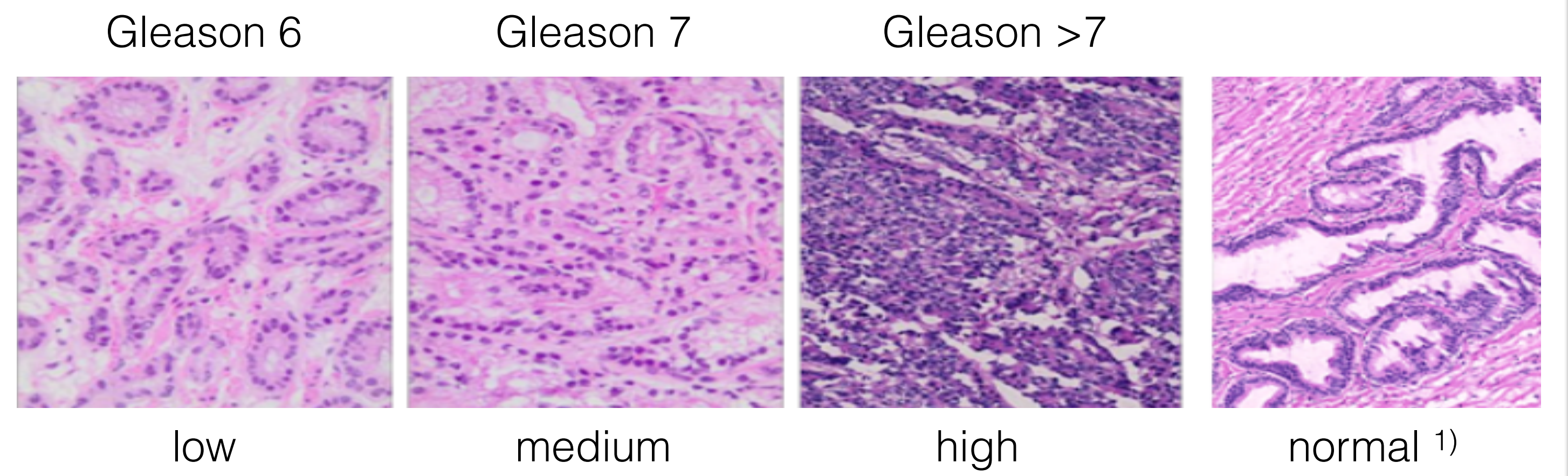


Motivation

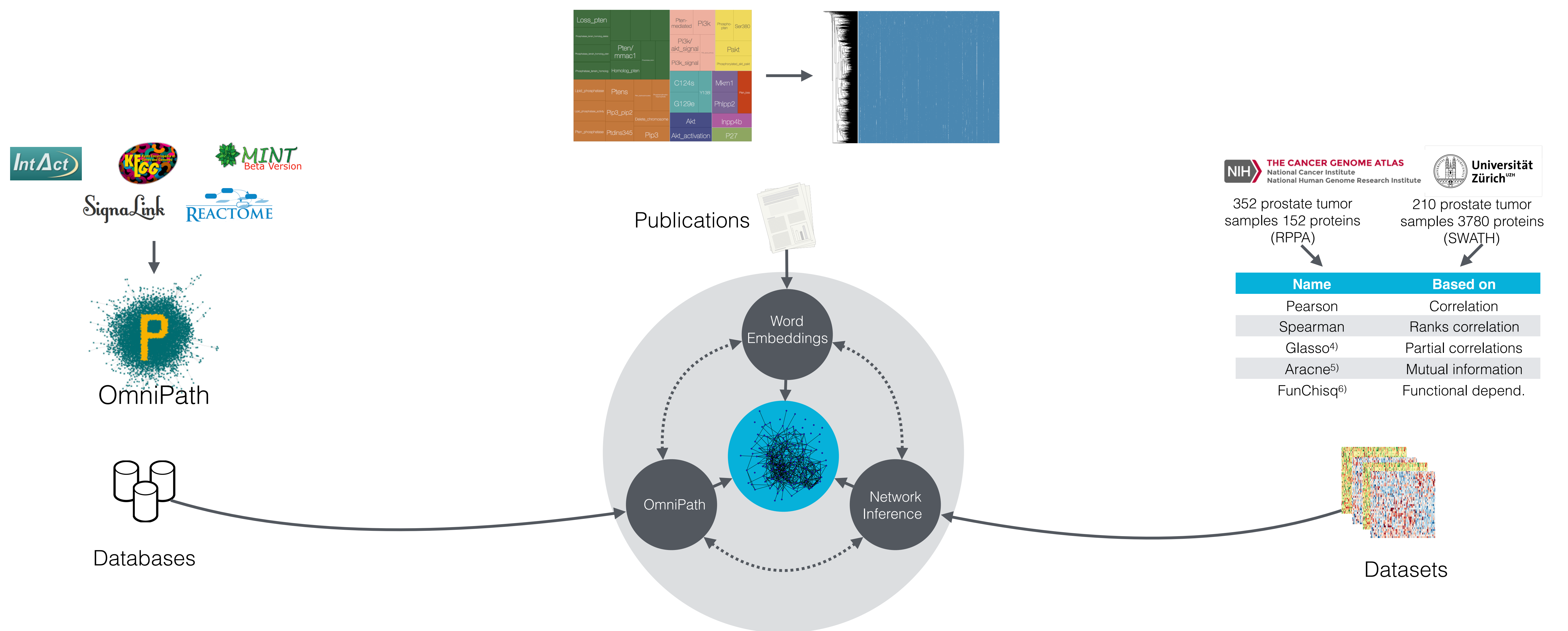
Prostate cancer is a leading cause of cancer death amongst men, however the molecular-level understanding of disease onset and progression are largely unknown. Specifically, stratification of intermediate prostate tumor states based on current markers is difficult. To improve stratification we

