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# **Systematic evaluation of the consensus across pathway databases**

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# Introduction

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- Deconvoluting biology into networks

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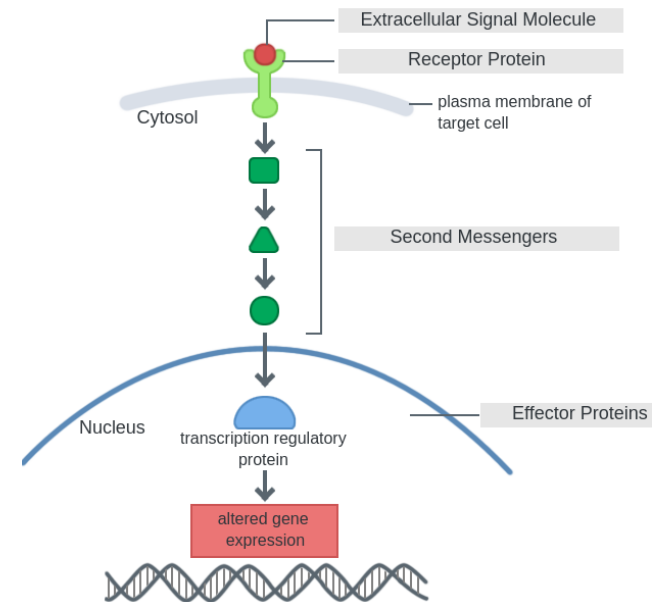
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- Deconvoluting biology into networks
  - Nodes represent biological entities (e.g, proteins, chemicals, etc.)
  - Edges represent the type of interaction (e.g., activation, inhibition, etc.)

# Introduction

## ■ Deconvoluting biology into networks

- Nodes represent biological entities (e.g., proteins, chemicals, etc.)
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Taken from:  
<https://theory.labster.com/signaling-cascade/>

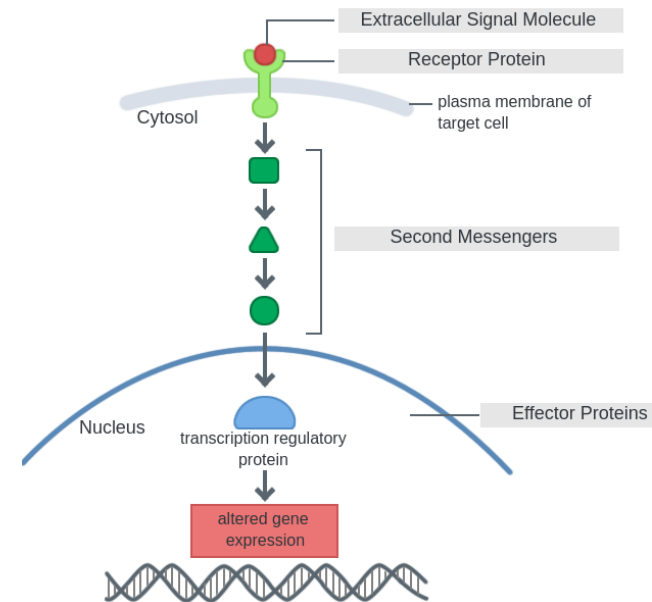
# Introduction

## ■ Deconvoluting biology into networks

- Nodes represent biological entities (e.g, proteins, chemicals, etc.)
- Edges represent the type of interaction (e.g., activation, inhibition, etc.)

## ■ Deconvoluting biological networks into pathways

- Pathways correspond to “small networks” representing a specific molecular process (e.g., inflammation, glucose-uptake, DNA repair, etc.)
- For researchers, these “small networks” or pathways represent how a specific part of biology works



Taken from:  
<https://theory.labster.com/signaling-cascade/>

# The Problem (1)



- Numerous pathway databases [1...3]
- Pathways are represented using different schemas and stored in data-silos
- By integrating this information, we can evaluate whether there is consensus among these resources true to the biology

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# The Problem (2)

