

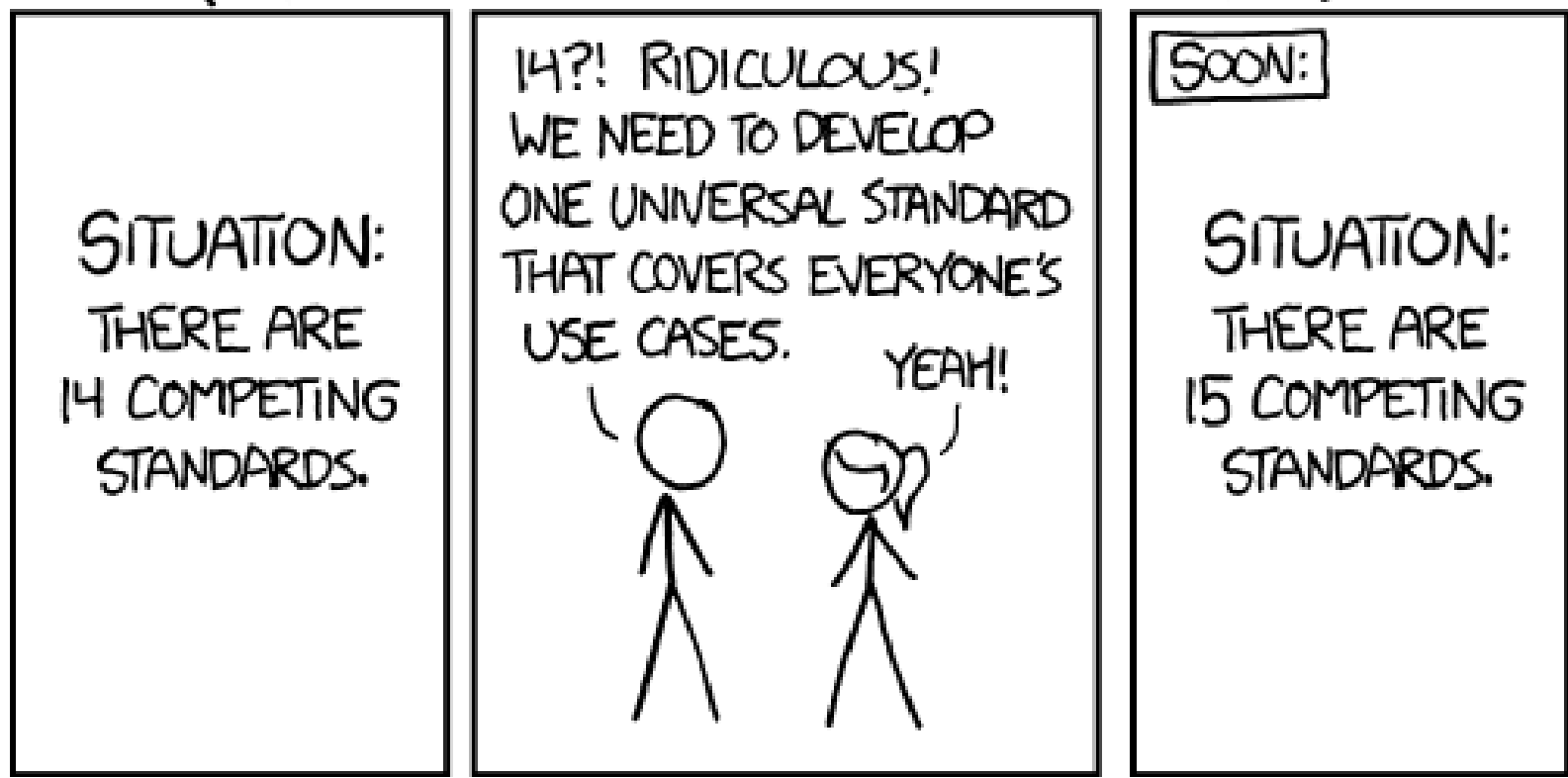
NEST Modeling Language: A modeling language for spiking neuron and synapse models for NEST

I.Blundell, D.Plotnikov, J.M.Eppler

Software Engineering
RWTH Aachen
FZ Jülich

Wait, yet another modeling language?

HOW STANDARDS PROLIFERATE:
(SEE: A/C CHARGERS, CHARACTER ENCODINGS, INSTANT MESSAGING, ETC.)

