

xml2jupyter: Mapping parameters between XML and Jupyter widgets

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Software

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Summary

Jupyter Notebooks (Kluyver et al. 2016, Perkel (2018)) provide executable documents (in a variety of programming languages) that can be run in a web browser. When a notebook contains graphical widgets, it becomes an easy-to-use graphical user interface (GUI). Many scientific simulation packages use text-based configuration files to provide parameter values. Manually editing these files, to explore how different values affect a simulation, can be burdensome. `xml2jupyter` is a Python package that bridges this gap. It provides a mapping between configuration files, formatted in the Extensible Markup Language (XML), and Jupyter widgets. Widgets are automatically generated from the XML file and these can, optionally, be incorporated into a larger GUI for a simulation package. Users modify parameter values via the widgets and the values are written to the XML configuration file which is input to the simulation's command-line interface. `xml2jupyter` has been tested using PhysiCell (Ghaffarizadeh et al. 2018), an open source, agent-based simulator for biology, and is being used by students for classroom and research projects. In addition, we use `xml2jupyter` to help create Jupyter GUIs for PhysiCell-related applications running on nanoHUB (Madhavan et al. 2013).

A PhysiCell configuration file defines model-specific `<user_parameters>` in XML. Each parameter element consists of its name with attributes, defining its data *type*, *units* (optional), *description* (optional), whether the widget should be *hidden* (optional), and the parameter's default value. The attributes will determine the appearance and behavior of the Jupyter widget. For numeric widgets (the most common type for PhysiCell), `xml2jupyter` will calculate a delta step size as a function of the default value and this step size will be used by the widget's graphical increment/decrement feature.

To illustrate, we show the following simple XML example, containing each of the four allowed data types (currently) and the various attributes:

```
<PhysiCell_settings>
  <user_parameters>
    <radius type="double" units="micron"
      description="initial tumor radius">250.0
    </radius>
    <threads type="int">8</threads>
    <color type="string" hidden="true">red</color>
    <fix_persistence type="bool">True</fix_persistence>
  </user_parameters>
</PhysiCell_settings>
```

When we map this into Jupyter widgets, we obtain the rendered results in Figure 1. Notice the `color` parameter is not displayed since we specified it should be hidden in the XML.

radius	250	micron	initial tumor radius
threads	8		
fix_persistence	<input checked="" type="checkbox"/>		

Figure 1: Simple example of XML parameters as Jupyter widgets.

The name of the other parameters, their values, and attributes, if present, are displayed in rows (as disabled Jupyter button widgets). Using alternating row colors (“zebra stripes”) helps visually match associated fields and avoid changing the wrong parameter value. For numeric widgets (type “int” or “double”), we compute a delta step value based on the magnitude (log) of the initial value. For example, the `radius` widget will have a step value of 10, whereas `threads` will have a step value of 1.

For a more realistic example, consider the `config_biorobots.xml` configuration file (found in the `config_samples` directory). The XML elements in the `<user_parameters>` block include the (optional) `description` attribute which briefly describes the parameter and is displayed in another widget. To demonstrate `xml2jupyter` on this XML file, one would: 1) clone or download the repository, 2) copy the XML configuration file to the root directory, and 3) run the `xml2jupyter.py` script, providing the XML file as a argument.

```
$ cp config_samples/config_biorobots.xml .
$ python xml2jupyter.py config_biorobots.xml
```

The `xml2jupyter.py` script parses the XML and generates a Python module, `user_params.py`, containing the Jupyter widgets, together with methods to populate their values from the XML and write their values back to the XML. To “validate” the widgets were generated correctly, one could, minimally, open `user_params.py` in an editor and inspect it.

But to actually see the widgets rendered in a notebook, we provide a simple test:

```
$ python xml2jupyter.py config_biorobots.xml test_user_params.py
$ jupyter notebook test_gui.ipynb
```

This should display a minimal notebook in your browser and, after selecting **Run all** in the **Cell** menu, you should see the notebook shown in Figure 2.

PhysiCell Jupyter GUI

Our ultimate goal is to generate a fully functional GUI for PhysiCell users. `xml2jupyter` provides one important piece of this - dynamically generating widgets for custom user parameters for a model. With other Python modules that provide additional components (tabs) of the GUI, common to all PhysiCell models, a user can configure, run, and visualize output from a simulation. Two tabs that provide visualization of output files are shown below with results from the *biorobots* simulation. Note that some of the required modules are not available in the Python standard library, e.g., Matplotlib (Hunter 2007) and SciPy. We provide instructions for installing these additional dependencies in the repository README.

