

# Connectivity Measures for Signaling Pathway Topologies



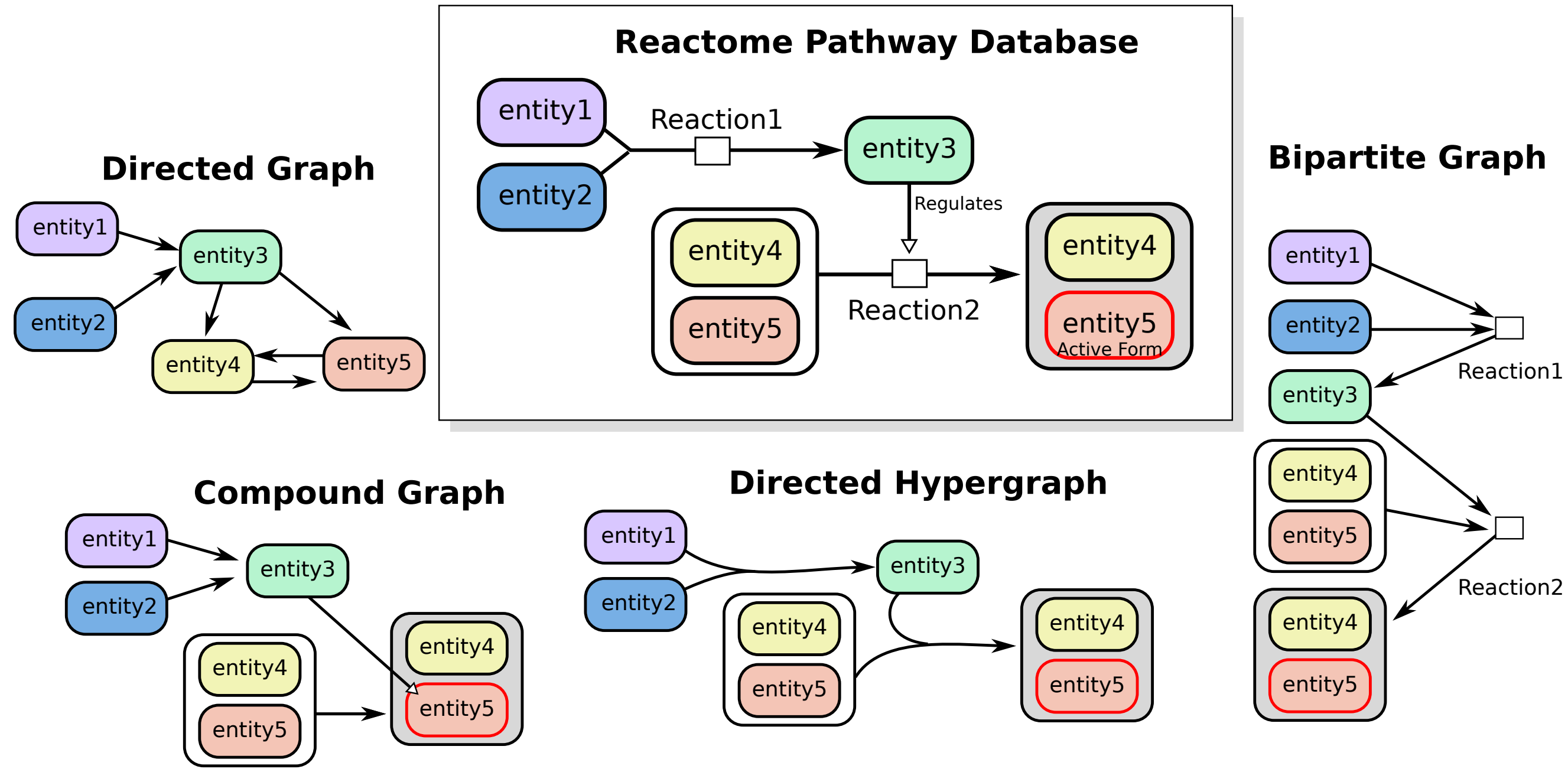
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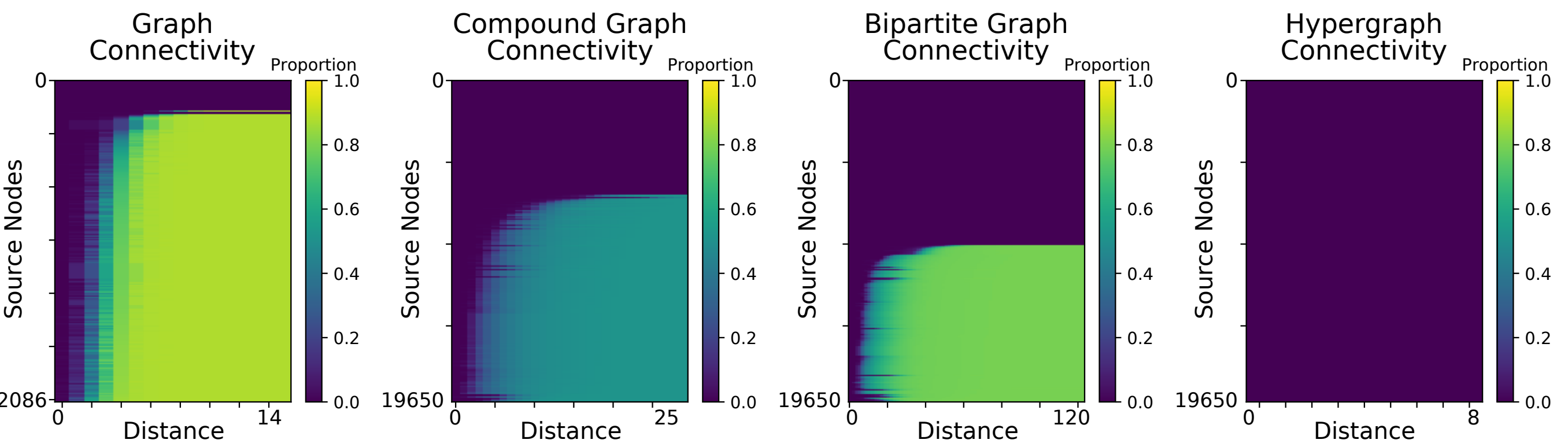
