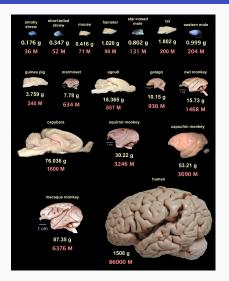
The NeuroML ecosystem for standardised multi-scale modelling in neuroscience

Ankur Sinha Silver Lab Department of Neuroscience, Physiology, & Pharmacology University College London

2024-02-26

An understanding of the brain

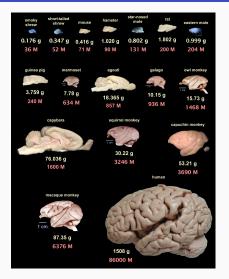


- ~86B neurons
- ~100T synapses
- also ~85B glia

¹Herculano-Houzel, S. The human brain in numbers: a linearly scaled-up primate brain. Frontiers in human neuroscience **3**, 31 (2009)

¹von Bartheld, C. S. et al. The search for true numbers of neurons and glial cells in the human brain: A review of 150 years of cell counting. Journal of Comparative Neurology 524, 3865–3895. ISSN: 1096-9861 (June 2016)

An understanding of the brain



- specialised circuits
- different neuronal types
- synaptic connections
- complex sub-cellular processes

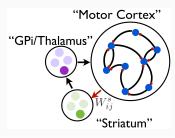
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Experiments provide a window into the brain

Multiple scales of experiments/data sources go here

Models test & unify experimental results; generate hypotheses





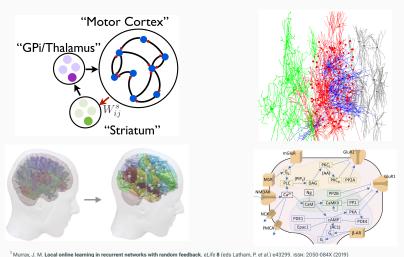
¹ Murray, J. M. Local online learning in recurrent networks with random feedback. eLife 8 (eds Latham, P. et al.) e43299. ISSN: 2050-084X (2019)

¹ Schirner, M. et al. Learning how network structure shapes decision-making for bio-inspired computing. Nature Communications 14. ISSN: 2041-1723 (May 2023)

¹Yao, H. K. et al. Reduced inhibition in depression impairs stimulus processing in human cortical microcircuits. Cell Reports 38. ISSN: 2211-1247. https://doi.org/10.1016/j.celrep.2021.110232 (Jan. 2022)

¹ Mäki-Marttunen, T. et al. A unified computational model for cortical post-synaptic plasticity. eLife 9 (eds Shouval, H. Z. et al.) e55714. ISSN: 2050-084X. https://doi.org/10.7554/eLife.55714 (July 2020)

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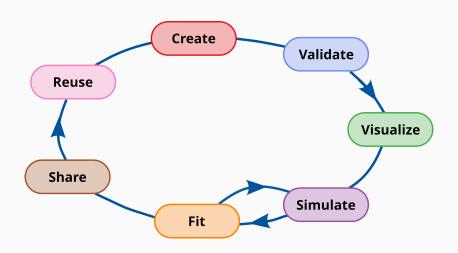
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A *mechanistic* understanding of the brain requires biophysically detailed modelling

The model life cycle



- many specialist tools:
 - NEURON, NEST, Brian, GENESIS, MOOSE, STEPS, ANNarchy, TVB, LFPy, NeuroLib, EDEN, Arbor, NetPyNE...

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 - custom machine readable internal representations:
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- but:
 - different APIs, syntax:
 - · increased difficulty for users
 - not well defined model descriptions:
 - · models cannot be validated
 - custom machine readable internal representations:
 - · cannot be easily inspected/analysed
 - ad-hoc utilities:
 - · cannot be used with all tools

Makes computational neuroscience models

less

FAIR

(Findable, Accessible, Interoperable, Reusable)

Standards enable FAIR neuroscience





COMBINE

¹Abrams, M. B. et al. A Standards Organization for Open and FAIR Neuroscience: the International Neuroinformatics Coordinating Facility. Neuroinformatics 20, 25–36. ISSN: 1559-0089. https://doi.org/10.1007/s12021-020-09509-0 (2022): https://incf.org

