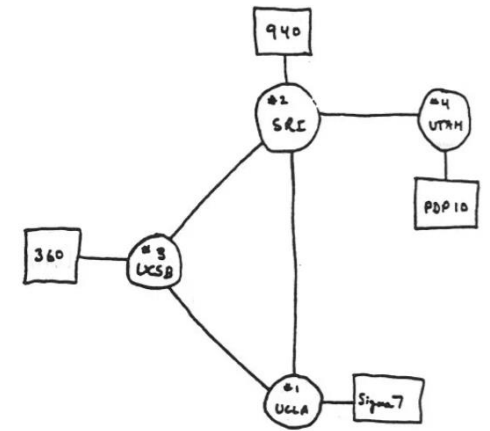
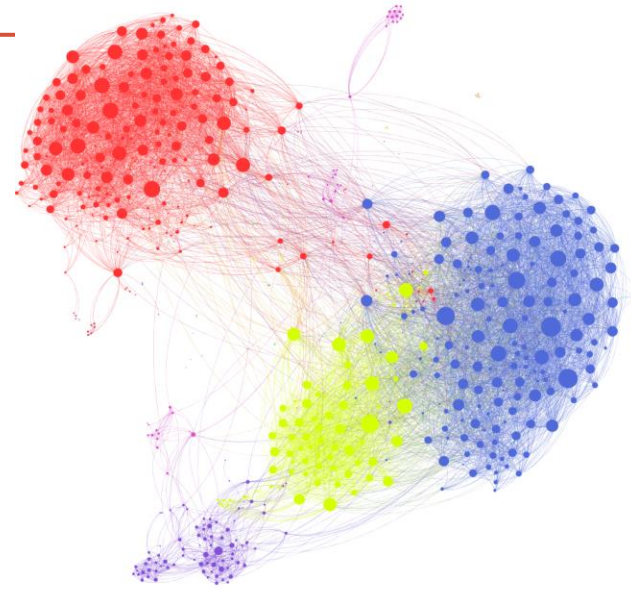
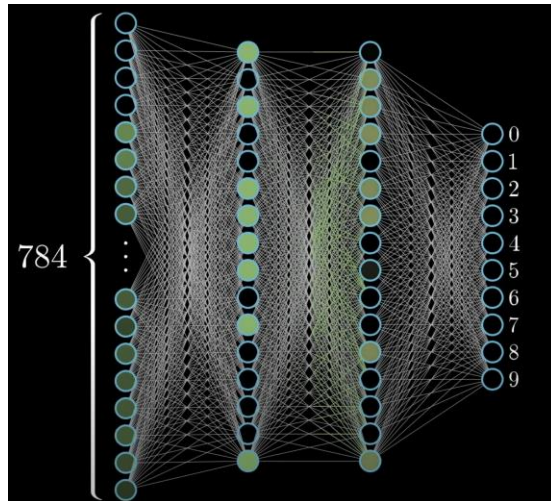


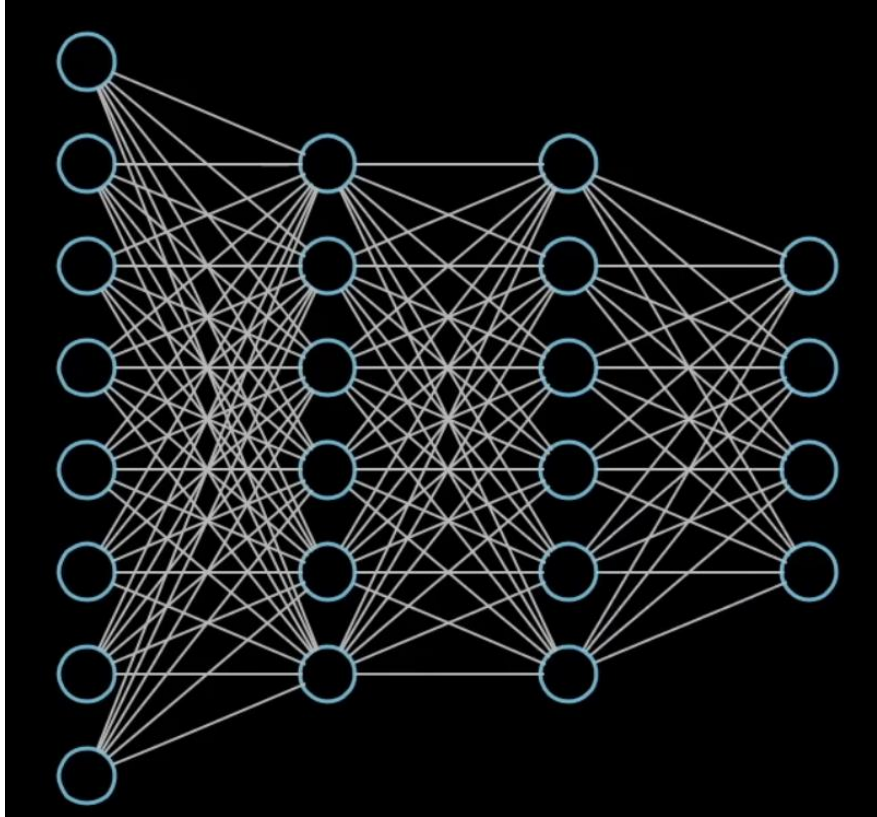
Handout: <https://bit.ly/NeuralNets-GraphRepresentation>

Book your mock interview today!

