

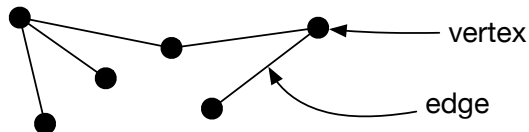
## Chapter 3

## Graph Theory

- 3.1 Foundations of Graph Theory
- 3.2 Graph Embeddings
- 3.3 Fundamental Graph Properties

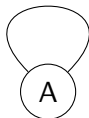
# Graphs, Vertices and Edges

- ▶ A **graph**  $G$  consists of a collection  $V$  of **vertices** and a collection  $E$  of **edges**, for which we write  $G = (V, E)$ .
  - ▶ Vertices and edges are also called
    - ▶ nodes and links (computer science)
    - ▶ sites and bonds (physics)
    - ▶ actors and ties (sociology)
- ▶ The set of vertices associated with graph  $G$  is denoted by  $V(G)$ .
- ▶ The set of edges associated with graph  $G$  is denoted by  $E(G)$ .
- ▶ Each edge  $e \in E$  is said to **join** two vertices, which are called its **end points**.
- ▶ If  $e$  joins  $u, v \in V$ , we write  $e = \langle u, v \rangle$ . Vertex  $u$  and  $v$  in this case are said to be **adjacent**. Edge  $e$  is said to be **incident** with vertices  $u$  and  $v$ , respectively.

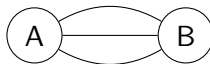


## Loops and Multiple Edges

- ▶ A **loop** (also called **self-edge**) is an edge that connects a vertex to itself.



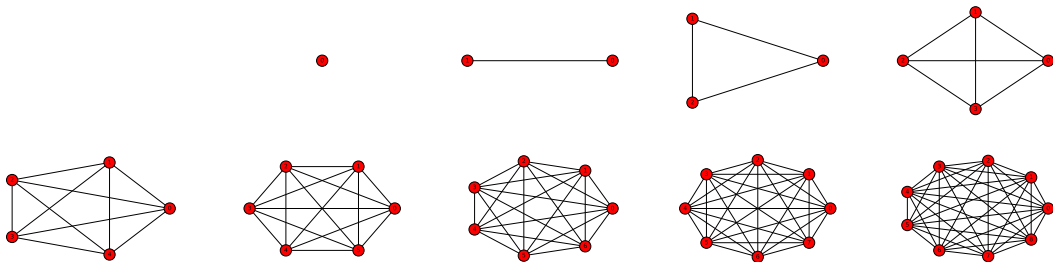
- ▶ **Multiple edges** (also called a **multi-edge**) are two or more edges that are incident to the same two vertices.



- ▶ A graph that does not have loops or multiple edges is called **simple**.

## Empty and complete Graphs

- ▶ A graph having no vertices and edges is called **empty**.
- ▶ A simple graph having  $n$  vertices, with each vertex being adjacent to every other vertex is called a **complete graph**. A complete graph with  $n$  vertices is commonly denoted as  $K_n$ .



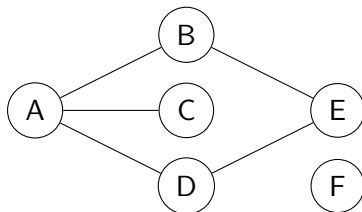
The graphs  $K_0, K_1, K_2, K_3, K_4, K_5, K_6, K_7, K_8$ , and  $K_9$ .

**Question:** how many edges in  $K_n$ ?



## Degree

- ▶ Let  $G = (V, E)$  be a simple graph. The number of edges incident with a vertex  $v \in V$  is called the **degree** of  $v$ , denoted as  $\delta(v)$ .
- ▶ The **maximal degree** of a vertex in graph  $G$  is denoted as  $\Delta(G)$ .
- ▶ Note: for any graph, the number of vertices with odd degree is even.



$$\Delta(G) = 3$$

$$\delta(A) = 3$$

$$\delta(E) = 2$$

$$\delta(F) = 0$$

# Mathematical representation of Graphs

There are a number of different ways to represent a graph.

