

# Great Ideas in Computing

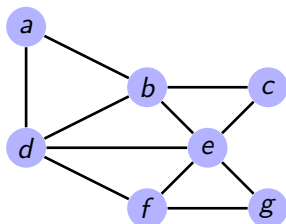
University of Toronto CSC196  
Fall 2021

Week 11: Appendix

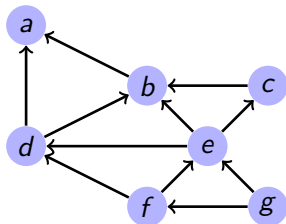
# Appendix: Network (graph) definitions and examples

Graphs come in two varieties

- 1 **undirected graphs** (“graph” usually means an undirected graph.)

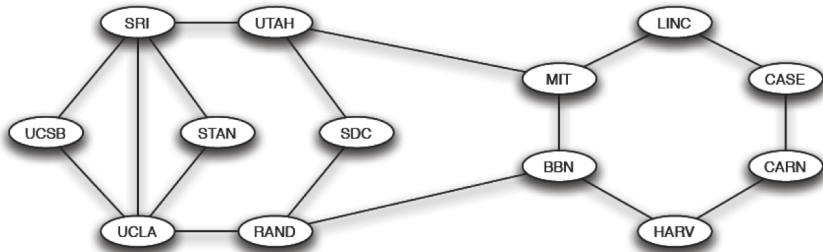


- 2 **directed graphs** (often called di-graphs).



# Visualizing Networks as Graphs

- **nodes**: entities (people, countries, companies, organizations, ...)
- **links** (may be **directed** or **weighted**): relationship between entities
  - ▶ friendship, classmates, did business together, viewed the same web pages, ...
  - ▶ membership in a club, class, political party, ...



**Figure:** Internet: Dec. 1970 [E&K, Ch.2]

## Adjacency matrix for graph induced by eastern sites ) in 1970 internet graph: another way to represent a graph

$$A(G) = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$

- This **node induced subgraph** (for the sites MIT = 1, LINC = 2, CASE = 3, CARN = 4, HARV = 5, BBN = 6) is a 6 node **regular graph** of **degree 2**. It is a **simple graph** in that there are no self-loops or multiple edges.
- Note that the adjacency matrix of an (undirected) simple graph is a symmetric matrix (i.e.  $A_{i,j} = A_{j,i}$ ) with  $\{0,1\}$  entries.
- To specify distances, we would need to give weights to the edges to represent the distances.

## The matrix $A^2$ where $A = A(G)$

Consider squaring the previous matrix  $A = A(G)$ . That is,  $A^2 = A * A$ .

$$A^2 = \begin{pmatrix} 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 0 & 1 \\ 0 & 1 & 0 & 1 & 0 & 0 \end{pmatrix}$$

Draw a visualization of the graph represented by  $A^2$ . If we let  $c_{i,j}$  be the  $i,j$  entry in  $A^2$ , can you describe the meaning of  $c_{i,j}$ ?

## The matrix $B = A + I$

Consider the  $6 \times 6$  identity matrix  $I = (\iota_{i,j})$ . That is,  $\iota_{i,i} = 1$  for  $1 \leq i \leq 6$  and  $\iota_{i,j} = 0$  for  $1 \leq i, j \leq 6$  and  $i \neq j$ .

Let  $B = A + I$  (as above). That is,  $b_{i,j} = a_{i,j} + \iota_{i,j}$  for all  $i, j$ . We have

$$B(G) = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 1 & 1 \end{pmatrix}$$

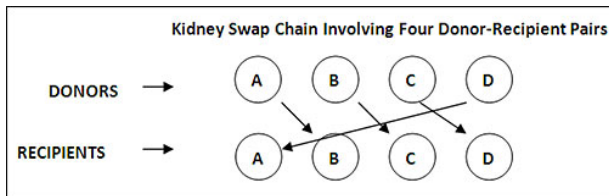
Note that now the matrix  $B$  has self loops and hence is not a simple graph.