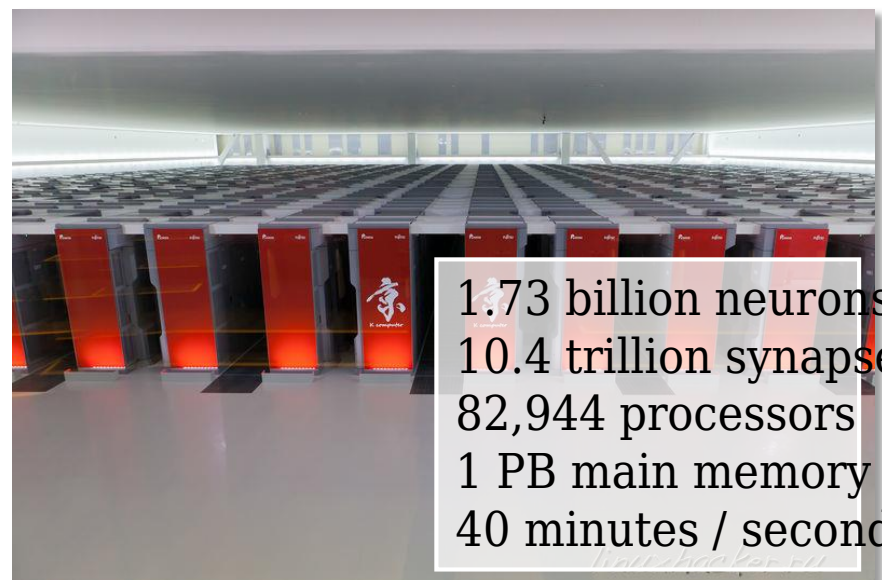


Wait, yet another modeling language?



The neural simulation tool NEST

- NEST is a hybrid parallel (OpenMP+MPI) simulator for spiking neural networks, written in C++, but with a Python frontend
- Neuron models are mainly point neurons and phenomenological synapse models (STDP, STP, neuromodulation)
- NEST supports large-scale models on the largest supercomputers
- Still the code also runs fine on laptops and workstations
- Get publication and source code on <http://nest-simulator.org>



The zoo of models

NEST 2.10.0 has 36 neuron models built in

19 are simple integrate-and-fire models

2 are based on the Hodgkin&Huxley formalism

11 have alpha-shaped post-synaptic responses

10 use exponentially decaying post-synaptic responses

15 with current-based dynamics solved exactly

9 conductance-based neurons using different solvers

plus some more exotic specimen

... and the situation gets worse each release
and each new modelling study

The zoo of models

NEST 2.10.0 has 36 neuron models built in

19 are simple integrate-and-fire models

2 are based on the Hodgkin&Huxley formalism

11 have alpha-shaped post-synaptic responses

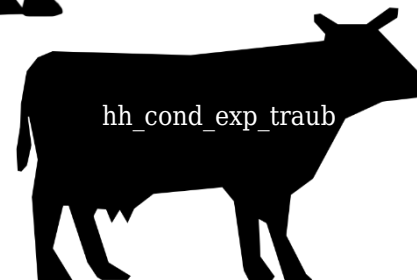
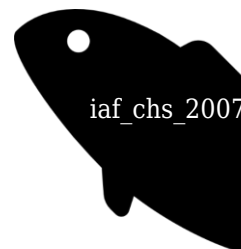
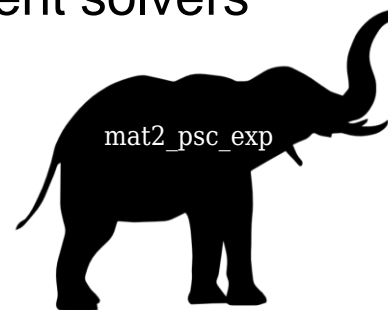
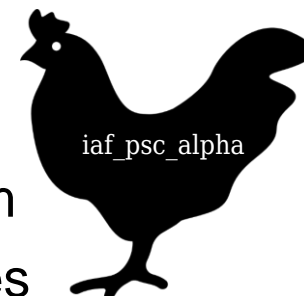
10 use exponentially decaying post-synaptic responses

