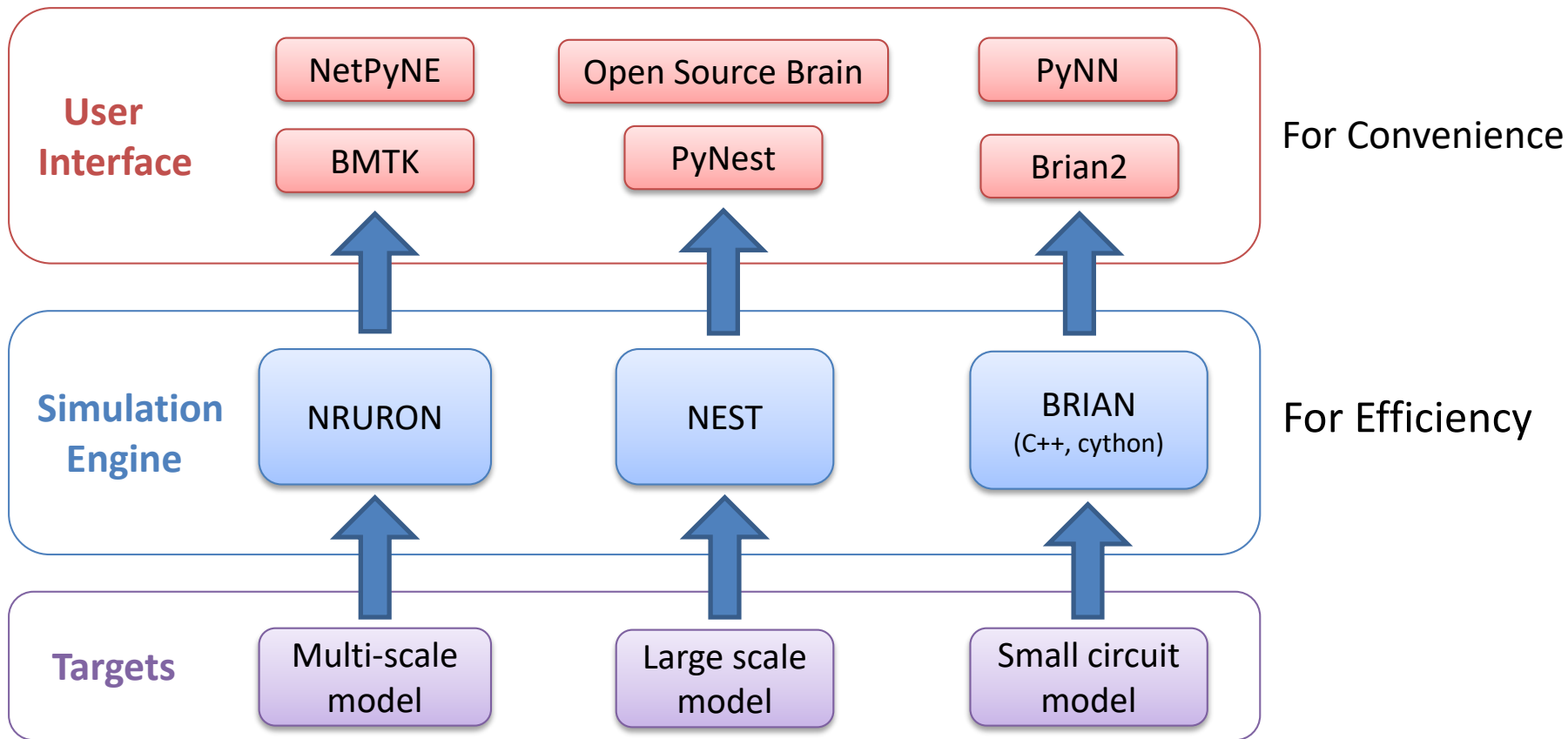


BrainPy (灵机): A *flexible* and *extensible* framework for brain modeling

<https://github.com/PKU-NIP-Lab/BrainPy>

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The programming paradigm of current neural simulators

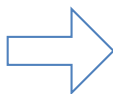


Model definition, building and simulation are separated from each other.

