PhysiCell data analysis with physicell data loader



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PhysiCell

classical mechanics based cell and tissue simulator

Agent based modeling:

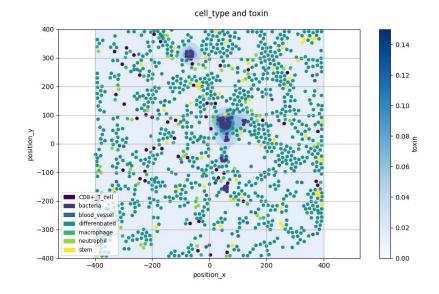
- each cell is an agent.
- agents have states (parameters and features).
- agents run functions.

Diffusion reaction solver:

- substrate source, sink, and decay.
- finite volume method (BioFVM).

Multiscale modeling

- space: intracellular < cell < domain (tissue)
- time: diffusion 0.01[min] < mechanics 0.1[min] < phenotype 6[min]



python data loader (2019-9-28)

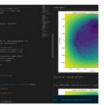
MathCancer Blog

MULTICELLDS, BIOFVM, PHYSICELL, MULTICELLULAR SYSTEMS BIOLOGY, AND CANCER

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PhysiCell Tools: python-loader

Posted in coding, MathCancer, MultiCelIDS, open source, PhysiCell, programming, tutorial, XML - October 3, 2019 - 0 Comment



The newest tool for PhysiCell provides an easy way to load your PhysiCell output data into python for analysis. This builds upon previous work on loading data into MATLAB. A post on that tool can be found at:

http://www.mathcancer.org/blog/working-with-physicell-snapshots-in-matlab/.

PhysiCell stores output data as a MultiCell Digital Snapshot (MultiCellDS) that consists of several files for each time step and is probably stored in your ./output directory. pyMCDS is a python object that is initialized with the .xml file

What you'll need

- python-loader, available on GitHub at
 - · https://github.com/PhysiCell-Tools/python-loader

pcdl - physicell data loader

load PhysiCell output into python3

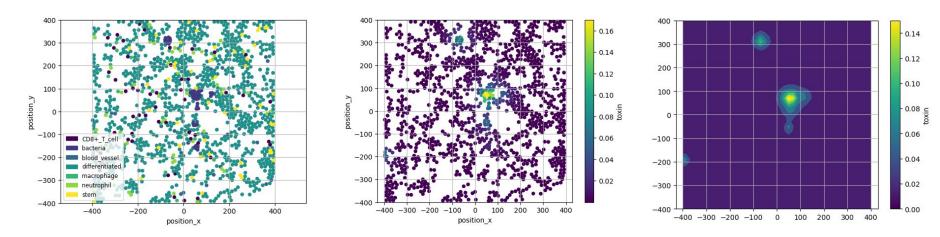


Aim:

- platform independent (Windows, MacOS, Linux).
- backwards compatible (pyMCDS.py).
- **pip** installable.
- pcdl is not analysis software!
- pcdl is your default connector between PhysiCell output and analysis software.

pcdl plots (battery included)

- mcds.plot scatter(), mcds.plot contour(), mcds.plot timeseries()
- matpletlib plots!



• mcdsts.make_gif(), and mcdsts.make movies()









- R Vector ≡ pandas Series
- R DataFrame = pandas DataFrame
 - df.loc[row, column]
 - df.iloc[row_number, column_number]

- if you miss ggplot plotnine might serve you:
 - https://plotnine.readthedocs.io/en/stable/index.html#

pcdl and pandas

fetch dataframes

- df = mcds.get cell df() # agent
- df = mcds.get_conc_df() # substrat

filter dataframes

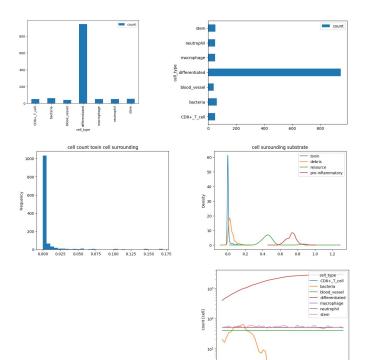
- df = mcdsts.get_cell_df_features(feature_values=2) # agent
- df = mcdsts.get_conc_df_features(feature_values=2) # substrat

pandas plots (battery included)

one line of code:

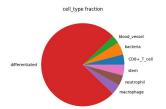
- df.plot(kind='bar')
- df.plot(kind='bah')
- df.plot(kind='pie', ...)
- df.plot(kind='hist')
- df.plot(kind='kde')
- df.plot(kind='box')
- df.plot(kind='line')
- df.plot(kind='area')

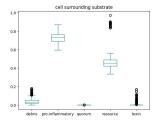
matpletlib plots!

