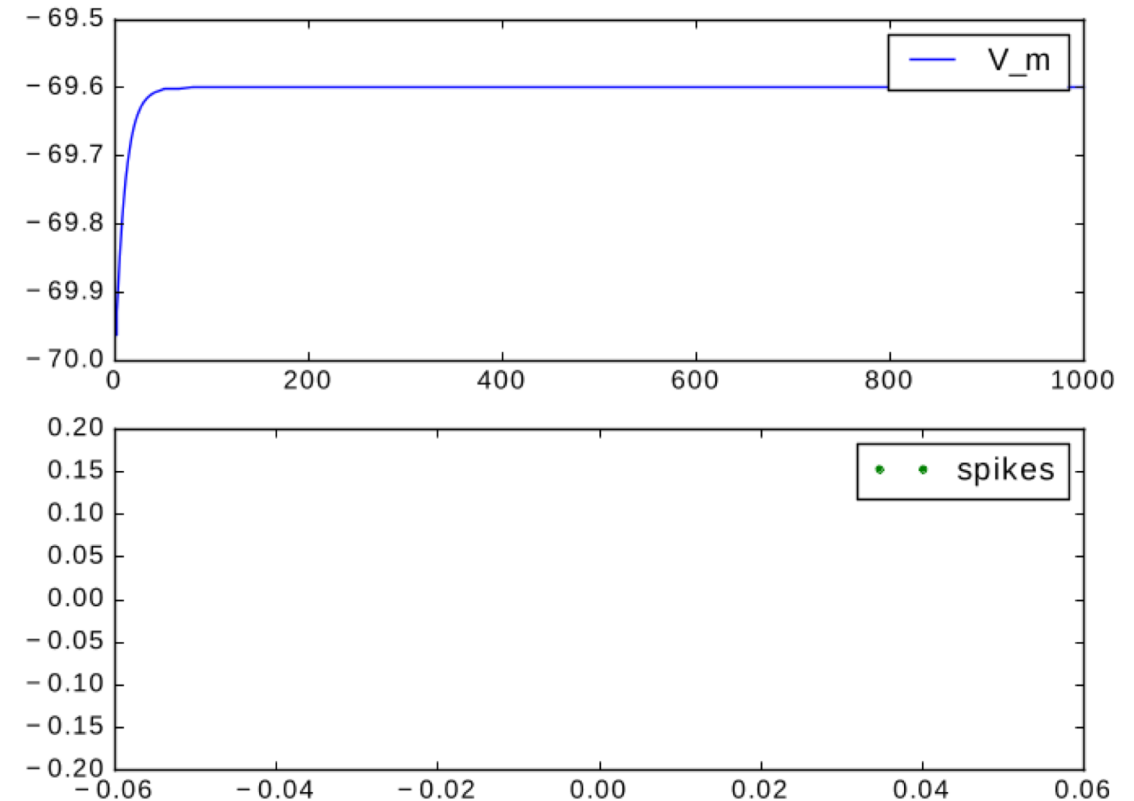
```
neuron rc_neuron_rest: NESTML
  initial_values:      <<rc_neuron_rest.nestml>>
     $V_m \text{ mV} = E_L$ 
  end

  equations:
     $V_m' = -(V_m - E_L) / \tau_m + I_{\text{syn}} / C_m$ 
  end

  parameters:
     $E_L \text{ mV} = -70 \text{ mV}$ 
  end

  ...
end
```