1. integrate data, literature and public databases,
2. extract interaction groups from interactome and cluster patients for each interaction group,
3. train a classifier using the patient profiles to identify marker interactions.



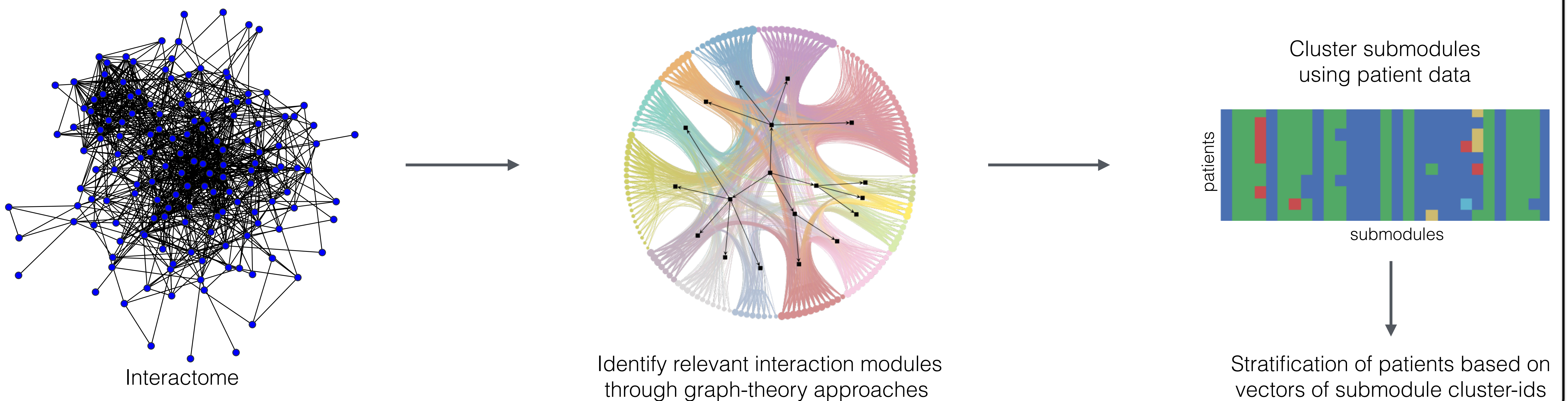
Interactome

Multiple inference methods were applied to a given data-set to construct a data-driven interactome. To enrich the data driven interaction network, separate interactomes were extracted from publications and databases using OmniPath²⁾ and word embeddings³⁾. By combining the interaction networks derived from data, public databases and publications a consensus interactome was built.



Stratification

From an interaction network we identify relevant interaction modules through graph-theory approaches. For each interaction module we cluster the patients based on molecular states measurements. The patient-specific cluster assignment vectors serve as a personalized interaction signatures and is used to stratify patients.



1) Image source: Peter Wild Lab, Zurich University Hospital
 2) D Turei, T Korcsmaros and J Saez-Rodriguez (2016) OmniPath: guidelines and gateway for literature-curated signaling pathway resources. Nature Methods 13(12)
 3) Mikolov, Tomas, et al. "Efficient estimation of word representations in vector space." arXiv Margolin,
 4) Adam A., et al. "ARACNE: an algorithm for the reconstruction of gene regulatory networks in a mammalian cellular context." BMC bioinformatics 7.Suppl 1 (2006): S7.
 5) Friedman, Jerome, Trevor Hastie, and Robert Tibshirani. "Sparse inverse covariance estimation with the graphical lasso." Biostatistics 9.3 (2008): 432-441.
 6) Yang Zhang and Mingzhou Song. "Deciphering interactions in causal networks without parametric assumptions." arXiv preprint arXiv:1311.2707, 2013.