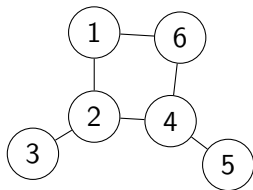
- ▶ Example: for a graph with  $n$  vertices and  $m$  edges we can label the vertices with integer labels  $1 \dots n$ . An edge between vertices  $i$  and  $j$  is denoted by  $(i, j)$ . Then the whole network can be specified by the value of  $n$  and a list of all edges. This is also called an **edge list**.
- ▶ Edge lists are sometimes used to store the network structure on computers, but for mathematical developments they are rather cumbersome.

A better representation is the **adjacency matrix**, which allows us to express calculations on the graph in terms of matrix operations.

## Adjacency matrix of a Graph

$$\mathbf{A}_{ij} := \begin{cases} 1 & \text{if there is an edge between vertices } i \text{ and } j \\ 0 & \text{otherwise} \end{cases}$$

**Example:** Let  $G = (V, E)$  be an undirected simple graph with  $V = \{1, 2, 3, 4, 5, 6\}$  and  $E = \{\langle 1, 2 \rangle, \langle 2, 3 \rangle, \langle 2, 4 \rangle, \langle 4, 5 \rangle, \langle 4, 6 \rangle, \langle 1, 6 \rangle\}$



The corresponding adjacency matrix<sup>1</sup>:

$$\mathbf{A} = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 \end{pmatrix}$$

Remember: the adjacency matrix of an undirected graph is symmetric.

<sup>1</sup> $\mathbf{A}_{ij}$  denotes the value in the  $i$ -th row,  $j$ -th column of the matrix  $\mathbf{A}$ .  $i$  is the row index of  $\mathbf{A}_{ij}$  and  $j$  is the column index of  $\mathbf{A}_{ij}$ . Remember:  $\mathbf{A}$  is a matrix, whereas  $\mathbf{A}_{ij}$  is a number.

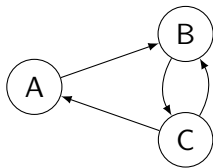
## A note on list and matrix representation

- ▶ For practical considerations, we will use a matrix notation for graphs throughout this course, because it allows to give concise and comprehensible definitions.
- ▶ However, when implementing a graph algorithm on a computer, it can be much more efficient to utilize list representations, particularly when the graph consists of many vertices but has comparatively few edges.



## Directed Graphs

- ▶ A **directed graph** (also called **digraph**)  $D$  consists of a collection vertices  $V$ , and a collection of **arcs**  $A$ , for which we write  $D = (V, A)$ .
- ▶ Each arc  $a = \langle u, \vec{v} \rangle$  is said to join vertex  $u \in V$  to another (not necessarily distinct) vertex  $v \in V$ .
- ▶ Vertex  $u$  is called the **tail** of  $a$ , whereas  $v$  is its **head**.



The degree is defined analogously for directed graphs:

- ▶ For a vertex  $v$  of digraph  $D$ , the number of arcs with head  $v$  is called the **in-degree**  $\delta_{in}(v)$  of  $v$ .
- ▶ Likewise, the **out-degree**  $\delta_{out}(v)$  is the number of arcs having  $v$  as their tail.

## Adjacency matrix of a directed graph

For a directed graph we define the adjacency matrix as follows

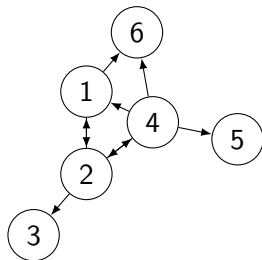
$$\mathbf{A}_{ij} := \begin{cases} 1 & \text{if there is an arc that joins } j \text{ to } i \\ 0 & \text{otherwise} \end{cases}$$

- Notice the direction of the arc here – it runs *from* the second index *to* the first.
- Remember:  $\mathbf{A}_{ij}$  denotes the value in the  $i$ -th row,  $j$ -th column of the matrix  $\mathbf{A}$ .  $i$  is the row index of  $\mathbf{A}_{ij}$  and  $j$  is the column index of  $\mathbf{A}_{ij}$ .  $\mathbf{A}$  is a matrix, whereas  $\mathbf{A}_{ij}$  is a number.