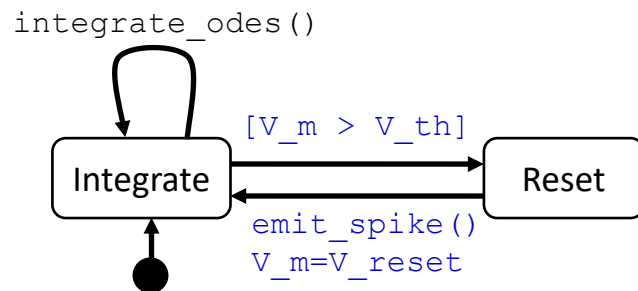
→ Still no spikes



Spiking and reset

Model

AD

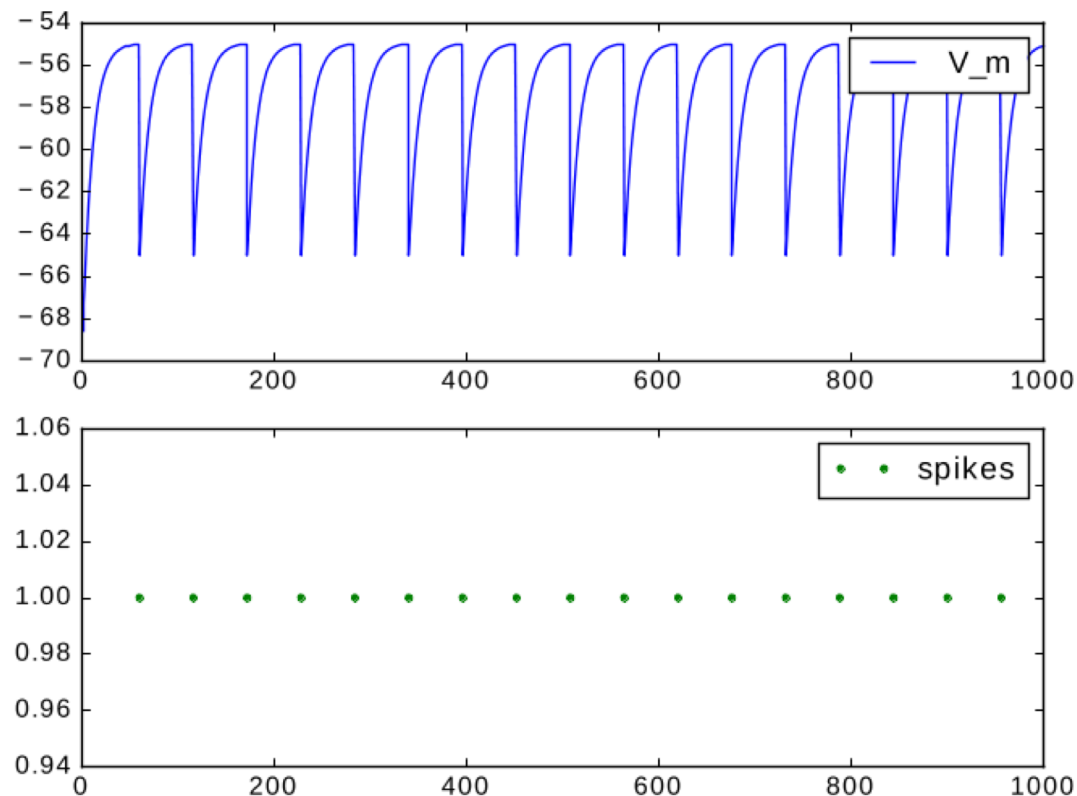


```

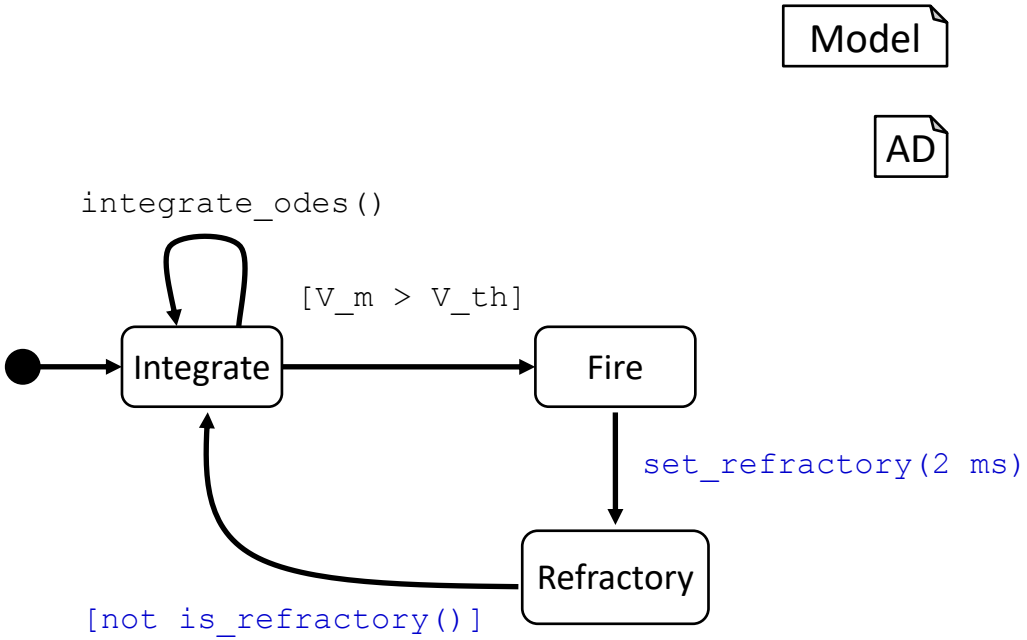
neuron rc_fire:
  parameters:
    V_th mV = -55 mV - E_L <<rc_fire.nestml>>
  end
  update:
    integrate_odes()
    if V_m >= V_th:
      V_m = V_reset
      emit_spike()
    end
  end
  ...
end
  