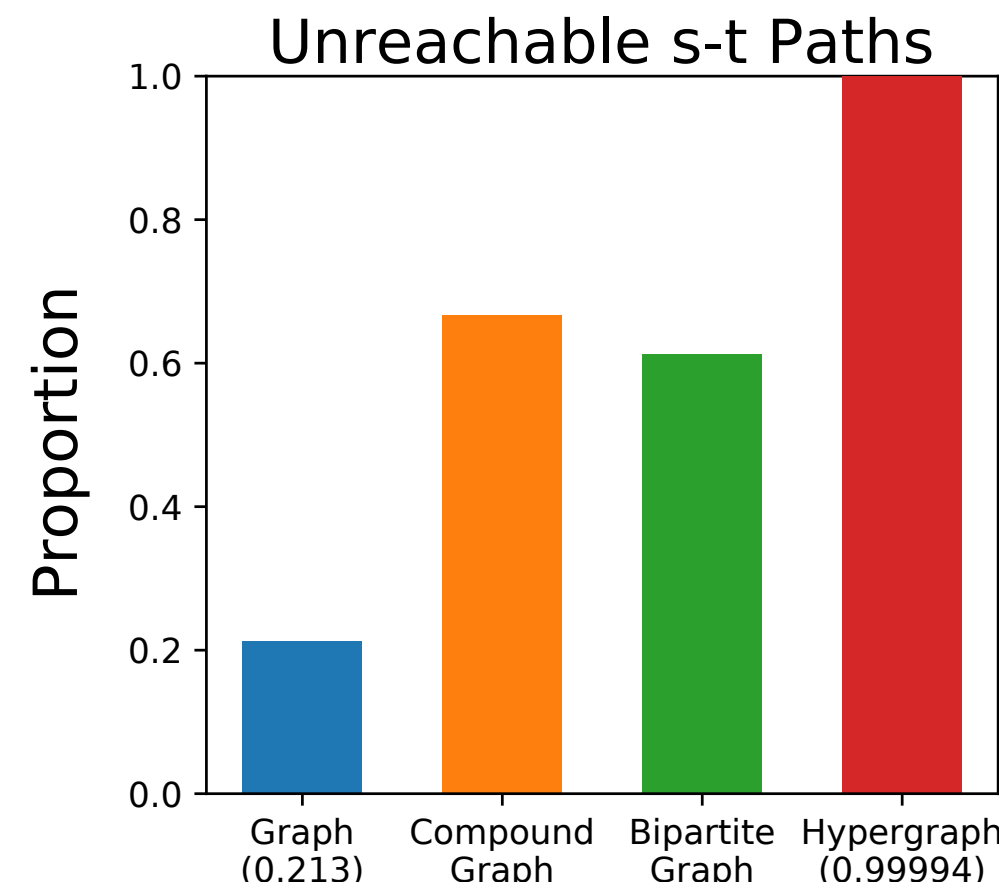
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## Background: Signaling Pathway Representations