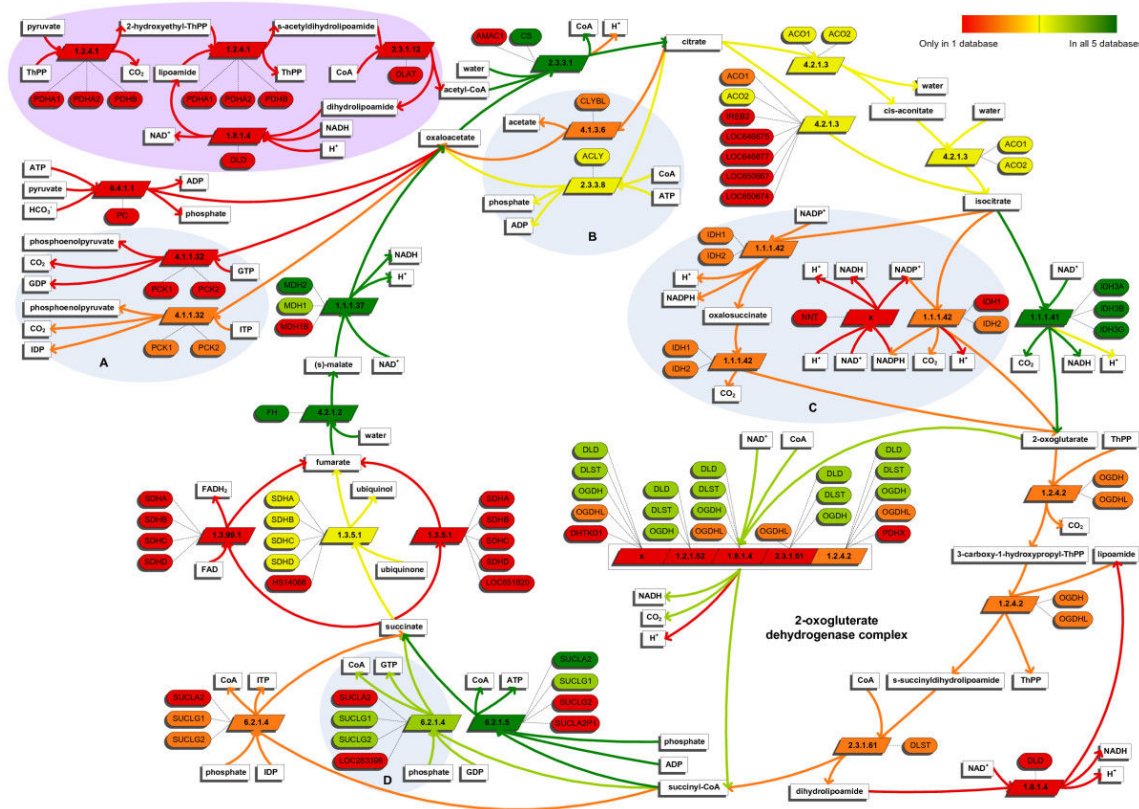
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*Krebs cycle* pathway (glucose metabolism) network from 5 different databases [4]

# The Problem (2)

Krebs cycle pathway (glucose metabolism) network from 5 different databases [4]