```

NESTML

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<<Runtime>>



Refractoriness



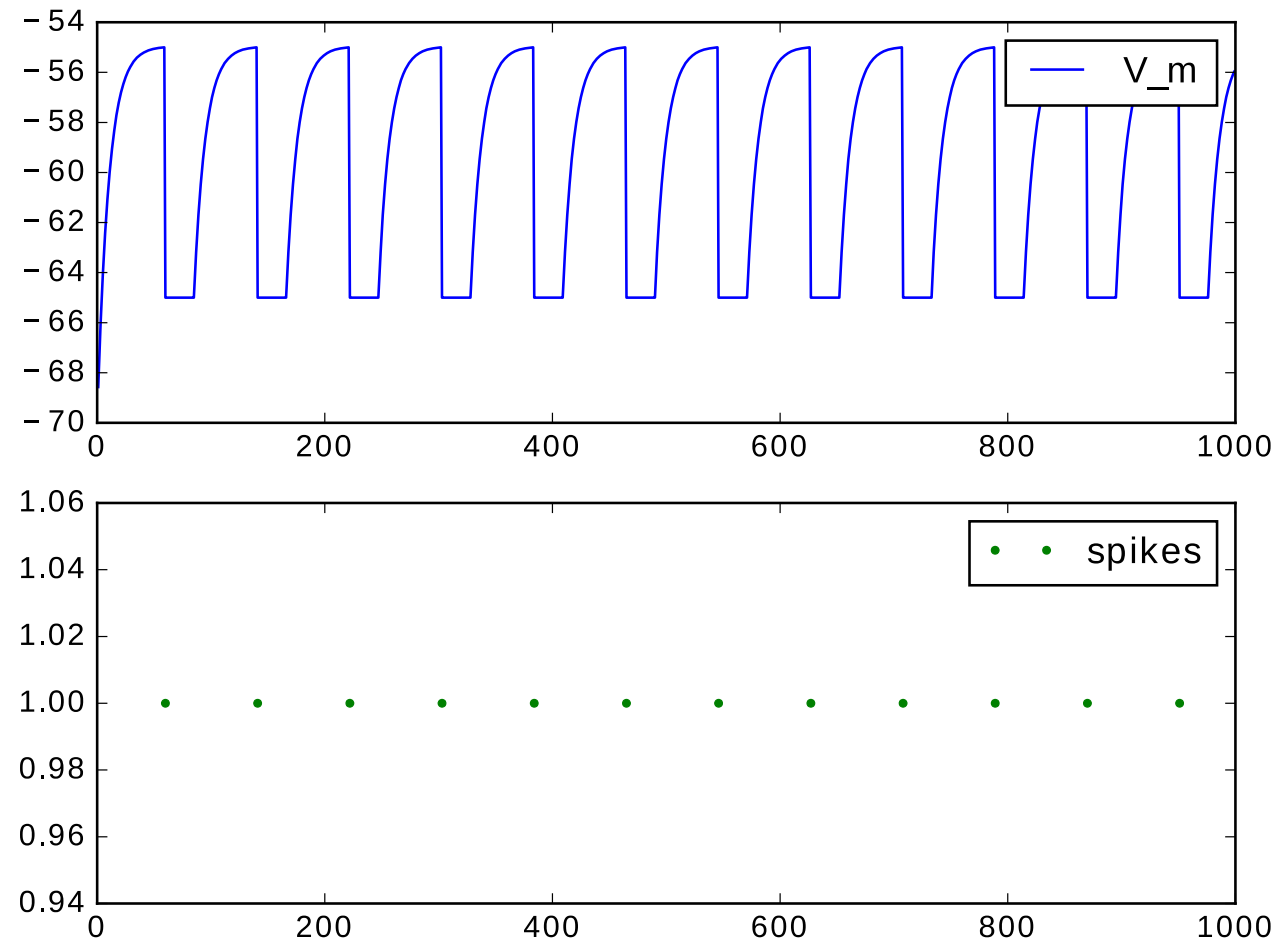
NESTML

```
<<rc_refractory>>
...

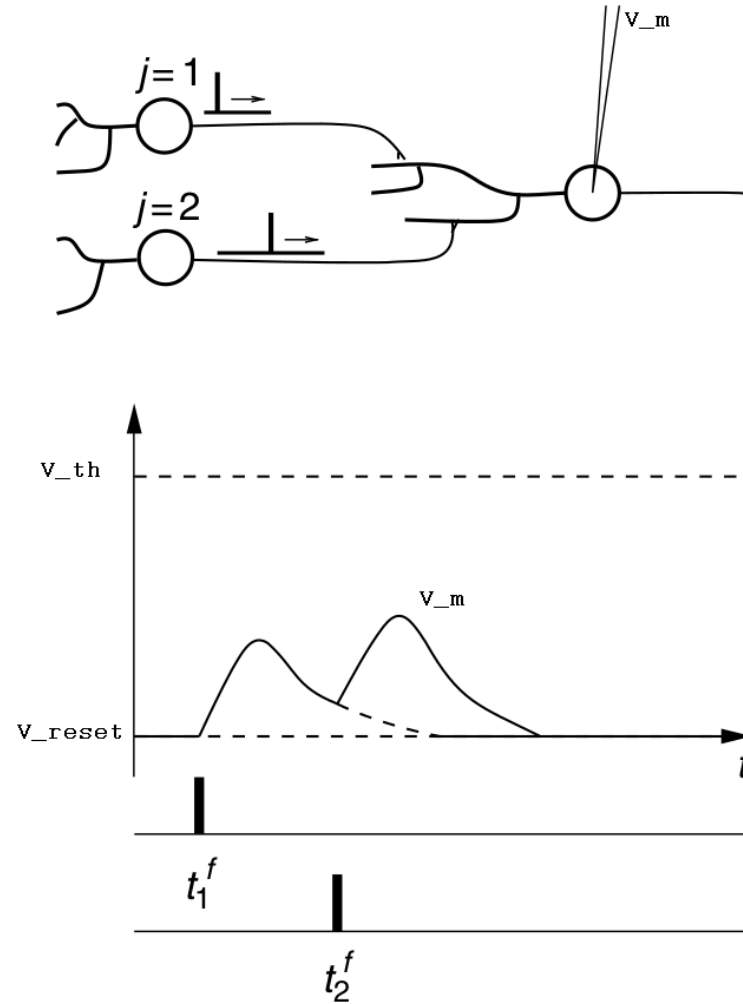
neuron rc_refractory:
  parameters:
    ref_counts integer = 0
    ref_timeout ms = 2 ms
  end
  internals:
    timeout_ticks integer = steps(ref_timeout)
  end
  update:
    if ref_counts == 0:
      integrate_odes()
      if V_m >= V_th:
        emit_spike()
        ref_counts = timeout_ticks
        V_m = V_reset
      end
    else:
      ref_counts -= 1
    end
  end
end
```

Simulating rc_refractory

NEST
<<Runtime>>



Input handling



(Source: [Wulfram Gerstner, Werner M. Kistler, Richard Naud, Liam Paninski-Neuronal Dynamics From Single Neurons to Networks and Models of Cognition](#))

