

# Algorithms and Data Structures

**Graphs: Introduction** 

**Ulf Leser** 

#### Content of this Lecture

- Graphs
- Definitions
- Representing Graphs
- Traversing Graphs
- Connected Components

## Graphs

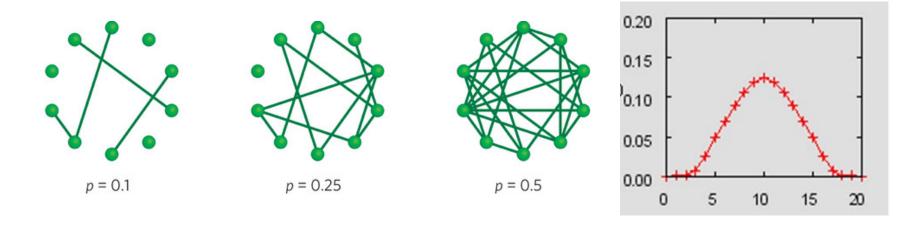
- There are objects and there are relations between objects
- Directed trees can represent hierarchical relations
  - Relations that are asymmetric, cycle-free, binary
  - Examples: parent\_of, subclass\_of, smaller\_than, ...
- Undirected trees can represent cycle-free, binary relations
- This excludes many (cyclic) real-life relations
  - friend\_of, similar\_to, reachable\_by, html\_linked\_to, ...
- (Classical) Graphs can represent all binary relationships
- N-ary relationships: Hypergraphs
  - exam(student, professor, subject), borrow(student, book, library)

## Types of Graphs

- Most graphs you will see are binary
- Most graphs you will see are simple
  - Simple graphs: At most one edge between any two nodes
  - Extension: multigraphs
- Some graphs you will see are undirected, some directed
- This lecture: Binary, simple (finite) graphs

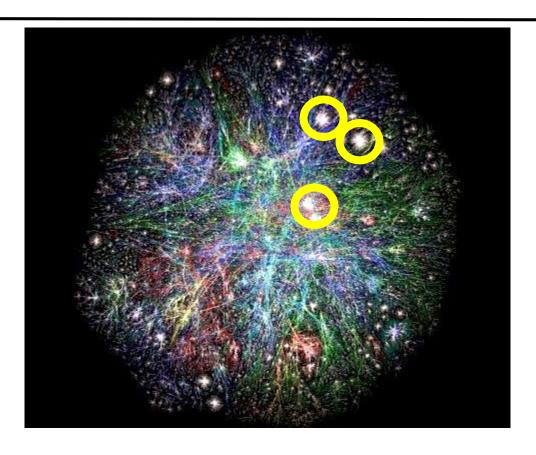
### **Exemplary Graphs**

- Classical theoretical model: Random Graphs
  - Create every possible edge with a fixed probability p



 In a random graph, the degree of every node has expected value p\*n, and the degree distribution follows a Poisson distribution

## Web Graph



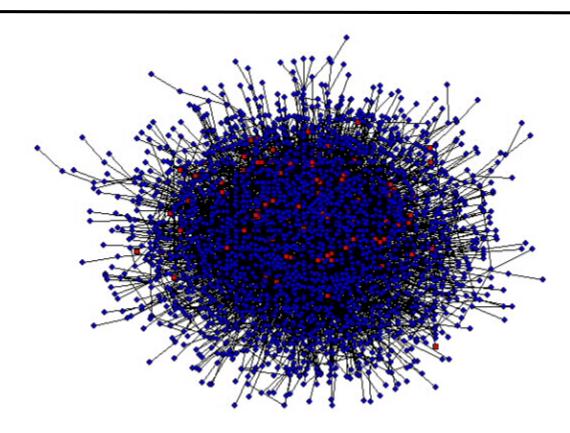
Note the strong local clustering

This is not a random graph

Graph layout is difficult

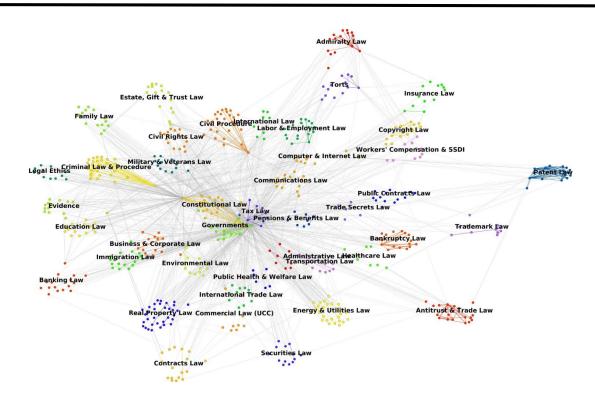
[http://img.webme.com/pic/c/chegga-hp/opte\_org.jpg]