**Legend**  
**Red** = only in one database  
**Green** = multiple databases



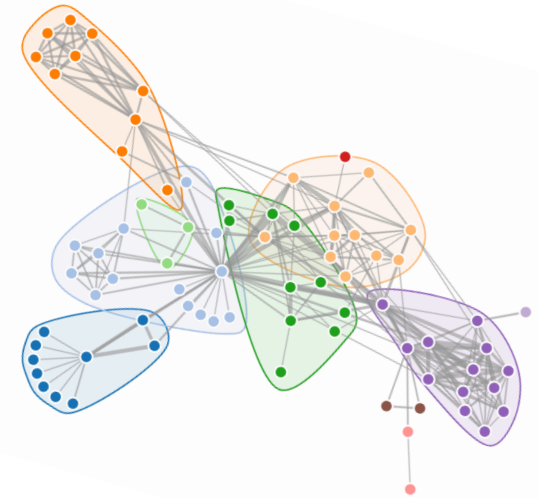


# Integrating pathway databases

Harmonization of resources



Integration in an unified schema

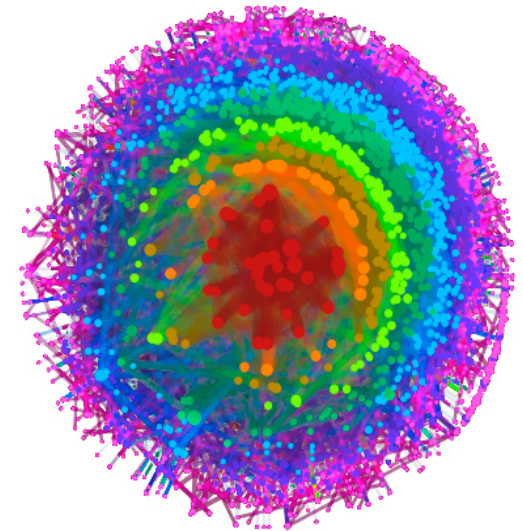


# Integrating pathway databases

Harmonization of resources



Integration in an unified schema



Network representing biological interactions in multiple databases



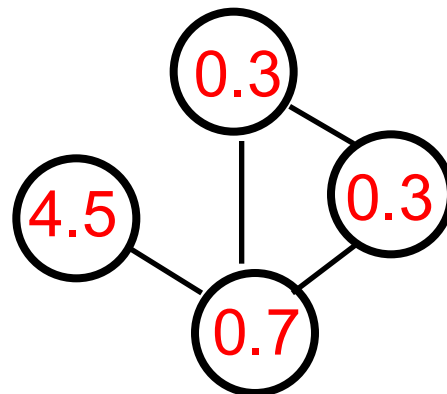
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# Use patients' signatures to predict treatments

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- Simulate combination of treatment in specific patients
- Use the trained classifier looking for a positive prediction

