

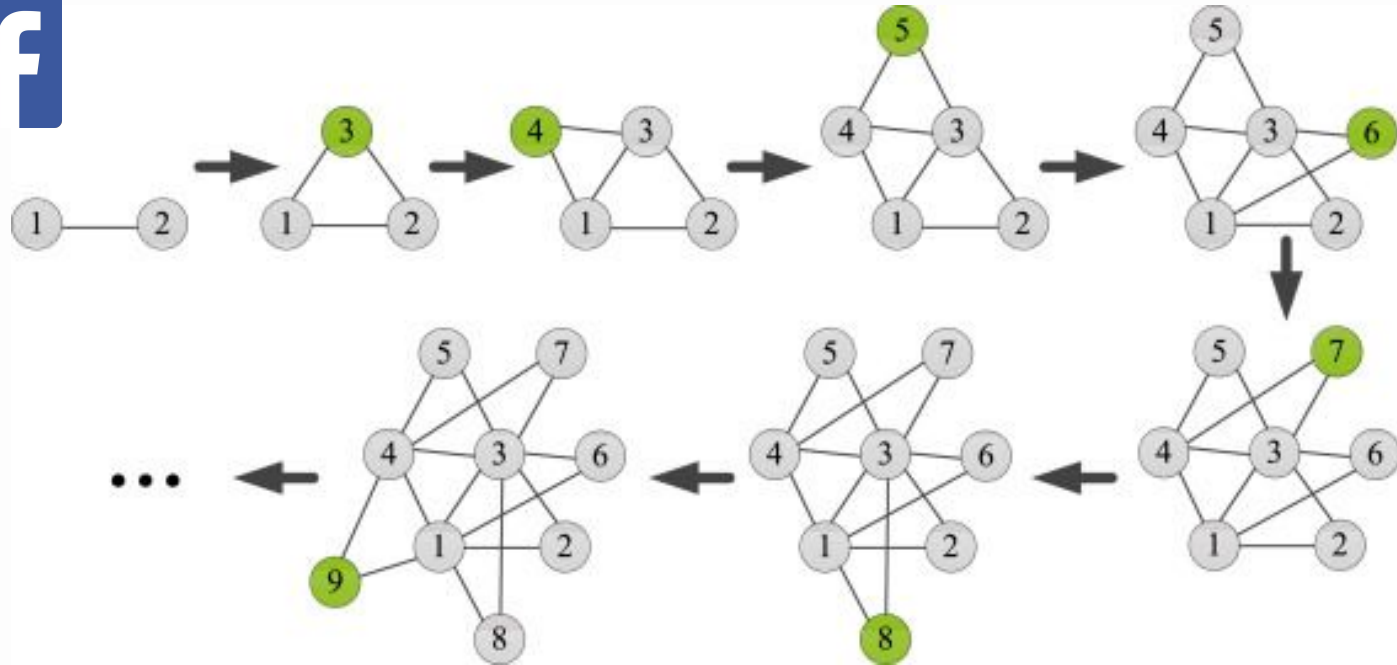
# **Uncovering Temporal Patterns in *Saccharomyces cerevisiae* Gene Co-expression Networks Through Dynamic Community Detection**