Spike input

```
neuron rc_input:
  initial_values:
    V_m mV = E_L
  end
```

NESTML
<<rc_input>>

```
equations:
  V_m' = -(V_m-E_L)/tau_m + I_syn/C_m
end
```

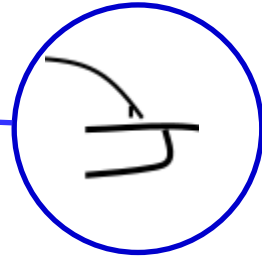
```
parameters:
  E_L mV = -70 mV
  ...
end
```

```
input:
  I_syn pA <- spike
end
```

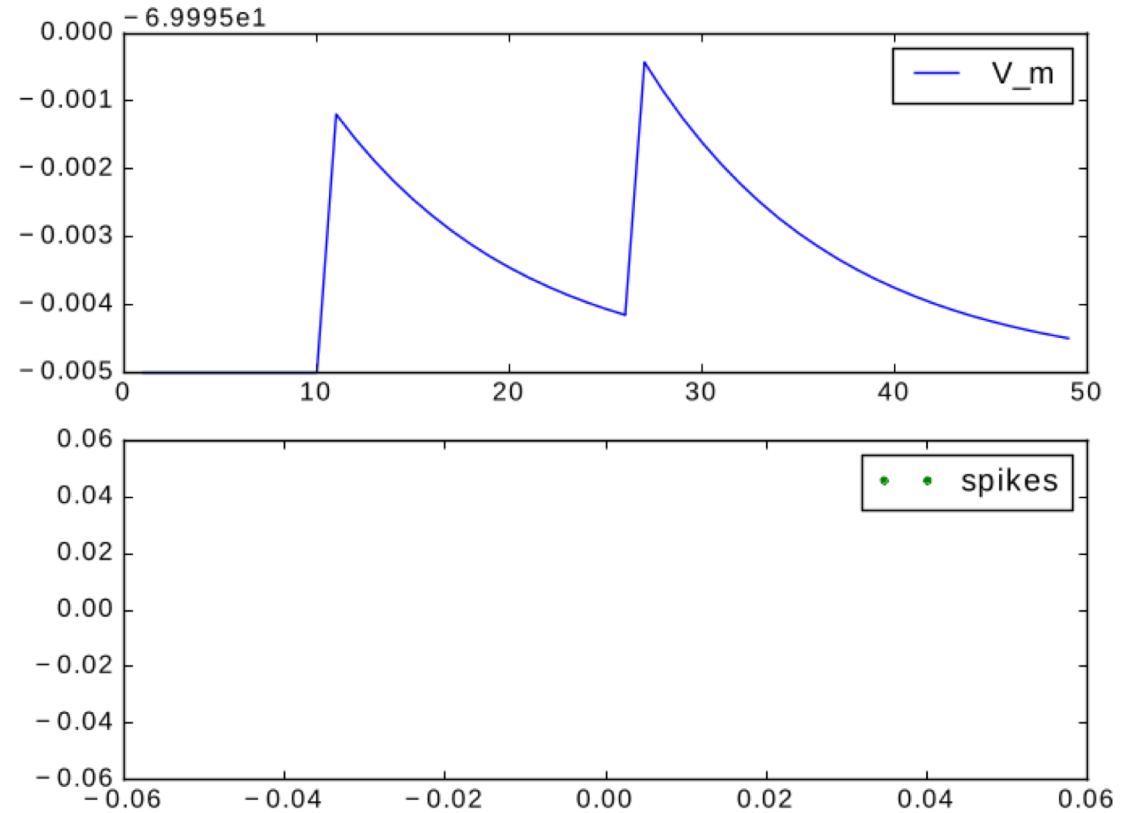
```
output: spike
```

```
end
```

buffer can be inhibitory,
excitatory or both (if nothing else
stated)