**When are two molecules connected in a network?** We computed distances between all pairs of nodes in Reactome representations.

1. **Directed graphs** and **bipartite graphs** use breadth first search.
2. **Compound graphs** use a traversal with biologically-inspired rules [1].
3. **Directed Hypergraphs** use the notion of  $B$ -connectivity [2]: a hyperedge is traversable w.r.t. a node  $s$  if all nodes in the tail (inputs) are reachable from  $s$  [2,3].

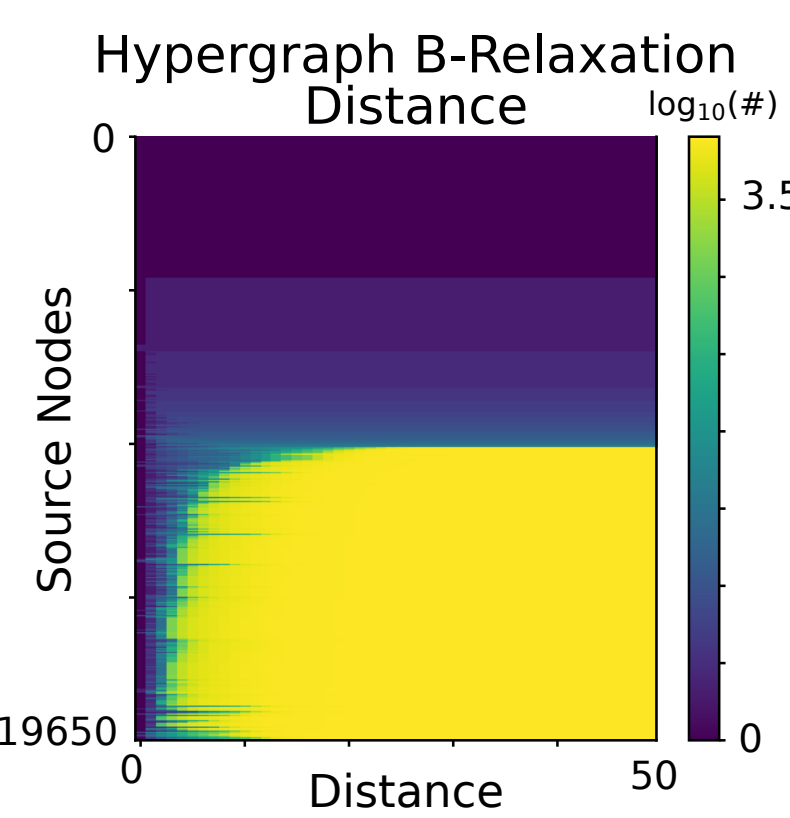


**Hypergraphs are more informative than other representations, but  $B$ -connectivity is overly strict for Reactome's topology.**