CS 198 Proposal: Atienza & Maximo

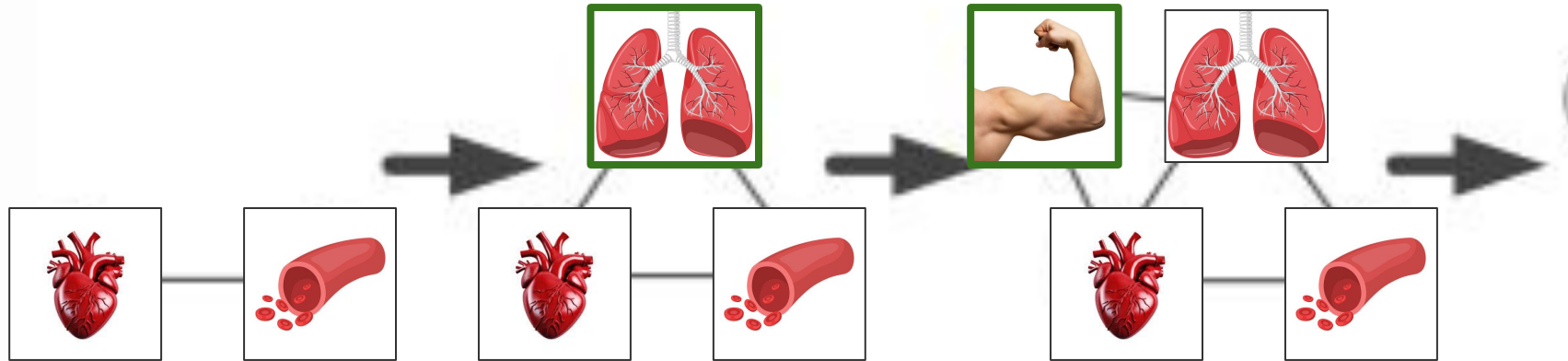
**01**

# **Overview of Biological Networks, Dynamic Community Detection, and Gene Co-Expression Networks**

# To put into perspective...



# To put into perspective...



## ...the same goes for genes!

# Overview of Biological Networks



## **Biological systems are inherently complex.**

They involve numerous interacting components, such as genes, proteins, and metabolites, all working together to regulate life processes (Barabási and Oltvai, 2004).



## **They exhibit community structures.**

These may pertain to functional modules which work to perform a biological role. It is an interesting area of research for analyzing how certain molecules are grouped within cells (Yu et al., 2013).



## **They are not static.**

The dynamic nature is evident in gene co-expression networks. Studying the structure within these networks can provide insights on the regulation of biological functions and the interactions between genes.

# Overview of Dynamic Community Detection



**Most of the literature is focused on static relations.**

This does not accurately represent the evolutionary world we live in.



**Time is an important component in studying these patterns.**

The dynamic nature of real-world networks complicates the task of community detection. The need to analyze the knowledge from the temporal dimension has



**Over time, the structure of a dynamic network changes.**

There are insights to be gained from understanding these changes and their transformations over time.

# Overview of Gene Co-expression Networks



## **They model interactions between genes.**

Nodes represent genes, and edges signify co-expression relationships, where the strength of an edge correlates with the similarity in gene expression.



## **They are dynamic in nature.**

There is a need to capture the temporal dynamics of GCNs because gene co-expression patterns fluctuate over time, which also affects the community structure.