#### Human Protein-Protein-Interaction Network



- Still terribly incomplete
- Proteins that are close in the graph likely share function [http://www.estradalab.org/research/index.html]

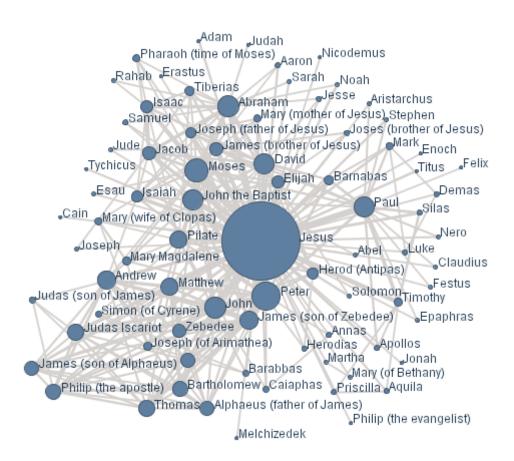
#### Word Co-Occurrence



- Words that are close have similar meaning
  - Close: Appear in the same contexts
- Words cluster into topics

[http://www.michaelbommarito.com/blog/]

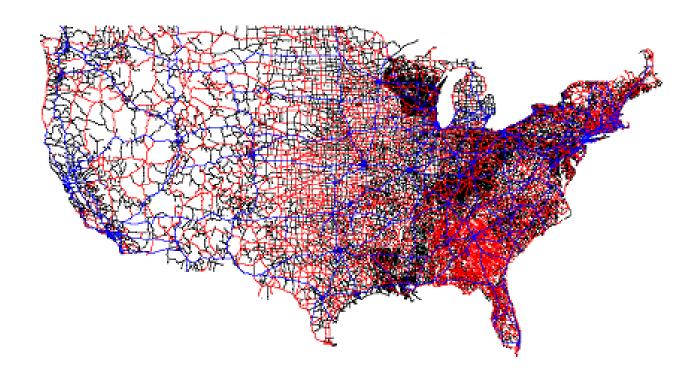
#### Social Networks



Six degrees of separation

[http://tugll.tugraz.at/94426/files/-1/2461/2007.01.nt.social.network.png]

### Road Network



## Specific property: Planar graphs

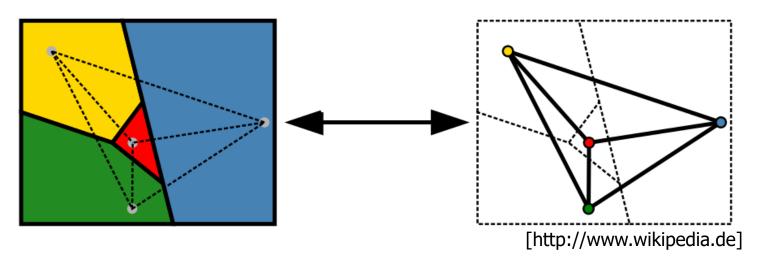
[Sanders, P. &Schultes, D. (2005). Highway Hierarchies Hasten Exact Shortest Path Queries. In *13th European Symposium on Algorithms (ESA)*, *568-579.*]

## More Examples

Graphs are also a wonderful abstraction

## Coloring Problem

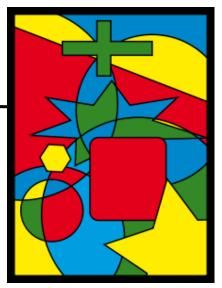
 How many colors do one need to color a map such that never two colors meet at a border?