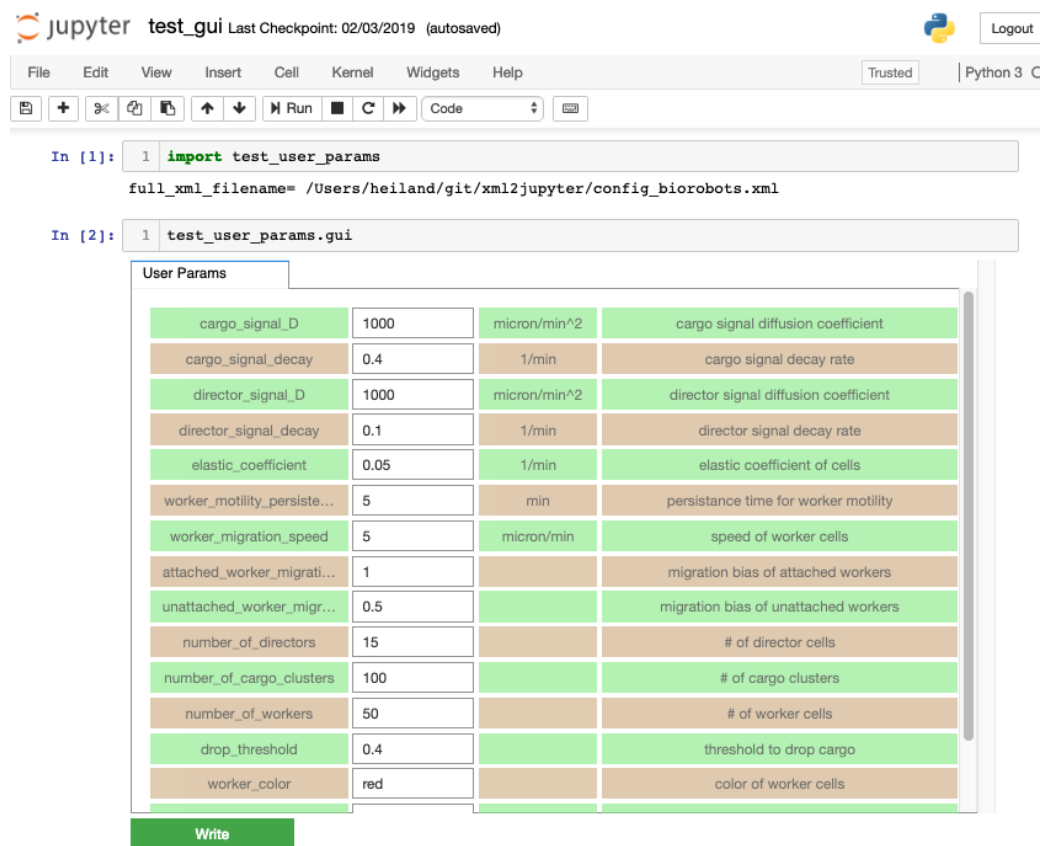


Figure 2: The biorobots parameters rendered as Jupyter widgets.