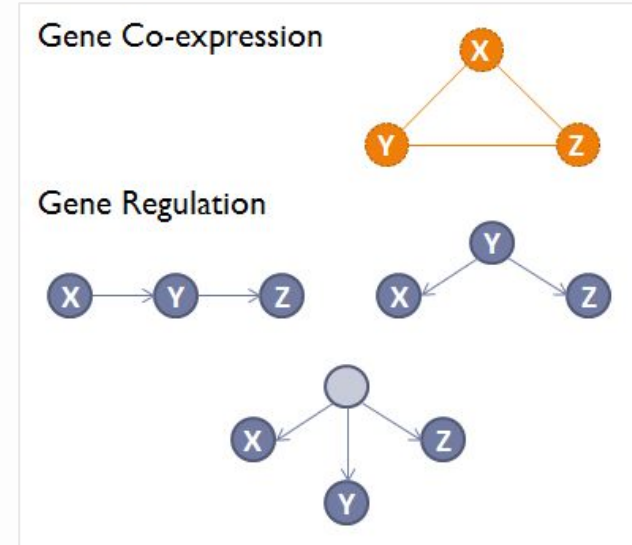
**02**

# **Preliminaries**



# Gene Co-expression Networks

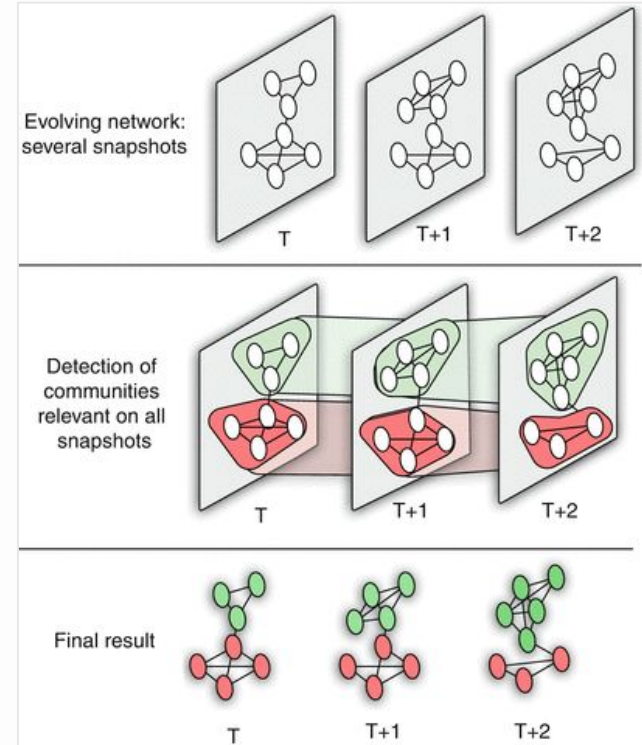
- Biological Network
- Nodes - Genes; Edges - Coexpression Relationship
  - A measure of similar patterns of activity (i.e. Expression Levels) across different conditions.
  - If a pair of genes have their activity go up or down together, they are said to be coexpressed.
    - They must work together!



[https://en.wikipedia.org/wiki/Gene\\_co-expression\\_network](https://en.wikipedia.org/wiki/Gene_co-expression_network)

# Community Detection

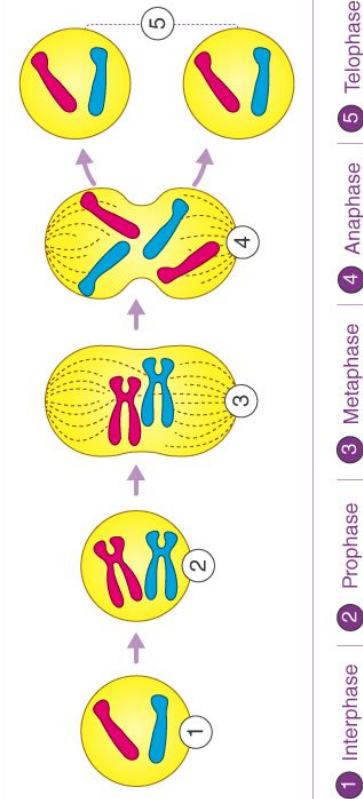
- Static Community Detection
  - Assumes the network is fixed throughout different time periods
  - Can reveal hierarchies in networks
- Dynamic Community Detection
  - Detect how communities vary at each time step
  - Can reveal temporal patterns
  - Our method of interest



[https://link.springer.com/referenceworkentry/10.1007/978-1-4614-6170-8\\_383](https://link.springer.com/referenceworkentry/10.1007/978-1-4614-6170-8_383)

# ***Saccharomyces cerevisiae***

- Well documented yeast cells, 6000 genes
- Dataset from Cho et al (1998)
  - Genes across the different phases of the mitotic cell cycle
  - Time-series data
  - How do the gene networks change at different phases of mitosis?

