

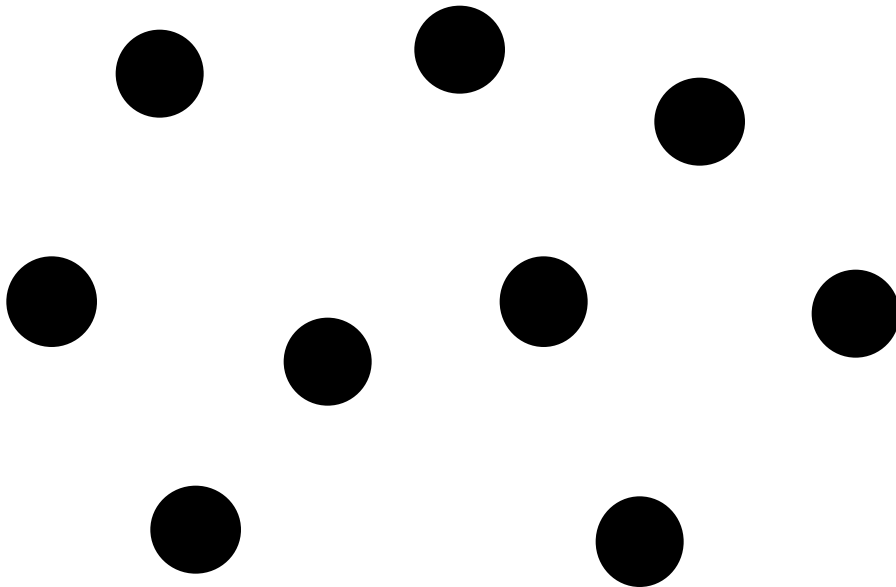
# Representation Learning on Networks

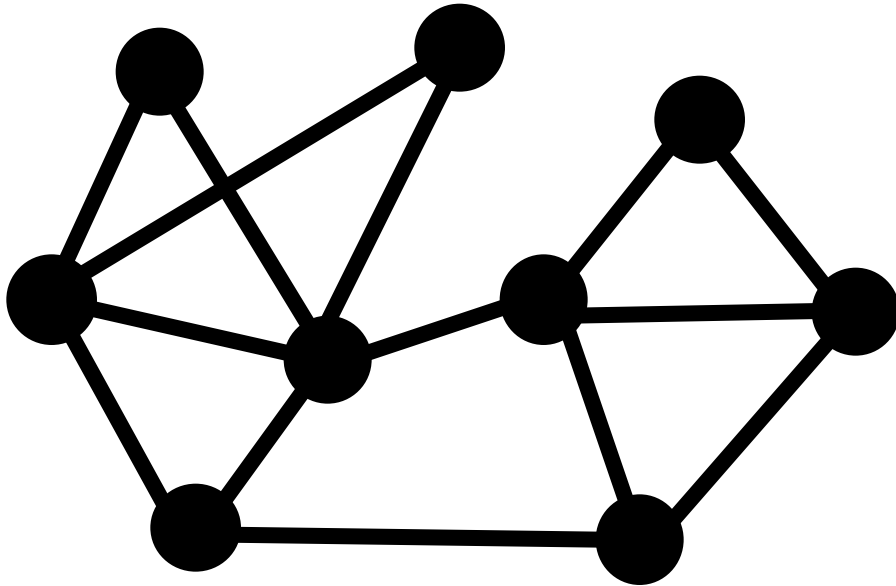
Jure Leskovec, William L. Hamilton, Rex Ying, Rok Sosis  
Stanford University



# Why networks?

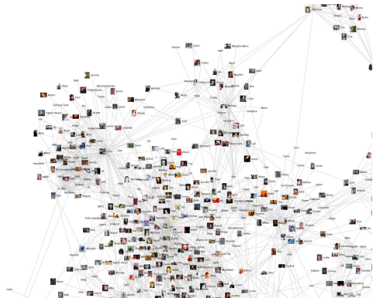
Networks are a general language for describing and modeling complex systems