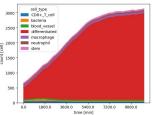


1800.0

7200.0









scverse - the single cell universe

Fabian Theis Lab

homepage: https://scverse.org/

data formats:

- anndata: https://anndata.readthedocs.io/en/latest/
- spatialdata: https://spatialdata.scverse.org/en/latest/ mudata: https://mudata.readthedocs.io/en/latest/

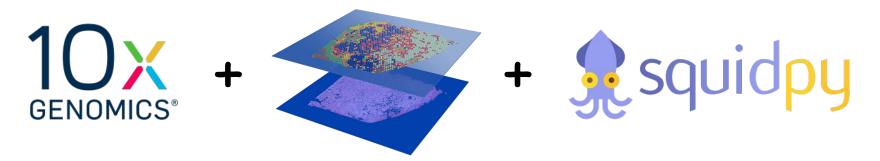


data analysis:

- **scanpy**: sincle cell analysis: https://scanpy.readthedocs.io/en/latest/squidpy: https://squidpy.readthedocs.io/en/stable/
- squidpy embeded napari: https://napari.org/stable/
- scvi-tools: single-cell deep learning: https://scvi-tools.org/
- muon: multimodal omics analysis: https://muon.scverse.org/
- scirpy: single cell immune sequencing analysis: https://scirpy.scverse.org/en/latest/
- scverse ecosystem: https://scverse.org/packages/#ecosystem



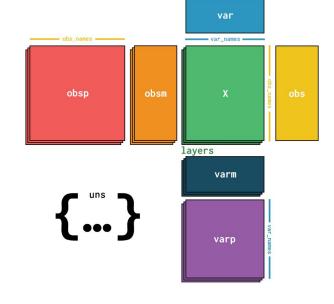
scverse & visum x10 data



squidpy:

- https://squidpy.readthedocs.io/en/stable/notebooks/tutorials/tutorial_visium_hne.html
- https://squidpy.readthedocs.io/en/stable/notebooks/tutorials/tutorial_visium_fluo.html
- https://squidpy.readthedocs.io/en/stable/notebooks/tutorials/tutorial-tangram.html

pcdl & scverse

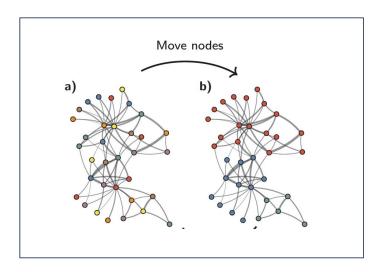


fetch anndata objects

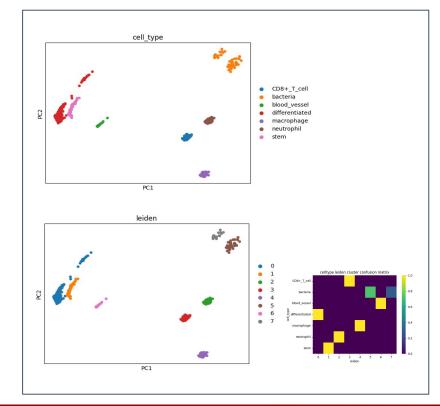
- ann = mcds.get_anndata()
- ann = mcdsts.get_anndata()
- ann = mcdsts.get_anndata(collapsed=False)

scanpy & neigberhood graph clustering

- sc.pp.neighbors(ann, n neighbors=15) # preprocess
- sc.tl.leiden(ann, resolution=0.01) #tool



Traag et al. From Louvain to Leiden: guaranteeing well-connected communities. Sci Rep 9, 5233 (2019). https://doi.org/10.1038/s41598-019-41695-z



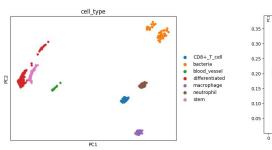
scanpy & dimensional reduction

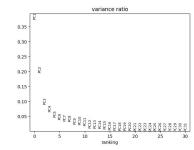
two lines of code:

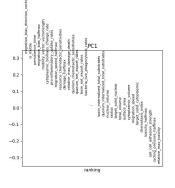
- sc.tl.tsne(ann) # tool
- sc.tl.tsne(ann) # plot

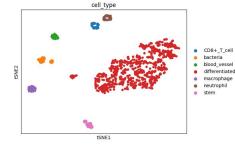
methods:

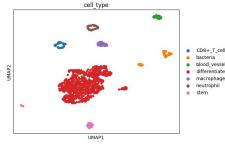
- Distributed Recursive
- ForceAtlas2
- Fruchterman Reingold
- Grid Fruchterman Reingold
- Kamadi Kawai
- Large Graph
- PCA
- Reingold Tilford Tree
- Reingold Tilford Circular
- tSNE
- UMAP

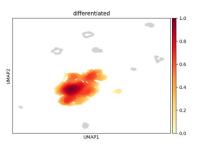






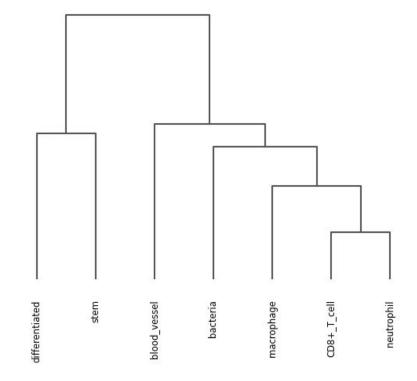






scanpy & dendrogram

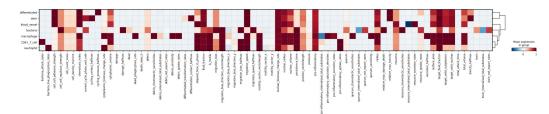
- sc.tl.dendrogram(ann, groupby='cell type') # tool
- sc.pl.dendrogram(ann, groupby='cell type') # plot

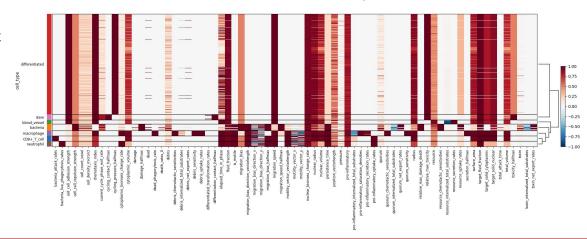


scanpy & heatmaps

one lines of code:

- sc.pl.matrixplot(ann, ...) # plot
- sc.pl.dotplot(ann, ...) # plot
- sc.pl.stacked violin(ann, ...)#plot
- sc.pl.tracksplot(ann, ...) # plot
- sc.pl.heatmap(ann, ...) # plot
- sc.pl.clustermap(ann, ...)#plot





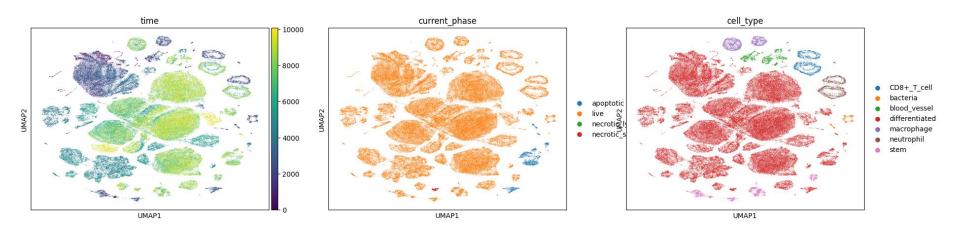


scanpy & timeseries

two lines of code:

- sc.tl.umap(ann, ...) # tool
- sc.tl.umap(ann, ...) # plot

this is physicell output!

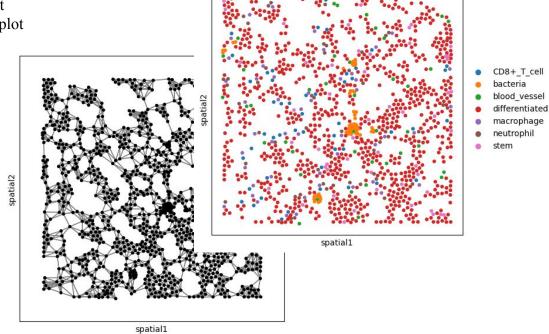




squidpy & spatial

one line of code:

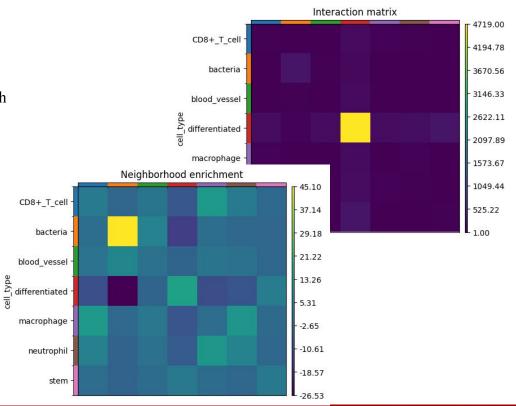
- sq.pl.spatial_scatter(ann, ...) # plot
- sq.pl.spatial neighbors(ann, ...) # plot



cell type

squidpy & neigbourhood

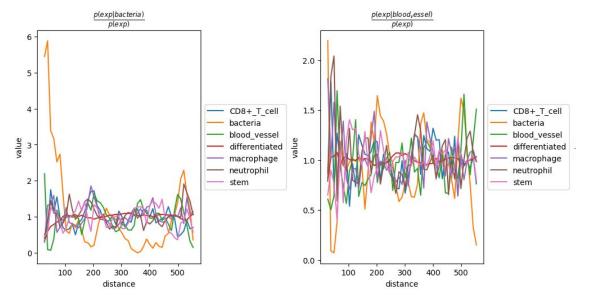
- sq.gr.interaction matrix(ann, ...) # graph
- sq.pl.interaction matrix(ann, ...) # plot
- sq.gr.nhood enrichment(ann, ...) # graph
- sq.gr.nhood_enrichment(ann, ...) # plot





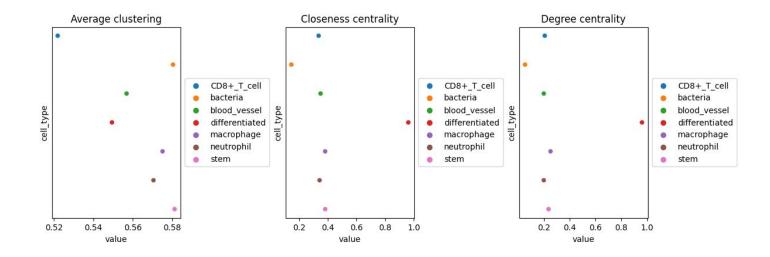
squidpy & co_occurrence

- sq.gr.co_occurrence(ann, ...) # graph
- sq.pl.co_occurrence(ann, ...) # plot



squidpy & centrality_scores

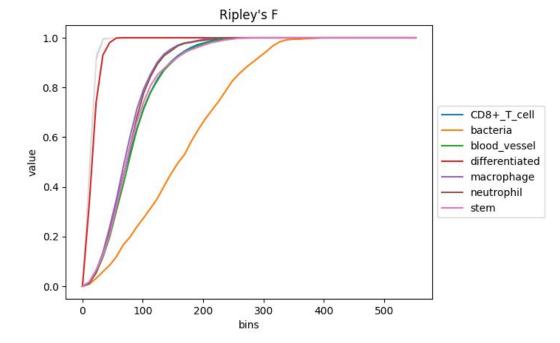
- sq.gr.centrality scores(ann, ...) # graph
- sq.pl.centrality_scores(ann, ...) # plot



squidpy & Ripley's statistics

for point processes

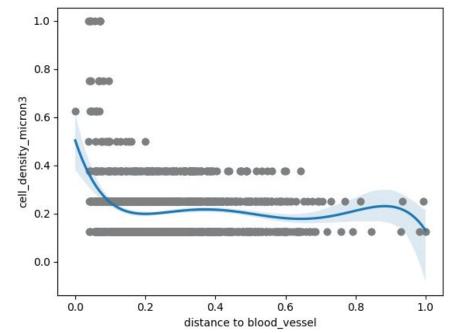
- sq.gr.ripley(ann, ...) # graph
- sq.gr.ripley(ann, ...) # plot



squidpy & var_by_distance

plot one **numerical feature (var; y-axis)** against a **categorical parameter (obs; x-axis)**. e.g. voxel cell density against distance to blood vessel.