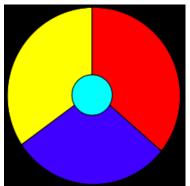


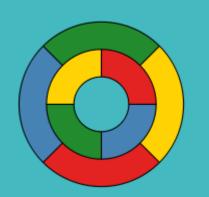
- Chromatic number: Number of colors sufficient to color a graph such that no adjacent nodes have the same color
- Every planar graph has chromatic number of at most 4

### **History** [Wikipedia.de]

- This is not simple to proof
- It is easy to see that one sometimes needs at least four colors
- It is easy to show that one may need arbitrary many colors for general graphs
- First conjecture which until today was proven only by computers
  - Falls into many, many subcases try all of them with a program

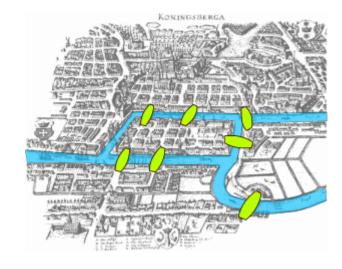






## Königsberger Brückenproblem

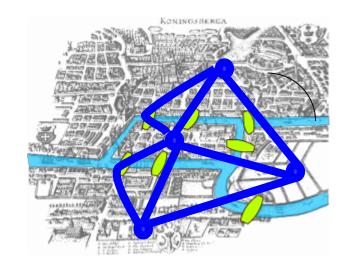
- Given a city with rivers and bridges: Is there a cycle-free path crossing every bridge exactly once?
  - Euler-Path



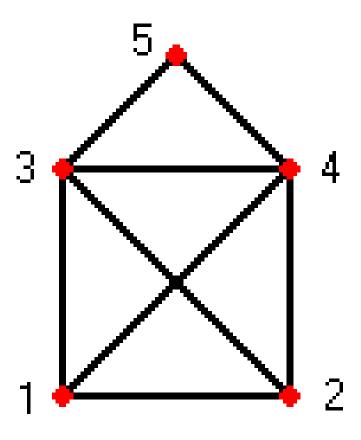
Source: Wikipedia.de

## Königsberger Brückenproblem

- Given a city with rivers and bridges: Is there a cycle-free path crossing every bridge exactly once?
  - A graph has an Euler-Path iff at contains 0 or 2 nodes with odd degree
- Hamiltonian path
  - ... visits each vertex exactly once
  - NP complete



## Recall?



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#### **Recall from Trees**

#### Definition

A graph G=(V, E) consists of a set of vertices (nodes) V and a set of edges  $(E\subseteq VxV)$ .

- A sequence of edges  $e_1$ ,  $e_2$ , ...,  $e_n$  is called a path iff  $\forall 1 \le i < n$ :  $e_i = (v', v)$  and  $e_{i+1} = (v, v)$ ; the length of this path is n
- A path  $(v_1, v_2)$ ,  $(v_2, v_3)$ , ...,  $(v_{n-1}, v_n)$  is acyclic iff all  $v_i$  are different
- G is acyclic, if no path in G contains a cycle; otherwise it is cyclic
- A graph is connected if every pair of vertices is connected by at least one path
- G is called undirected, if ∀(v,v')∈E ⇒(v',v)∈E. Otherwise it is called directed.

#### More Definitions

#### Definition

Let G=(V, E) be a directed graph. Let  $v \in V$ 

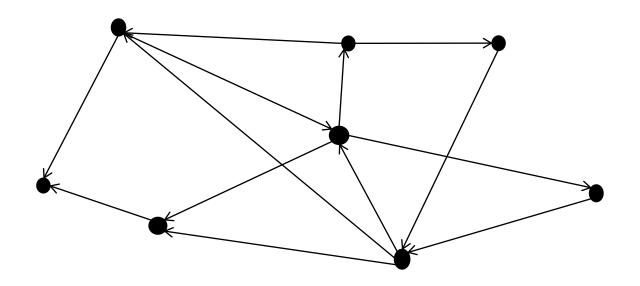
- The outdegree out(v) is the number of edges with v as start point
- The indegree in(v) is the number of edges with v as end point
- G is edge-labeled, if there is a function w:E→L that assigns an element of a set of labels L to every edge
- If L are numbers (real, int, ...), G is called weighted

#### Remarks

- Labels / weights max be assigned to edges or nodes (or both)
- Indegree and outdegree are identical for undirected graphs and called degree (number of neighbors)

