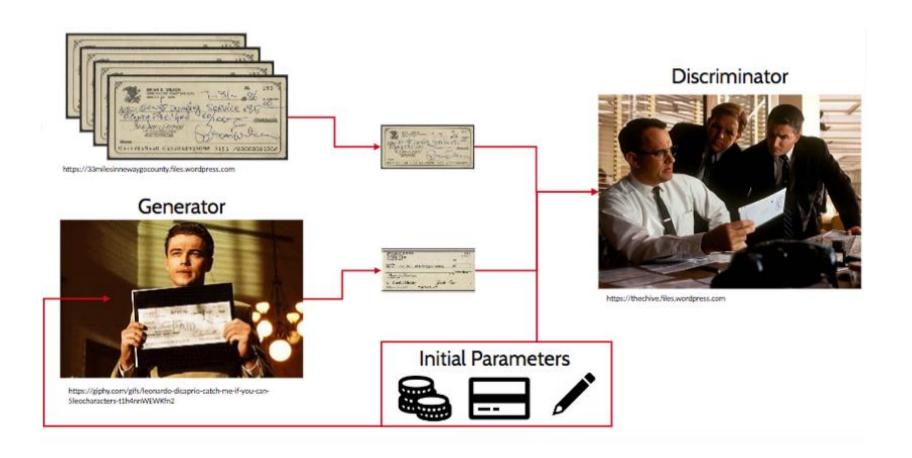


PH451, PH551 April 3, 2025

#### **Generative Adversarial Networks**



### **GAN**

#### **Generative Adversarial Networks**

- Co-trained networks
  - first gradient ascent for discriminator
  - then gradient descent for generator
- Game Theory:
  - Nash Equilibrium, Minimax game
  - Watch out for "mode collapse"

### **GAN**

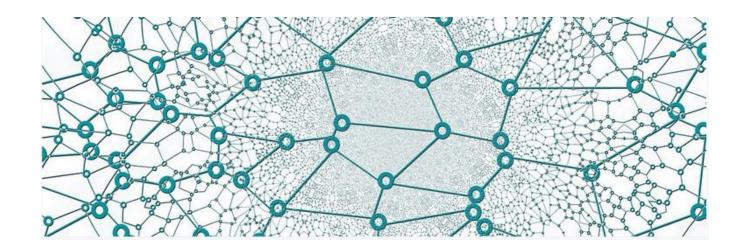
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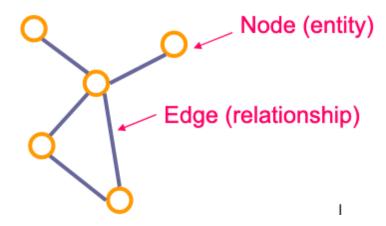
 04/03/25
 Sergei Gleyzer
 PH451/PH551 Lecture
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## **Graph Neural Networks**

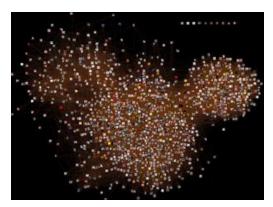


#### **Network definition:**

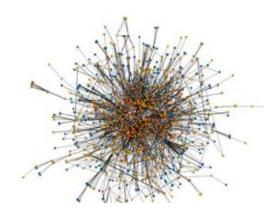
- collection of entities (nodes) joined by relationships (edges)
- Network = "graph"