Example: NEST

底层定义和构建模型，Python调用模型

```
1 import nest
2
3 nest.ResetKernel()
4 nest.SetKernelStatus(
5     {"resolution": 0.05})
6 neuron_param = {"tau_m": Tau,
7                 "t_ref": TauR,
8                 "tau_syn_ex": Tau_psc,
9                 "tau_syn_in": Tau_psc,
10                "C_m": C,
11                "V_reset": E_L,
12                "E_L": E_L,
13                "V_m": E_L,
14                "V_th": Theta}
15 neurons = nest.Create("iaf_psc_exp", 2)
16 neurons.set(neuron_param)
```



```
23 #include "iaf_psc_exp.h"
24
25 // C++ includes:
26 #include <limits>
27
28 // Includes from libnestutil:
29 #include "dict_util.h"
30 #include "numerics.h"
31 #include "propagator_stability.h"
32
33 // Includes from nestkernel:
34 #include "event_delivery_manager_impl.h"
35 #include "exceptions.h"
36 #include "kernel_manager.h"
37 #include "universal_data_logger_impl.h"
```



Flexibility Problem

	NEURON	NEST
Neuron model without dynamics	M	M
Neuron model with simplified and discontinuous dynamics <i>Examples: Leaky Integrate-and-Fire (LIF), Izhikevich or Quadratic LIF; Exponent Leaky Integrate-and-Fire (eLIF)</i>	M	M
Neuron model with simplified and continues dynamics <i>Examples: FitzHugh–Nagumo, Morris–Lecar</i>	M	M
Single compartment, conductance-based model—temporal integration (point neuron) <i>Examples: Single-Compartment Hodgkin–Huxley model</i>	YG	M
Can conductance-based descriptions of ion channels be added to the neuron model? <i>Example: h-channel</i>	YG/M	m
Neuron model with simplify morphology (2-compartment model) <i>Example: Pinsky–Rinzel model</i>	YG	M
New model of chemical synapse	M	m
New model of electrical synapse	M	m
New model of learning rule	m	M

Descriptive Language: Python描述(定义)模型，底层构建和运行模型

Example: BRIAN2

```
G = NeuronGroup(10, '''dv/dt = I_leak / Cm : volt
                      I_leak = g_L*(E_L - v) : amp''')
```



Code Generation

```
for(int _idx=0; _idx<_num_idx; _idx++)
{
    double v = _array_neurongroup_v[_idx];
    double w = _array_neurongroup_w[_idx];
    const double _w = -(dt) * w / tau_w + w;
    const double _v = dt * (-(v) + w) / tau_v + v;
    w = _w;
    v = _v;
    _array_neurongroup_v[_idx] = v;
    _array_neurongroup_w[_idx] = w;
}
```

(frontiers, 2014; eLife, 2019)

Example: NetPyNE

i) `popParams['EXC_L2'] = {
 'cellType': 'PYR',
 'cellModel': 'simple',
 'yRange': [100, 400],
 'numCells': 50}`

ii) `popParams['EXC_L5'] = {
 'cellType': 'PYR',
 'cellModel': 'complex',
 'yRange': [700, 1000],
 'density': 80e3}`

(eLife, 2019)

iii) `cellParams['PYR_simple'] = {
 'conds': {'cellType': 'PYR',
 'cellModel': 'simple'},
 'secs': {'soma':
 'geom': {'diam': 18, 'L': 18},
 'mechs': {'hh':
 {'gnabar': 0.12,
 'gkbar': 0.036,
 'gl': 0.003,
 'el': -70}}}}`

iv) `importCellParams(
 label = 'PYR_complex',
 conds = {'cellType': 'PYR',
 'cellModel': 'complex'},
 fileName = 'L5_pyr_full.hoc',
 cellName = 'PYR_L5')`