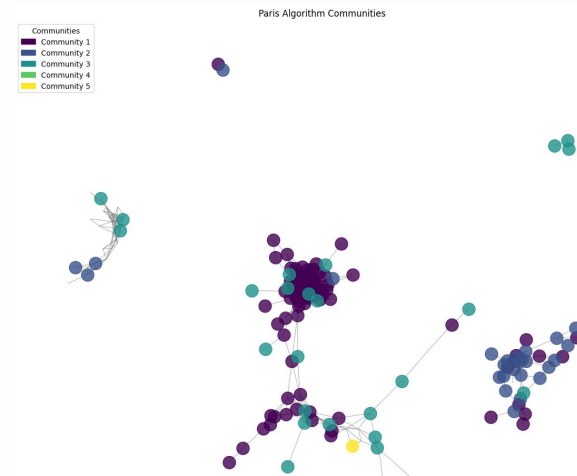


**03**

# **Related Works**

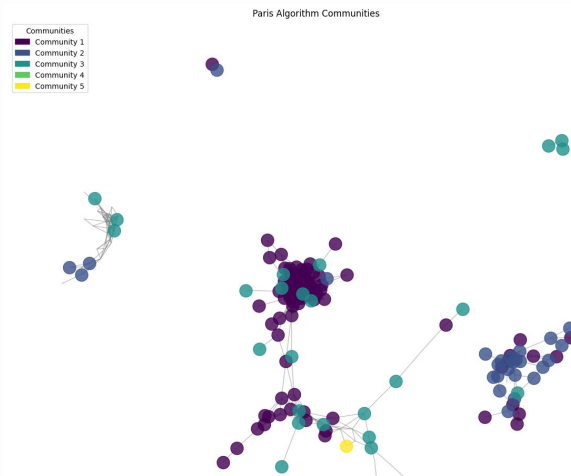
# Hierarchical CD on GCNs

- Calderon and Ventures (2024)
  - Explored hierarchical CD in *S. cerevisiae* to enhance functional classification of cell-cycle regulated genes
  - Analyzed the performance of three algorithms:
    - Girvan-Newman
    - Paris
    - LFM



# Hierarchical CD on GCNs

- Calderon and Ventures (2024)
  - Hierarchical CD is appropriate for GCNs due to their hierarchical nature.
  - Valuable insights into the static structure of GCNs
  - Open problem: dynamic nature of gene interactions



# Dynamic Community Detection Algorithms

- C-Blondel Algorithm
  - Proposed by Seifikar et al. (2020)
  - Built upon the Louvain algorithm, and leverages historical information to improve execution time
  - Compresses the network into supernodes and superedges
  - Identifies **destructive nodes** through a degree centrality heuristic
  - Faster than other DCD algorithms (D-Blondel, S-Blondel)
  - Comparable performance in terms of modularity

# Dynamic Community Detection Algorithms

- Semi-supervised ENMF (sE-NMF) Algorithm
  - Proposed by Ma and Dong (2017)
  - Combines evolutionary nonnegative matrix factorization (ENMF) and spectral clustering for dynamic community detection, balancing clustering accuracy and temporal consistency
  - Key advantage: Leveraging **a priori information**, avoiding local optima without affecting the run time complexity



# Dynamic Community Detection Algorithms

- Semi-supervised ENMF (sE-NMF) Algorithm
  - Evaluated against three well-known DCD algorithms (FacetNet, Kim-Han, DYNMOGA)
  - Particularly interesting network: breast cancer progression based on gene expression data
  - Outperformed the other algorithms in terms of specificity and sensitivity
  - Great potential as a tool for assessing temporal networks of disease progression

**04**

# References

# References

Barabási, A.-L., & Oltvai, Z. N. (2004). Network biology: Understanding the cell's functional organization. *Nature Reviews Genetics*, 5 (2), 101–113. <https://doi.org/10.1038/nrg1272>

Calderon, J. K., & Ventures, P. A. (2024). Hierarchical community detection on co-expression networks for functional classification of cell-cycle regulated genes of *saccharomyces cerevisiae*. Algorithms and Complexity Lab, University of the Philippines Diliman.

Cho, R. J., Campbell, M. J., Winzeler, E. A., Steinmetz, L., Conway, A., Wodicka, L., Wolfsberg, T. G., Gabrielian, A. E., Landsman, D., Lockhart, D. J., & Davis, R. W. (1998). A genome-wide transcriptional analysis of the mitotic cell cycle. *Molecular Cell*, 2 (1), 65–73. [https://doi.org/10.1016/s1097-2765\(00\)80114-8](https://doi.org/10.1016/s1097-2765(00)80114-8)

Dorogovtsev, S. N., & Mendes, J. F. F. (2002). Evolution of networks. *Advances in Physics*, 51 (4), 1079–1187. <https://doi.org/10.1080/00018730110112519>