## Directed Graph Example

Let  $D = (V, A)$  be a directed simple graph with

- ▶  $V = \{1, 2, 3, 4, 5, 6\}$
- ▶  $A = \{\langle 1, 2 \rangle, \langle 1, 6 \rangle, \langle 2, 1 \rangle, \langle 2, 3 \rangle, \langle 2, 4 \rangle, \langle 4, 1 \rangle, \langle 4, 2 \rangle, \langle 4, 5 \rangle, \langle 4, 6 \rangle\}$





The corresponding adjacency matrix:


$$\mathbf{A} = \begin{pmatrix} 0 & \mathbf{1} & 0 & \mathbf{1} & 0 & 0 \\ \mathbf{1} & 0 & 0 & \mathbf{1} & 0 & 0 \\ 0 & \mathbf{1} & 0 & 0 & 0 & 0 \\ 0 & \mathbf{1} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \mathbf{1} & 0 & 0 \\ \mathbf{1} & 0 & 0 & \mathbf{1} & 0 & 0 \end{pmatrix}$$

In general the adjacency matrix of a directed graph is asymmetric.

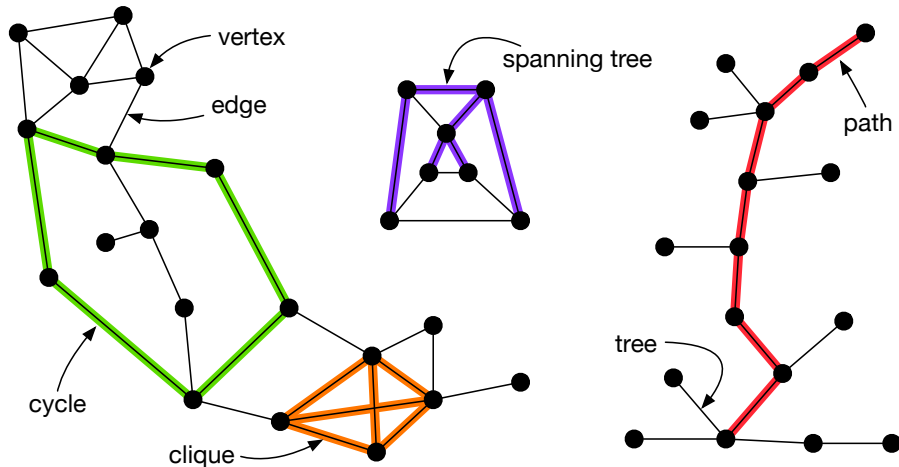
## Walks, Trails, Paths, and Cycles

- ▶ Consider a graph  $G$ . A  $(v_0, v_k)$ -**walk** in  $G$  is an alternating sequence   $[v_0, e_1, v_1, e_2, \dots, v_{k-1}, e_k, v_k]$  of vertices and edges from  $G$  with  $e_i = \langle v_{i-1}, v_i \rangle$ .
- ▶ In a **closed walk**,  $v_0 = v_k$ .
- ▶ A **trail** is a walk in which all edges are distinct.
- ▶ A **path** is a trail in which also all vertices are distinct .
- ▶ A **cycle** is a closed trail in which all vertices excepts  $v_0$  and  $v_k$  are distinct.
- ▶ A simple, connected graph having no cycles is called a **tree** (also called **acyclic graph**).
- ▶ A simple graph having only trees as its components, is called a **forest**.
- ▶ A **spanning tree** of a connected graph  $G$  is an acyclic connected subgraph of  $G$  containing all of  $G$ 's vertices.

## Connectedness and Components

- ▶ Two distinct vertices  $u$  and  $v$  in graph  $G$  are **connected** if there exists a  $(u, v)$ -path in  $G$ .
- ▶ A graph  $G$  is **connected** if all pairs of distinct vertices are connected.
- ▶ A graph  $H$  is a **subgraph** of  $G$  if  $V(H) \subseteq V(G)$  and  $E(H) \subseteq E(G)$  such that for all  $e \in E(H)$  with  $e = \langle u, v \rangle$ , we have that  $u, v \in V(H)$ .
- ▶ When  $H$  is a subgraph of  $G$ , we write  $H \subseteq G$ .
- ▶ A subgraph  $H$  of  $G$  is called a **component** of  $G$  if  $H$  is connected and not contained in a connected subgraph of  $G$  with more vertices or edges. 
- ▶ The **number of components** of  $G$  is denoted as  $\omega(G)$ .
- ▶ The adjacency matrix of a network with more than one component can be written in block diagonal form, meaning that the non-zero elements of the matrix are confined to square blocks along the diagonal of the matrix, with all other elements being zero.

