

SPRAS: A Workflow for Streamlining Network-based Pathway Reconstruction

Adam Shedivy¹, Nina Young², Christopher Magnano^{1,3}, Pramesh Singh², Tobias Rubel⁴, Anna Ritz², and Anthony Gitter¹



¹Morgridge Institute and University of Wisconsin-Madison, Madison, WI, USA; ²Biology Department, Reed College, Portland, OR, USA; ³Center for Computational Biomedicine, Harvard Medical School, Cambridge, MA, USA; ⁴Computer Science Department, University of Maryland - College Park, College Park, MD, USA



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Pathway Reconstruction

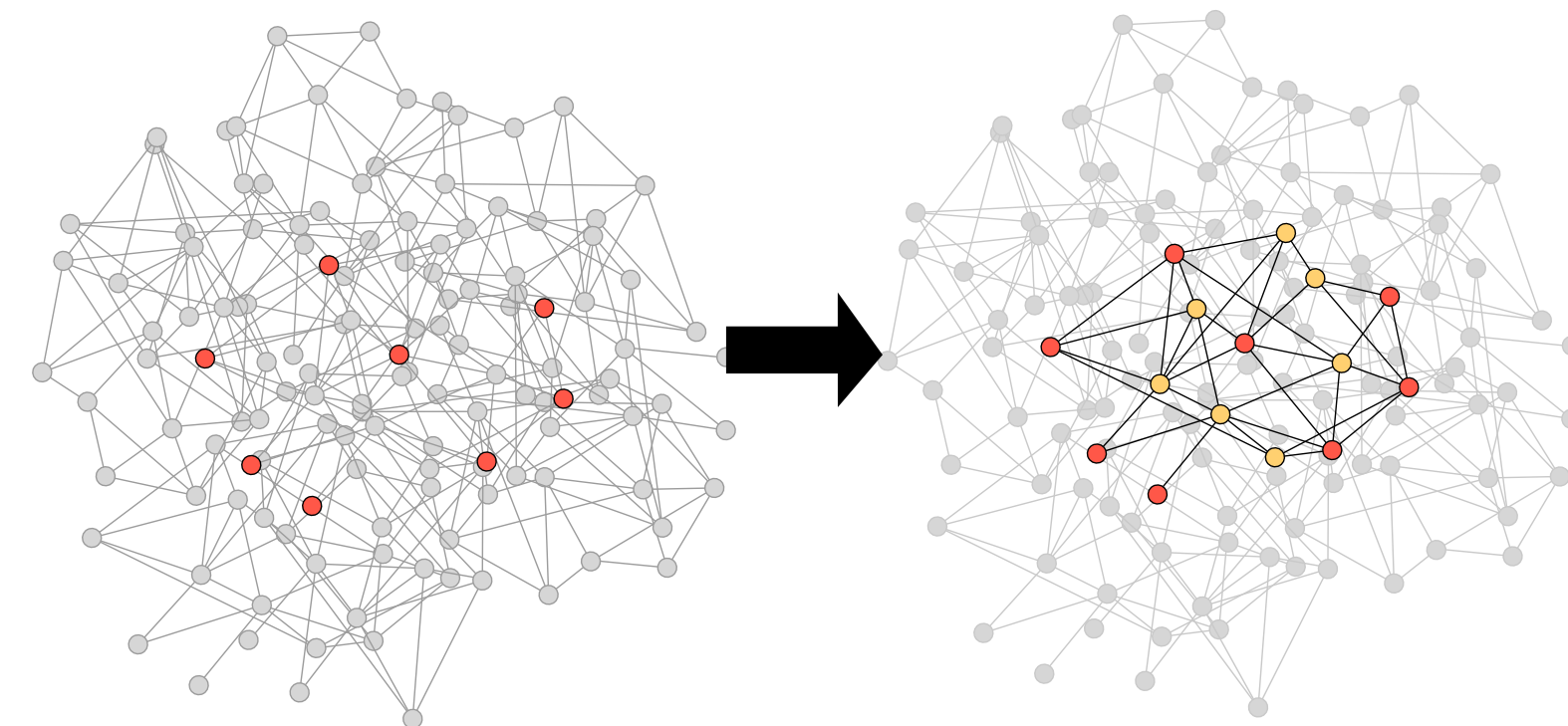
Analyzing data from transcriptomic, proteomic, and other high-throughput assays in a biological network context provides a systems-level understanding of individual events. **Pathway reconstruction** aims to connect genes or proteins of interest (nodes) by selecting relevant interactions (edges) from a background network that connect these genes. Dozens of network biology methods exist that solve the pathway reconstruction problem (examples include [1, 2, 3, 4]).