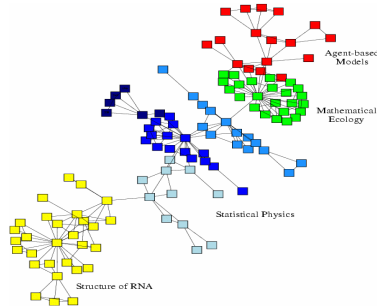


# Network!

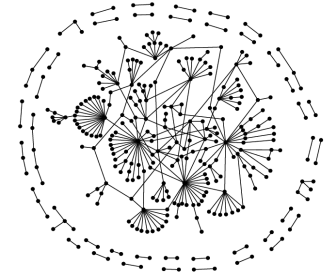
# Many Data are Networks



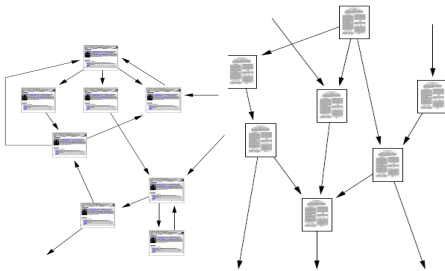
Social networks



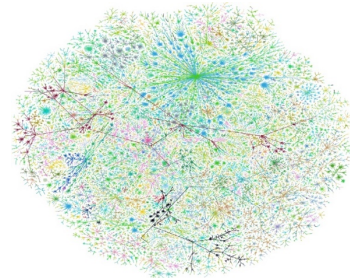
Economic networks



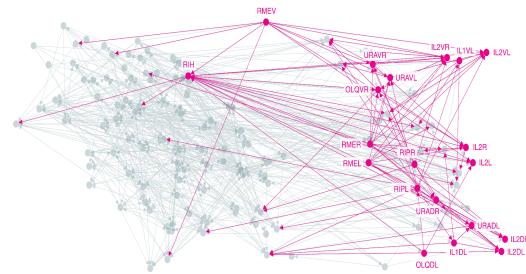
Biomedical networks



Information networks:  
Web & citations



Internet



Networks of neurons

# Why Networks? Why Now?

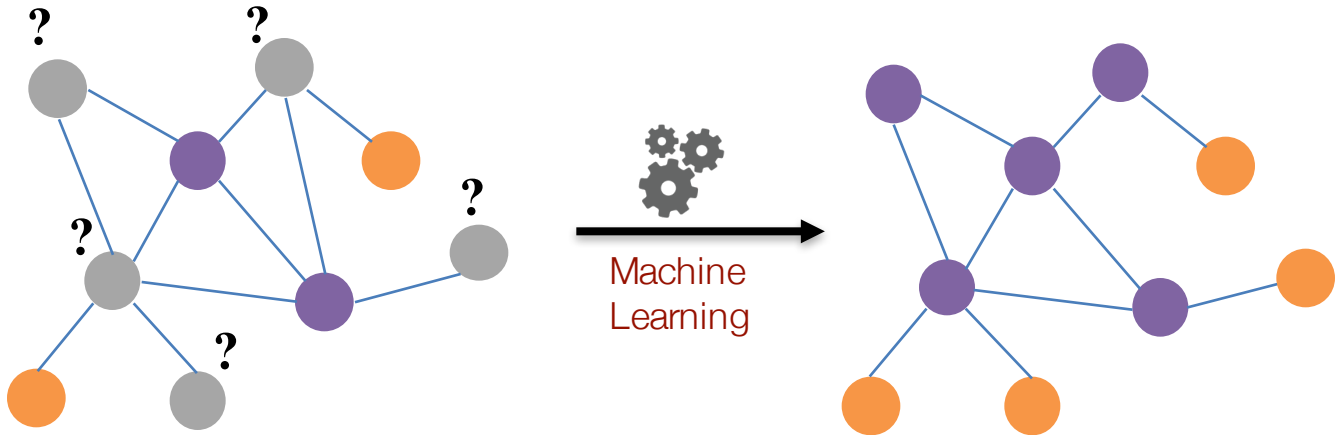
- Universal language for describing complex data
  - Networks from science, nature, and technology are more similar than one would expect
- Shared vocabulary between fields
  - Computer Science, Social science, Physics, Economics, Statistics, Biology
- Data availability (+computational challenges)
  - Web/mobile, bio, health, and medical
- Impact!
  - Social networking, Social media, Drug design

# Machine Learning with Networks

## Classical ML tasks in networks:

- Node classification
  - Predict a type of a given node
- Link prediction
  - Predict whether two nodes are linked
- Community detection
  - Identify densely linked clusters of nodes
- Network similarity
  - How similar are two (sub)networks