Girvan, M., & Newman, M. E. J. (2002). Community structure in social and biological networks. *Proceedings of the National Academy of Sciences*, 99 (12), 7821–7826. <https://doi.org/10.1073/pnas.122653799>

Greene, D., Doyle, D., & Cunningham, P. (2010). Tracking the evolution of communities in dynamic social networks. 2010 International Conference on Advances in Social Networks Analysis and Mining, 176–183. <https://doi.org/10.1109/ASONAM.2010.17>

He, J., Chen, D., Sun, C., Fu, Y., & Li, W. (2017). Efficient stepwise detection of communities in temporal networks. *Physica A: Statistical Mechanics and its Applications*, 469, 438–446. <https://doi.org/https://doi.org/10.1016/j.physa.2016.11.019>

# References

Lau, L. Y., Reverter, A., Hudson, N. J., Naval-Sanchez, M., Fortes, M. R. S., & Alexandre, P. A. (2020). Dynamics of gene co-expression networks in time-series data: A case study in drosophila melanogaster embryogenesis. *Frontiers in Genetics*, 11. <https://doi.org/10.3389/fgene.2020.00517>

Ma, X., & Dong, D. (2017). Evolutionary nonnegative matrix factorization algorithms for community detection in dynamic networks. *IEEE Transactions on Knowledge and Data Engineering*, 29 (5), 1045-1058. <https://doi.org/10.1109/TKDE.2017.2657752>

Ovens, K., Eames, B. F., & McQuillan, I. (2021). Comparative analyses of gene co-expression networks: Implementations and applications in the study of evolution. *Frontiers in Genetics*, 12. <https://doi.org/10.3389/fgene.2021.695399>

Parapouli, M., Vasileiadi, A., Afendra, A.-S., & Hatziloukas, E. (2020). *Saccharomyces cerevisiae* and its industrial applications. *AIMS Microbiology*, 6 (1), 1-32. <https://doi.org/10.3934/microbiol.2020001>

Redekar, S. S., & Varma, S. L. (2022). A survey on community detection methods and its application in biological network, 1030-1037. <https://doi.org/10.1109/ICAAIC53929.2022.9792913>

Rossetti, G., & Cazabet, R. (2018). Community discovery in dynamic networks: A survey. *ACM Comput. Surv.*, 51 (2). <https://doi.org/10.1145/3172867>

Sattar, N. S., Buluc, A., Ibrahim, K. Z., & Arifuzzaman, S. (2023). Exploring temporal community evolution: Algorithmic approaches and parallel optimization for dynamic community detection. *Applied Network Science*, 8 (1), 64. <https://doi.org/10.1007/s41109-023-00592-1>

# References

Seifikar, M., Farzi, S., & Barati, M. (2020). C-blondel: An efficient louvainbased dynamic community detection algorithm. IEEE Transactions on Computational Social Systems, 7 (2), 308–318. <https://doi.org/10.1109/TCSS.2020.2964197>

Singh, D. K., Haraty, R. A., Debnath, N. C., & Choudhury, P. (2020). An analysis of the dynamic community detection algorithms in complex networks. 2020 IEEE International Conference on Industrial Technology (ICIT), 989–994. <https://doi.org/10.1109/ICIT45562.2020.9067224>

Stuart, J. M., Segal, E., Koller, D., & Kim, S. K. (2003). A gene-coexpression network for global discovery of conserved genetic modules. Science, 302 (5643), 249–255. <https://doi.org/10.1126/science.1087447>

Yu, D., Kim, M., Xiao, G., & Hwang, T. H. (2013). Review of biological network data and its applications. Genomics Inform, 11 (4), 200–210. <https://doi.org/10.5808/GI.2013.11.4.200>

# Thanks!

**Do you have any questions?**

Carmelo Ellezandro Atienza | Calvin James Maximo  
[cratienza1@up.edu.ph](mailto:cratienza1@up.edu.ph) | [ctmaximo1@up.edu.ph](mailto:ctmaximo1@up.edu.ph)

CREDITS: This presentation template was created by [Slidesgo](#), and includes icons by [Flaticon](#), and infographics & images by [Freepik](#)

Please keep this slide for attribution