Node Inputs:

Gene Expression,
Protein Expression,
GWAS,
...

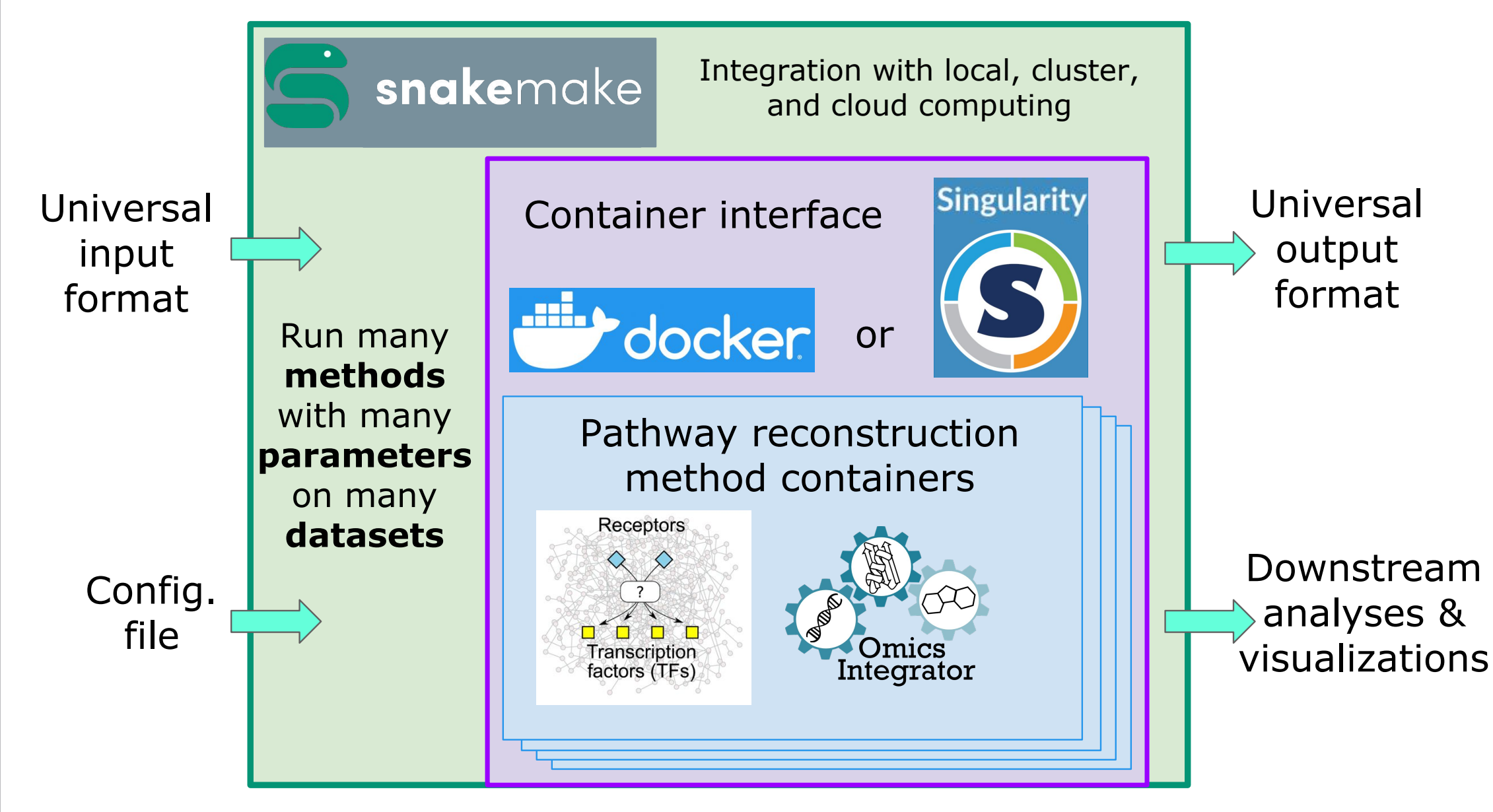
Network Input:

Protein interactions,
Functional similarities,
...



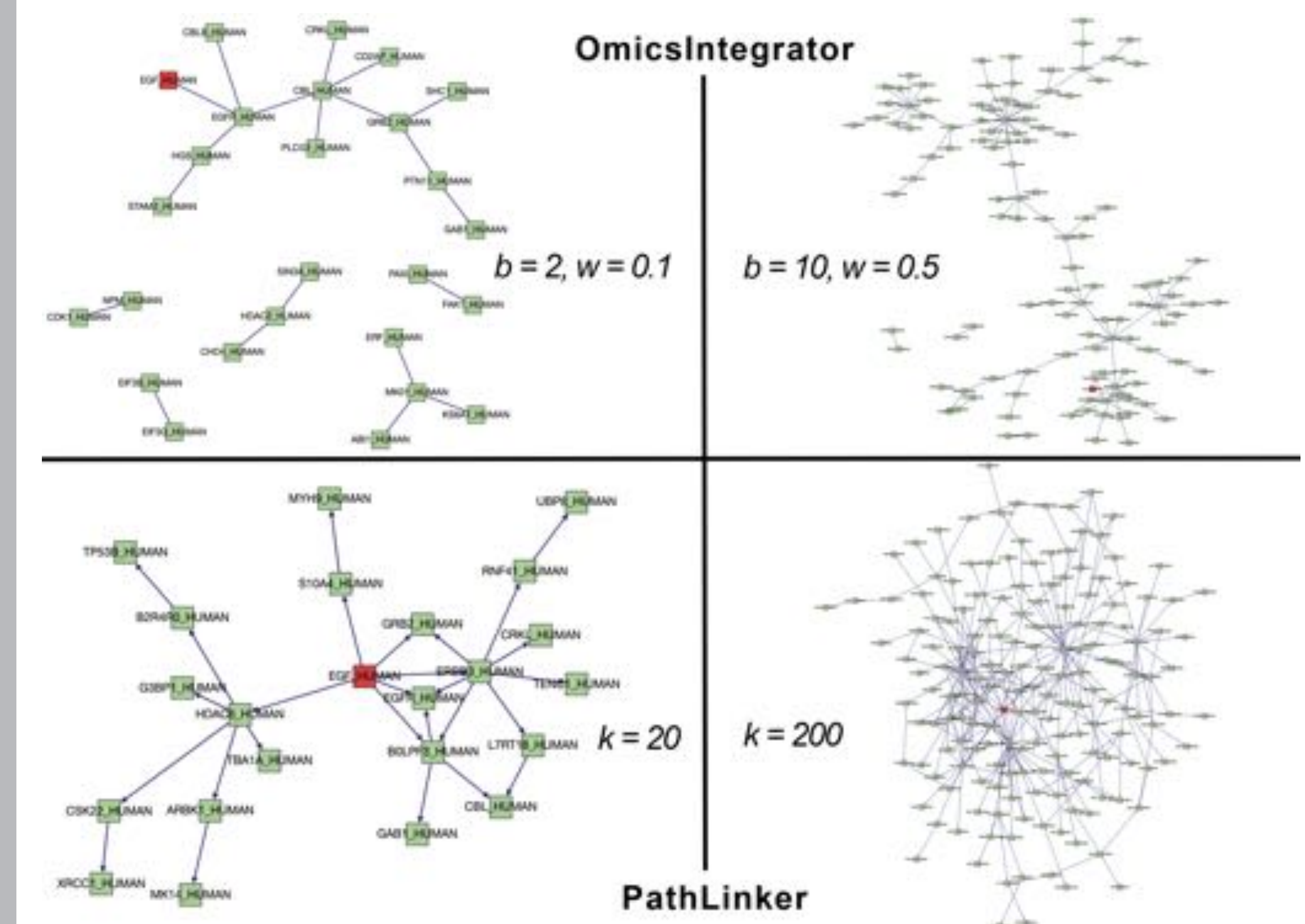
SPRAS: Signaling Pathway Reconstruction Analysis Streamliner