# GRAPH REPRESENTATION



# Neural Networks: Biologically inspired structure



**Neural Network:** Collection of connected neurons modeled after the brain

Human brain has billions of neurons  
Each neuron is connected to thousands of other neurons

Video intro to neural networks: <https://youtu.be/aircAruvnKk?feature=shared>

# What is an artificial neuron?



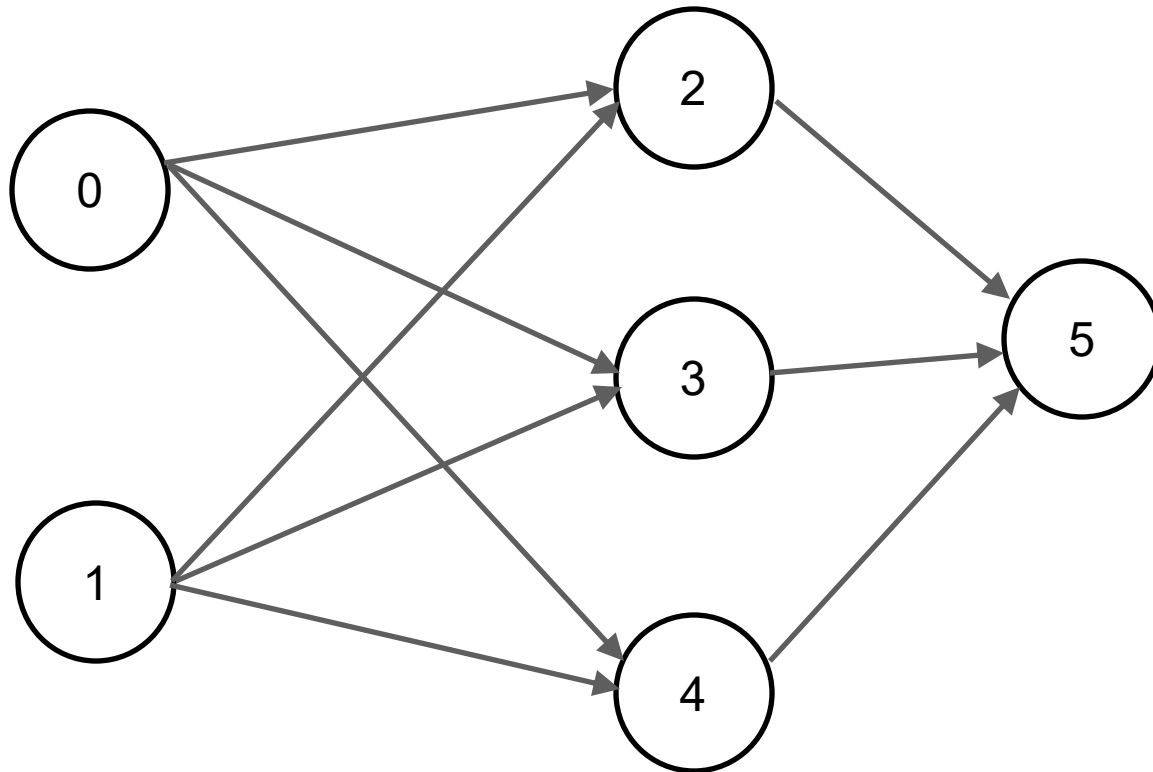
Neuron  $\rightarrow$  Thing that holds a number

If a neuron just holds a number...where does that number come from?

Terminology: the number stored in the neuron is also called its activation value or value for short

# What decides a neuron's value?

A neuron's activation is computed based on the activations of neurons connected to it.  
Think of connections as pathways of information flow!



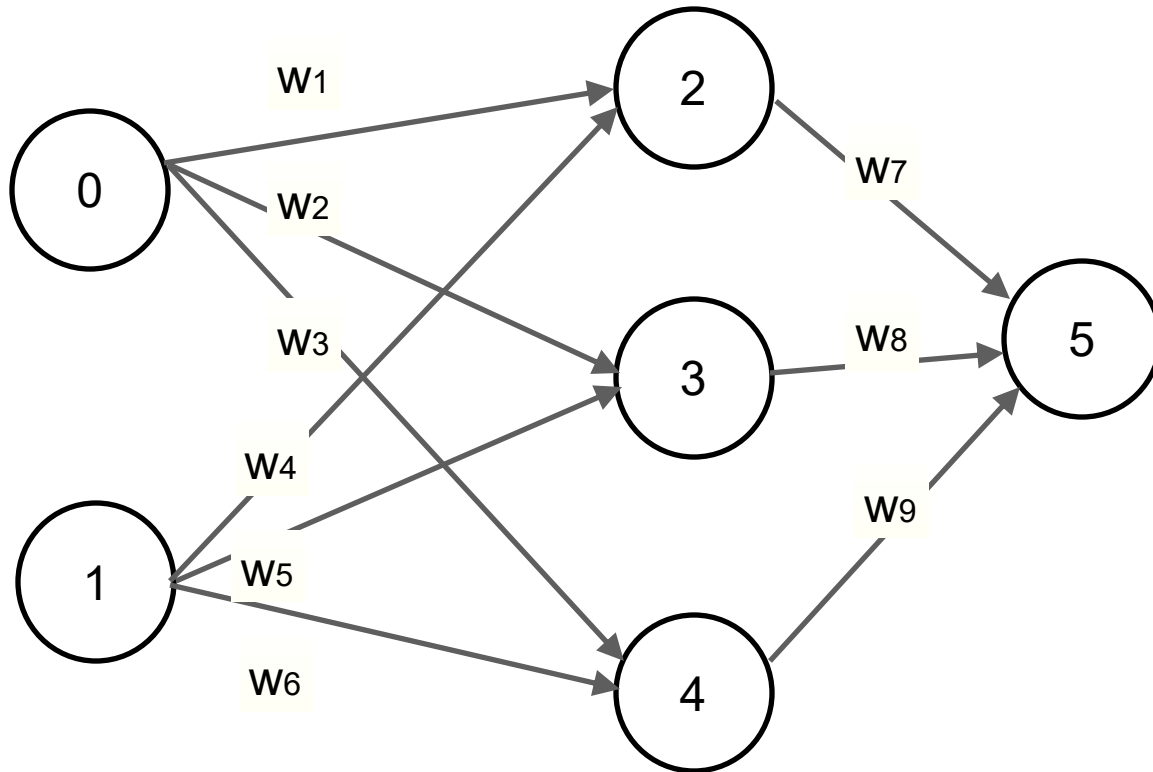
Which neurons will determine the activation value of neuron 3?