15 with current-based dynamics solved exactly

9 conductance-based neurons using different solvers

plus some more exotic specimen

... and the situation gets worse each release
and each new modelling study



Creating neuron models

```
void lat_psc_alpha::update(Time const & origin, const long_t from, const long_t to)
{
    assert(to >= 0 && (delay) from < Scheduler::get_min_delay());
    assert(from < to);

    for ( long_t lag = from; lag < to; ++lag )
    {
        if ( S_r_ == 0 )
        {
            // neuron not refractory
            S_y3_ = V_P30_ * S_y0_ + P_1_e_
                + V_P31_ex_ * S_y1_ex_ + V_P32_ex_ * S_y2_ex_
                + V_P31_in_ * S_y1_in_ + V_P32_in_ * S_y2_in_
                + V_expm1_tau_m_ * S_y3_ + S_y3_;

            // lower bound of membrane potential
            S_y3_ = ( S_y3_ < P_LowerBound_ ? P_LowerBound_ : S_y3_ );
        }
        else // neuron is absolute refractory
            -S_r_;

        // alpha shape EPSCs
        S_y2_ex_ = V_P21_ex_ * S_y1_ex_ + V_P22_ex_ * S_y2_ex_;
        S_y1_ex_ *= V_P11_ex_;

        // Apply spikes delivered in this step: spikes arriving at T+1 have
        // an immediate effect on the state of the neuron
        V_weighted_spikes_ex_ = B_ex_spikes_get_value(lag);
        S_y1_ex_ += V_EPSCInitialValue_ * V_weighted_spikes_ex_;

        // alpha shape EPSCs
        S_y2_in_ = V_P21_in_ * S_y1_in_ + V_P22_in_ * S_y2_in_;
        S_y1_in_ *= V_P11_in_;

        // Apply spikes delivered in this step: spikes arriving at T+1 have
        // an immediate effect on the state of the neuron
        V_weighted_spikes_in_ = B_in_spikes_get_value(lag);
        S_y1_in_ += V_EPSCInitialValue_ * V_weighted_spikes_in_;

        // threshold crossing
        if ( S_y3_ >= P_Theta_ )
        {
            S_r_ = V_RefractoryCounts_;
            S_y3_ = P_V_reset_;
            // A supra-threshold membrane potential should never be observable.
            // The reset at the time of threshold crossing enables accurate integration
            // independent of the computation step size, see [2,3] for details.

            set_spikeTime(Time::step(origin.get_steps()+lag+1));
            SpikeEvent se;
            network()->send("this, se, lag);
        }

        // set new input current
        S_y0_ = B_currents_get_value(lag);

        // log state data
        B_logger_record_data(origin.get_steps() + lag);
    }
}
```