NEST
<<Runtime>>



Synaptic response

```
neuron rc_alpha_response:
```

```
  initial_values:
```

```
    V_m mV = E_L
```

```
    I_a pA = 0 pA
```

```
    I_a' pA/ms = e/tau_syn * pA
```

```
  end
```

```
  equations:
```

```
    shape I_a'' = -2 * I_a' / tau_syn - I_a / tau_syn**2
```

```
    V_m' = -(V_m-E_L)/tau_m + convolve(I_a, spikes)/C_m
```

```
  end
```

```
  input:
```

```
    spikes pA <- spike
```

```
  end
```

```
  output: spike
```

```
  update:
```

```
    integrate_odes()
```

```
    ...
```

```
  end
```

```
end
```

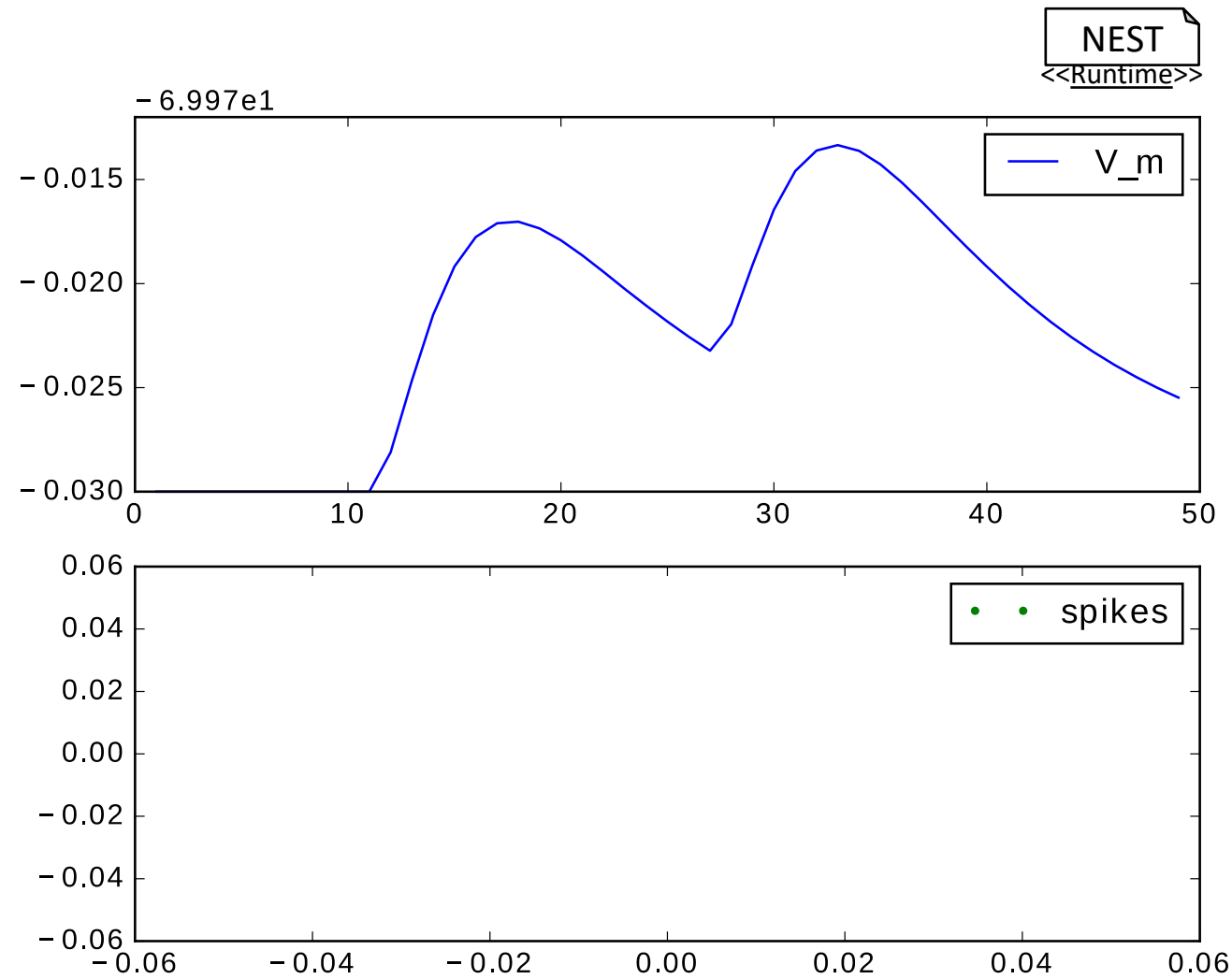
NESTML
<<rc_alpha>>

ODEs of order n require
all initial values of the
derivatives from 0 to $n-1$

$$\sum_{t_i \leq t, i \in \mathbb{N}} \sum_{w \in W} w \cdot I_a(t_i - t)$$

$$= \sum_{t_i \leq t, i \in \mathbb{N}} I_a(t_i - t) \sum_{w \in W} w$$

Simulating rc_alpha_response



Shape notation

```
neuron rc_alpha_response_shape:
  state:
    V_m mV = E_L
  end
```

NESTML
<<rc_shape>>

```
equations:
  shape I_a = (e/tau_syn) * t * exp(-t/tau_syn)
  V_m' = -(V_m-E_L)/tau_m + convolve(I_a, spikes)/C_m
end
```

initial values
computed automatically

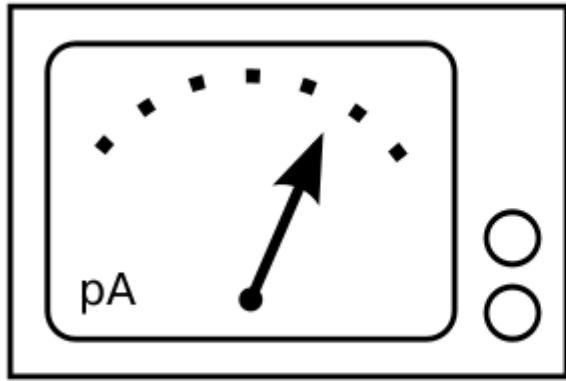
```
input:
  spikes pA <- spike
end
```

```
output: spike
```

```
update:
  integrate_odes()
  ...
end
```

```
end
```

Injecting currents



DC Generator

```
neuron rc_neuron:
```

```
...
```

```
equations:
```

```
function I_syn pA = I_e + convolve(I_a, spikes) + currents
```

```
V_m' = -V_m/tau_m + I_syn/C_m
```

```
end
```

```
input:
```

```
currents pA <- current
```

```
spikes pA <- spike
```

```
end
```

```
output: spike
```

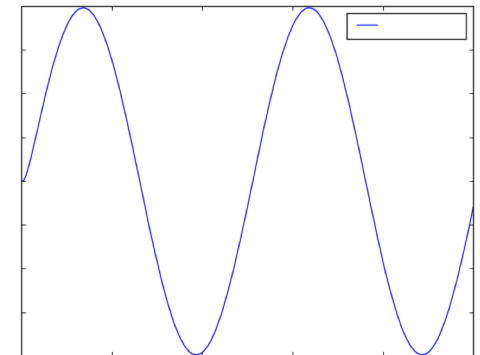
```
...
```

```
end
```

```
currents = nest.Create('dc_generator', 1,  
                        {'amplitude': 100.0})  
nest.Connect(currents, rc_neuron)
```

PyNEST

NEST
<<Runtime>>



PyNEST API of generated NEST module

```
import nest

nest.Install("nestmlmodels")

neuron = Create("rc_neuron")
nest.SetStatus(neuron, {"V_m": -72.0,
                        "C_m": 300.0})

mm = nest.Create('multimeter')
nest.SetStatus(mm, {"record_from": ["V_m"]})
Connect(mm, neuron)
```

PyNEST

```
neuron rc_neuron:
    initial_values:
        V_m mV = -70mV
    end

    equations:
        V_m' = -(V_m - 70mV) / tau_m + I_syn / C_m
    end

    parameters:
        C_m pF = 250pF
        tau_m ms = 10ms
        I_syn = 10pA
    end
end
```