## Methods: $B$ -Relaxation Distance [4]

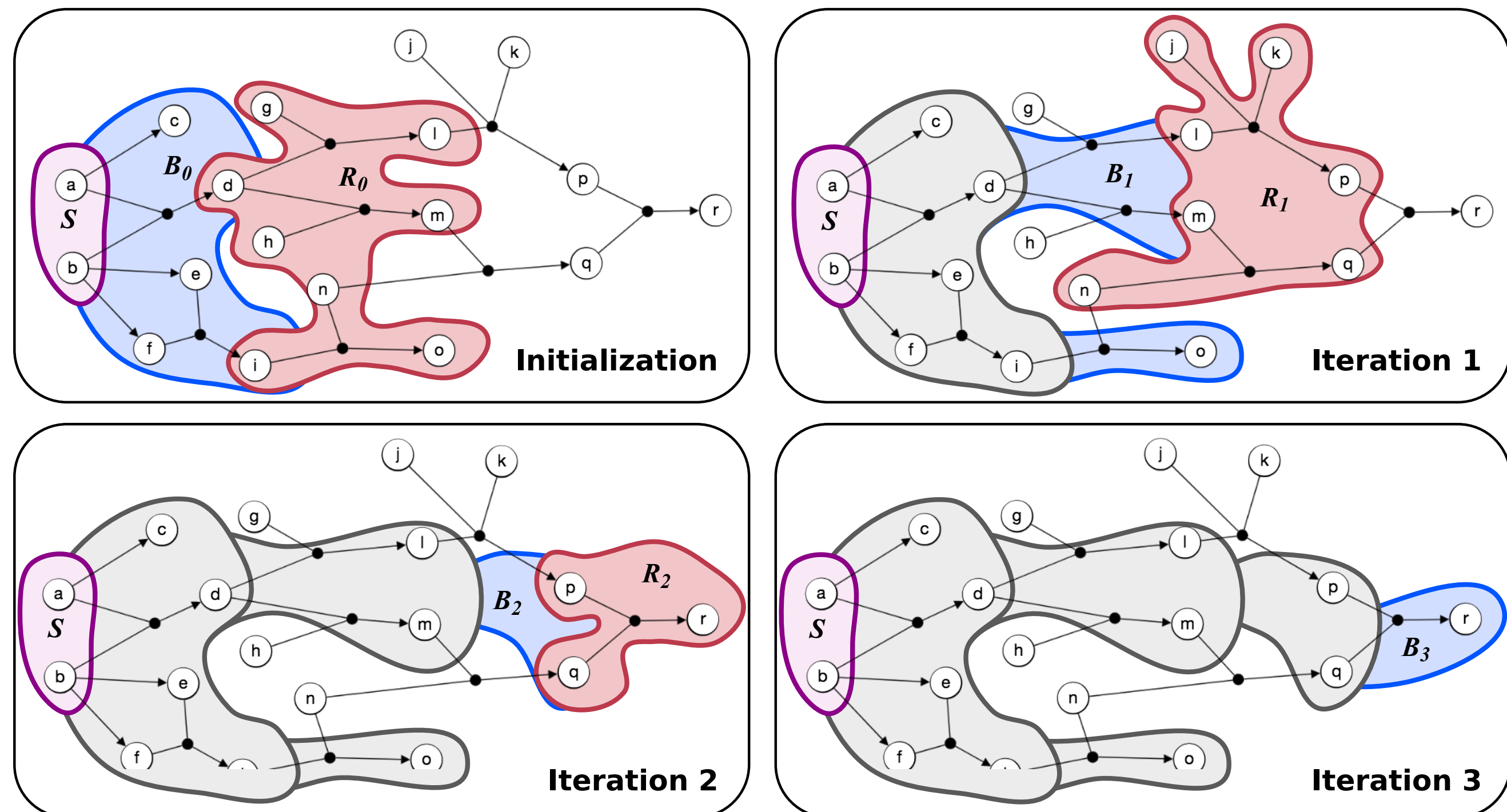
$B$ -relaxation distance relaxes  $B$ -connectivity in hypergraphs while preserving hypergraph topology by iteratively lifting the  $B$ -connectivity restriction on hyperedges. Given a set  $S$  of source nodes, the algorithm computes two sets:

1.  $B_0$ : nodes  $B$ -connected to  $S$ .
2.  $R_0$ : restrictive hyperedges where at least one tail node is in  $B_0$  but it was not traversed.



The algorithm then computes  $B_1$  and  $R_1$  from the heads of each restrictive hyperedge in the previous iteration (e.g.,  $R_0$ ). This process continues until no new nodes are reachable; this set is exactly the set of reachable nodes in the bipartite graph.

