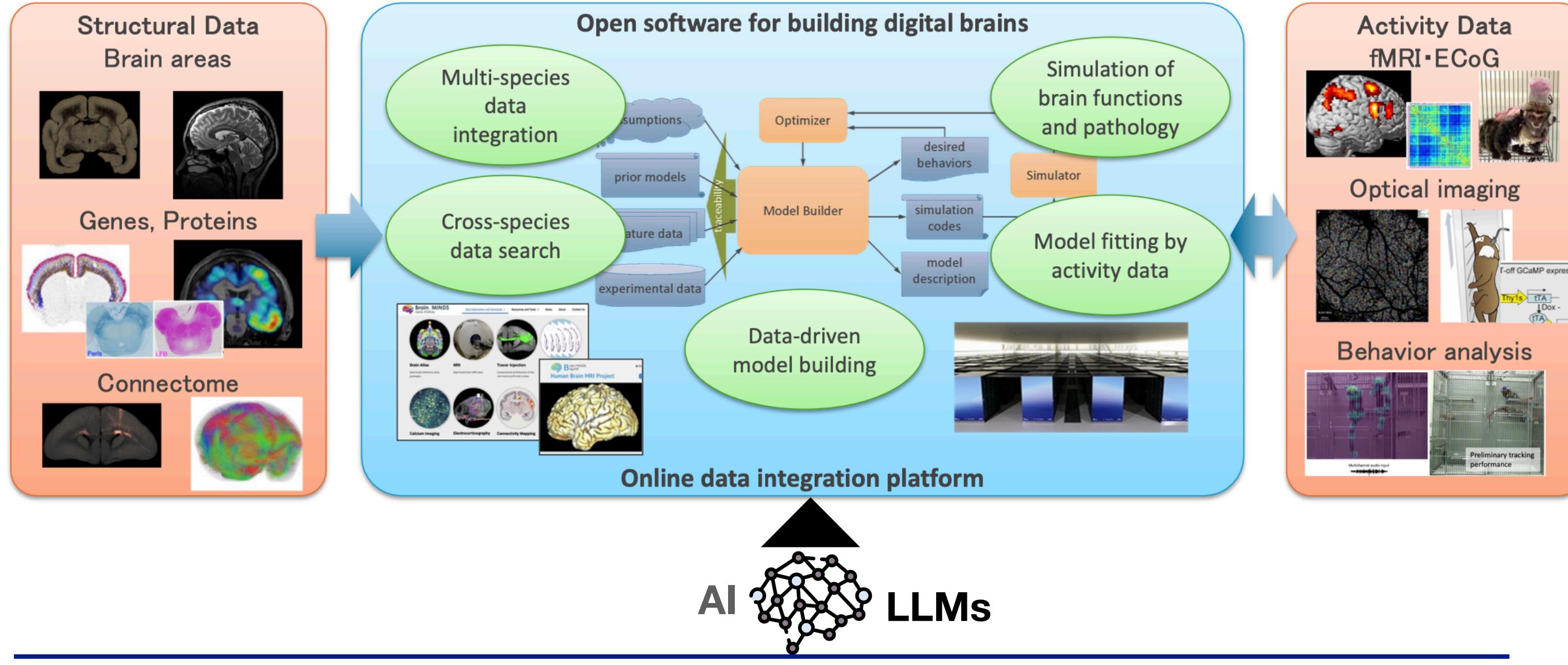


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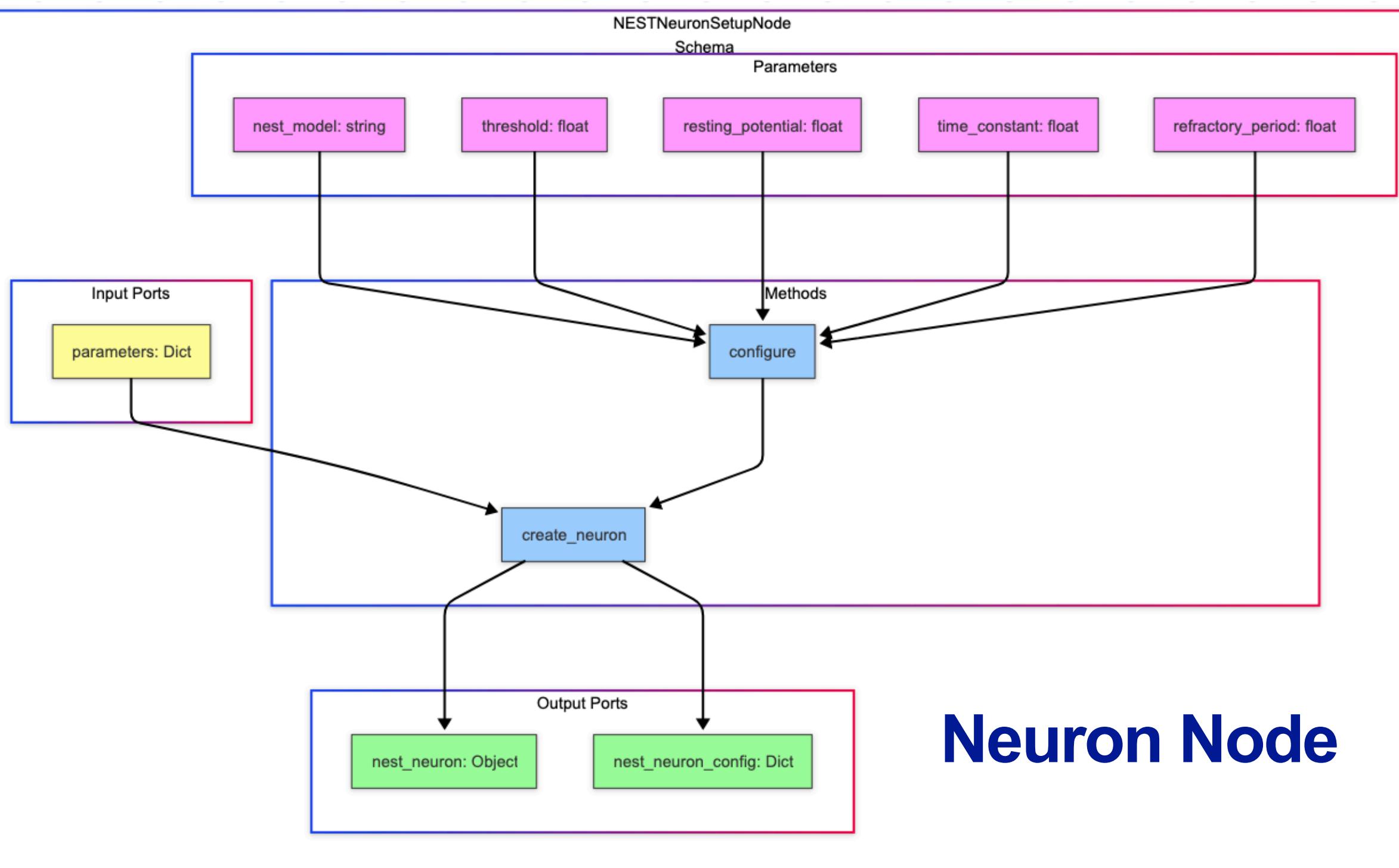
## Brain/MINDS 2.0 - The Digital Brain Project

**Goals:** open software for building digital brains  
-online platform for data integration, model building and simulations-



### NeuroWorkflow Node's Core Components

- The node is the fundamental building block of NeuroWorkflow. It defines **inputs, outputs, parameters, and processing steps** (functions). It handles data flow and type validation and it can be **extended to create specialized/custom nodes**.
- The diagram below illustrates the structure of a NEST Neuron Setup Node class in the NeuroWorkflow package which **defines the node as a python class with a certain schema**. Here's a breakdown of the components:



### In Python a Node is defined as a Class:

```
class NESTNeuronSetupNode(Node):
    """Node for creating and configuring neuron models.
    This class represents a neuron model with configurable parameters.
    In a real implementation, this could be a more complex model or
    interface with external neural simulators."""

    NODE_DEFINITION = NodeDefinitionSchema(
        type="neuron_setup",
        description="Creates and configures a neuron model in NEST",
        parameters={
            "nest_model": ParameterDefinition(
                default_value="iaf_psc_alpha",
                description="Neuron model name in NEST"
            ),
            "threshold": ParameterDefinition(
                default_value=-55.0,
                description="Firing threshold (mV)",
                constraints={"min": -70.0, "max": -40.0},
                optimizable=True,
                optimization_range=(-65.0, -45.0)
            ),
            "resting_potential": ParameterDefinition(
                default_value=-70.0,
                description="Resting membrane potential (mV)",
                constraints={"min": -90.0, "max": -60.0},
                optimizable=True,
                optimization_range=(-90.0, -60.0)
            ),
            "time_constant": ParameterDefinition(
                default_value=20.0,
                description="Membrane time constant (ms)",
                constraints={"min": 5.0, "max": 50.0},
                optimizable=True,
                optimization_range=(10.0, 30.0)
            ),
            "refractory_period": ParameterDefinition(
                default_value=2.0,
                description="Refractory period (ms)",
                constraints={"min": 1.0, "max": 5.0}
            )
        },
        inputs={
            "parameters": PortDefinition(
                type=PortType.DICT,
                description="Parameters to configure the neuron",
            )
        }
    )

    def create_neuron(self, parameters: Optional[Dict[str, Any]] = None) -> Dict[str, Any]:
        """Create and configure a neuron model.

        Args:
            parameters: Optional parameters to override defaults

        Returns:
            Dictionary with neuron model and configuration
        """
        # If parameters are provided, configure the node
        if parameters:
            self.configure(**parameters)

        # Create nest neuron object
        neuro_params = {
            "V_L": self._parameters["threshold"],
            "E_L": self._parameters["resting_potential"],
            "tau_m": self._parameters["time_constant"],
            "t_ref": self._parameters["refractory_period"],
        }

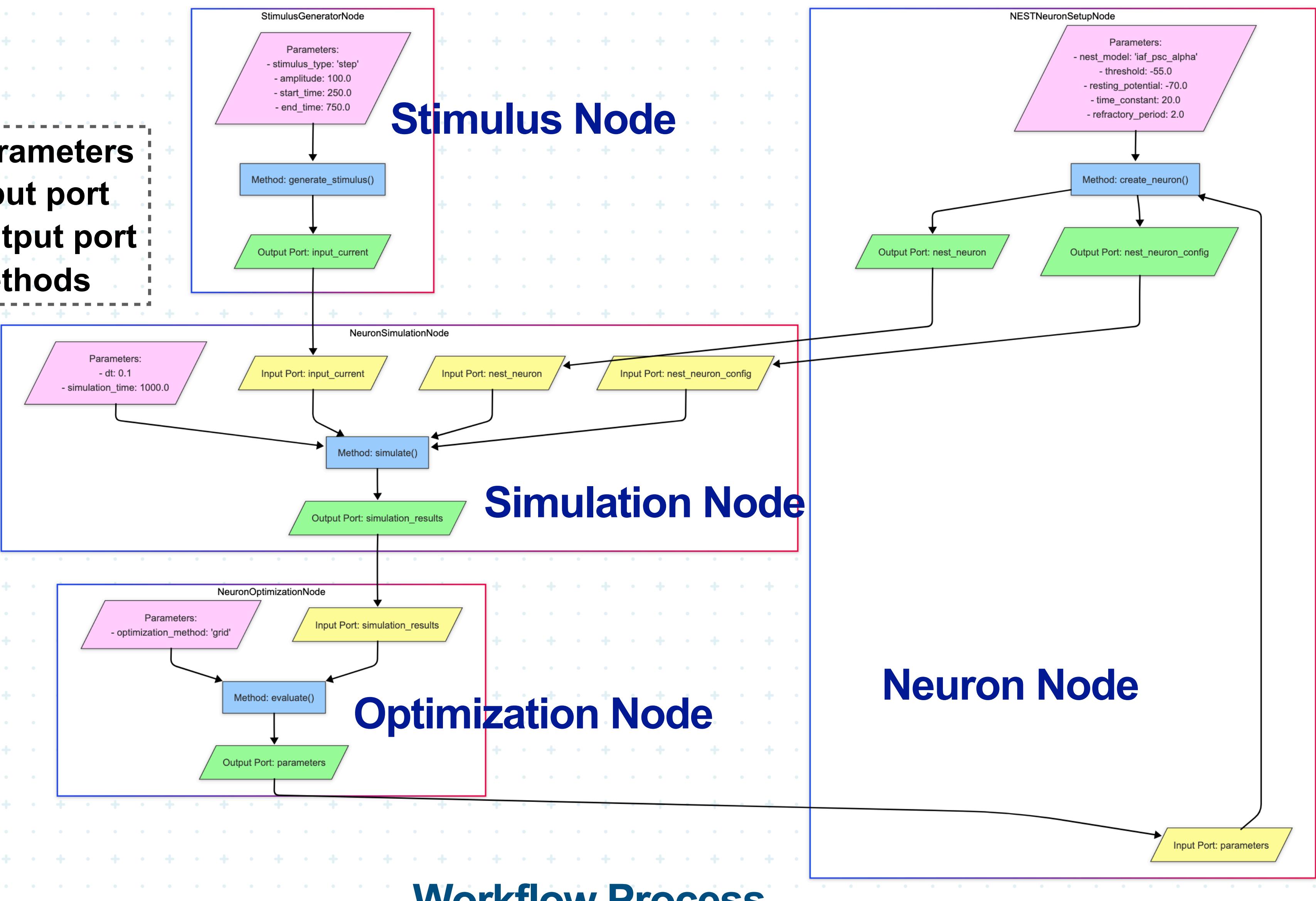
        nest.set_verbosity("M_ERROR")
        nest.ResetKernel()
        neuron = nest.Create(self._parameters["nest_model"], params=neuro_params)
```