# Graph Terminology



An undirected graph with 3 components. One of the components is a tree (right)<sup>1</sup>


<sup>1</sup>Depiction based on: R. Segewick, Algorithms in Java - Part 5 Graph Algorithms, Addison Wesley, 2004

## Vertex Eccentricity, Graph Radius and Diameter

- ▶ Let  $G$  be a directed or undirected graph and  $u, v \in V(G)$ :
- ▶ The **(geodesic) distance** between  $u$  and  $v$ , denoted as  $d(u, v)$ , is the length of a shortest  $(u, v)$ -path.
- ▶ The **eccentricity**  $\epsilon(u)$  of a vertex  $u$  in  $G$  is defined as  $\max\{d(u, v) \mid v \in V(G)\}$ .
- ▶ The **radius** of  $G$ , denoted as  $rad(G)$ , is equal to  $\min\{\epsilon(u) \mid u \in V(G)\}$ .
- ▶ The **diameter** of  $G$ , denoted as  $diam(G)$ , is the maximal shortest path between any two vertices:  $diam(G) := \max\{d(u, v) \mid u, v \in V(G)\}$ .
- ▶ The eccentricity of a vertex  $u$  tells us how far the farthest vertex from  $u$  is positioned in the network. The radius of a network, defined as the minimum over all eccentricity values, is an indication of how disparate the vertices in a network actually are. Finally, the diameter simply tells what the maximal distance in a network is.


## How to find shortest paths?

There exist various path-finding algorithms for graphs with  $n = |V|$  nodes and  $m = |E|$  edges:

- ▶ Find all shortest paths between all nodes
  - ▶ **Floyd's algorithm** solves the *all-pairs-shortest-path* problem in  $\mathbf{O}(n^3)$  
- ▶ Find all shortest paths from one node to all others
  - ▶ **Dijkstra's algorithm** solves the *single-source-shortest-path* problem in  $\mathbf{O}(m \times \log(n))$ .



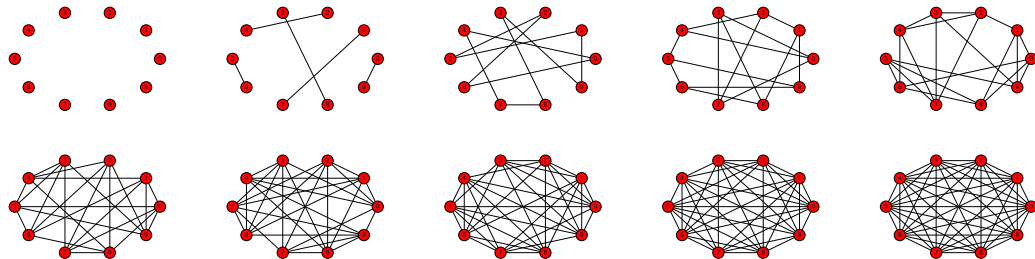
# Cliques and Clans

- ▶ Consider an undirected simple graph  $G$ . A **(maximal) clique** of  $G$  is a complete subgraph  $H$  of at least three vertices such that  $H$  is not contained in a larger complete subgraph of  $G$ . A clique with  $k$  vertices is called a  **$k$ -clique**. 
- ▶ Let  $G$  be an undirected simple graph. A  **$k$ -distance-clique** of  $G$  is a maximal subgraph  $H$  of  $G$  such that for all vertices  $u, v \in V(H)$ , the distance  $d_G(u, v) \leq k$ .
- ▶ Let  $G$  be an undirected simple graph. A  **$k$ -clan** of  $G$  is a  $k$ -distance-clique  $H$  of  $G$  such that for all vertices  $u, v \in V(H)$ , the distance  $d_H(u, v) \leq k$ .
- ▶ Let  $G$  be an undirected simple graph. A  **$k$ -core** of  $G$  is a maximal subgraph  $H$  of  $G$  such that for all vertices  $u \in V(H)$ , the degree  $\delta(u) \geq k$ .