**The  $B$ -Relaxation Distance of a node  $v$  from a source set  $S$  is the smallest value of  $k$  where  $v$  is in  $B_k$ .**



## References & Acknowledgements

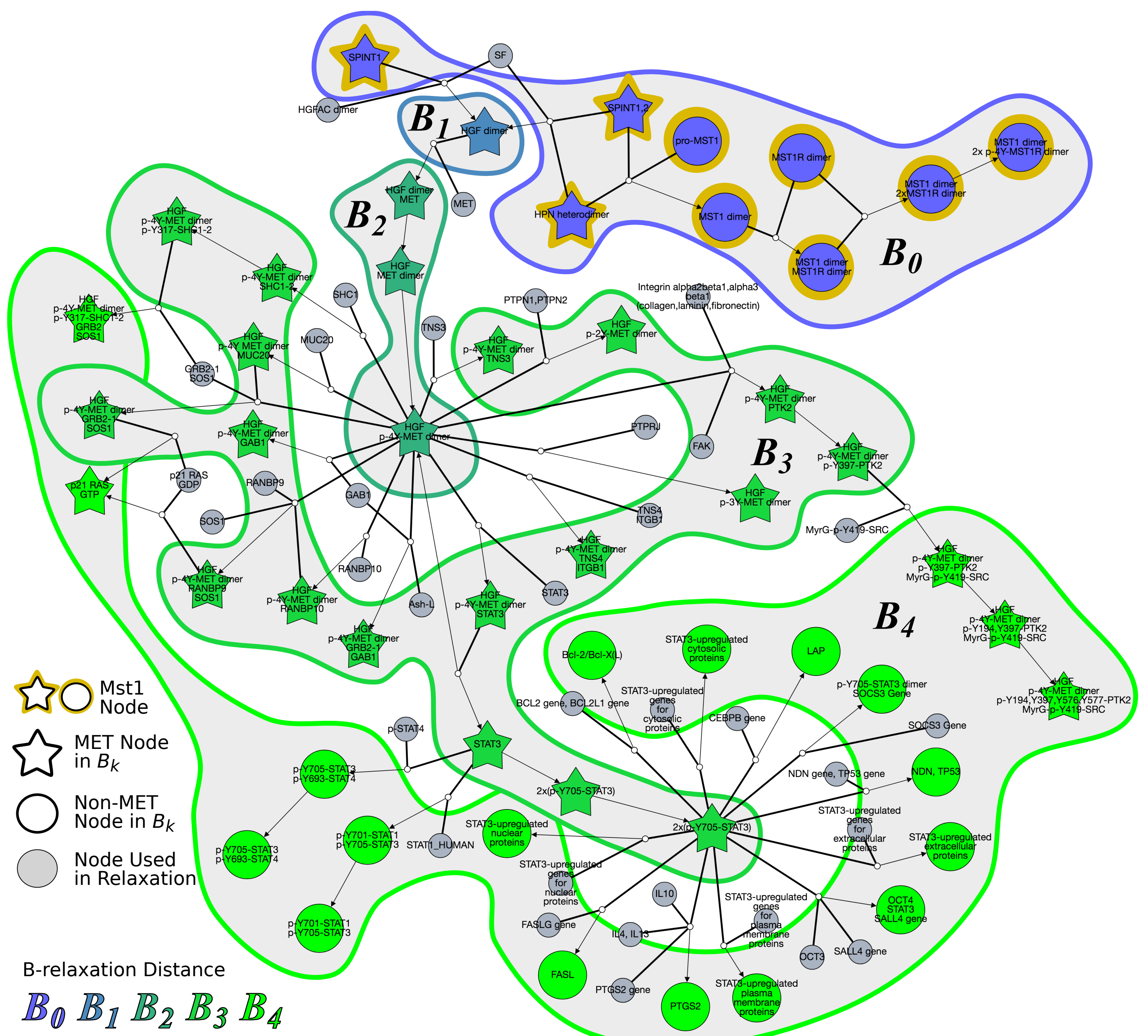
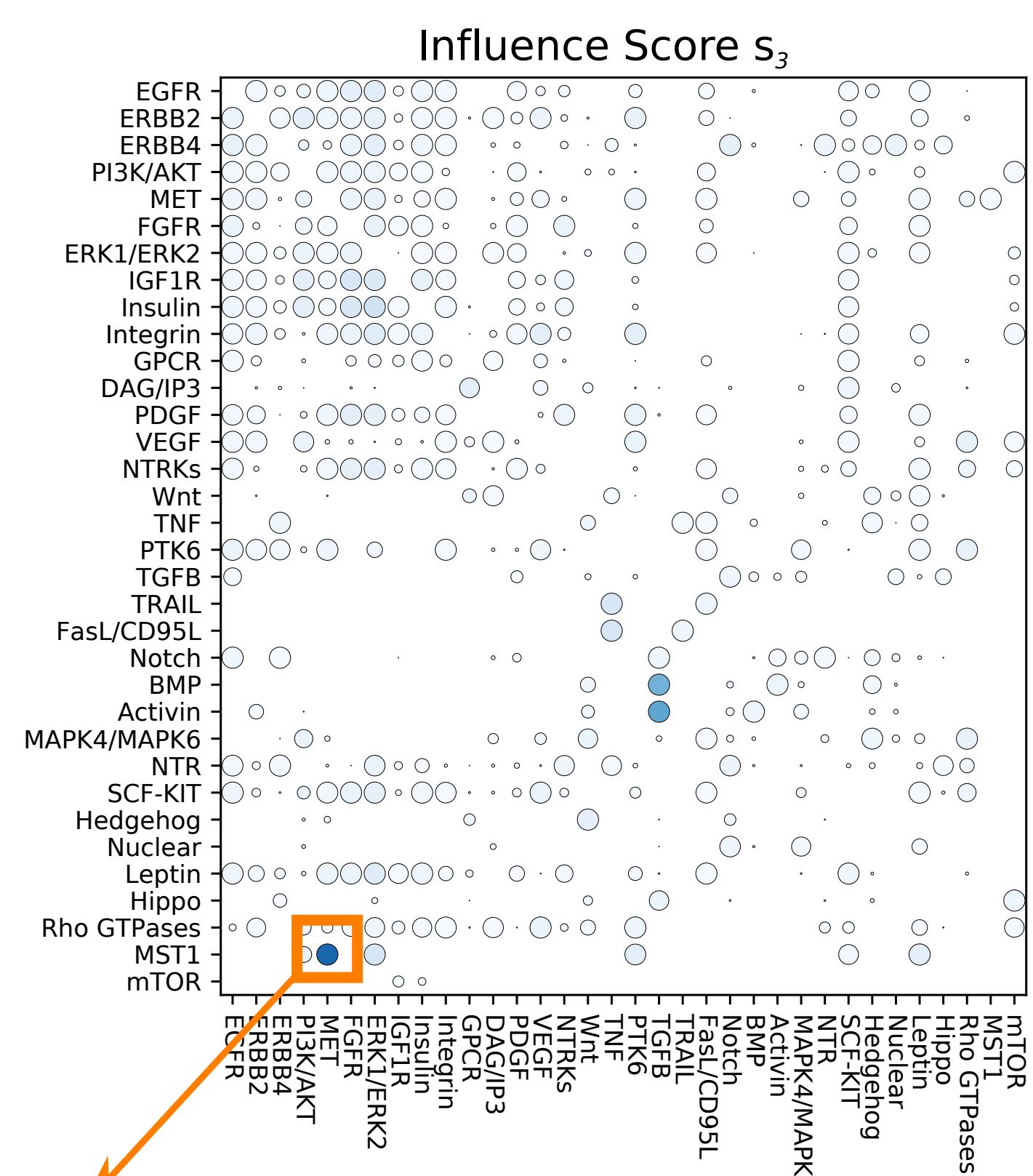
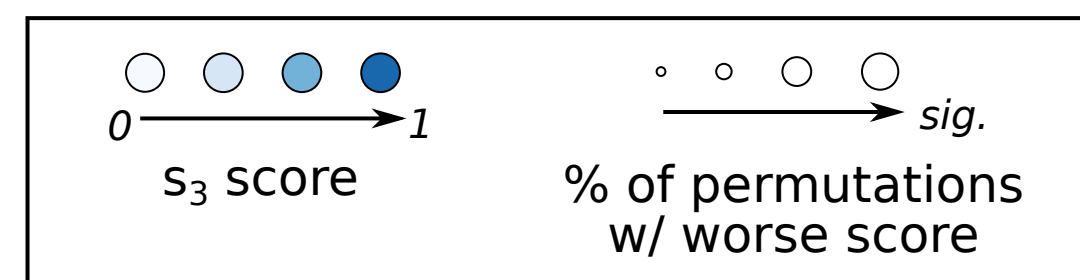
**Refs:** [1] Dogrusoz et al, *BMC Bioinf.* 2009; [2] Gallo et al., *Discrete Appl. Math.* 1993; [3] Ritz et al., *IEEE TCBB* 2017; [4] Ritz et al., *bioRxiv* 2019. **Acknowledgements:** Work is supported by NSF DBI-1750981 (to AR) and NSF CCF-1617678 (to TMM).