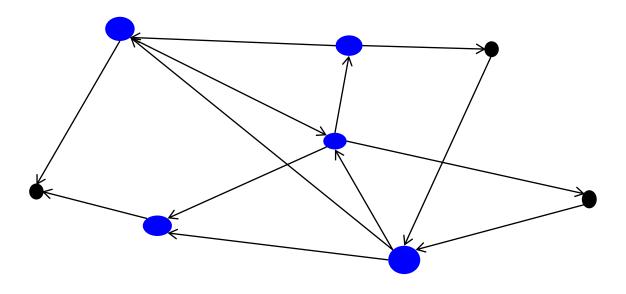
#### Some More Definitions

- Definition. Let G=(V, E) be a directed graph.
  - Any G'=(V', E') is called a subgraph of G, if  $V'\subseteq V$  and  $E'\subseteq E$  and  $\forall (v_1, v_2) \in E'$ :  $v_1, v_2 \in V'$
  - For any V'⊆V, the graph (V', E∩(V'×V')) is called the induced subgraph of G (induced by V')



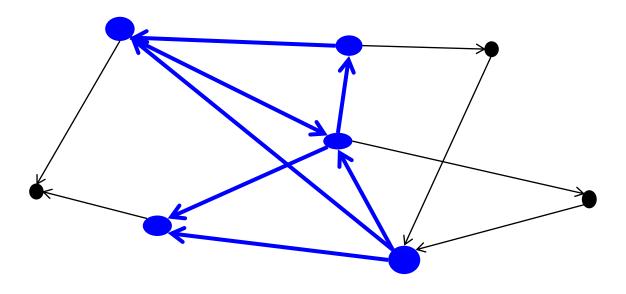
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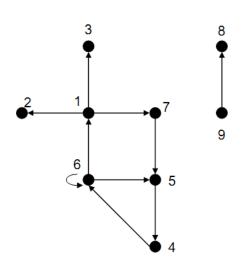
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#### **Data Structures**

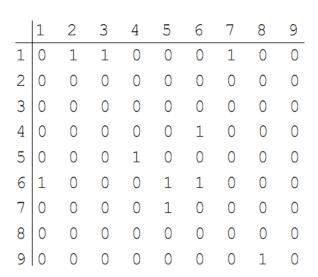
- From an abstract point of view, a graph is a list of nodes and a list of (weighted, directed) edges
- Two fundamental implementations
  - Adjacency matrix
  - Adjacency lists
- As usual, the representation determines the complexity of primitive operations
  - E.g. find node, find edge, find neighbors, ...
- Suitability depends on the specific problem under study and the nature of the graphs
  - Shortest paths, transitive hull, cliques, spanning trees, ...
  - Random, sparse/dense, scale-free, planar, ...

## Example [OW93]

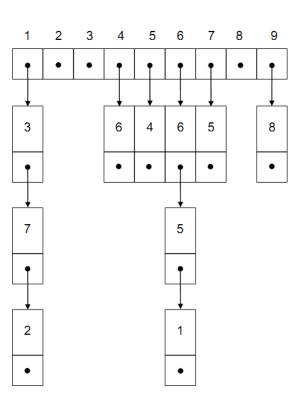
### Graph



#### Adjacency Matrix



### Adjacency List



### Adjacency Matrix

#### Definition

Let G=(V, E) be a simple graph. The adjacency matrix  $M_G$  for G is a two-dimensional matrix of size  $|V|^*|V|$ , where M[i,j]=1 iff  $(v_i,v_i)\in E$ 

#### Remarks

- Allows to test existence of a given edge in O(1)
- Requires O(|V|) to obtain all incoming (outgoing) edges of a node
- For large graphs, M is too large to be of practical use
- If G is sparse (much less edges than  $|V|^2$ ), M wastes a lot of space
- If G is dense, M is a very compact representation (1 bit / edge)
- In labeled graphs, M[i,j] contains the label
- Since M must be initialized with zero's, without further tricks all algorithms working on adjacency matrices are in  $\Omega(|V|^2)$