## Kidney Exchange: Swap Cycles

- Live kidney donation common in N.A. to get around waiting list problems: donor-recipient pairs are nodes and links are directed.
- Exchange: supports willing pairs who are incompatible
  - ① allows multiway-exchange
  - ② supported by sophisticated algorithms to find matches

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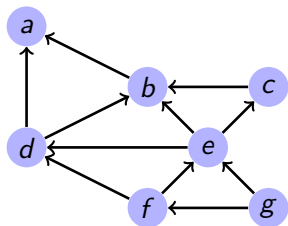
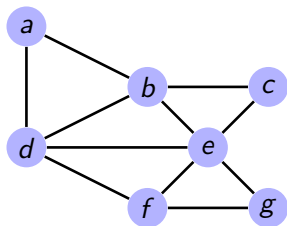
- Live kidney donation common in N.A. to get around waiting list problems: **donor-recipient pairs** are nodes and links are directed.
- Exchange: supports willing pairs who are incompatible
  - 1 allows multiway-exchange
  - 2 supported by sophisticated algorithms to find matches
- But what if someone reneges?  $\Rightarrow$  require **simultaneous transplantation!** Non-cyclic paths can be started by an altruistic donor!



**Figure:** Dartmouth-Hitchcock Medical Center, NH, 2010



## Recall: undirected graphs vs. directed graphs



## More definitions and terminology

- In order to refer to the nodes and edges of a graph, we define graph  $G = (V, E)$ , where
  - ▶  $V$  is the set of **nodes** (often called vertices)
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- **Undirected graph**: an edge  $(u, v)$  is an **unordered** pair of nodes.
  
- **Directed graph**: a directed edge  $(u, v)$  is an **ordered pair** of nodes  $\langle u, v \rangle$ .
  - ▶ However, we usually know when we have a directed graph and just write  $(u, v)$ .

## Basic definitions continued

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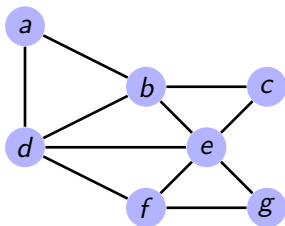
- First start with **undirected** graphs  $G = (V, E)$ .
- A **path** between two nodes, say  $u$  and  $v$  is a sequence of nodes, say  $u_1, u_2, \dots, u_k$ , where for every  $1 \leq i \leq k - 1$ ,
  - ▶ the pair  $(u_i, u_{i+1})$  is an edge in  $E$ ,
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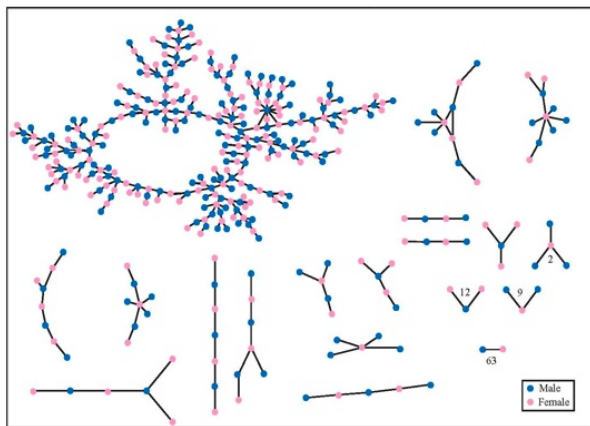
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- The **length** of a path is the number of edges on that path.
- A graph is a **connected** if there is a path between every pair of nodes. For example, the following graph is connected.





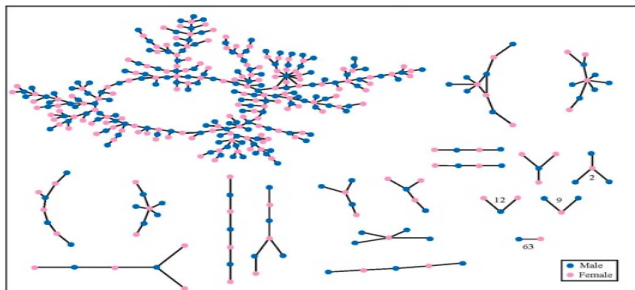
# Romantic Relationships [Bearman et al, 2004]



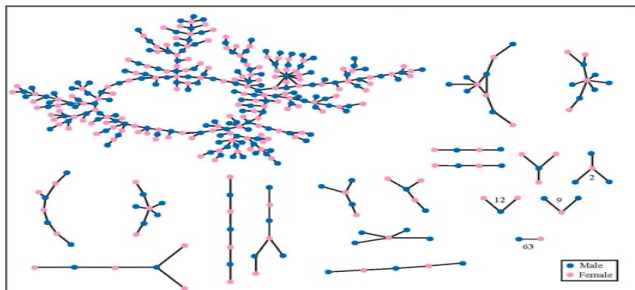
**Figure:** Dating network in US high school over 18 months.