**An example NN with 6 neurons and 9 connections**

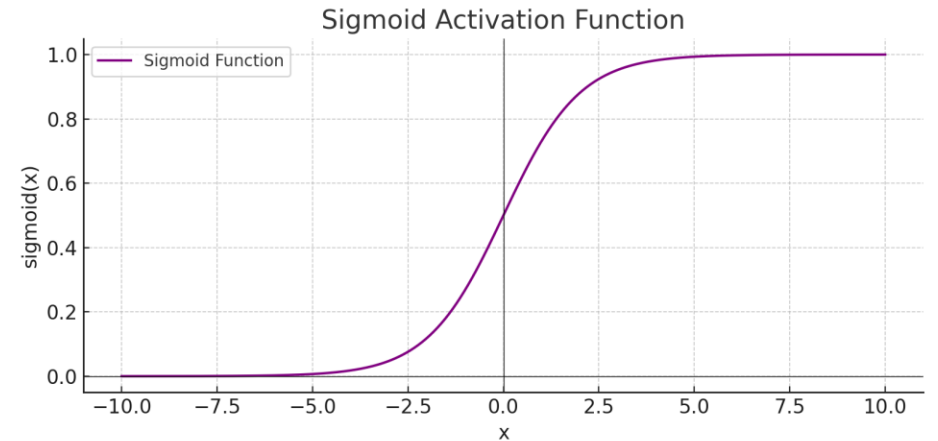
# What decides a neuron's value?

The activation for a neuron is computed as the weighted sum of its input neurons (with some adjustments).