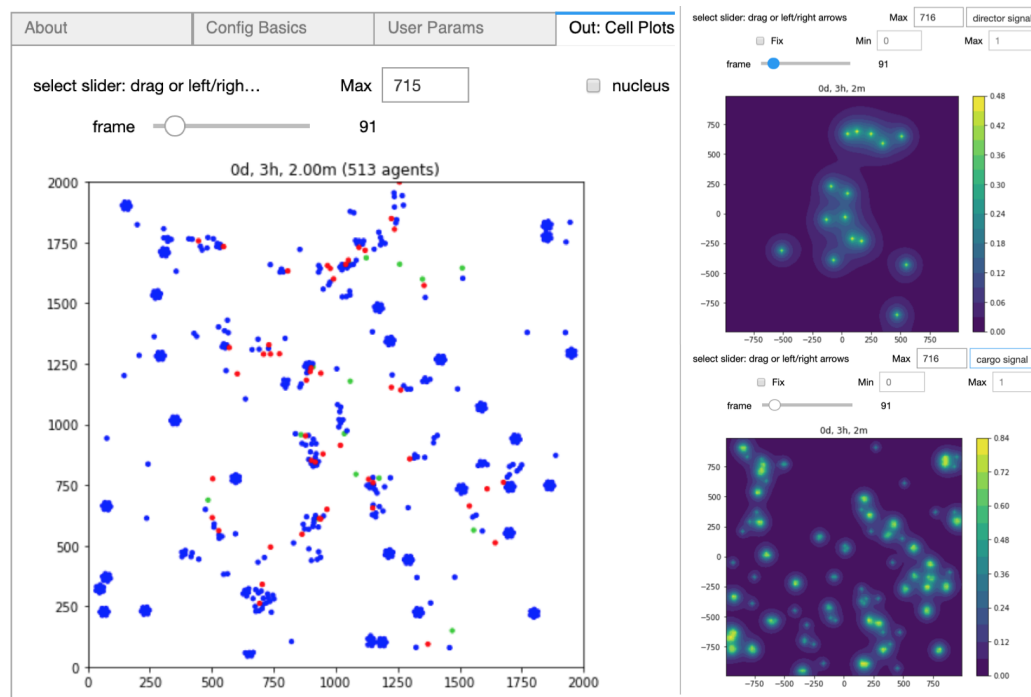
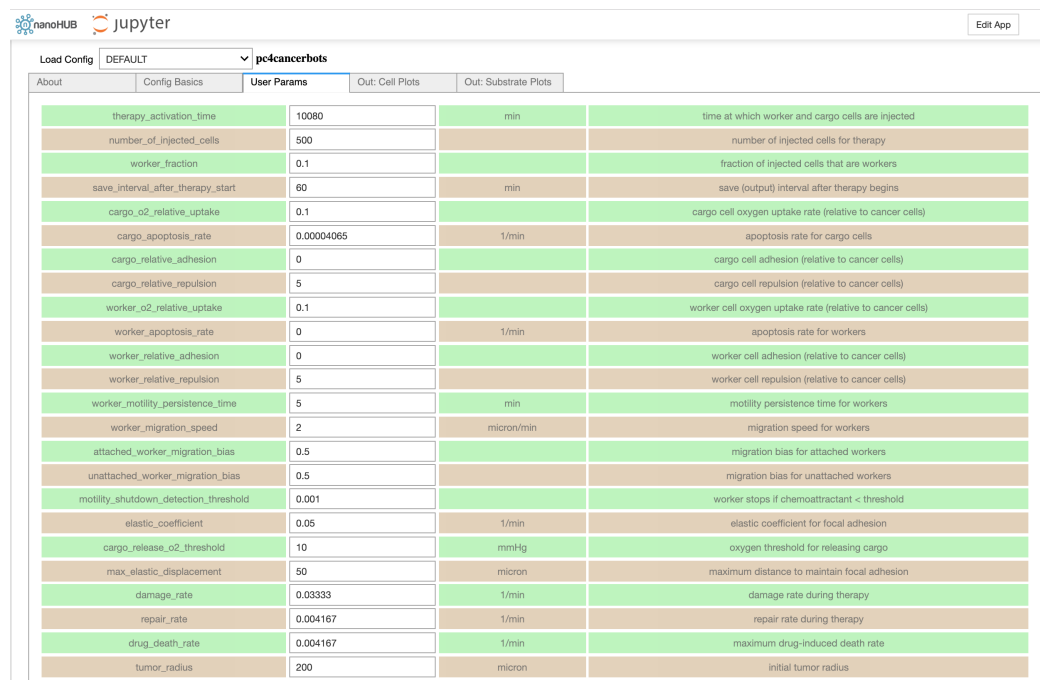


Figure 3: Plotting the biorobots (cells; left) and signals (substrates; right).



Parameter	Value	Unit	Description
therapy_activation_time	10080	min	time at which worker and cargo cells are injected
number_of_injected_cells	500		number of injected cells for therapy
worker_fraction	0.1		fraction of injected cells that are workers
save_interval_after_therapy_start	60	min	save (output) interval after therapy begins
cargo_o2_relative_uptake	0.1		cargo cell oxygen uptake rate (relative to cancer cells)
cargo_apoptosis_rate	0.00004065	1/min	apoptosis rate for cargo cells
cargo_relative_adhesion	0		cargo cell adhesion (relative to cancer cells)
cargo_relative_repulsion	5		cargo cell repulsion (relative to cancer cells)
worker_o2_relative_uptake	0.1		worker cell oxygen uptake rate (relative to cancer cells)
worker_apoptosis_rate	0	1/min	apoptosis rate for workers
worker_relative_adhesion	0		worker cell adhesion (relative to cancer cells)
worker_relative_repulsion	5		worker cell repulsion (relative to cancer cells)
worker_motility_persistence_time	5	min	motility persistence time for workers
worker_migration_speed	2	micron/min	migration speed for workers
attached_worker_migration_bias	0.5		migration bias for attached workers
unattached_worker_migration_bias	0.5		migration bias for unattached workers
motility_shutdown_detection_threshold	0.001		worker stops if chemoattractant < threshold
elastic_coefficient	0.05	1/min	elastic coefficient for focal adhesion
cargo_release_o2_threshold	10	mmHg	oxygen threshold for releasing cargo
max_elastic_displacement	50	micron	maximum distance to maintain focal adhesion
damage_rate	0.03333	1/min	damage rate during therapy
repair_rate	0.004167	1/min	repair rate during therapy
drug_death_rate	0.004167	1/min	maximum drug-induced death rate
tumor_radius	200	micron	initial tumor radius

Figure 4: The cancer biorobots parameters in a nanoHUB Jupyter application.