- Illustrates common “structural” properties of many networks
- What predictions could you use this for?

## More basic definitions



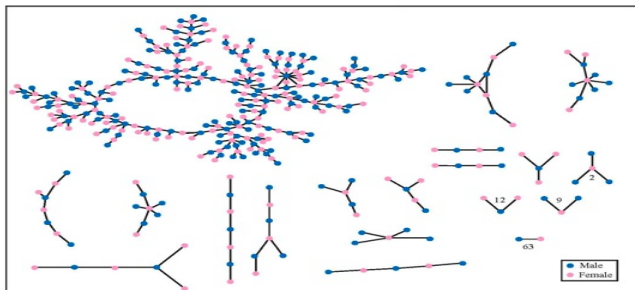
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### Observation

Many **connected components** including one “**giant component**”

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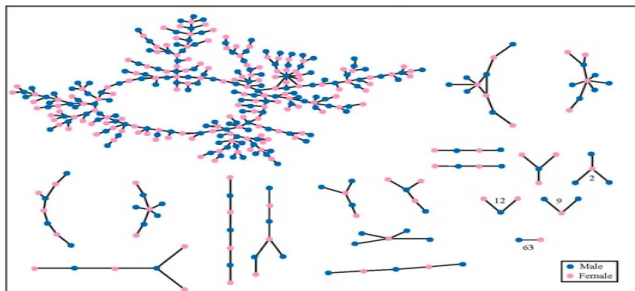
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- We will use this same graph to illustrate some other basic concepts.
- A **cycle** is path  $u_1, u_2, \dots, u_k$  such that  $u_1 = u_k$ ; that is, the path **starts and ends at the same node**.

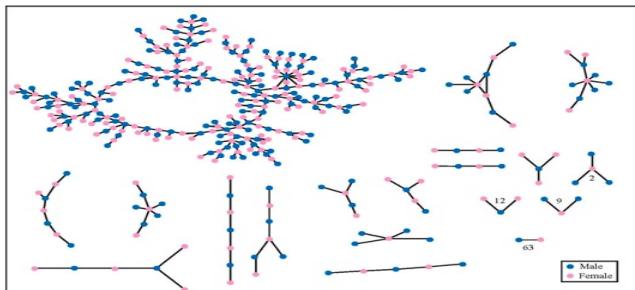
# Simple paths and simple cycles

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# Simple paths and simple cycles

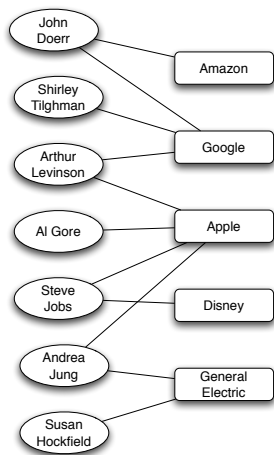
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## Observation

- There is one big simple cycle and (as far as I can see) three small simple cycles in the “giant component”.
- Only one other connected component has a **cycle**: a **triangle** having three nodes. Note: this graph is “almost” **bipartite** and “almost” **acyclic**.

## Example of an acyclic bipartite graph



**Figure:** [E&K, Fig 4.4] One type of affiliation network that has been widely studied is the memberships of people on corporate boards of directors. A very small portion of this network (as of mid-2009) is shown here.