1. Copy & paste
2. Modify parts of the code
3. Ideally adapt the comments ;-))
4. Add to Makefiles
5. Re-compile and test
6. Goto 2...



```
void nest::lat_cond_alpha::update(Time const & origin, const long_t from, const long_t to)
{
    assert(to >= 0 && (delay) from < Scheduler::get_min_delay());
    assert(from < to);

    for ( long_t lag = from; lag < to; ++lag )
    {
        double t = 0.0;

        // numerical integration with adaptive step size control:
        // gsl_odev_evolve_apply performs only a single numerical
        // integration step, starting from t and bounded by step;
        // the while-loop ensures integration over the whole simulation
        // step (0, step) if more than one integration step is needed due
        // to a small integration step size.
        // note that (t+IntegrationStep > step) leads to integration over
        // (t, step) and afterwards setting t to step, but it does not
        // enforce setting IntegrationStep to step: this is of advantage
        // for a consistent and efficient integration across subsequent
        // simulation intervals
        while ( t < B_step_ )
        {
            const int status = gsl_odev_evolve_apply(B_e_, B_c_, B_s_,
                &B_sys_, // system of ODE
                &t, // from t
                B_step_, // to t+step
                &B_IntegrationStep_, // integration step size
                S_y); // neuronal state

            if ( status != GSL_SUCCESS )
                throw GSLSolverFailure(get_name(), status);

            // refractoriness and spike generation
            if ( S_r_ )
            { // neuron is absolute refractory
                -S_r_;
                S_y[State::V_M] = P_V_reset; // clamp potential
            }
            else
            { // neuron is not absolute refractory
                if ( S_y[State::V_M] >= P_V_th )
                {
                    S_r_ = V_RefractoryCounts_;
                    S_y[State::V_M] = P_V_reset;
                }

                // log spike with Archiving Node
                set_spikeTime(Time::step(origin.get_steps()+lag+1));
                SpikeEvent se;
                network()->send("this, se, lag);
            }

            // add incoming spikes
            S_y[State::DG_EXC1] += B_spike_exc_get_value(lag) * V_PSCConInit_E;
            S_y[State::DG_INH1] += B_spike_inh_get_value(lag) * V_PSCConInit_I;

            // set new input current
            B_i_stim_ = B_currents_get_value(lag);

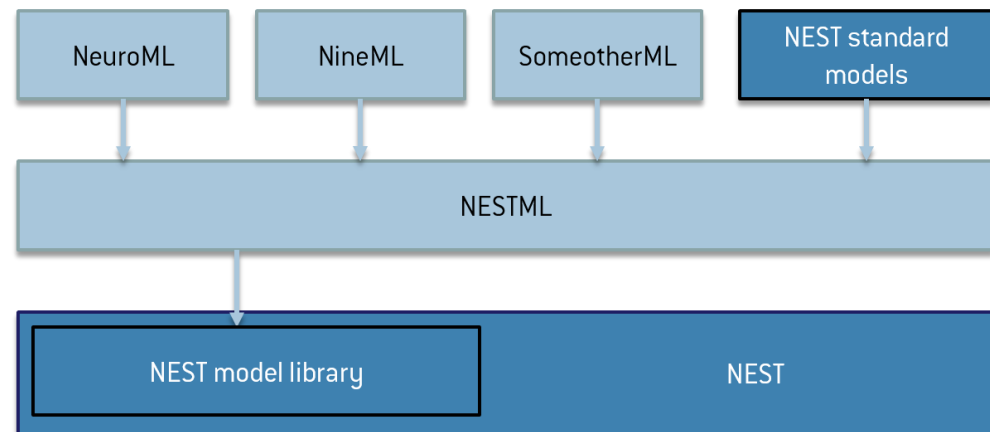
            // log state data
            B_logger_record_data(origin.get_steps() + lag);
        }
    }
}
```

The current process for model creation and the diversity leads to problems

- Copy & paste leads to errors and bad maintainability
- Implementation by non-programmers, often by trial and error

Basic NESTML features

- Semantic model checking and automatic choice of solver
- Automatic adaptation to new API versions
- Library for commonly used neuron dynamics and synaptic responses
- Ease of use

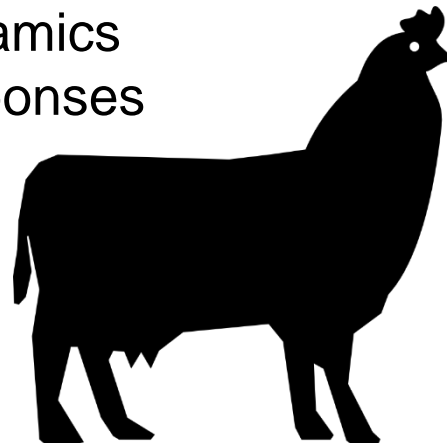


The current process for model creation and the diversity leads to problems

- Copy & paste leads to errors and bad maintainability
- Implementation by non-programmers, often by trial and error

Basic NESTML features

- Semantic model checking and automatic choice of solver
- Automatic adaptation to new API versions
- Library for commonly used neuron dynamics and synaptic responses
- Ease of use



Introductory Example: An IaF PSC model with alpha shape

```
neuron iaf_neuron:

  state:
    y0, y1, y2, V_m mV [V_m >= -99.0]
    # Membrane potential
    alias V_rel mV = V_m + E_L
  end

  function set_V_rel(v mV):
    V_m = v - E_L
  end

  parameter:
    # Capacity of the membrane.
    C_m pF = 250 [C_m > 0]
  end

  internal:
    h ms = resolution()
    P11 real = exp(-h / tau_syn)
    ...
    P32 real = 1 / C_m * (P33 - P11)
              / (-1/tau_m - -1/tau_syn)
  end

end
```

Fist class domain
concepts

SI Units

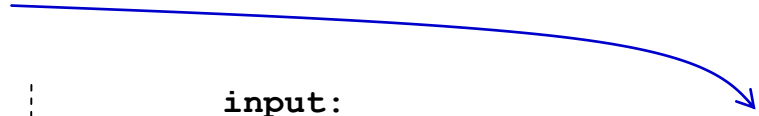
Gaurds

```
input:
  spikeInh <- inhibitory spike
  spikeExc <- excitatory spike
  currentBuffer <- current
end

output: spike

dynamics timestep(t ms):
  if r == 0: # not refractory
    V_m = P30 * (y0 + I_e) + P31 *
          y1 + P32 * y2 + P33 * V_m
  else:
    r = r - 1
  end
  # alpha shape PSCs
  V_m = P21 * y1 + P22 * y2
  y1 = y1 * P11
  y0 = currentBuffer.getSum(t);
end

end
```