Extensions and Discussion

We hope others will be inspired to extend the core idea of this project to other text-based configuration files. XML is only one of several data-interchange formats. It just happens to be the one of interest to us for PhysiCell. And while the additional Python modules that provide visualization are also tailored to PhysiCell output, they can serve as templates for other file formats and provide similar functionality.

xml2jupyter has helped us port PhysiCell-related Jupyter tools to nanoHUB, a scientific cloud for nanoscience education and research that includes running interactive simulations in a browser. For example, Figure 4 shows the xml2jupyter-generated *User Params* tab in our [pc4cancerbots](#) tool running on nanoHUB. Figure 5 shows the cells (upper-left) and three different substrate plots for this same tool. This particular model and simulation is described in this [video](#).

Other PhysiCell-related nanoHUB tools that have been created using xml2jupyter include [pc4heterogen](#), [pcISA](#), and [pc4cancerimmune](#). Readers can create an account on nanoHUB and run these simulations for themselves. We encourage students to use xml2jupyter to create their own nanoHUB tools of PhysiCell models that can 1) be run and evaluated by the instructor, 2) be shared with others, and 3) become part of a student's living portfolio. (Another repository, <https://github.com/rheiland/tool4nanobio>, provides instructions and scripts to help generate a full GUI from an existing PhysiCell model.)

We welcome suggestions and contributions to xml2jupyter. For example, currently, we arrange the generated parameter widgets vertically, one row per parameter. This is an appropriate layout for an educational setting. But if a GUI will be used by researchers who are already familiar with the parameters, it may be preferable to generate a more compact layout of widgets, e.g., in a matrix with only the parameter names and values.

Also, we currently provide just 2-D visualizations of (spatial) data. In the near future, we will provide visualizations of 3-D models and welcome suggestions from the community.

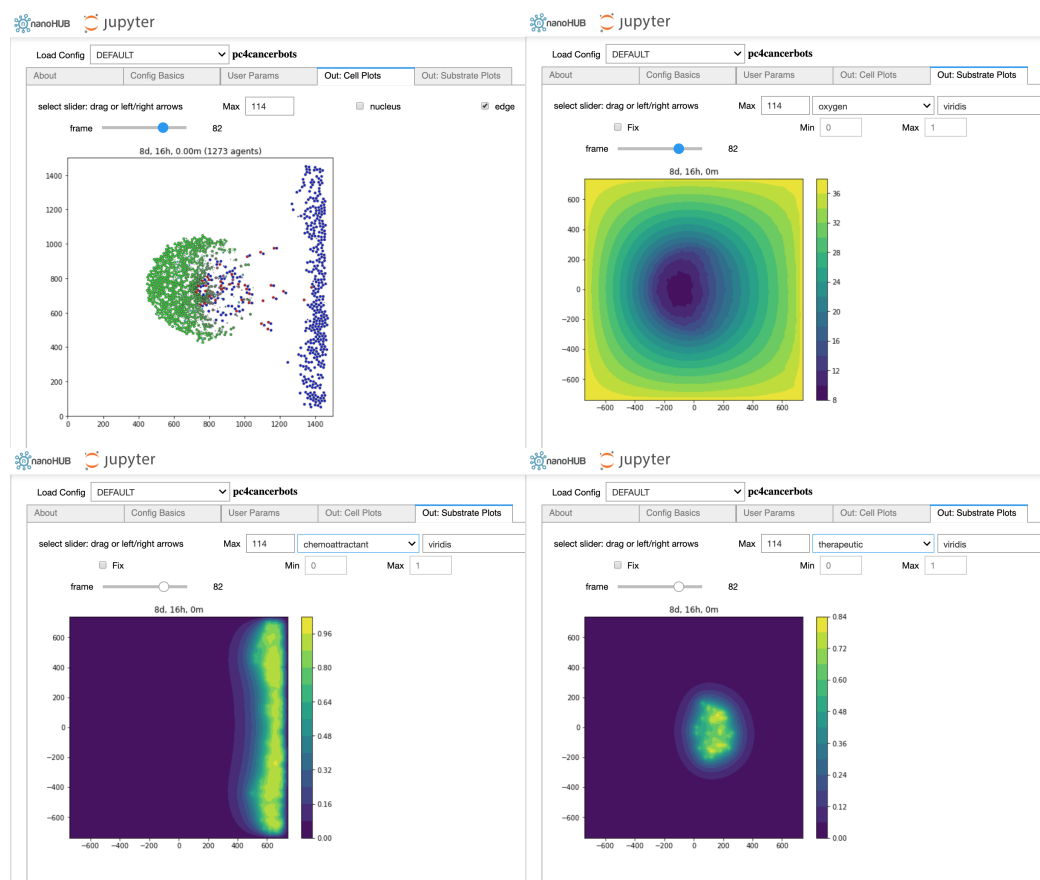


Figure 5: The cancer biorobots Jupyter notebook on nanoHUB.

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