Example: BMTK

```
from bmtk.builder import NetworkBuilder
from bmtk.builder.auxiliary.node_params import positions_columinar

net = NetworkBuilder("ei")

net.add_nodes(N=8000, pop='exc', model_name='Scnn1a',
              morphology='Scnn1a_473845048_m',
              model_type='biophysical',
              model_template='nml:Cell_472363762.cell.nml',
              positions=positions_columinar(8000, radius=1000, height=500))

net.add_nodes(N=2500, pop='inh', model_name='PV',
              morphology='Pvalb_470522102_m',
              model_type='biophysical',
              model_template='nml:Cell_472912177.cell.nml',
              positions=positions_columinar(2500, radius=1000, height=500))

net.add_edges(source={'pop': 'e'}, target={'pop': 'i'},
              connection_rule=gaussian_distance,
              dynamics_params='AMPA_ExcToInh.json',
              model_template='Exp2Syn',
              syn_weight=0.1,
              delay=1.5,
              target_sections=['somatic', 'basal'])

net.add_edges(source={'pop': 'i'}, target={'pop': 'e'},
              connection_rule=random_connections,
              dynamics_params='GABA_InhToExc.json',
              model_template='Exp2Syn',
              syn_weight=0.5,
              delay=1.5,
              target_sections=['apical'],
              distance_range=[50.0, 100.0])

net.build()
net.save(output_dir='ei_network')
```

(PLOS CB, 2020)

Transparency and Extensibility Problems

1. String description is pseudo programming, greatly reducing the program expressive power

```
dm/dt = alpha_m*(1-m)-beta_m*m : 1
dn/dt = alpha_n*(1-n)-beta_n*n : 1
dh/dt = alpha_h*(1-h)-beta_h*h : 1
```

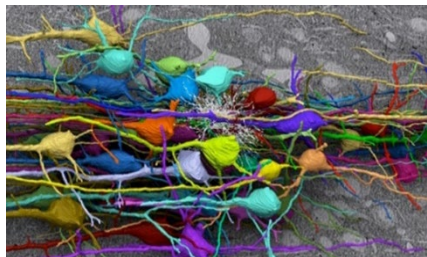
```
m = np.clip(int_m(ST['m'], _t_, ST['V']), 0., 1.)
h = np.clip(int_h(ST['h'], _t_, ST['V']), 0., 1.)
n = np.clip(int_n(ST['n'], _t_, ST['V']), 0., 1.)
```

2. 代码对用户隐藏，不支持debug，不知道是否生成用户想到的逻辑。一旦发现错误用户无法纠正代码。这往往导致用户定义新模型时捉襟见肘。

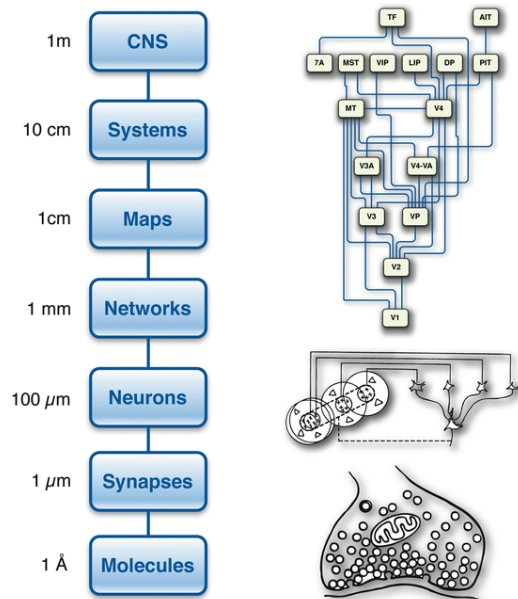
3. 对不满足假设与规定的模型不支持

Challenges of Programming in Computational Neuroscience

多尺度建模



大尺度建模



1. Efficiency Problem

2. Flexibility & Transparency Problem

新模型

- Pre-defined models is not sufficient
- New models at every scale

3. Extensibility Problem

新方法

- New differential equations
- New numerical integrators
- Numerical continuation