Compute the activation of neuron 3

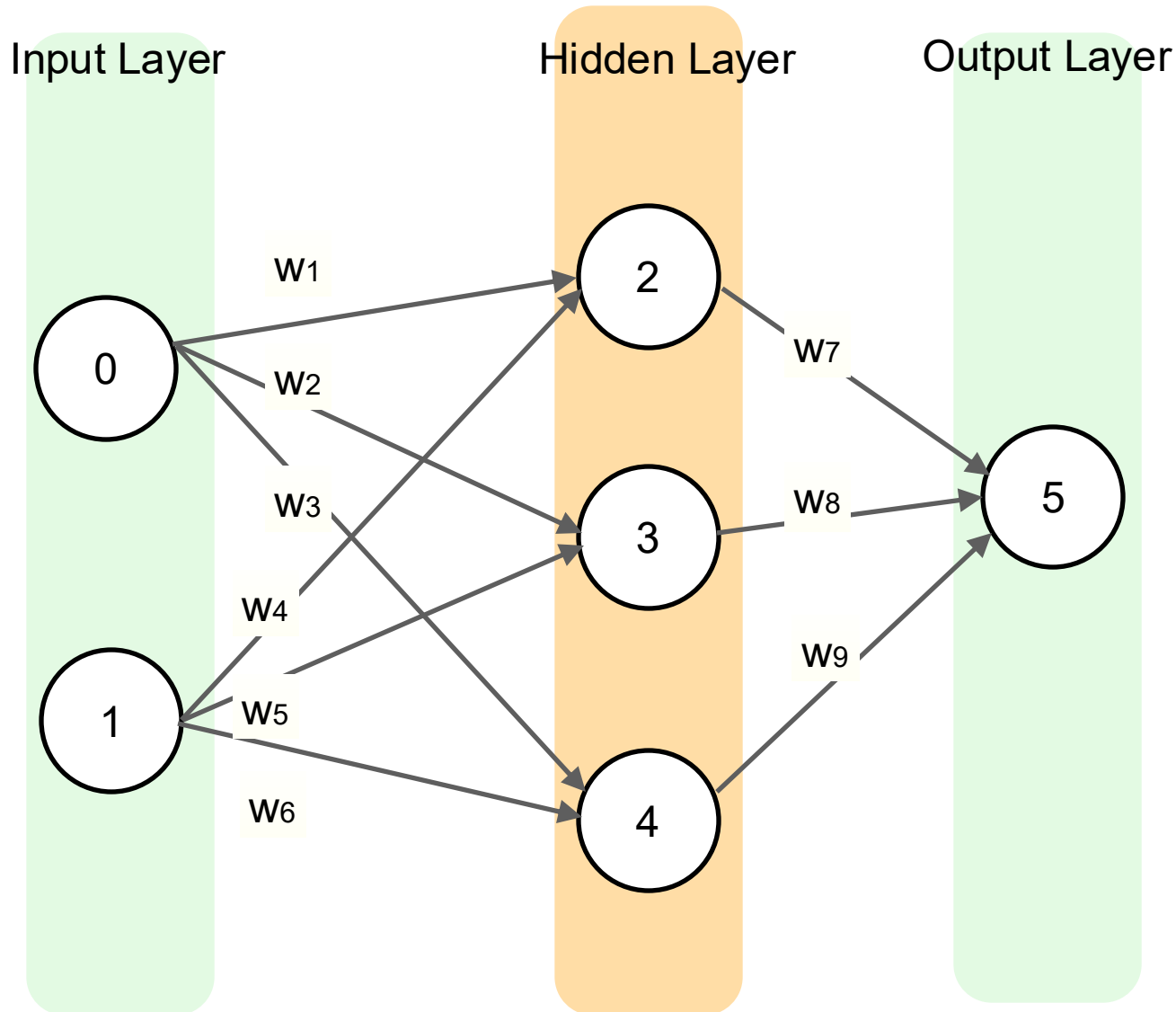


**An example NN with 6 neurons and 9 connections**





# How does a feed-forward neural network compute?



A **feedforward neural network (FNN)** is the simplest type of neural network where data flows in one direction — from input to output — without looping back.

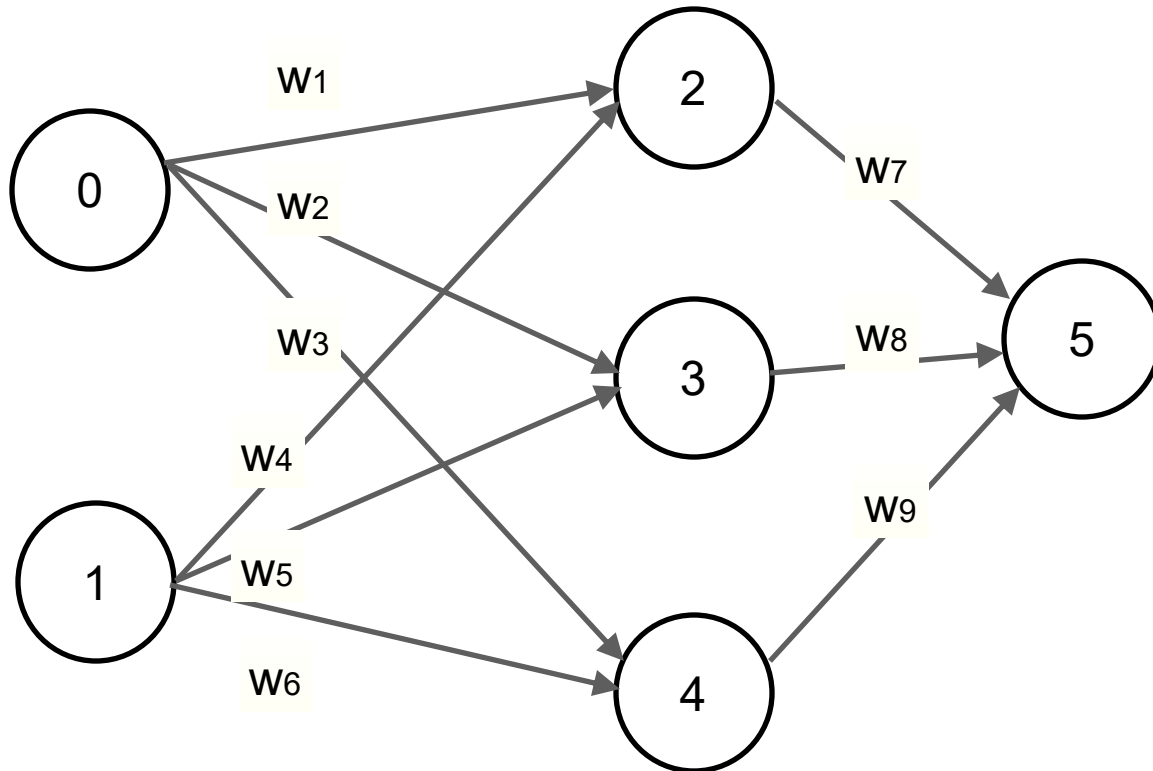
## NN Terminology

- Neurons
- Connections
- Input Layer
- Hidden Layer(s)
- Output Layer
- Neuron Info: activation value, bias, activation function

How would you represent the neural net using the data structures learned so far?

