Introduction to NEST

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https://github.com/tomtetzlaff/2023_eitnfallschool

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Overview and general features

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Built-in synapse models

Custom neuron and synapse models with NESTML

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Parametrization

“Hello world!”

Example: balanced random network

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  - Linux
  - Mac OS X
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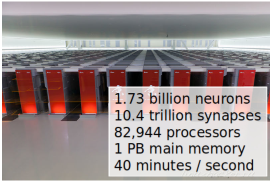

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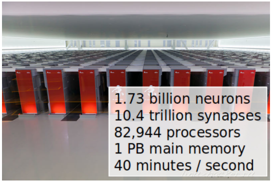

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- sources:
  - web: `https://www.nest-simulator.org`
  - code: `https://github.com/nest/nest-simulator`
  - docs: `https://nest-simulator.readthedocs.io`

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- citing NEST: see Zenodo, e.g. (Sinha et al. 2023)

Built-in neuron models

- integrate-and-fire models (`iaf_*`, `*if_*`)
  - with current-based (`iaf_psc_*`)
  - and conducance based synapses (`iaf_cond_*`)
  - and different synaptic kernels (`*_delta`, `*_exp`, `*_alpha`)
- adaptive exponential IaF models (AdEx; `aeif_*`)
- generalized leaky integrate-and-fire models (glif; `gif_*`)
- multi-timescale adaptive threshold model (`mat2_*`,`amat2_*`)
- Izhikevich model (`izhikevich`)
- Hodgkin-Huxley type models (`hh_*`)
- neuron models with multiple (few) compartments (`*_mc_*`)
- firing rate neurons (`tanh_*`, `threshold_lin_*`, `sigmoid_*`, `siegert_*`)
- binary neuron models (`mcculloch_pitts_*`, `ginzburg_*`)
- …and many more

(for an overview of all built-in models, see https://nest-simulator.readthedocs.io/en/stable/models)

Built-in synapse models

- current-based and conductance-based synapses (see neuron models)
- gap junctions (`gap_junction`)
- static synapses (`static_synapse*`)
- various types of synaptic plasticity
  - various forms of short term plasticity (`tsodyks*`, `quantal_stp_*`)
  - various forms of spike-timing-dependent plasticity (`stdp_*`)
  - STDP plus third factors (`clopath_*`, `stdp_dopamine_*`)
  - structural plasticity (Butz and Ooyen 2013)
- stochastic synapses (synaptic failure; `bernoulli_synapse`)
- …and many more

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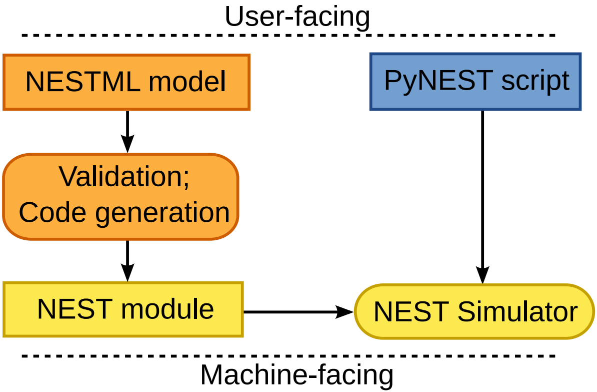
Custom neuron and synapse models with NESTML

- domain-specific language for neuron and synapse models



- code: `https://github.com/nest/nestml`
- docs: `https://nestml.readthedocs.io`

Stimulation devices

Spike generators

- spikes at prescribed points in time: `spike_generator`
- realizations of homogeneous or inhomogeneous Poisson point processes:
  - `poisson_generator`
  - `inhomogeneous_poisson_generator`
  - `sinusoidal_poisson_generator`
- realizations of homogeneous or inhomogeneous Gamma point processes:
  - `gamma_sup_generator`
  - `sinusoidal_gamma_generator`

Current generators

- constant current: `dc_generator`
- sinusoidal current: `ac_generator`
- step-wise constant current: `step_current_generator`
- noisy current: `noise_generator`

(see https://nest-simulator.readthedocs.io/en/stable/devices/stimulate_the_network.html)