# Example: Node Classification





# Example: Node Classification

**Classifying the function of proteins in the interactome!**

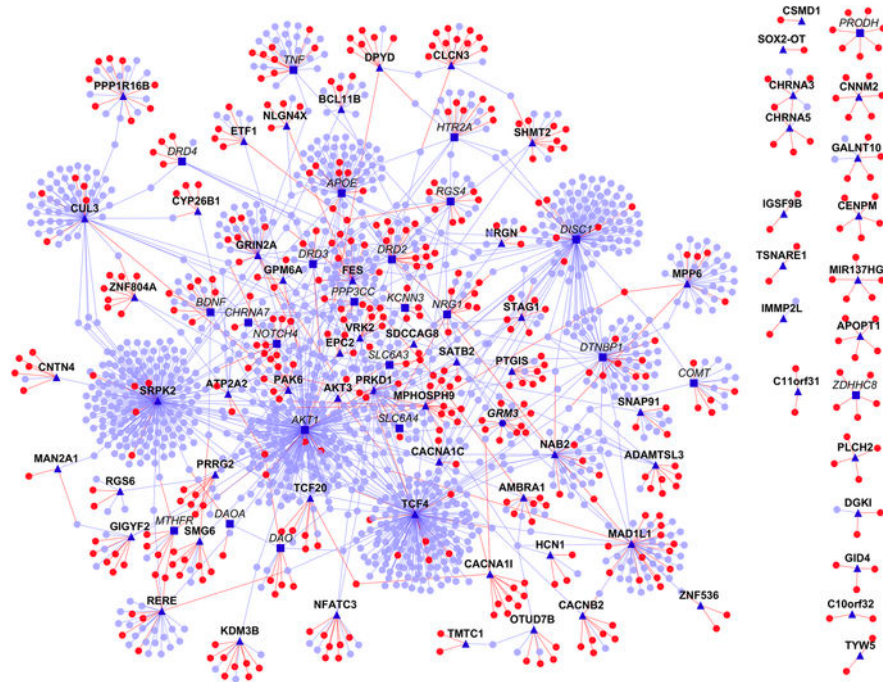
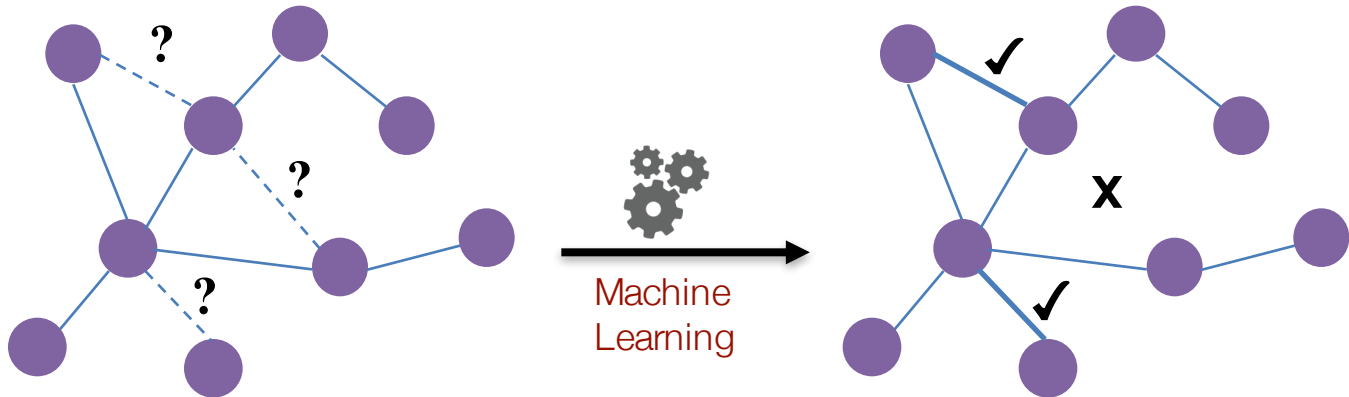


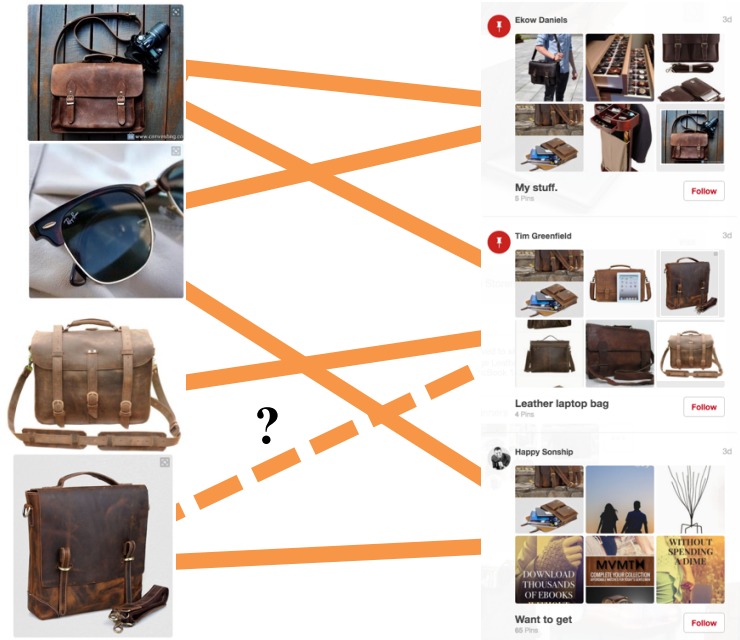
Image from: Ganapathiraju et al. 2016. [Schizophrenia interactome with 504 novel protein-protein interactions](#). *Nature*.