¹COmputational Modeling in Blology NEtwork (COMBINE): https://co.mbine.org/

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COMBINE





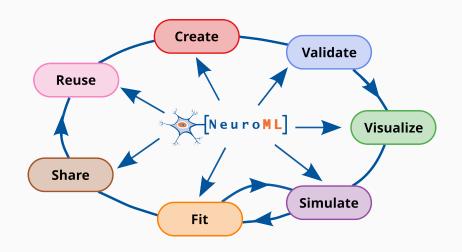




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NeuroML ecosystem supports all stages of the model cycle



NeuroML ecosystem

- standard/specification
- software ecosystem

NeuroML schema/standard

Model specification (XSD)

- elements
- attributes
- hierarchical relationships

NeuroML schema/standard

Model specification (XSD)

- elements
- attributes
- hierarchical relationships

Dynamics (LEMS)

• dynamical behaviour

NeuroML schema/standard: XSD

Way of specifying the structure of an XML document.

 allows defining types and extensions/restrictions on types to create new types.

NeuroML schema/standard: XSD

Way of specifying the structure of an XML document.

 allows defining types and extensions/restrictions on types to create new types.

One can validate a model description against the schema before simulation

NeuroML schema/standard: specification: XSD

```
<xs:simpleType name="Nml2Quantity voltage"> <!-- For params with dimension voltage -->
 <xs:restriction base="xs:string">
   <xs:pattern value="-?([0-9]*(\.[0-9]+)?)([eE]-?[0-9]+)?[\s]*(V|mV)"/>
  </r></restriction>
</xs:simpleType>
<xs:complexType name="Izhikevich2007Cell">
  <xs:annotation>
    <xs:documentation>Cell based on ...>
    </r></re>
  </r></r></re></re>
  <xs:complexContent>
   <xs:extension base="BaseCellMembPotCap">
      <xs:attribute name="v0" type="Nml2Quantity_voltage" use="required"/>
      <xs:attribute name="k" type="Nm12Quantity_conductancePerVoltage" use="required"/>
     <xs:attribute name="vr" type="Nml2Quantity_voltage" use="required"/>
     <xs:attribute name="vt" type="Nml2Quantity_voltage" use="required"/>
     <xs:attribute name="vpeak" type="Nml2Quantity voltage" use="required"/>
     <xs:attribute name="a" type="Nml2Quantity_pertime" use="required"/>
     <xs:attribute name="b" type="Nm12Quantity_conductance" use="required"/>
     <xs:attribute name="c" type="Nml2Quantity voltage" use="required"/>
     <xs:attribute name="d" type="Nml2Quantity_current" use="required"/>
    </r></re></re>
  </xs:complexContent>
</xs:complexType>
```

NeuroML schema/standard: LEMS

Low Entropy Model Specification language

- domain independent
- machine readable
- allows creation of "Component Types" (classes) from which "Components" (objects) can be instantiated by providing the necessary parameters
- provides a reference implementation/simulator

¹ Cannon, R. C. et al. LEMS: a language for expressing complex biological models in concise and hierarchical form and its use in underpinning NeuroML 2. Frontiers in Neuroinformatics 8 (2014)

NeuroML schema/standard: LEMS

Low Entropy Model Specification language

- domain independent
- machine readable
- allows creation of "Component Types" (classes) from which "Components" (objects) can be instantiated by providing the necessary parameters
- provides a reference implementation/simulator
- but is not aware of the cable equation (required for multi-compartmental neuron models)

¹ Cannon, R. C. et al. LEMS: a language for expressing complex biological models in concise and hierarchical form and its use in underpinning NeuroML 2. Frontiers in Neuroinformatics 8 (2014)

NeuroML schema/standard: dynamics (LEMS)

```
<ComponentType name="izhikevich2007Cell" extends="baseCellMembPotCap"</pre>
    description="Cell based ...">
    <Parameter name="v0" dimension="voltage" description="Initial membrane potential"/>
    Defined in baseCellMembPotCap:
    <Parameter name="C" dimension="capacitance"/>
    <Parameter name="k" dimension="conductance_per_voltage"/>
    <Parameter name="vr" dimension="voltage" description="Resting membrane potential"/>
    <Parameter name="vt" dimension="voltage" description="Spike threshold"/>
    <Parameter name="vpeak" dimension="voltage" description="Peak action potential value"/>
    <Parameter name="a" dimension="per_time" description="Time scale of recovery variable u"/>
    <Parameter name="b" dimension="conductance" description="Sensitivity of recovery variable u to subthreshold</pre>

    fluctuations of membrane potential v"/>

    <Parameter name="c" dimension="voltage" description="After-spike reset value of v"/>
    <Parameter name="d" dimension="current" description="After-spike increase to u"/>
    <a href="attachments">Attachments</a> name="synapses" type="basePointCurrent"/>
    <Exposure name="u" dimension="current" description="Membrane recovery variable"/>
    <Dynamics><!-- snipped --> </Dynamics>
</ComponentType>
```