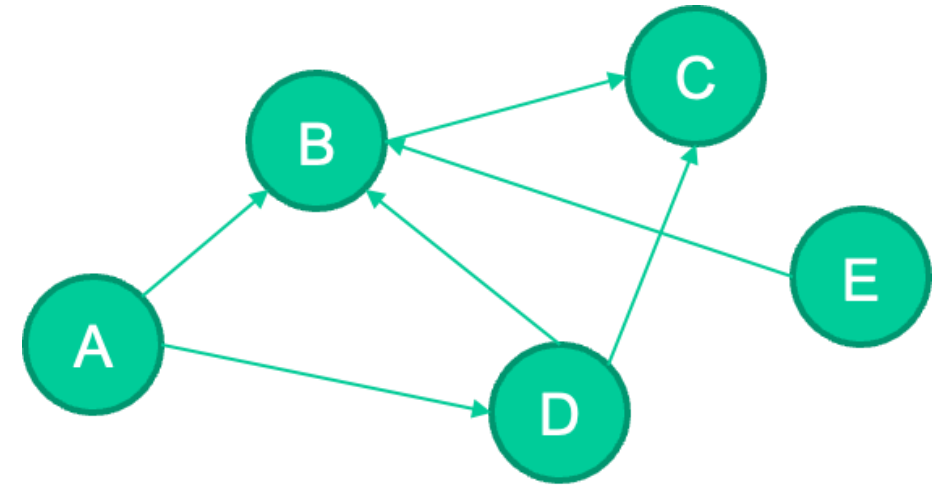
# Neural Networks as Graphs

(PA03) Model a neural network as a \_\_\_\_\_ graph.



NN is a set of **neurons and connections**

- Connections are directed (one-way)
- Connections have weights (strength of connection)



Graph is a set of nodes (vertices) and edges

- Directed graph: Edges go one way
- Undirected graph: Edges go both ways
- Weighted graph: Edges have weights

# Graph Terminology and Notation

Graph  $G = \{V, E\}$

- Vertices  $V = \{0, 1, 2, 3, \dots, n - 1\}; |V| = n$
- Edges  $E = \{(u, v) \mid u \in V, v \in V\}; |E| = m$

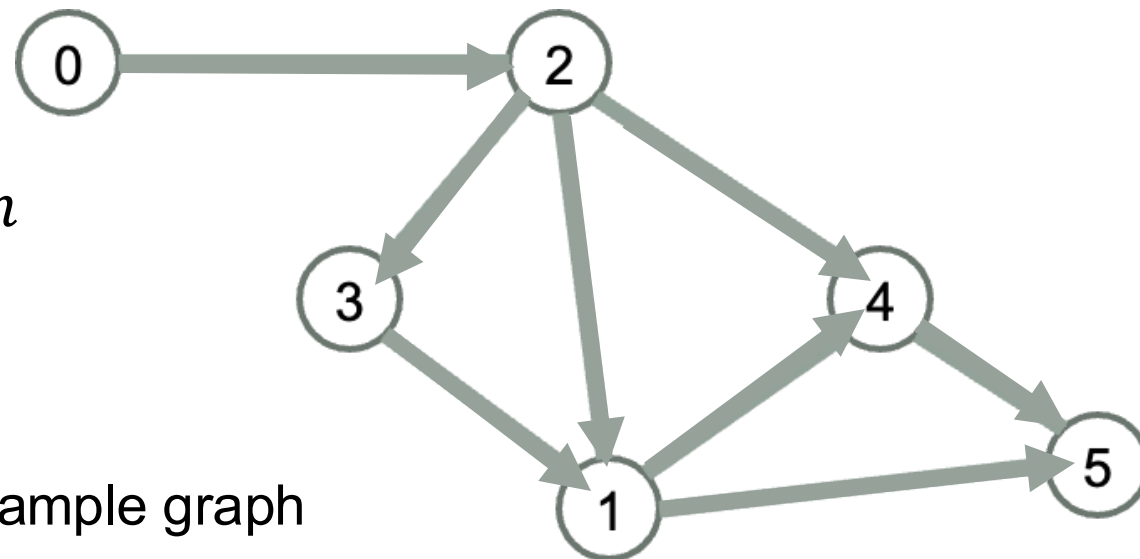
Activity 1: Write the vertices and edges for the example graph

$V =$

$E =$

$n =$

$m =$





# Representing vertices

Graph  $G = \{V, E\}$

$$V = \{0, 1, 2, 3, \dots, n - 1\}$$

Activity 2: Your graph has vertices labeled 0 through  $n - 1$ .

What's the best way to store the set of vertices?

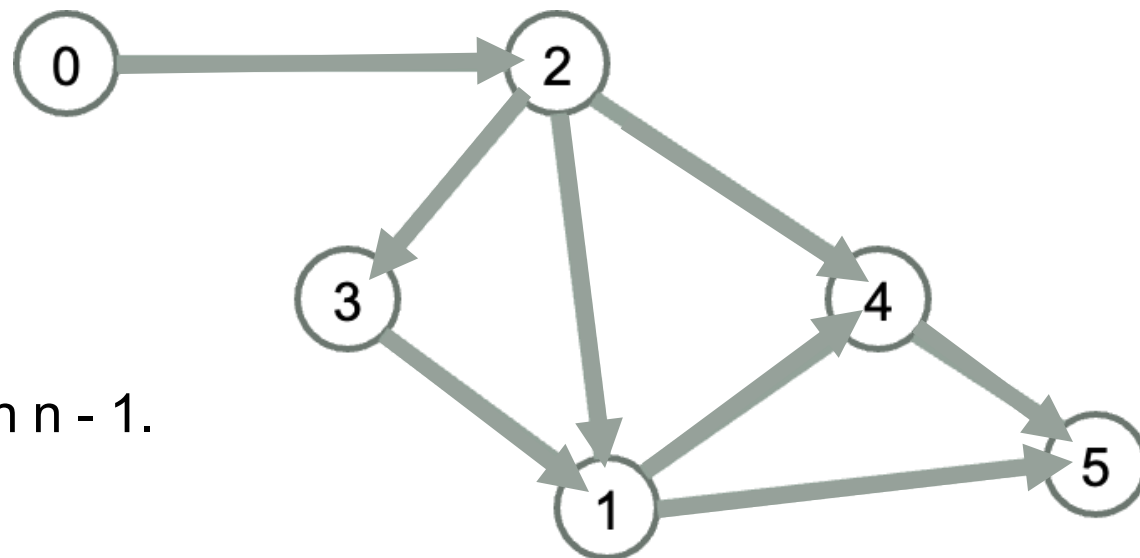
- A. Vector, why?
- B. Hashtable, why?
- C. BST, why?
- D. Something else?