Recording devices

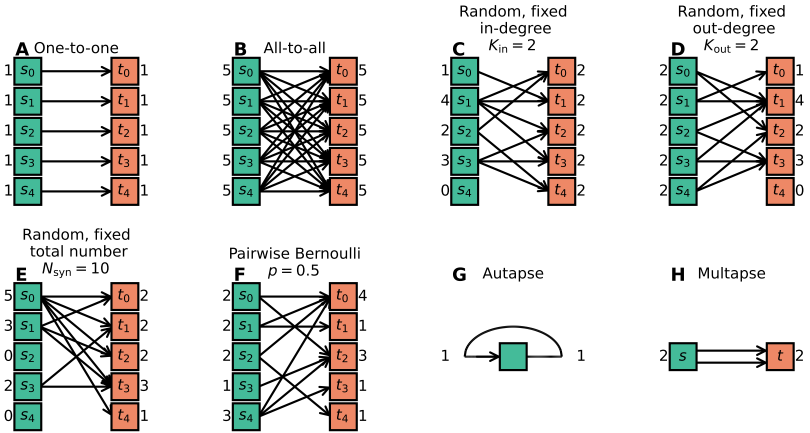
- spikes : `spike_recorder`
- analog quantities (membrane potentials, conductances, . . .): `multimeter`
- synaptic weights: `weight_recorder`

(see https://nest-simulator.readthedocs.io/en/stable/devices/record_from_simulations.html)

Connection management

- large repertoire of efficient built-in connection routines (Ippen et al. 2017), incl.
  - various forms of deterministic and random connectivity patterns
  - spatially structured networks

(see `https://nest-simulator.readthedocs.io/en/stable/synapses/connection_management.html`)



(Senk et al. 2022)

Parametrization

- parameters: specification of, e.g.,
  - initial conditions
  - neuron or device properties
  - positions in physical space
  - connection probabilities, synaptic weights, delays

(see `https://nest-simulator.readthedocs.io/en/stable/neurons/parametrization.html`)

Parametrization

- parameters: specification of, e.g.,
  - initial conditions
  - neuron or device properties
  - positions in physical space
  - connection probabilities, synaptic weights, delays
- parameter types:
  - random parameters (built-in RNGs)

    (see https://nest-simulator.readthedocs.io/en/stable/nest_behavior/random_numbers.html)
  - spatial parameters
  - spatially distributed parameters
  - mathematical functions
  - clipping, redrawing, and conditional parameters
  - combination of parameters

(see https://nest-simulator.readthedocs.io/en/stable/neurons/parametrization.html)

“Hello world!”

```
import nest     # import NEST module
import matplotlib.pyplot as plt # for plotting

nest.ResetKernel() # reset simulation kernel

neuron=nest.Create('iaf_psc_exp') # create LIF neuron with exponential synaptic currents

# create a spike generator, and set it up to create two spikes at 10 and 30ms
spikegenerator=nest.Create('spike_generator', params={'spike_times': [10.,30.]})

# create multimeter and set it up to record the membrane potential V_m
multimeter=nest.Create('multimeter', {'record_from': ['V_m']})

# connect spike generator with neuron with synaptic weight 100 pA
nest.Connect(spikegenerator, neuron,syn_spec={'weight': 50.0})

nest.Connect(multimeter, neuron)  # connect multimeter to the neuron

nest.Simulate(100.) # run simulation for 100ms

# read out recording time and voltage from voltmeter
times=multimeter.get('events')['times']
voltage=multimeter.get('events')['V_m']

# plot results
plt.figure(1)
plt.clf()
plt.plot(times,voltage,'k-',lw=2)
plt.xlabel('time (ms)')
plt.ylabel('membrane potential (mV)')
plt.savefig('./figures/hello_world.pdf')
```

(see `hello_world.py`)

“Hello world!”