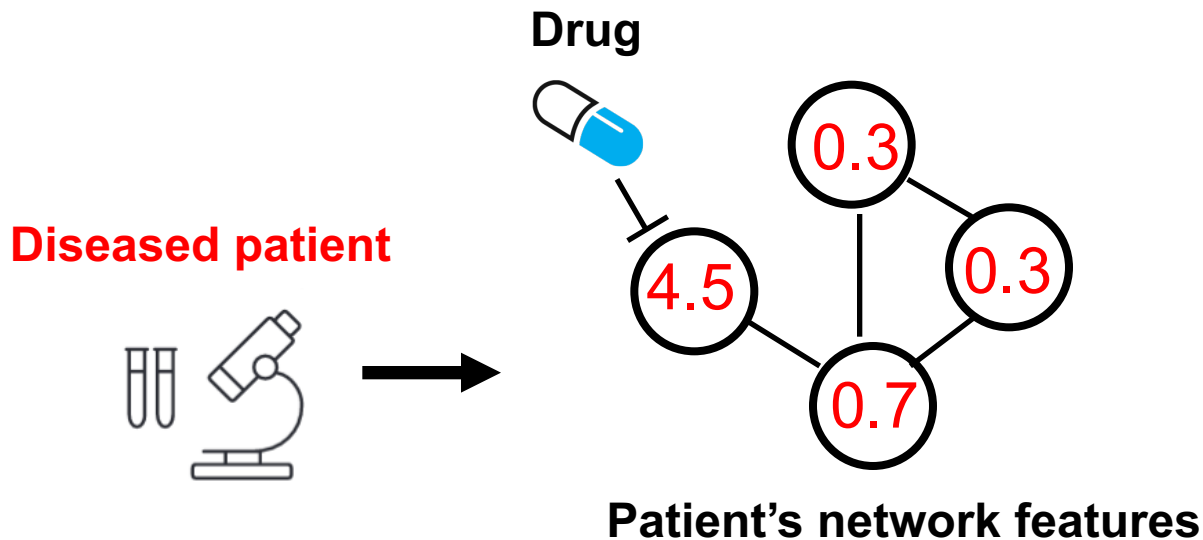
**Diseased patient**



**Patient's network features**

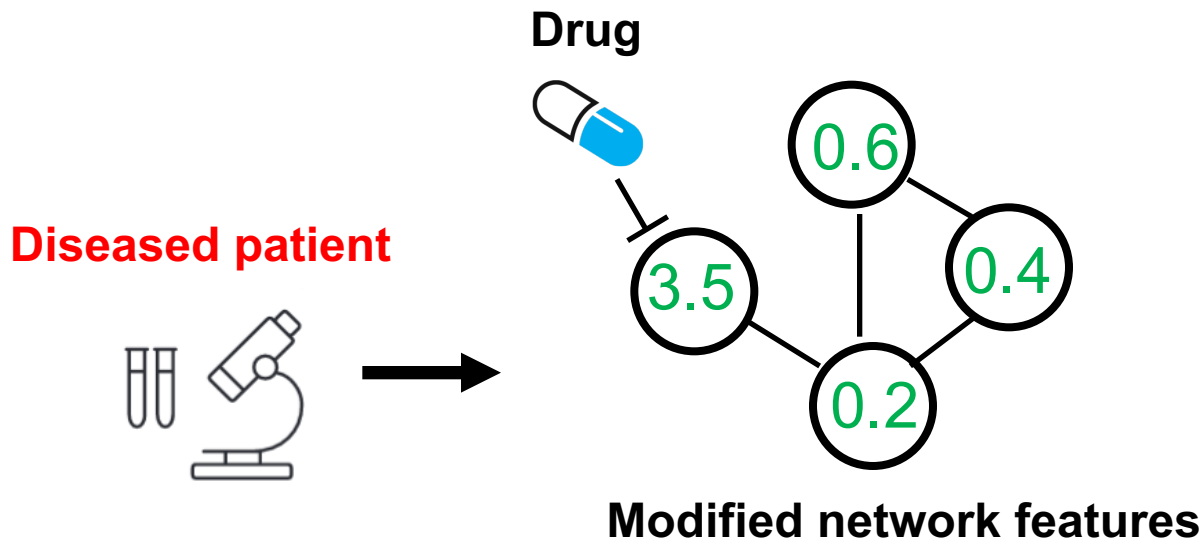
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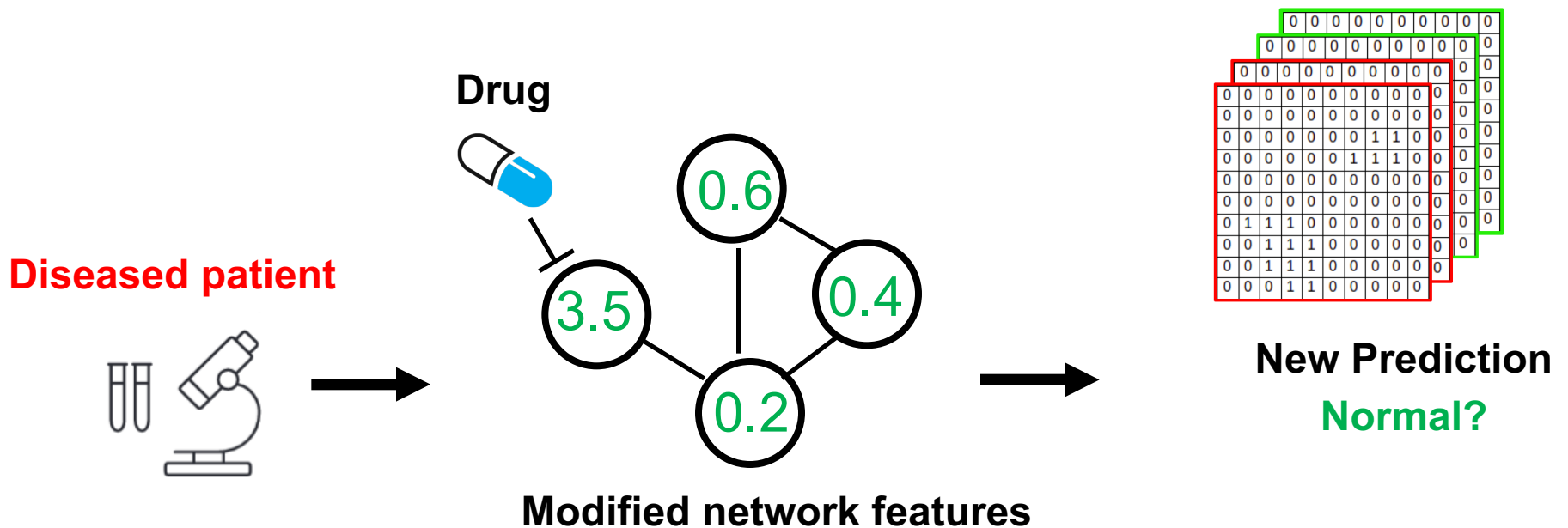
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# Use patients' signatures to predict treatments

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# Other applications

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## ■ Machine Learning

- Train classifiers for each database and compare performance (which databases/pathways predict better? use pathways as markers?)
- Remove random "parts" of the network to evaluate changes in performance for prioritization (non relevant in the disease context? noise?)
- ...

## ■ Graph theory

- Biological contradictions (e.g., drug A **activates**/**inhibits** protein B)
- Explore biological landscape (where is science heading to?)
- ...



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# Summary

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- Biology can be deconvoluted as a network
- The entire network is divided into smaller subnetworks that represent specific biological processes -> pathways
- We can interrogate this network with data and in a combination with ML for
  - Prediction
  - Treatment simulation
  - Pathway prioritization
  - ...

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**Thank you!**

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# Acknowledgments

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- Charles Tapley Hoyt

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# References

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# Evaluating biological knowledge (ongoing work)

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