**Consider:**

**1. What do you need to do with the vertices?**

(Look them up? Loop through them?)

**2. Does it help that the vertex labels are just  $0, 1, 2, \dots, n - 1$ ?**



**Convince your neighbor** why your pick is the best.

# Representing vertices with associated info

```
struct NodeInfo {  
    // The value of this Node before activation.  
    double preActivationValue;  
    // The value of this Node after activation.  
    double postActivationValue;  
    double bias; //bias  
    // other fields associated with a neuron  
};  
  
std::vector<NodeInfo> nodes(n); // where n is known ahead of time  
  
OR  
  
std::vector<NodeInfo*> nodes(n);
```

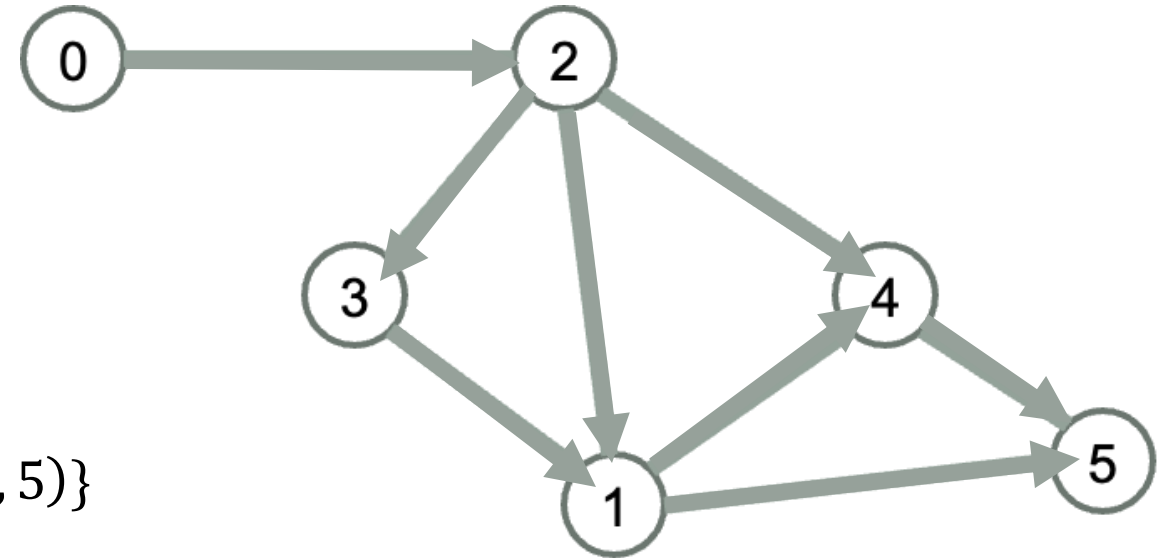
# Representing the graph using an adjacency list

An **adjacency list** is a way to represent a graph where each vertex stores a list of its neighbors (the ones its connected to by an edge).

$$G = (V, E)$$

$$V = \{0, 1, 2, 3, 4, 5\}$$

$$E = \{(0,2), (2,3), (2,1), (2,4), (3,1), (1,4), (1,5), (4,5)\}$$



Guiding question: Given a vertex ( $v$ ), how efficiently can we find all its neighbors?