Major Building blocks Blocks 1/2

- State block
 - Variables describing the neuron's state
 - `alias` to express a dependency (also in another block possible)
- Parameter block
 - Values adjustable during instantiation
 - Guard checks
- Internal
 - Capture helper variables

```
state:  
  V_m mV [V_m >= -99.0]  
  # Membrane potential  
  alias V_rel mV = V_m + E_L  
end
```

```
parameter:  
  # Capacity of the membrane.  
  C_m          pF = 250 [C_m > 0]  
end
```

```
internal:  
  h    ms = resolution()  
  P11 real = exp(-h / tau_syn)  
  ...  
  P32 real = 1 / C_m * (P33 - P11)  
           / (-1/tau_m - -1/tau_syn)  
end
```

Major Building blocks Blocks 2/2

- Neuron's dynamic is modelled in a predefined dynamics function

- timestamp, event based

```
dynamics timestep(t ms):  
  if r == 0: # not refractory  
    V_m = P30 * (y0 + I_e) ...  
  else:  
    r = r - 1  
  end  
end
```

- Auxiliary helper functions

```
function set_V_rel(v mV):  
  V_m = v - E_L  
end
```

- Buffers:

- First-order language concept
- Semantic checks

```
input:  
  spikeInh    <- inhibitory spike  
  spikeExc    <- excitatory spike  
  cur <- current  
end
```

```
output: spike
```

- ODE Blocks

- Dynamics can be defined declaratively

```
ODE:  
  G := E/tau_syn) * t * exp(-1/tau_syn*t)  
  d/dt V := -1/Tau * V + 1/C_m * G + I_e + cur  
end
```