NESTML

<<rc_neuron.nestml>>

Practical exercise: Izhikevich model

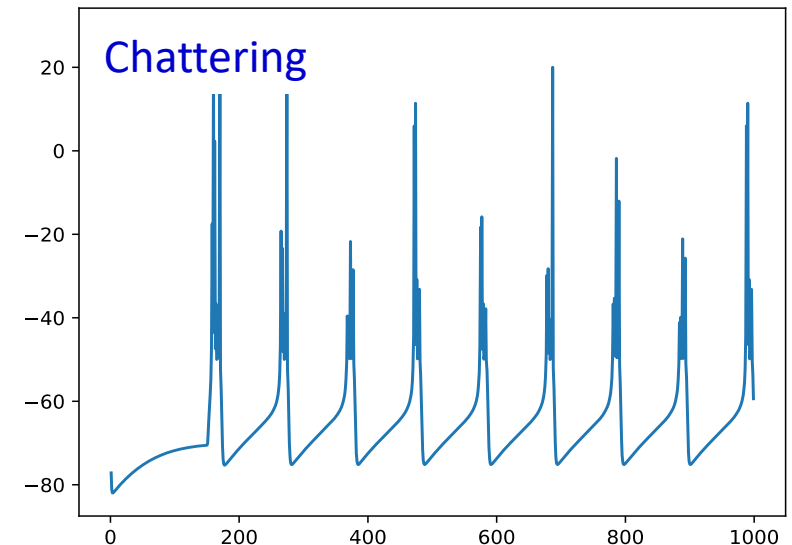
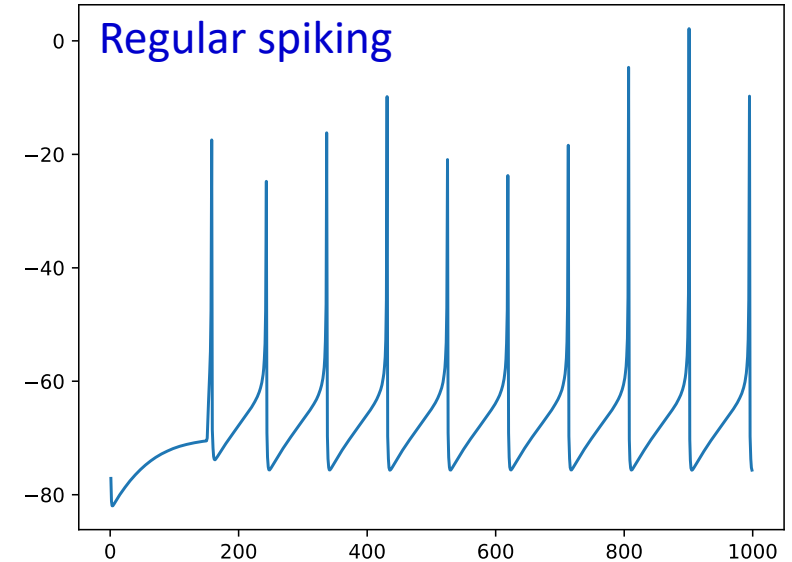
- Izhikevich: simple model for spiking neurons

$$v' = 0.04v^2 + 5v + 140 - u + I$$

$$u' = a(bv - u)$$

$$\text{if } v \geq 30\text{mV} \text{ then } \begin{cases} v = c \\ u = u + d \end{cases}$$

- Tutorial task:
 - Finish the model
 - Change parameters to produce chattering behaviour
- See E. Izhikevich, IEEE Transactions on Neural Networks (2003) 14:1569-1572



Practical exercise: Izhikevich model

- Go to <https://jupyter.cscs.ch/>
- Under "Piz Daint node type", select "mc" (for multicore) and click "Spawn"
- On the welcome screen, scroll down and click the "Terminal" button
- ```
git clone https://github.com/jougs/HPAC_Training --depth=1 && pip3
install nestml --user && cp HPAC_Training/.jupyterhub.env ~
```
- Restart the Jupyter server:
  - Go to File → Hub control panel → "Stop my server"
  - Click "Start my server" → "Launch server" and use the same details as in the first step
- Now, everything is installed and we are ready to go! Double click the "HPAC\_Training" folder in the left panel, then open the "NESTML" folder and open the notebook "NESTML-izhikevich-tutorial"
- Step through the notebook and convince yourself that everything works. Change `izhikevich_solution.nestml` to `izhikevich_task.nestml`, and finish the model!