NeuroML schema/standard: XSD and LEMS

```
<xs:attribute name="v0" type="Nml2Quantity_voltage" use="required"/>
<xs:attribute name="k" type="Nml2Quantity conductancePerVoltage" use="required"/>
<xs:attribute name="vr" type="Nml2Quantity_voltage" use="required"/>
<xs:attribute name="vt" type="Nm12Quantity_voltage" use="required"/>
<xs:attribute name="vpeak" type="Nml2Quantity voltage" use="required"/>
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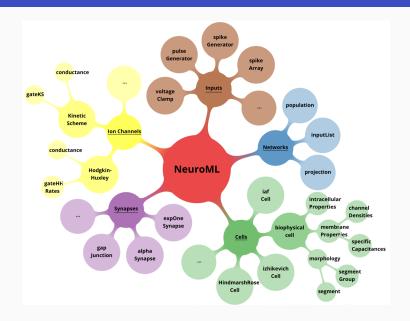
→ fluctuations of membrane potential v"/>

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<Parameter name="d" dimension="current" description="After-spike increase to u"/>
```

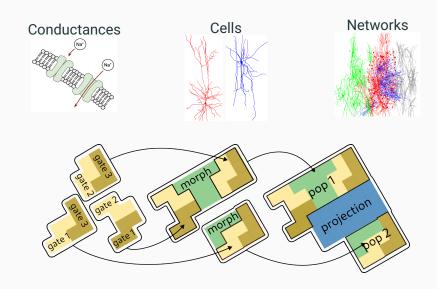
NeuroML schema/standard: dynamics (LEMS)

```
<ComponentType name="izhikevich2007Cell" extends="baseCellMembPotCap"</pre>
    description="Cell based ...">
    <!-- snipped -->
    <Dynamics>
        <StateVariable name="v" dimension="voltage" exposure="v"/>
        <StateVariable name="u" dimension="current" exposure="u"/>
        <DerivedVariable name="iSyn" dimension="current" exposure="iSyn" select="synapses[*]/i" reduce="add" />
        <DerivedVariable name="iMemb" dimension="current" exposure="iMemb" value="k * (v-vr) * (v-vt) + iSvn - u"/>
        <TimeDerivative variable="v" value="iMemb / C"/>
        <TimeDerivative variable="u" value="a * (b * (v-vr) - u)"/>
        <OnStart>
            <StateAssignment variable="v" value="v0"/>
            <StateAssignment variable="u" value="0"/>
        </OnStart>
        <OnCondition test="v .gt. vpeak">
            <StateAssignment variable="v" value="c"/>
            <StateAssignment variable="u" value="u + d"/>
            <EventOut port="spike"/>
        </Orderation>
    </Dynamics>
</ComponentType>
```

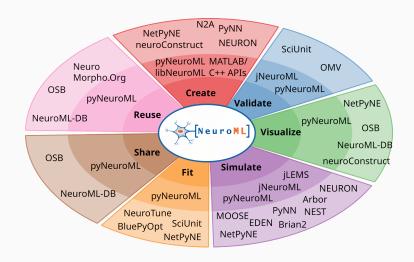
NeuroML provides a set of curated model elements



NeuroML is modular, structured and hierarchical



NeuroML: software ecosystem



NeuroML: software ecosystem: core tools

• Figure 4

NeuroML: create models

- Figure 5
- Code example

NeuroML: validate models

• Figure 6

NeuroML: visualise models

- Figure 7
- Figure 8
- Figure 9

NeuroML: simulate models

• Example simulation: neuron/netpyne

NeuroML: fit models

- Figure from docs
- Mention inspyred

NeuroML: share and re-use models

• GitHub, OSBv1, OSBv2, NeuroML-DB

NeuroML: the APIs

• Python API

NeuroML: Documentation

• Jupyterbook

NeuroML: projects

• GSoC, Outreachy, good computer science students

But, too many standards?

• XKCD here.