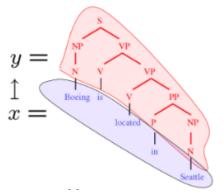
### **Examples**



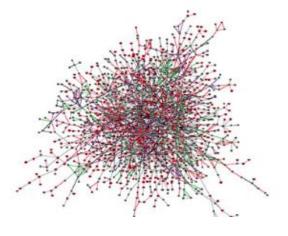
**Social Network** 



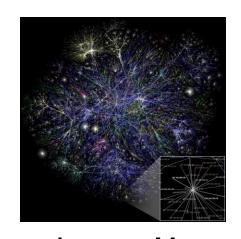
**Genomic Associations** 



Natural Language Parsing



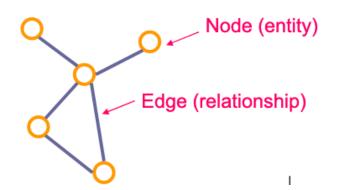
**Protein-Protein Interactions** 

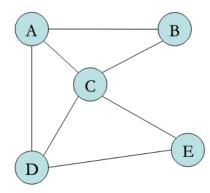


Internet Map

#### How to build a simple graph

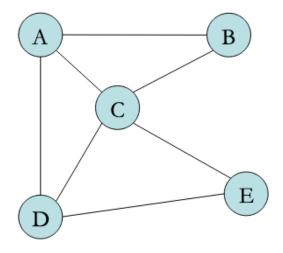
- Measure distance between pairs
- Connect each entity to its k nearest neighbors





 Define: Adjacency matrix: A<sub>ij</sub> weight of edge from i to j

	1	1	1	
1		1		
1	1		1	1
1		1		1
		1	1	

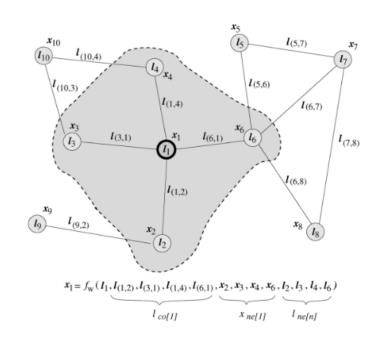


Adjacency matrix

### **Graph Neural Networks**

#### **Graph Neural Network:**

- State of the node depends on its neighbors
- Any neural network can be expressed as a graph
- Powerful approach for when your data is not inherently Euclidian
- Optimize by energy minimization

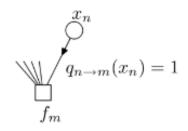


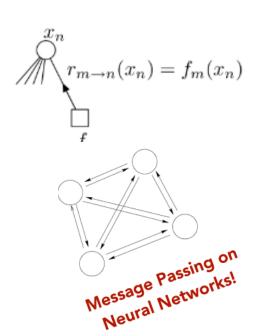
$$x_i = \sum_{j \in \mathcal{N}(i)} f(l_i, l_{i,j}, x_j, l_j)$$

### Message Passing

# Message Passing Graph Neural Network:

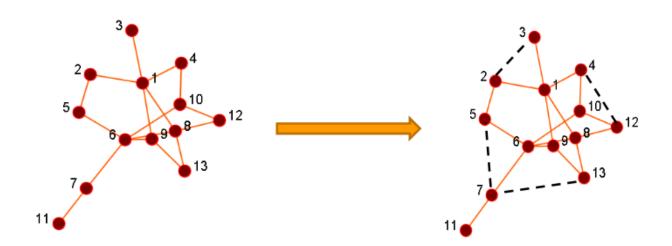
- Message Passing
- Key idea: each graph node has a feature vector hidden state
  - Update the hidden state with message from previous hidden state (possibly also edges)





#### **Example: link prediction**

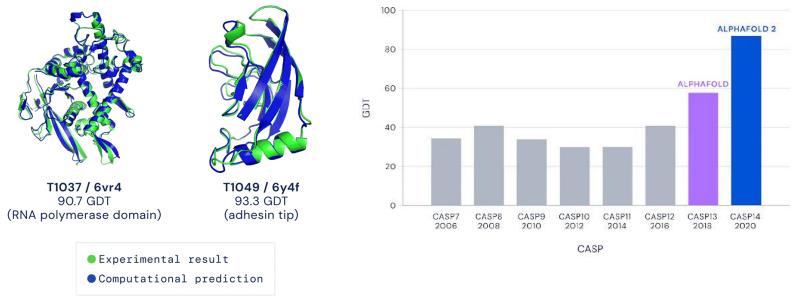
Which nodes are likely to get connected



i.e. Social network connection recommendations

#### **Example: protein folding**

Predicting which shapes proteins fold into



Alphafold 2 (DeepMind 2020)