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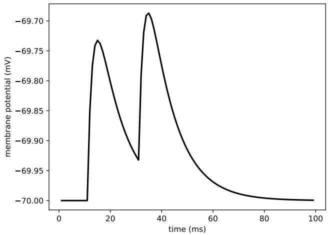
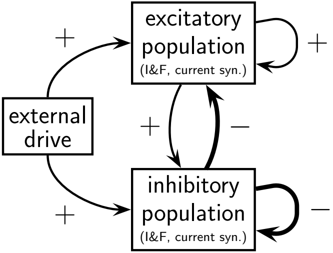
hello_world.pdf:



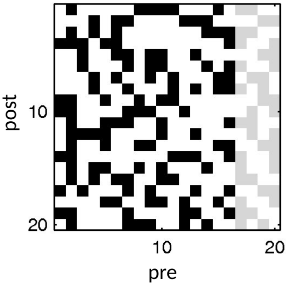
(see `hello_world.py`)

Example: balanced random network

simple model of a local cortex volume (Brunel 2000)
($\sim 1\text{mm}^3$, $\sim 10^{4\ldots5}$ neurons)



Connectivity matrix $J = \{J_{ij}\}$



- N_E excitatory, N_I inhibitory neurons, $N = N_\mathsf{E} + N_\mathsf{I} \sim 10^{4\ldots5}$
- sparse, random connectivity with fixed in-degree $K = N/10$
- LIF dynamics with current-based synapses with exponential shape:

$$\begin{aligned} \tau_\mathsf{m}\dot{V}_i &= -V_i(t) + R\left(I_i^{\mathsf{net}}(t) + I^{\mathsf{ext}}\right) \quad (i \in \{1,\ldots,N\}) \\ I_i^{\mathsf{net}}(t) &= \sum_j J_{ij}(h * s_j)(t-d) \quad \text{with} \quad h(t) = e^{-t/\tau_\mathsf{s}}\Theta(t) \end{aligned}$$

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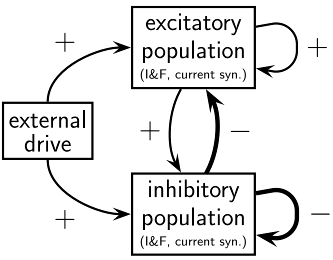
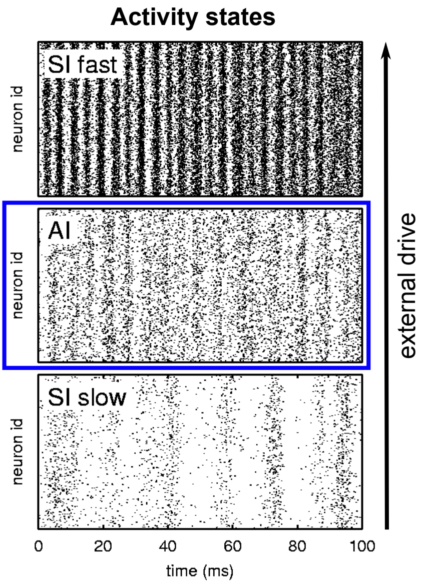
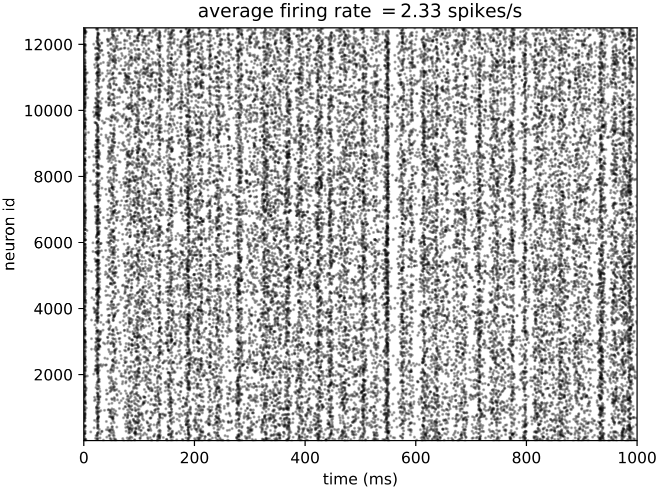


- in-vivo like activity
  - large membrane potential fluctuations
  - low firing rates
  - irregular spiking
- global oscillatory modes in various frequency bands



Example: balanced random network

implementation: see `balanced_random_network.py`



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- Charl Linssen
- Abigail Morrison
- Jannis Schuecker
- Johanna Senk
- Tom Tetzlaff
- Sacha van Albada
- Alexander van Meegen

Thanks

References I

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