An IaF PSC model with alpha shape ODE Approach

```
neuron iaf_neuron:
  internal:
    h ms = resolution()
    P11 real = exp(-h / tau_syn)
    ...
    P32 real = 1 / C_m * (P33 - P11)
              / (-1/tau_m - -1/tau_syn)
  end

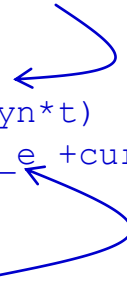
  dynamics timestep(t ms):
    if r == 0: # not refractory
      V_m = P30 * (y0 + I_e) + P31 *
            z1 + P32 * y2 + P33 * y3
    else:
      r = r - 1
    end
    # alpha shape PSCs
    V_m = P21 * y1 + P22 * V_m
    y1 = y1 * P11
  end
end
```

```
neuron iaf_neuron_ode:
  internal:
    h ms = resolution()
  end

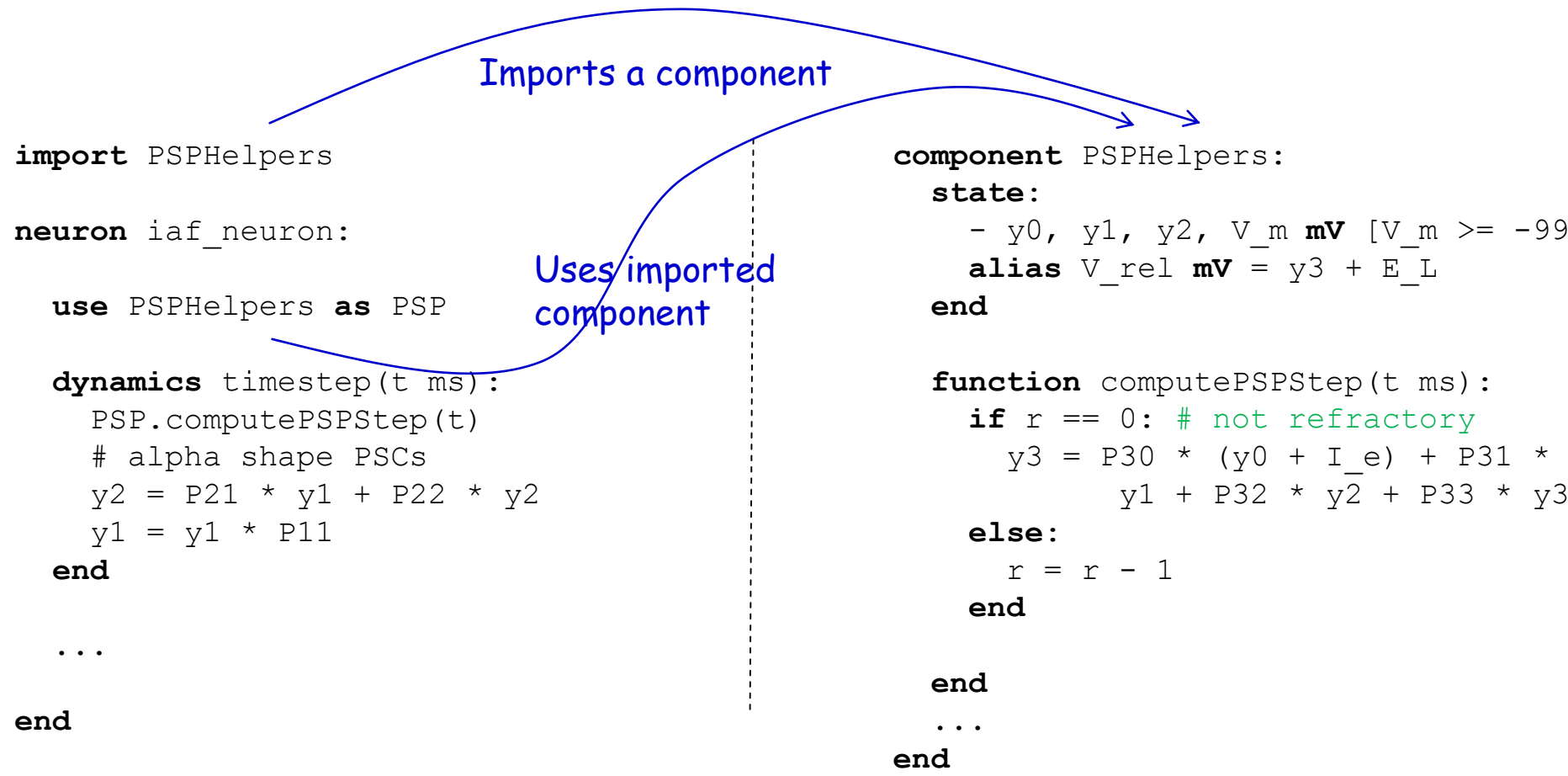
  dynamics timestep(t ms):
    if r == 0: # not refractory
      ODE:
        G := E/tau_syn) * t * exp(-1/tau_syn*t)
        d/dt V := -1/Tau * V + 1/C_m * G + I_e + cur
    end
    else:
      r = r - 1
    end
    ...
  end
  ...
end
```

Current equations

Membran potential



Model cross-referencing



MontiCore Language Workbench



- **Opensource and free** github project
- **Grammar** based
- Definition of **modular language fragments**
- Assistance for **analysis, transformations**
- Generates: parsers, symbol tables, language processing infrastructure

- Composition of languages:
 - **independent** language **development**
 - **composition of languages and tools**
 - **Language extension**
 - Language inheritance (allows replacement)

- Quick definition of domain specific languages (DSLs)
 - by reusing existing languages
 - **variability** in syntax, context conditions, generation, semantics

Language Architecture of NESTML

NESTML
Nest Modeling Language
Description of the neuron models



PL
Precedural Language:
Description of the imperative
parts (e.g. definition of the
dynamics function)