SPRAS provides a unified pathway reconstruction workflow through Snakemake. It also provides containers for popular algorithms [1, 2].



SPRAS Case Study

Output networks from the workflow on the left visualize phosphorylation changes in response to EGF treatment [3].



The example outputs illustrate reconstructed Steiner forests (Omics Integrator [1]) and k -shortest paths (PathLinker [2]) run with parameters that produce small and large reconstructions.

Pathway Reconstruction Challenges

Practical challenges have limited the adoption of pathway reconstruction methods.

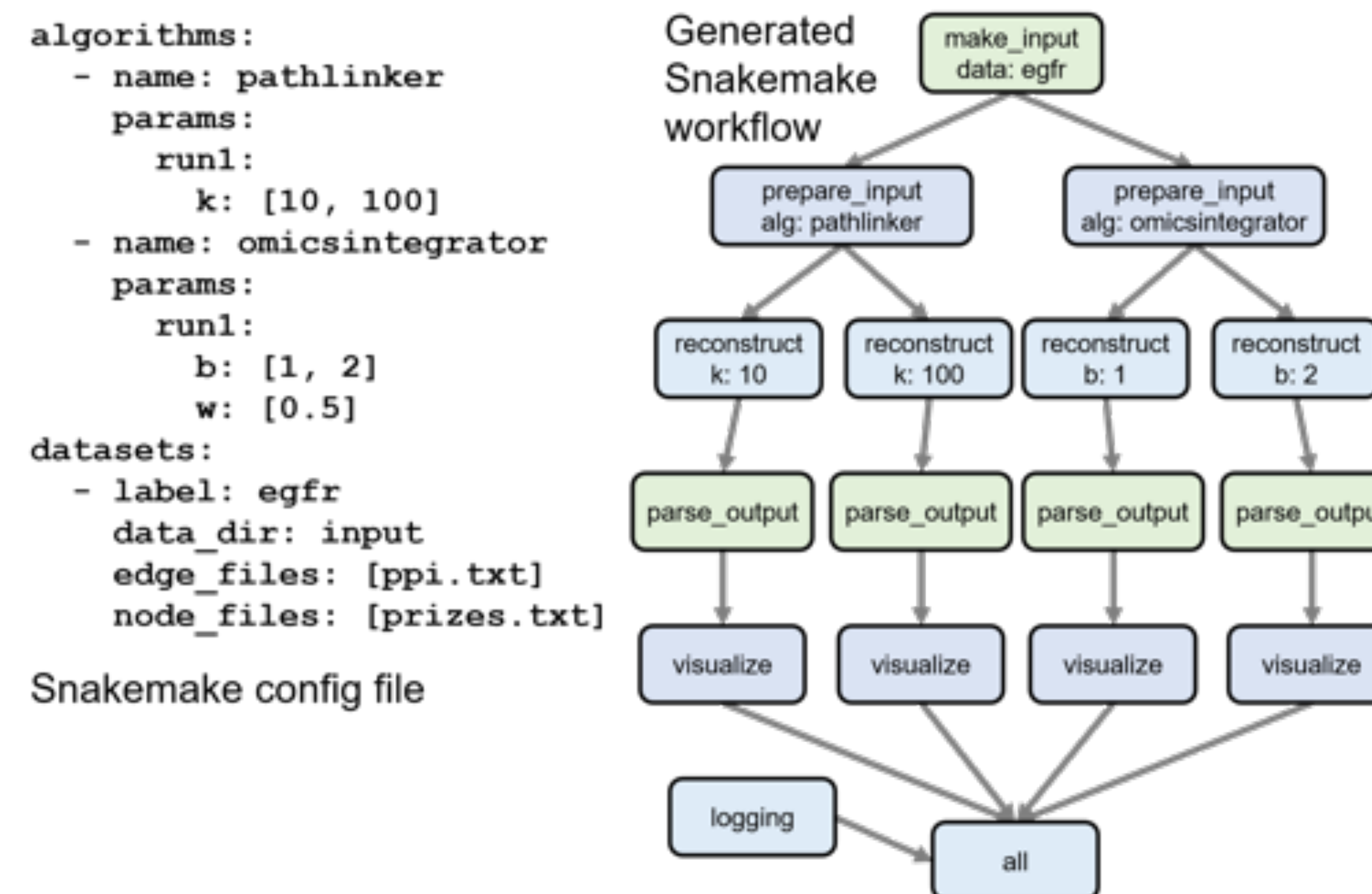
Each method has its own file formats, dependencies, installation process, and approach to setting parameters. These decisions are shown for two methods [1, 2].