# Example: Link Prediction



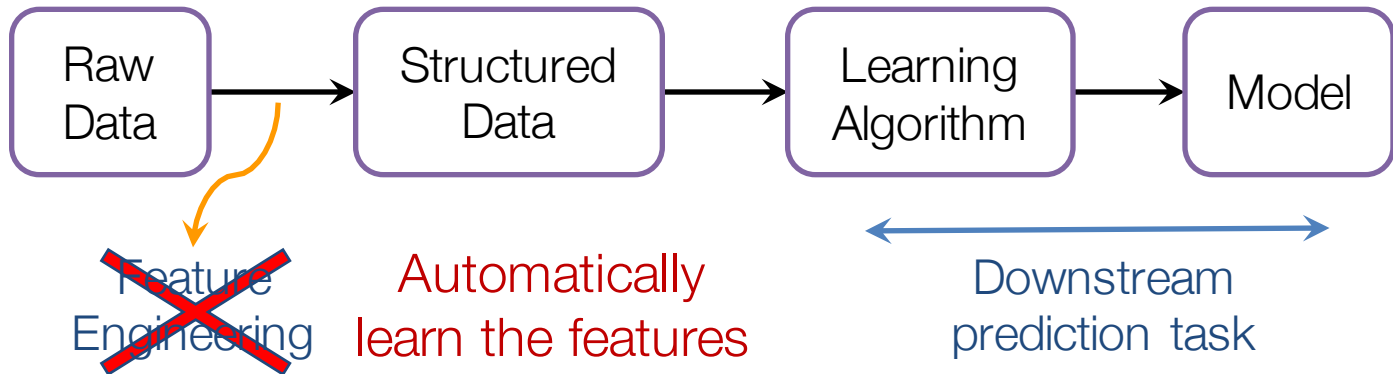
# Example: Link Prediction

**Content  
recommendation  
is link prediction!**



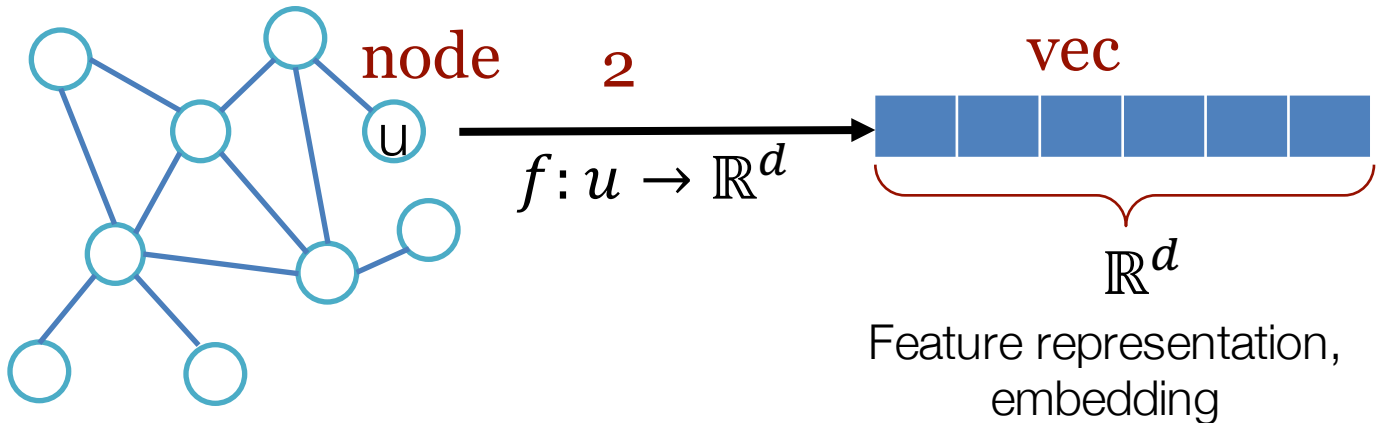
# Machine Learning Lifecycle

- (Supervised) Machine Learning Lifecycle: This feature, that feature.  
Every single time!