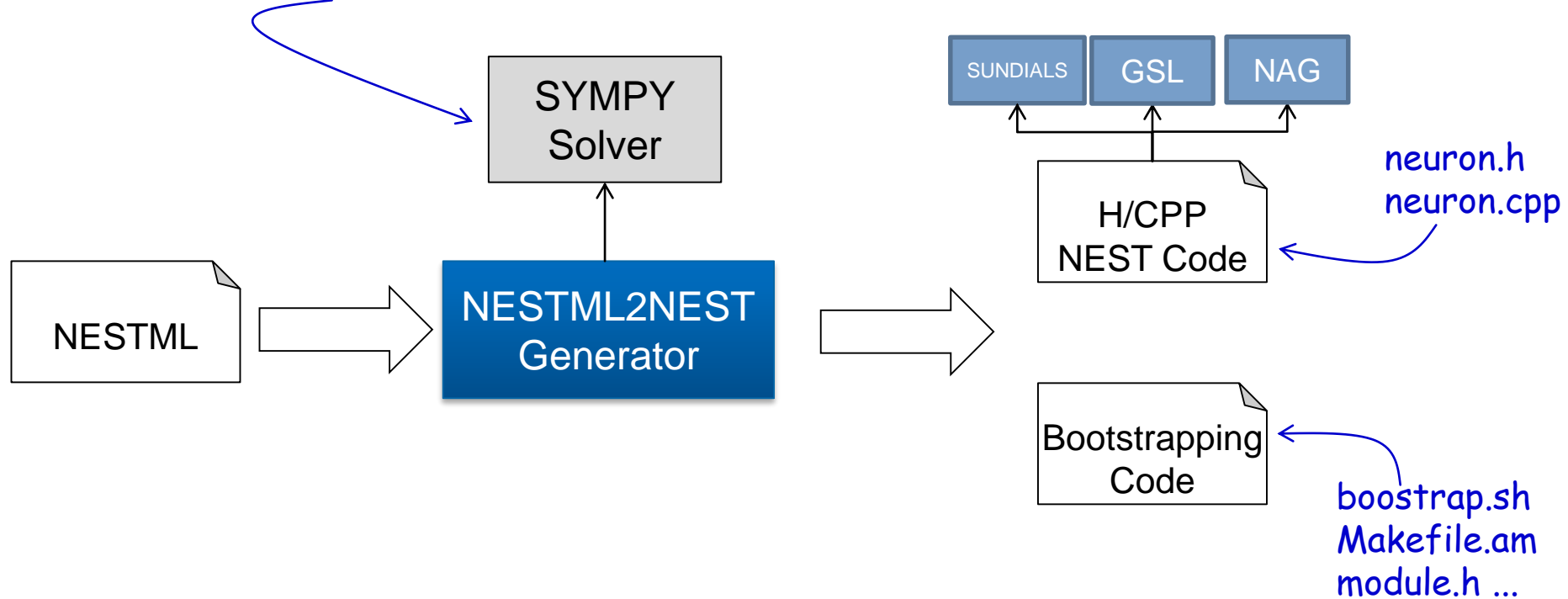
ODEDSL
Definition of
Ordinary Differential Equations