# Regular Graphs

- ▶ If every vertex has the same degree, the graph is called **regular**.
- ▶ In a  **$k$ -regular** graph each vertex has degree  $k$ .
- ▶ As a special case, 3-regular graphs are also called **cubic graphs**.

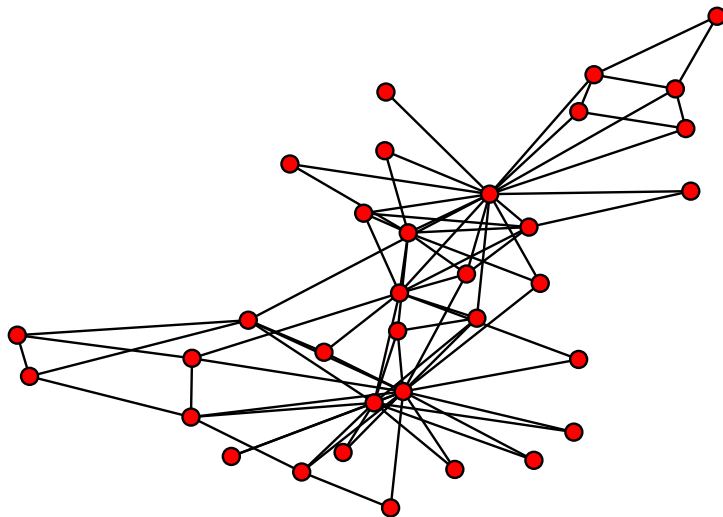


Regular graphs with  $|V| = 10$  nodes and  $k = \{0, 1, \dots, 9\}$ .

## Vertex Cuts and Edge Cuts

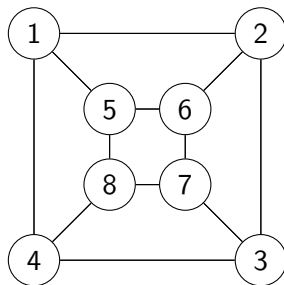
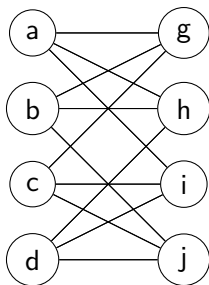
- ▶ For a graph  $G$  let  $V^* \subset V(G)$  and  $E^* \subset E(G)$ .  $V^*$  is called a **vertex cut** if  $\omega(G - V^*) > \omega(G)$ . If  $V^*$  consists of a single vertex  $v$ , then  $v$  is called a **cut vertex**. Likewise, if  $\omega(G - E^*) > \omega(G)$  then  $E^*$  is called an **edge cut**. If  $E^*$  consists of only a single edge  $e$ , then  $e$  is known as a **cut edge**.
- ▶ The size of a **minimal vertex cut** for graph  $G$  is denoted as  $\kappa(G)$ .
- ▶ The size of a **minimal edge cut** for graph  $G$  is denoted as  $\lambda(G)$ .
- ▶  $\kappa(G) \leq \lambda(G) \leq \min\{\delta(v) | v \in V(G)\}$
- ▶ A graph  $G$  for which  $\kappa(G) \geq k$  for some  $k$  is said to be **k-connected**. Likewise, graph  $G$  is **k-edge-connected** if  $\lambda(G) \geq k$ . Finally, a graph for which  $\kappa(G) = \lambda(G) = \min\{\delta(v) | v \in V(G)\}$  is said to be **optimally connected**.

## Discussion on vertex cuts and edge cuts



# Graph Isomorphism

- ▶ Consider two graphs  $G = (V, E)$  and  $G^* = (V^*, E^*)$ .
- ▶  $G$  and  $G^*$  are **isomorphic** if there exists a one-to-one mapping  $\phi : V \rightarrow V^*$  such that for every edge  $e \in E$  with  $e = \langle u, v \rangle$ , there is a unique edge  $e^* \in E^*$  with  $e^* = \langle \phi(u), \phi(v) \rangle$ .