# Florentine marriages and “centrality”

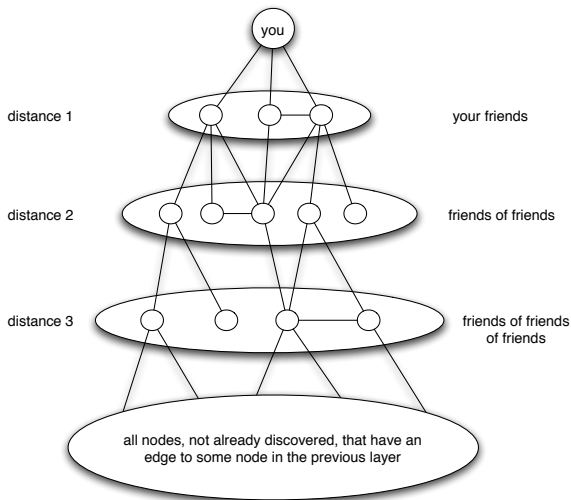
- Medici connected to more families, but not by much
- More importantly: lie between most pairs of families
  - ▶ **shortest paths** between two families: coordination, communication
  - ▶ Medici lie on 52% of all shortest paths; Guadagni 25%; Strozzi 10%



Figure: see [Jackson, Ch 1]



## Breadth first search and path lengths [E&K, Fig 2.8]



**Figure:** Breadth-first search discovers distances to nodes one “layer” at a time. Each layer is built of nodes adjacent to at least one node in the previous layer.

## Analogous concepts for directed graphs

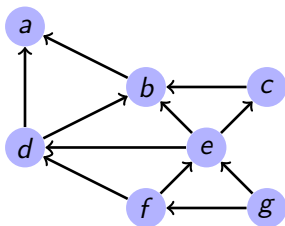
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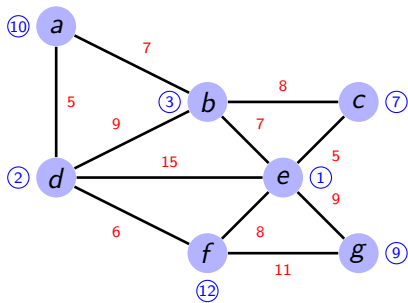
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  - ▶ However, it is usually clear from context if we are discussing undirected or directed graphs and in both cases most people just write  $(u, v)$ .
- We now have **directed paths** and **directed cycles**. Instead of connected components, we have **strongly connected components**.



# Weighted graphs

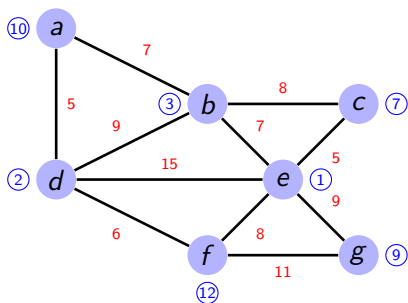
- We will often consider **weighted graphs**. Lets consider a (directed or undirected) graph  $G = (V, E)$ . Example:



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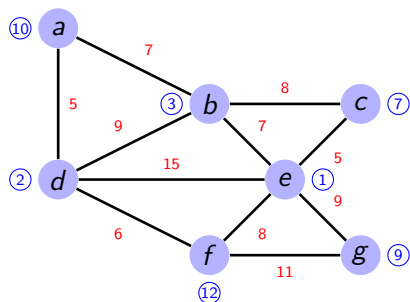


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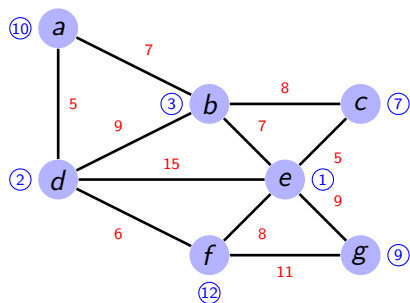


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- For example, in a social network whose nodes represent people, the weight  $w(v)$  of node  $v$  might indicate the importance of this person.
- The weight  $w(e)$  of edge  $e$  might reflect the strength of a friendship.



## Edge weighted graphs

- When considering **edge weighted** graphs, we often have edge weights  $w(e) = w(u, v)$  which are non negative (with  $w(e) = 0$  or  $w(e) = \infty$  meaning no edge depending on the context).
- In some cases, weights can be either positive or negative. A **positive** (resp. **negative**) weight reflects the **intensity** of connection (resp. **repulsion**) between two nodes (with  $w(e) = 0$  being a neutral relation).
- Sometimes (as in Chapter 3) we will only have a **qualitative** (rather than quantitative) weight, to reflect a strong or weak relation (tie).
- Analogous to shortest paths in an **unweighted** graph, we often wish to compute **least cost paths**, where the cost of a path is the sum of weights of edges in the path.