### Adjacency List

- Definition
   Let G=(V, E). The adjacency list L<sub>G</sub> for G is a list of all nodes v<sub>i</sub> of G. The entry representing v<sub>i</sub>∈V is a list of all edges outgoing (or incoming or both) from v<sub>i</sub>.
- Remarks (assume a fixed node v)
  - Let k be the maximal outdegree of G. Then, accessing an edge outgoing from v is O(log(k)) (if list is sorted; or use hashing)
  - Obtaining a list of all outgoing edges from v is in O(k)
    - If only outgoing edges are stored, obtaining a list of all incoming edges is O(|V|\*log(|E|)) – we need to search all lists
    - Therefore, usually outgoing and incoming edges are stored, which doubles space consumption
  - If G is sparse, L is a compact representation
  - If G is dense, L is wasteful (many pointers, many IDs)

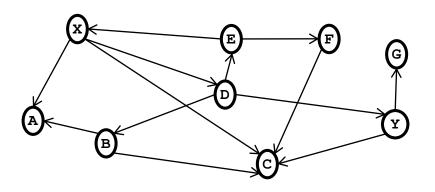
## Comparison

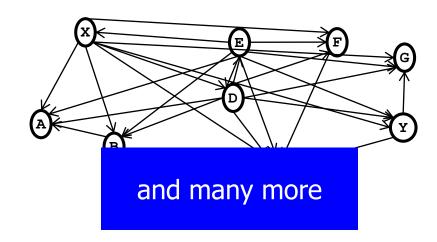
	Matrix	Lists
Test if a given edge exists	O(1)	O(log(k))
Find all outgoing edges of a given v	O(n)	O(k)
Space of G	O(n <sup>2</sup> )	O(n+m)

- With n=|V|, m=|E|
- We assume a node-indexed array
  - L is an array and nodes are uniquely numbered
  - We find the list for node v in O(1)
  - Otherwise, L has additional costs for finding v

#### **Transitive Closure**

- Definition
   Let G=(V,E) be a digraph and v<sub>i</sub>, v<sub>j</sub>∈V. The transitive
   closure of G is a graph G'=(V, E') where (v<sub>i</sub>, v<sub>j</sub>)∈E' iff G
   contains a path from v<sub>i</sub> to v<sub>j</sub>.
- TC usually is dense and represented as adjacency matrix
- Compact encoding of reachability information





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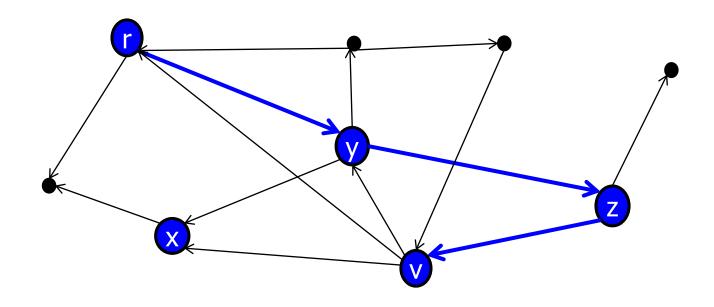
## **Graph Traversal**

- One thing we often do with graphs is traversal
- "Traversal" means: Visit every node exactly once in a sequence determined by the graph's topology
  - Not necessarily on one consecutive path (Hamiltonian path)
- Two popular orders
  - Depth-first: Using a stack
  - Breadth-first: Using a queue
  - The scheme is identical to that in tree traversal
- Difference
  - We have to take care of cycles
  - No root where should we start?

## **Breaking Cycles**

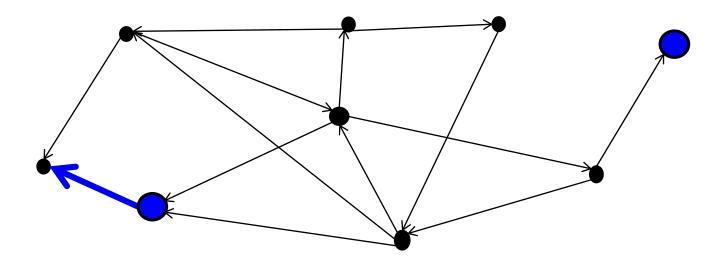
- Any naïve traversal will visit nodes more than once
  - If there is at least one node with more than one incoming edge
- Any naïve traversal will run into infinite loops
  - If the graphs contains at least one cycle (is cyclic)
- Breaking cycles / avoiding multiple visits
  - Assume we started the traversal at a node r
  - During traversal, we keep a list S of already visited nodes
  - Assume we are in v and aim to proceed to v' using e=(v, v')∈E
  - If v'∈S, v' was visited before and we are about to run into a cycle or visit v' twice
  - In this case, e is ignored