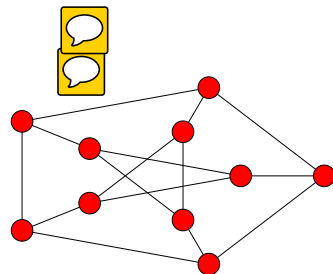
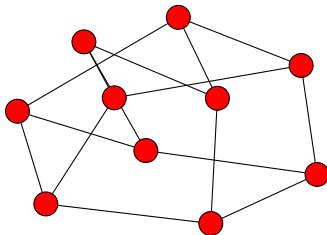
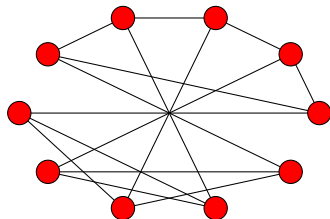
Two isomorphic graphs  $G_1 = (V_1, E_1)$  and  $G_2 = (V_2, E_2)$ . The one-to-one-mapping is given by:  $\phi(a) = 1, \phi(b) = 6, \phi(c) = 8, \phi(d) = 3, \phi(g) = 5, \phi(h) = 2, \phi(i) = 4, \phi(j) = 7$ .

## Graph Isomorphism II

- ▶ *Necessary* conditions for  $G$  and  $G^*$  to be isomorphic are that they have the same number of vertices and edges, and that they have the same degree sequence. However, none of these are *sufficient* conditions for a graph isomorphism.
- ▶ In general, graph isomorphism plays an important role in complexity theory. It is in  $NP$ , but it is not known to be in  $P$  and it is not known to be  $NP$ -complete.
- ▶ Put simply, for a given graph  $G$  it is computationally easy to create an isomorphic graph  $G^*$ . However, for two given graphs  $G$  and  $G^*$  it is computationally hard to reconstruct an isomorphic mapping function  $\phi$ .

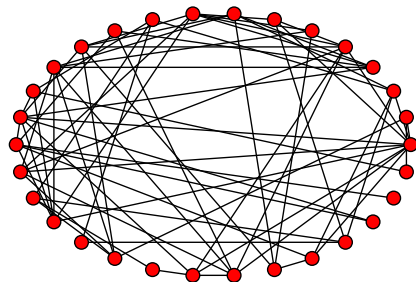
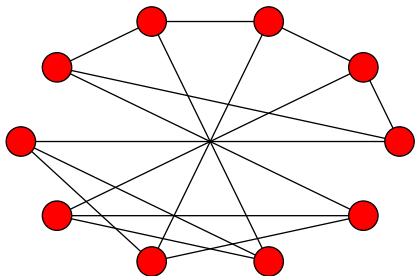
# Graph Embeddings

- ▶ A fundamental question in graph theory is **"how to draw a graph?"** and the visualization of a graph can be accomplished in various ways.
- ▶ Formally, we consider so-called **graph embeddings**. A graph embedding is a representation of a graph on a surface where every vertex is mapped to a particular position on the surface (three-dimensional embeddings are also possible).
- ▶ For example, consider the **Petersen graph**, which is a particular 3-regular graph. We can draw the Petersen graph in various ways and three embeddings are given below:



## Graph Embeddings: Circular Embedding

- ▶ The **circular embedding** is a widely used embedding, which places vertices at evenly spaced points on a surface:



- ▶ A circular embedding is particularly useful if one wants to *see all edges*. Because no three vertices ever lie on the same line, this allows every edge to stay (reasonably) visible, even if there is a large number of edges to draw.