- sq.tl.var by distance(ann, ...) #tool
- sq.pl.var_by_distance(ann, ...) # plot



squidpy & global autocorrelation statistic

Moran's I and Geary's C

one line of code:

- sq.gr.spatial autocorr(ann, mode='moran') # graph
- ann.uns["moranI"].head(10) # datafame

	1	pval_norm	var_norm	pval_norm_fdr_bh				
position_vectorlength	0.992993	0.0	0.000273	0.0				
quorum	0.934202	0.0	0.000273		С	pval_norm	var_norm	pval_norm_fdr_bh
debris	0.927604	0.0	0.000273	position_vectorlength	0.009540	0.0	0.000273	0.0
pro-inflammatory	0.870561	0.0	0.000273	quorum	0.030820	0.0	0.000273	0.0
toxin	0.853246	0.0	0.000273	debris	0.062550	0.0	0.000273	0.0
damage_halfmax	0.603295	0.0	0.000273	toxin	0.080504	0.0	0.000273	0.0
cell_cell_repulsion_strength	0.603295	0.0	0.000273	pro-inflammatory	0.128708	0.0	0.000273	0.0
relative_max_damage_death	0.603295	0.0	0.000273	cell_cell_repulsion_strength	0.340886	0.0	0.000273	0.0
resource_chemotactic_sensitivities	0.603295	0.0	0.000273	relative_max_damage_death	0.340886	0.0	0.000273	0.0
migration_speed	0.595514	0.0	0.000273	resource_chemotactic_sensitivities	0.340886	0.0	0.000273	0.0
				damage_halfmax	0.340886	0.0	0.000273	0.0
				migration_speed	0.346057	0.0	0.000273	0.0

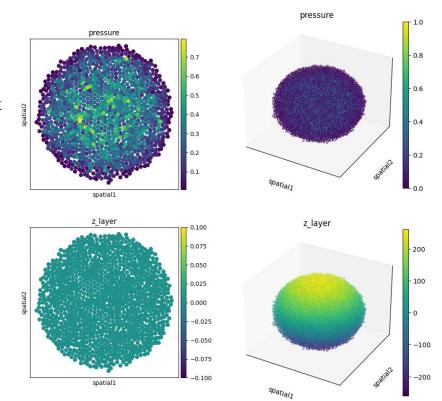


squidpy & z-stacks

one line of code:

- sq.pl.spatial_scatter(ann3d ...) # plot
- sc.pl.embedding(ann3d ...) # plot

this is physicell output!





Where to go from here?

the basics:

- learn core python3: https://www.python.org/ [resources: https://realpython.com/]
- learn numpy: https://numpy.org/ learn scipy: https://scipy.org/
- learn pandas: https://pandas.pydata.org
- learn matplotlib: https://matplotlib.org/

depending based on your needs, learn:

- statsmodels: https://www.statsmodels.org/stable/index.html
- sklearn: https://scikit-learn.org/stable/
- pytorch: https://pytorch.org/
- skimage: https://scikit-image.org/

at https://scverse.org:

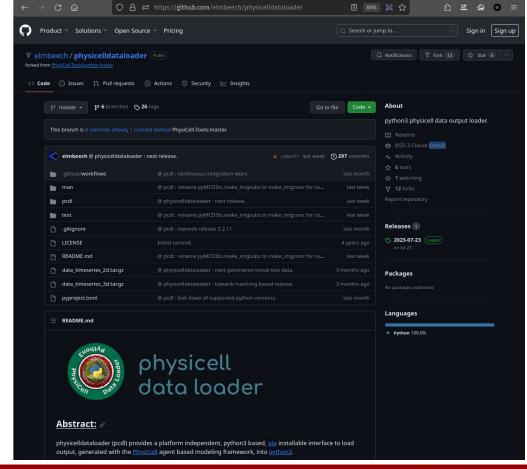
- read the anndata docs the basic data object

- read the scanpy docs single cell analysis read the squidpy docs spatial single cell analysis read the scvi-tools docs single cell machine learning



pcdl homepage

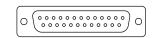
- work through the **TUTORIAL.md**.
- work through the Jupyter notebook.
- read the docstrings from the "workhorse functions" mentioned in the REFERENCE.md





Conclusion

pcdl is simply a connector, an interface.



• for software development: don't re-invent the wheel! learn, make use of, and contribute to in the filed well established high-level libraries.



Acknowledgement

The whole **MathCancer** lab!

python-loader original implementation:

- Pat Wall
- Randy Heiland
- Paul Macklin

pcdl 2022 / 2023 evolution:

- Ben Jacobs (get graph)
- Furkan Kurtoglu (get vtk)
- Heber Rocha (testing)
- Marshal Gress (plot_scatter)
 Thierry-Pascal Fleurant (plot timeseries)

OHSU:

- Jenny Eng (scanpy)
- Tina Ghodsi Asnaashari (abm)

Software - because pcdl is standing on the shoulder of giants:

- PhysiCell
- PhysiCell Studio (pyMCDS.py)
- Python3 core library
- numpy, scipy, pandas, matplotlib
- anndata
- vtk
- http: requests

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