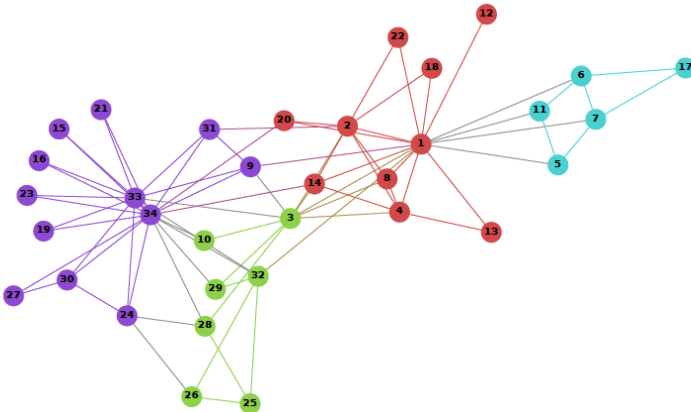
# Feature Learning in Graphs

**Goal:** Efficient task-independent feature learning for machine learning in networks!

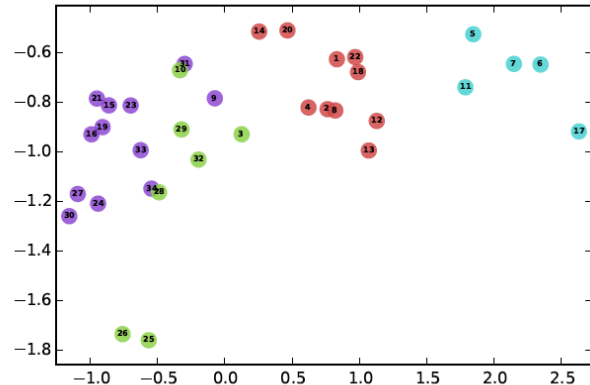


# Example

- Zachary's Karate Club Network:



**Input**

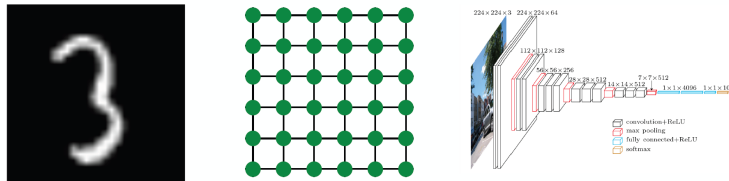


**Output**

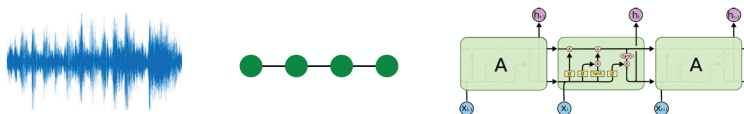
Image from: [Perozzi et al. 2014](#). DeepWalk: Online Learning of Social Representations. *KDD*.

# Why Is It Hard?

- Modern deep learning toolbox is designed for simple sequences or grids.
  - CNNs for fixed-size images/grids....



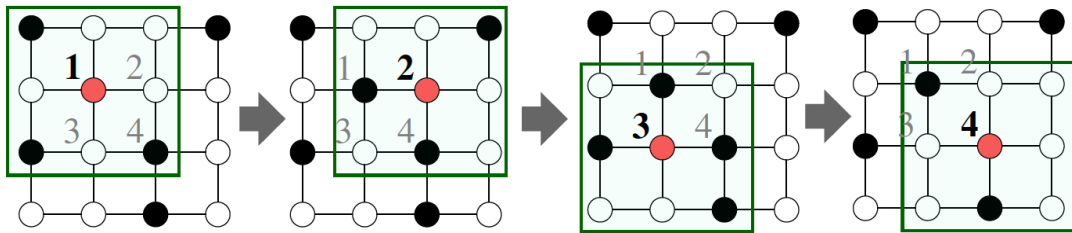
- RNNs or word2vec for text/sequences...



# Why Is It Hard?

- But networks are far more complex!

- Complex topographical structure (i.e., no spatial locality like grids)



- No fixed node ordering or reference point (i.e., the isomorphism problem)
- Often dynamic and have multimodal features.



# This talk

- 1) Node embeddings
  - Map nodes to low-dimensional embeddings.
- 2) Graph neural networks
  - Deep learning architectures for graph-structured data
- 3) Applications