Easy to learn and use

Flexible

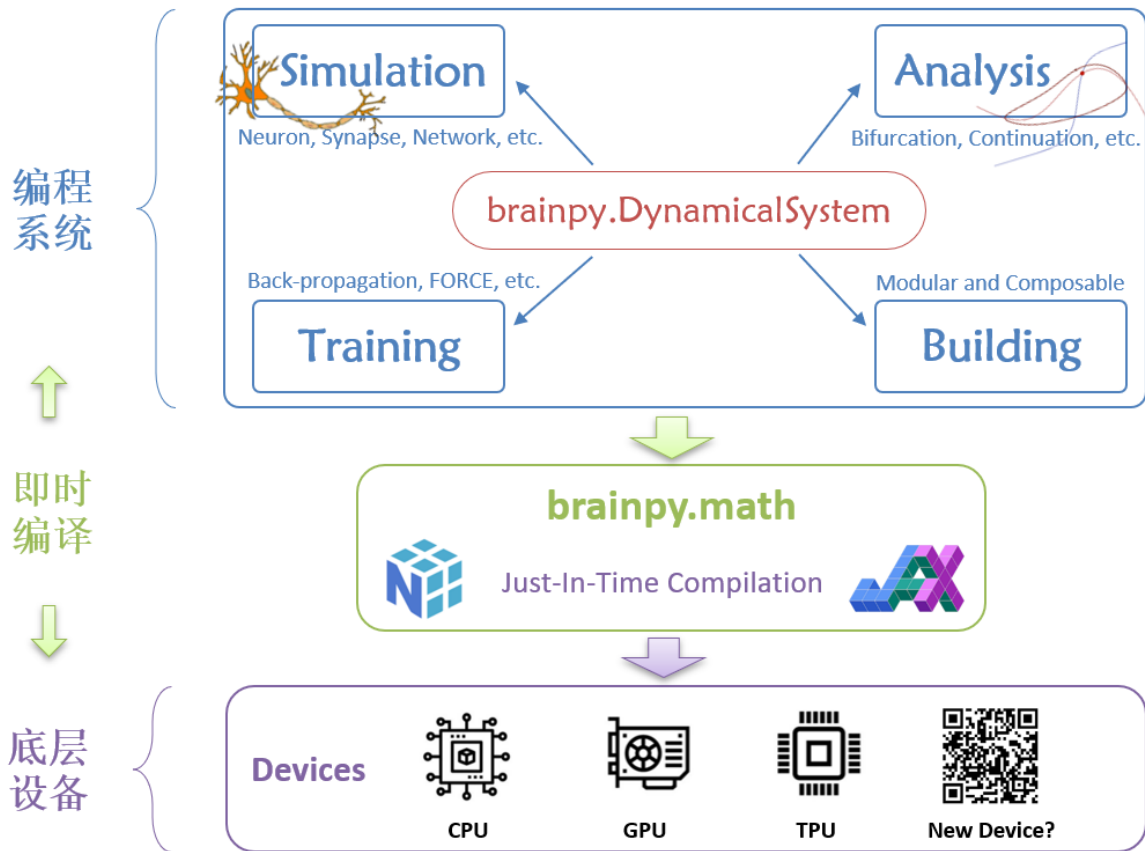
Transparent

Extensible

Efficient

- ✓ 到目前没有任何一个计算神经科学内的框架能像 PyTorch、TensorFlow 一样允许用户自主灵活编程
- ✓ 是时候改变计算神经科学的编程范式了！

A Just-In-Time compilation approach for brain modeling



Model definition, building and simulation are all done in Python.