## Spring Embedding

- The **spring embedding** was proposed by Eades in 1984. In a spring embedding, initially each vertex  $u$  is randomly positioned in a two-dimensional plane with coordinates  $(u_x, u_y)$ . Each vertex represents a ring and each edge  $e = \langle u, v \rangle$  represents a spring, which exerts an attracting force  $F_{att}(u, v)$  between the vertices  $u$  and  $v$  it joins as follows:

$$F_{att} \stackrel{def}{=} \begin{cases} 2\log(d(u, v)) & \text{if } u \text{ and } v \text{ are adjacent} \\ 0 & \text{otherwise} \end{cases}$$

where  $d(u, v)$  is the length of the spring between  $u$  and  $v$ . At the same time a repelling force  $F_{rep}(u, v)$  between nonadjacent vertices  $u$  and  $v$  is defined as

$$F_{rep} \stackrel{def}{=} \begin{cases} 0 & \text{if } u \text{ and } v \text{ are adjacent} \\ 1/\sqrt{d(u, v)} & \text{otherwise} \end{cases}$$

## Spring Embedding (algorithm)

1. Place all vertices randomly.
2. For each vertex  $u$  calculate the current forces in the  $x$  and  $y$  direction, respectively:

$$F_x(u) \stackrel{\text{def}}{=} \sum_{v \neq u} (F_{att,x}(u, v) - F_{rep,x}(u, v))$$

$$F_y(u) \stackrel{\text{def}}{=} \sum_{v \neq u} (F_{att,y}(u, v) - F_{rep,y}(u, v))$$

3. Reposition vertex  $u$  according to:

$$u_x \leftarrow u_x + 0.1 * F_x(u) \quad \text{and} \quad u_y \leftarrow u_y + 0.1 * F_y(u)$$

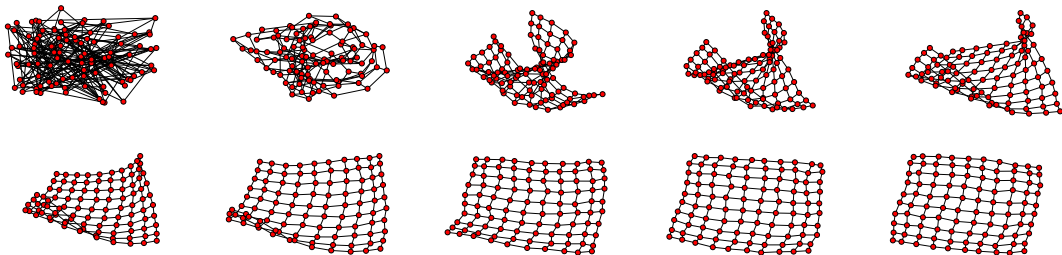
4. Goto step 2. Stop after  $M$  iterations.

$F_{att,x}(u, v)$  is the attracting force in the  $x$ -direction from neighboring vertex  $v$ :

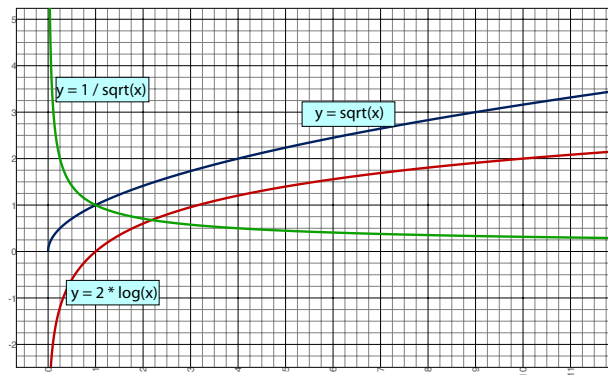
$$F_{att,x}(u, v) \stackrel{\text{def}}{=} F_{att}(u, v) * \frac{|v_x - u_x|}{d(u, v)}$$

## Spring Embedding: Example

- ▶ Input: a **2D grid graph** of  $10 * 10 = 100$  nodes.
- ▶ Starting from a random layout (top left), the nine other embeddings show the result of the spring embedding algorithm after increments of 10 iterations each.
- ▶ The last embedding (bottom right) is the result after 90 iterations.



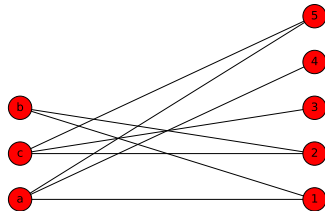
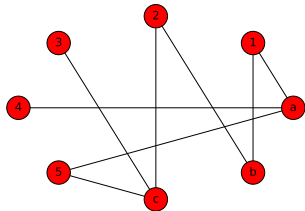
## Spring Embedding: a note on the dampening functions



Discussion: what could be alternative dampening functions for the attracting and repelling forces in the spring embedding algorithm?