## Results: Pathway influence across Reactome

We compute the *influence score*  $s_k$  of a source pathway  $S$  (w/ nodes  $P_S$ ) on target pathway  $T$  (w/ nodes  $P_T$ ) for  $B$ -relaxation distance up to  $k$  (denoted  $B_{\leq k}$ ):

$$s_k = \frac{|\overbrace{(B_{\leq k}(P_S) \cap P_T)}^{\text{"reachable" overlap}} \setminus \overbrace{(P_S \cap P_T)}^{\text{initial overlap}}|}{|B_{\leq k}(P_S) \setminus (P_S \cap P_T)|}$$

We compute the significance of  $s_k$  with a permutation test that preserves the node overlap of all 34 Reactome pathways.

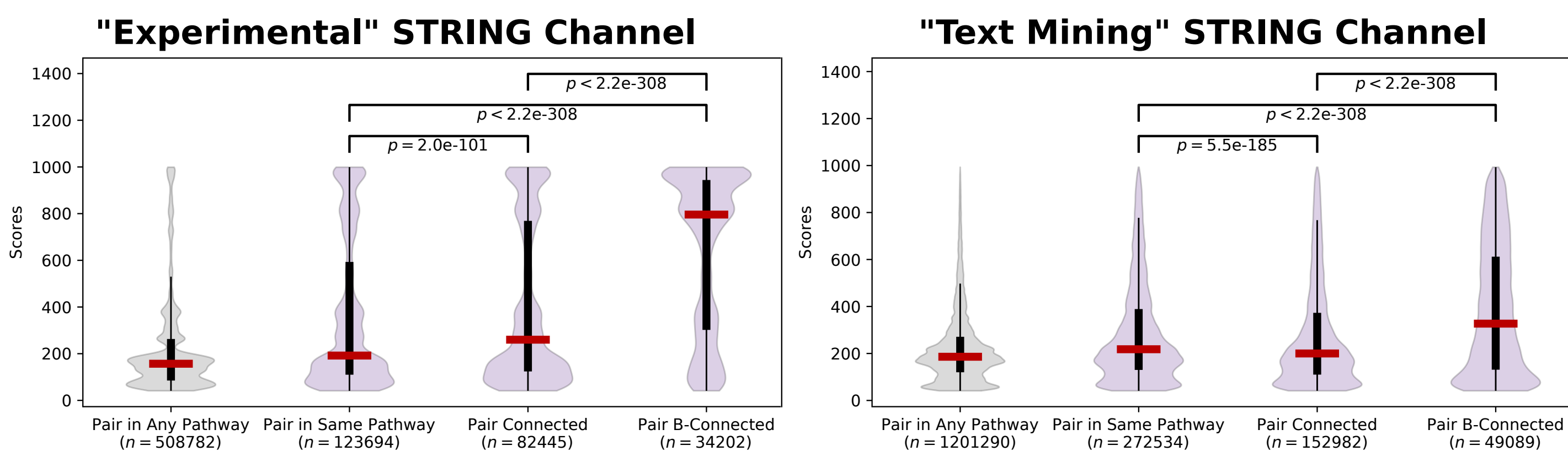


**Case Study: Mst1 Pathway Influence on MET Signaling**

## Results: Benchmarking STRING functional relationships

We examined the enrichment of STRING interaction scores for node pairs with different definitions of connectivity: (1) node pairs in any Reactome pathway; (2) node pairs in the same Reactome pathway; (3) node pairs connected in the Reactome bipartite graph; and (4) node pairs connected in the Reactome hypergraph using  $B$ -connectivity.

**Node pairs that are  $B$ -connected in Reactome are enriched for higher-scoring STRING interactions across different channels.**



## Availability

**Hypergraph Algorithms Package (python):**  
<http://murali-group.github.io/halp>

**Source Code for bioRxiv Paper:**

<https://github.com/annaritz/pathway-connectivity>