### Example



- Started at r and went S={r, y, z, v}
- Testing (v,y): y∈S, drop
- Testing (v, r): r∈S, drop
- Testing (v, x): x∉S, proceed

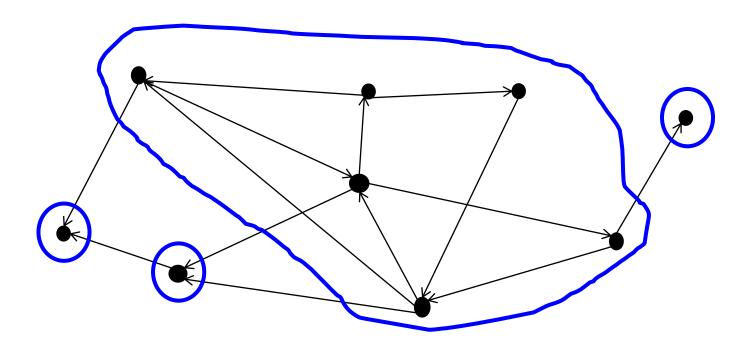
### Where do we Start?



#### Where do we Start?

- Definition
   Let G=(V, E). Let V'⊆V and G' be the subgraph of G
   induced by V'
  - G' is called connected if it contains a path between any pair v, v'∈V'
  - G' is called maximally connected, if no subgraph induced by a superset of V' is connected
  - If G is undirected, any maximal connected subgraph of G is called a connected component of G
  - If G is directed, any maximal connected subgraph of G is called a strongly connected component of G

# Example



#### Where do we Start?

- If a undirected graph falls into several connected components, we cannot reach all nodes by a single traversal, no matter which node we use as start point
- If a digraph falls into several strongly connected components, we might not reach all nodes by a single traversal
- Remedy: If the traversal gets stuck, we restart at unseen nodes until all nodes have been traversed

### Depth-First Traversal on Directed Graphs

```
func void DFS (G=(V,E)) {
  U := V;  # Unseen nodes
  while U≠Ø do
    v := getNextUnseen( U);
    traverse( G, v, U);
  end while;
}
```

Called once for every connected component

```
func void traverse (G, v node,
                      U set) {
  t := new Stack();
  t.put(v);
  U := U \setminus \{v\};
  while not t.isEmpty() do
    n := t.pop();
    print n;
    c := n.outgoingNodes();
    foreach x in c do
      if xEU then
        U := U \setminus \{x\};
        t.push(x);
      end if;
    end for;
  end while;
```

### **Analysis**

- We put every node exactly once on the stack
  - Once visited, never visited again
- We look at every edge exactly once
  - Outgoing edges of a visited node are never considered again
- U can be implemented as bitarray of size |V|, allowing O(1) operations
  - Add, remove, getNextUnseen
- Altogether: O(n+m)

```
func void traverse (G, v node,
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    foreach x in c do
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         t.push(x);
      end if;
    end for:
  end while;
```

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## In Undirected Graphs

- In an undirected graph, whenever there is a path from r to v and from v to v', then there is also a path from v' to r
  - Simply go the path  $r \rightarrow v \rightarrow v'$  backwards
- Thus, DFS (and BFS) traversal can be used to find all connected components of a undirected graph G
  - Whenever you call traverse(v), create a new component
  - All nodes visited during one call of traverse(v) form one connected component
- Obviously in O(n+m)

## In Digraphs

- The problem is considerably more complicated for digraphs
  - Previous conjecture does not hold
- Still: Tarjan's or Kosaraju's algorithm find all strongly connected components in O(n + m)
  - See next lecture

## Possible Examination Questions

- Let G be an undirected graph and S,T be two connected components of G. Proof that S and T must be disjoint, i.e., cannot share a node.
- Let G be an undirected graph with n vertices and m edges, m<=n². What is the minimal and what is the maximal number of connected components G can have?
- Let G be a positively edge-weighted digraph G. Design an algorithm which finds the longest acyclic path in G. Analyze the complexity of your algorithm.
- An Euler path through an undirected graph G is a cyclefree path from any start to any end node that hits every node of G (exactly once). Give an algorithm which tests for an input graph G whether it contains an Euler path.