**Activity 3:** Complete the adjacency list representation of the example graph.

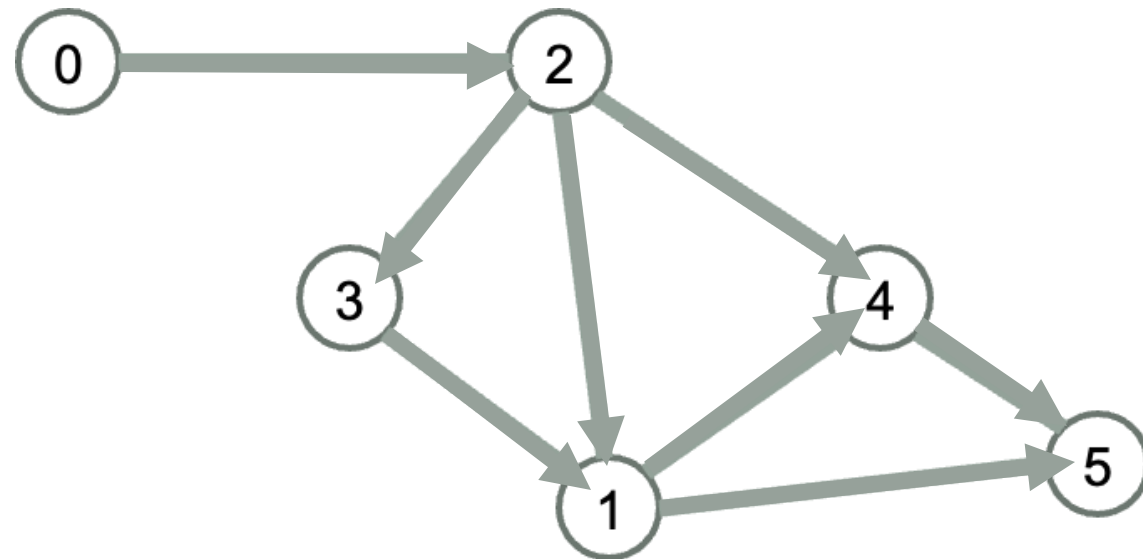

Does this look like any other data structure you have seen so far?

# Representing a graph using an adjacency list

```
class graph{  
    ...  
private:  
    vector<NodeInfo*> nodes;  
    _____ adjlist;  
};
```

Choose the ADT to represent adjlist

- A. `vector<int>`
- B. `vector<unordered_set<int>>`
- C. `list<vector<int>>`
- D. `vector<list<int>>`
- E. `set<list<int>>`

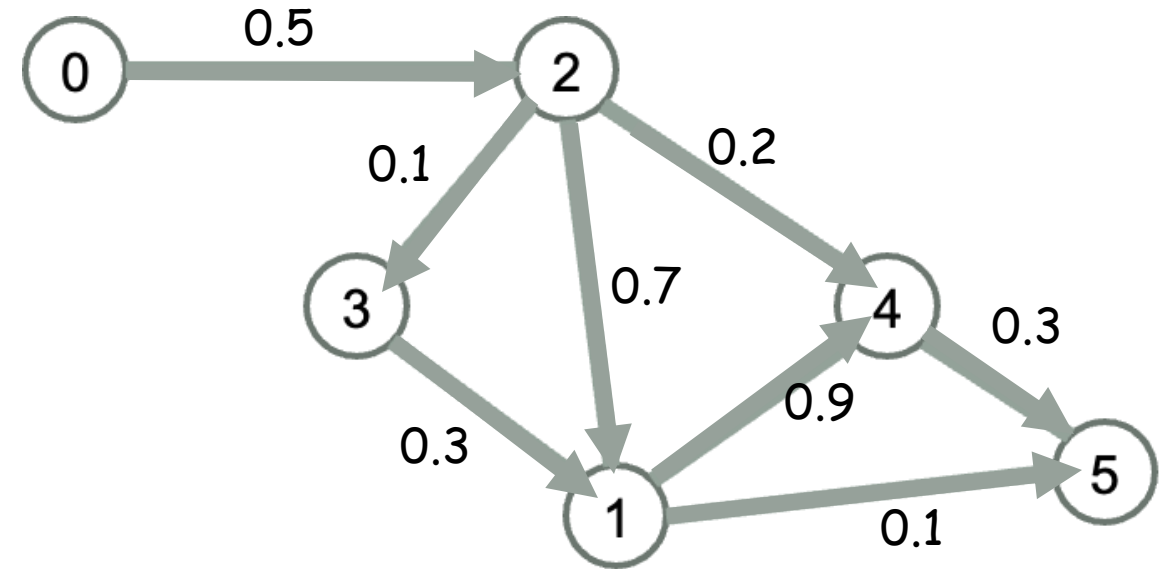


# What if edges had weights?

```
class graph{
  ...
private:
  vector<NodeInfo*> nodes;
  _____ adjlist;
};
```

How would you represent adjlist for a weighted graph?