- Easy to learn and use
- Efficient
- Flexible
- Transparent
- Extensible

HH model as an example for flexible customization

```
1 class HH(bp.NeuGroup):
2     def __init__(self, size, ENa=50., gNa=120., EK=-77., gK=36., EL=-54.387,
3        gL=0.03, V_th=20., C=1.0, method='exponential_euler', **kwargs):
4         # 初始化父类
5         super(HH, self).__init__(size=size, **kwargs)
6
7         # 定义神经元参数
8         self.ENa = ENa
9         self.EK = EK
10        self.EL = EL
11        self.gNa = gNa
12        self.gK = gK
13        self.gL = gL
14        self.C = C
15        self.V_th = V_th
```

Initialize Parameters

```
16
17        # 定义神经元变量
18        self.V = bm.Variable(-65. * bm.ones(self.num)) # 膜电位
19        self.m = bm.Variable(0.5 * bm.ones(self.num)) # 离子通道m
20        self.h = bm.Variable(0.6 * bm.ones(self.num)) # 离子通道h
21        self.n = bm.Variable(0.32 * bm.ones(self.num)) # 离子通道n
22        self.input = bm.Variable(bm.zeros(self.num)) # 神经元接收到的输入电流
23        self.spike = bm.Variable(bm.zeros(self.num, dtype=bool)) # 神经元的发放状态
24        self.t_last_spike = bm.Variable(bm.ones(self.num) * -1e7) # 神经元上次发放的时刻
```

Initialize
Variables

```
@bp.odeint(method='exponential_euler')
```

```
def integral(self, V, m, h, n, t, Iext):
```

```
alpha = 0.1 * (V + 40) / (1 - bm.exp(-(V + 40) / 10))
beta = 4.0 * bm.exp(-(V + 65) / 18)
dmdt = alpha * (1 - m) - beta * m
```

$$\frac{dm}{dt} = \alpha_m(1 - m) - \beta_m$$

```
alpha = 0.07 * bm.exp(-(V + 65) / 20.)
beta = 1 / (1 + bm.exp(-(V + 35) / 10))
dhdt = alpha * (1 - h) - beta * h
```

$$\frac{dh}{dt} = \alpha_h(1 - h) - \beta_h$$

```
alpha = 0.01 * (V + 55) / (1 - bm.exp(-(V + 55) / 10))
beta = 0.125 * bm.exp(-(V + 65) / 80)
dndt = alpha * (1 - n) - beta * n
```

$$\frac{dn}{dt} = \alpha_n(1 - n) - \beta_n$$

```
I_Na = (self.gNa * m ** 3.0 * h) * (V - self.ENa)
I_K = (self.gK * n ** 4.0) * (V - self.EK)
I_leak = self.gL * (V - self.EL)
dVdt = (- I_Na - I_K - I_leak + Iext) / self.C
```

$$C \frac{dV}{dt} = -(\bar{g}_{Na} m^3 h (V - E_{Na}) + \bar{g}_K n^4 (V - E_K) + g_{leak} (V - E_{leak})) + I(t)$$

```
return dVdt, dmdt, dhdt, dndt
```

```
def update(self, _t, _dt)  $\longleftrightarrow x(t + dt) = f(x(t), t, dt)$ 
```

```
V, m, h, n = self.integral(self.V, self.m, self.h, self.n, _t, self.input, dt=_dt) # 更新变量值
```

```
self.spike[:] = bm.logical_and(self.V < self.V_th, V >= self.V_th) # 判断神经元是否产生膜电位
```

```
self.t_last_spike[:] = bm.where(self.spike, _t, self.t_last_spike) # 更新神经元发放的时间
```

```
self.V[:] = V
```

```
self.m[:] = m
```

```
self.h[:] = h
```

```
self.n[:] = n
```

```
self.input[:] = 0. # 重置神经元接收到的输入
```

Channels	<u>INa</u>	IK
Ca	<u>ICaT</u>	<u>ICaN</u>

Neuron models	Hodgkin-Huxley Model	Morris–Lecar model
LIF model	Quadratic LIF model	Exponential LIF model
Resonate-fire model	Izhikevich model	Generalized LIF model
Wilson Cowan model	Hindmarsh Rose model	And others ...

Synapse models		Exponential synapse		Voltage <u>jump</u> synapse
Dual Exponential synapse		Alpha synapse		Short-term plasticity
GABA _A	GABA _B	NMDA	AMPA	And others ...

Network models	CANN network	E/I balanced network
Liquid-state machine	Echo state network	And others ...

Learning rules	STDP	Oja	BCM	And others ...
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