SPRAS unifies different methods so they can all be run in a single consistent workflow.

	PathLinker	Omics Integrator
Dependencies	Python packages python 3.5 networkx 1.11	Python packages python 2.7 networkx 1.11 numpy 1.13 ... Boost C++ libraries Compiled msgsteiner C++ code
Setting parameters	Command line python run.py \ -k 100	Configuration file w = 1 b = 2 D = 10 mu = 0.5
Input files	Node types file #Node Node type A source D target Network file #N1 N2 A B 0.1 A C 0.9 B D 0.1	Node prizes file name prize A 3.5 D 1.2 Network file p1 p2 weight dir A B 0.1 U A C 0.9 D B D 0.1 U
Output file	#tail head i cost A B 1 0.01 B D 1 0.01	A pp B B pp D

SPRAS Workflow

A single configuration file sets parameters for all methods. Snakemake runs all combinations of datasets, algorithms, and parameters and visualizes network outputs.



Ongoing and Future Work

Pathway visualization is a common downstream task. We now have a prototype Cytoscape implementation.



We plan to implement additional pathway reconstruction methods such as ResponseNet [4]. We also plan to develop evaluation criteria for benchmarking.

Interested? We are looking for collaborators and contributors! Email aritz@reed.edu or gitter@biostat.wisc.edu for more info.

References

- [1] Tuncbag *et al.* Network-Based Interpretation of Diverse High-Throughput Datasets through the Omics Integrator Software Package. *PLoS Comp. Biol.* **12**, e1004879 (2016).
- [2] Ritz *et al.* Pathways on demand: automated reconstruction of human signaling networks. *npj Systems Biology and Applications* **2**, 16002 (2016).
- [3] Köksal *et al.* Synthesizing Signaling Pathways from Temporal Phosphoproteomic Data. *Cell Reports* **24**, 3607–3618 (2018).
- [4] Lan *et al.* ResponseNet: revealing signaling and regulatory networks linking genetic and transcriptomic screening data. *Nucleic Acids Research* **39**, W424–W429 (2011).

Acknowledgements

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