UnitDSL:
definition and automatic
conversion of physical units

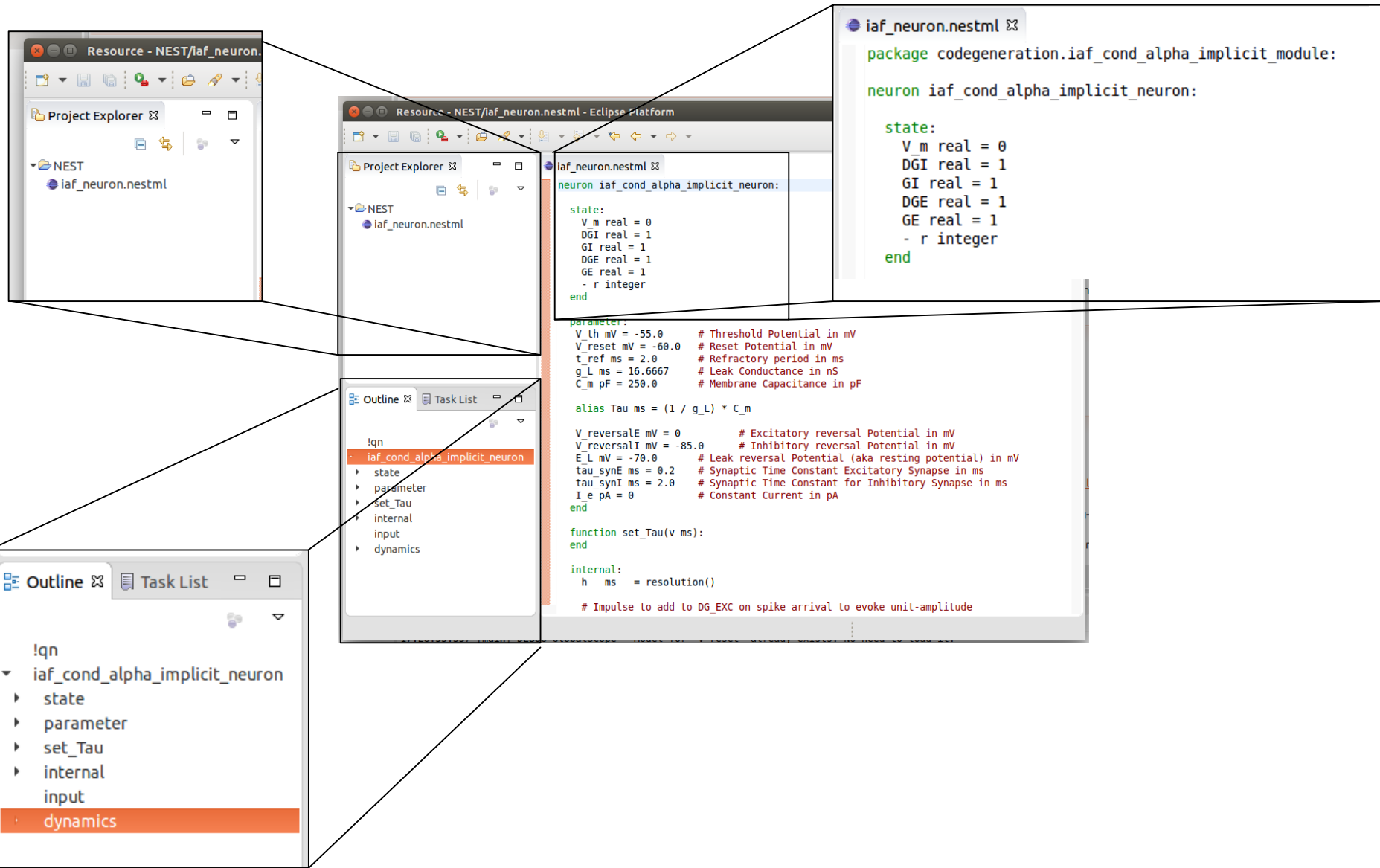
Generator Architecture for NEST

- Templated based code generation
 - Based on well founded mathematical theory
- Traceable model transformations
 - After transformations altered NESTML model is produced

SYMPY Solver
Computation of the exact solution



For Comfort: Editor in Eclipse for NESTML



Current State and Future Work

- Open-source github project
- First evaluation during a community workshop
 - Participant wrote NESTML models and ran them in NEST under 30 minutes
 - Also without preliminary experience with NEST or NESTML
- Publication: NESTML: a modeling language for spiking neurons
 - (to appear in spring 2016)
- Support for:
 - Explicit solvable models
 - E.g. PSC models in the NEST context
 - Numerical solvers
 - For now the GSL solver is already integrated
- New modeling concepts and optimisations
 - E.g. struct of arrays
 - Multi-compartment models
- Targeting new platforms
 - GPU
 - SpiNNaker

Backup

ODE Processing Workflow

NESTML

$$G == (E/\tau_{syn}) * t * \exp(-1/\tau_{syn}*t)$$
$$d/dt V == -1/\tau * V + 1/C_m * G$$

Text

```
...  
h*exp(-h/tau_in)# P10  
exp(-h/tau_in)# P11  
...
```

SymPy

For the ODE a SymPy-Solver is generated and executed.

NESTML

The matrix is parsed and a new NESTML Model with the solution matrix is created

```
internal:  
...  
  P10 = h*exp(-h/tau_in)  
  P11 = exp(-h/tau_in)  
...  
end
```

H/CPP
NEST Code

SI Units Specification

Größen-Name	Größen-Zeichen	Einheiten-Name	Einheiten-Zeichen
Länge	l	Meter	m
Masse	m	Kilogramm	kg
Zeit	t	Sekunde	s
Stromstärke	I	Ampere	A
Temperatur	T	Kelvin	K
Stoffmenge	n	Mol	mol
Lichtstärke	I_V	Candela	cd

- Every another unit is defined as a combination of base units:

$$Q = L^\alpha \cdot M^\beta \cdot T^\gamma \cdot I^\delta \cdot \Theta^\varepsilon \cdot N^\zeta \cdot J^\eta$$

- E.g. volt is defined as.

$$V = m^2 \cdot kg \cdot s^{-3} \cdot A^{-1} \cdot K^0 \cdot mol^0 \cdot cd^0$$