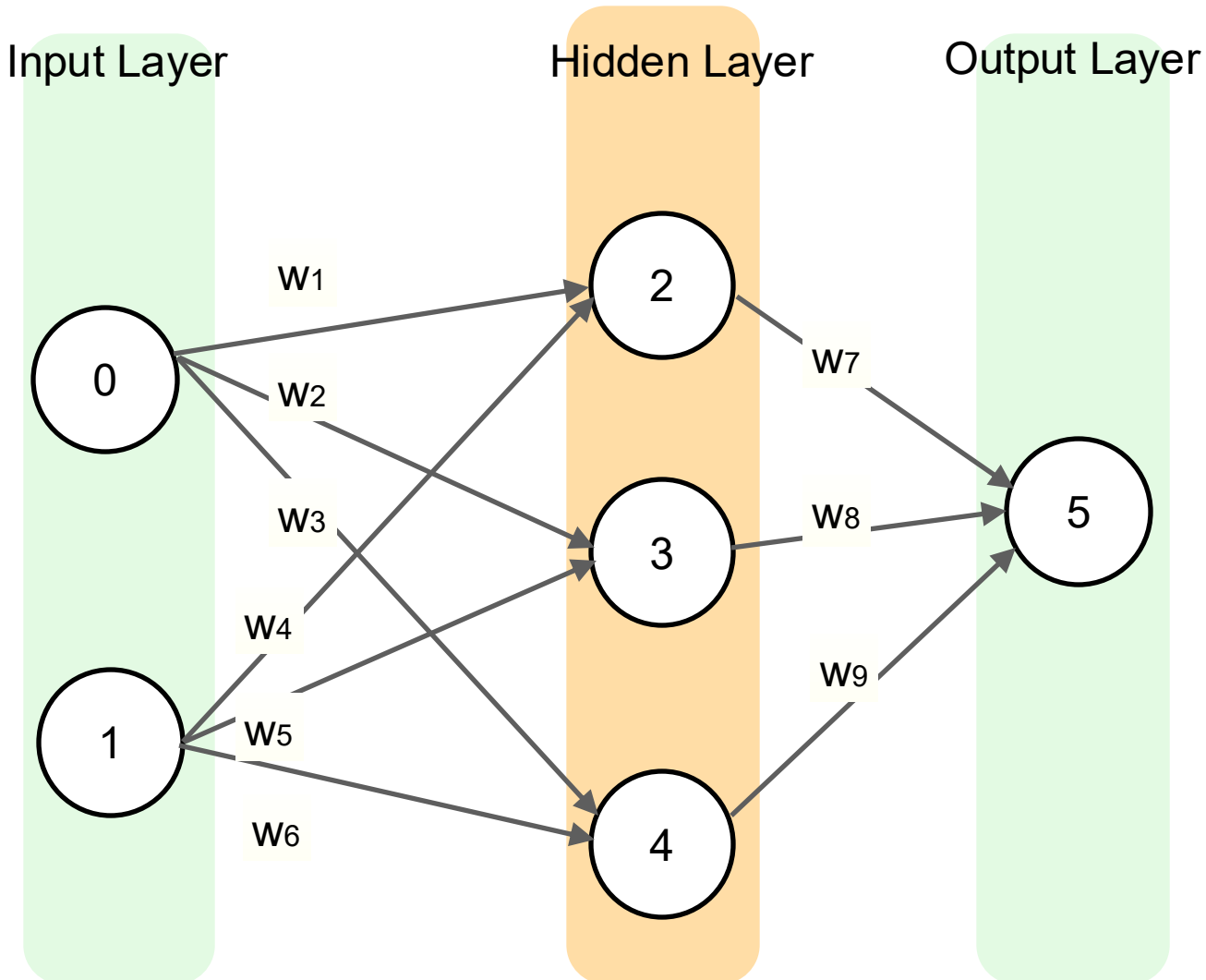
```
// Use when edges have no weights
vector<unordered_set<int>> adjlist;
```



adjlist


# Neural Network structure for upcoming assignment

```
typedef std::vector<std::unordered_map<int, Connection> > AdjList;
```



```
class Graph {  
    std::vector<NodeInfo*> nodes;  
    AdjList adjacencyList;  
};
```



# Understanding the Graph and NeuralNetwork classes

```
typedef std::vector<std::unordered_map<int, Connection> > AdjList;
```

```
class Graph {
public:
    Graph();
    Graph(int size);
    // Constructors and destructor

    // TODO: graph methods
    void updateNode(int id, NodeInfo n);
    NodeInfo* getNode(int id) const;
    void updateConnection(int v, int u, double w);

protected:
    // protected to give NeuralNetwork access

    // adjacency list containing weights for edges.
    AdjList adjacencyList;

    // vector storing node info
    std::vector<NodeInfo*> nodes;

    //Other functions
};
```

```
class NeuralNetwork : public Graph {
public:
    // Constructors and public functions

private:
    // each index of layers holds a vector which
    // contains the id's of every node in that layer.
    std::vector<std::vector<int> > layers;

    // contains ids of input nodes
    std::vector<int> inputNodeIds;

    // contains ids of output nodes
    std::vector<int> outputNodeIds;

    // since NeuralNetwork inherits from Graph, you can imagine all of the
    // graph members here as well...

};
```

```

void test_algorithm() {
    cout << "test_algorithm" << endl;
    NeuralNetwork nn(6);

    NodeInfo n0("ReLU", 0, -0.2);
    NodeInfo n1("ReLU", 0, 0.2);
    NodeInfo n2("identity", 0, 0);
    NodeInfo n3("sigmoid", 0, 0.98);
    NodeInfo n4("ReLU", 0, 0.11);
    NodeInfo n5("identity", 0, 0);

    nn.updateNode(0, n0);
    nn.updateNode(1, n1);
    nn.updateNode(2, n2);
    nn.updateNode(3, n3);
    nn.updateNode(4, n4);
    nn.updateNode(5, n5);

    nn.updateConnection(2, 1, 0.1);
    nn.updateConnection(2, 4, 0.2);
    nn.updateConnection(2, 0, 0.3);
    nn.updateConnection(5, 1, 0.4);
    nn.updateConnection(5, 4, 0.5);
    nn.updateConnection(5, 0, 0.6);
    nn.updateConnection(1, 3, 0.7);
    nn.updateConnection(4, 3, 0.8);
    nn.updateConnection(0, 3, 0.9);

    nn.setInputNodeIds({2, 5});
    nn.setOutputNodeIds({3});
}

```

## Activity 5: Draw the final neural net and its representation in memory

# Next lecture preclass activities

- Review pa03 tutorial: <https://ucsb-cs24.github.io/s25/pa/pa03-tutorial/>
- Finish watching the neural net intro video:  
<https://youtu.be/aircAruvnKk?feature=shared>
- Do the assigned reading: Breadth First Search and Depth First Search on graphs.