## Embeddings for bipartite graphs

- ▶ A graph  $G$  is **bipartite** if  $V(G)$  can be partitioned into two disjoint subsets  $V_1$  and  $V_2$  such that each edge  $e \in E(G)$  has one end point in  $V_1$  and the other in  $V_2$ :  
 $E(G) \subseteq \{e = \langle u_1, u_2 \rangle \mid u_1 \in V_1, u_2 \in V_2\}$ .
- ▶ A common example for a bipartite graph is an **affiliation network**, for example, a graph of soccer players and clubs, where there is an edge between a player and a club if the player has played for that club.



Two embeddings of the same bipartite graph  $G$  with disjoint subsets  $V_1 = \{a, b, c\}$  and  $V_2 = \{1, 2, 3, 4, 5\}$ .

# Planar Graphs

- ▶ A **plane graph** is a specific embedding of a graph  $G$  such that **no two edges intersect**. If such an embedding exists,  $G$  is said to be **planar**.
- ▶ When considering a plane graph, we will observe a number of **regions** (also called **faces**), which are enclosed by the edges of the graph. Each region (except the exterior region) is enclosed by a cycle.
- ▶ (**Euler's formula**): For a plane graph  $G$  with  $n$  vertices,  $m$  edges, and  $r$  regions, we have that  $n - m + r = 2$ .

# Graph Coloring

- ▶ Consider a connected, loop-less graph  $G$ .  $G$  is  **$k$ -edge colorable** if there exists a partitioning of  $E(G)$  into  $k$  disjoint sets  $E_1, \dots, E_k$  such that no two edges from the same  $E_i$  are incident with the same vertex.
- ▶ Consider a simple connected graph  $G$ .  $G$  is  **$k$ -vertex colorable** if there exists a partitioning of  $V(G)$  into  $k$  disjoint sets  $V_1, \dots, V_k$  such that no two vertices from the same  $V_i$  are adjacent.
- ▶ The **vertex-coloring problem** for a given graph  $G$  is finding the minimal  $k$  for which  $G$  is  $k$ -vertex colorable. This minimal  $k$  is called the **chromatic number** of  $G$ , denoted as  $\chi(G)$ .
- ▶ For any simple connected graph  $G$ ,  $\chi(G) \leq \Delta(G) + 1$ .
- ▶ For any planar graph  $G$ ,  $\chi(G) \leq 4$ .

## Euler tour

- ▶ A **tour** of a graph  $G$  is a  $(u, v)$ -walk in which  $u = v$  (i.e., it is a **closed walk**) and that traverses each edge in  $G$ .
- ▶ An **Euler tour** is a tour in which all edges are traversed exactly once.
- ▶ A connected graph  $G$  (with more than one vertex) has an Euler tour if and only if it has no vertices of odd degree.
- ▶ Consider a weighted graph  $G$  in which each edge has a nonnegative weight. The **Chinese postman problem** is to find a closed walk  $W = [v_0, e_1, v_1, \dots, v_{n-1}, e_n, v_n]$  that covers all edges of  $G$ , but with minimal weight. In other words,  $E(W) = E(G)$  and  $\sum_{i=1}^n w(e_i)$  is minimal.



## Hamiltonian Paths and Cycles

- ▶ Consider a connected graph  $G$ . A **Hamiltonian path** of  $G$  is a path that contains every vertex of  $G$ . A Hamiltonian path is by definition self-avoiding. Likewise, a **Hamiltonian cycle** is a cycle that contains every vertex of  $G$ .
- ▶  $G$  is called **Hamiltonian** if it has a Hamiltonian cycle.
- ▶ Where Euler tours focus on traversing every edge in a graph, **Hamiltonian walks** deal with traversing every vertex in a graph. There is no known efficient procedure in general to determine whether a graph is Hamiltonian or not. The problem is also known as the **traveling salesman problem**.
- ▶ A **spanning tree** of a connected graph  $G$  is an acyclic connected subgraph of  $G$  containing all of  $G$ 's vertices. In a weighted connected graph  $G$ , a spanning tree  $T$  with minimal weight among all spanning trees of  $G$  can be found by **Kruskal's algorithm**.