## The Digital Brain as a graph-based framework

- The Brain/MINDS 2.0 Digital Brain aims to create a framework that supports **interoperability with open data and popular neuroscience tools** such as TVB, NEST, NEURON, and others, leveraging their Python interfaces.
- Modern brain modeling and simulation workflows often require complex sequences of data processing, model configuration, and analysis steps. While **powerful tools** exist, they frequently **demand advanced programming knowledge, limiting accessibility and collaboration potential**. Besides that, **shared models are frequently difficult to understand, execute and expand**.
- We present a python-based graph framework that **transforms complex scientific workflows into modular, interpretable, reusable components** through a node-edge architecture.

### Exploring workflows for model's optimization

- This diagram illustrates a **preliminary workflow for the optimization of a single neuron parameters**, showing how different nodes interact via ports (edges).



### Workflow Process

- Initialization:**
  - Nodes are configured with initial parameters
- Execution Loop:**
  - Neuron Node creates a neuron model
  - Stimulus Node generates an input current
  - Simulation Node runs the simulation with the neuron and input current
  - Optimization Node evaluates the results and suggests new parameters
  - The new parameters are fed back to Neuron Node
  - The loop continues until optimization criteria are met

This simple workflow **demonstrates the power of the node-based architecture**, where each node handles a specific aspect of the neural simulation process, and data flows between nodes through well-defined ports.

### Community-based approach for building digital brains

- The model builder aims to be **composed by several custom nodes built by BM2 researchers and collaborators**.
- Several **easy-to-understand complex brain models** and workflows may be built and **deployed easily**.
- Data integration and data analysis workflows can be deployed as well.
- Machine learning / AI dedicated nodes can also be created and integrated.

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