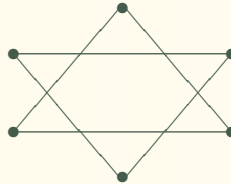
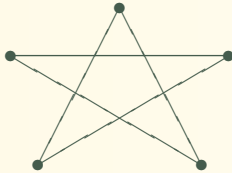


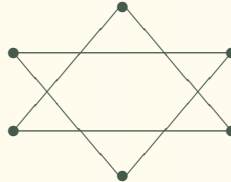
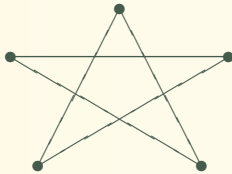
2 Connectivity in Graphs

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Brief outline of this lecture

- Walks in a graph, the definition, connected components.
- Exploring a graph, search algorithms – BFS and DFS.
- Higher levels of connectivity, 2-connected graphs, Menger’s theorem.
- Connectivity in directed graphs, strong components.
- Eulerian tours and trails, “Seven Bridges” and the even degree cond.

2.1 Graph Connectivity and Components

Definition: A *walk* of length n in a graph G is a sequence of alternating vertices and edges

$$(v_0, e_1, v_1, e_2, v_2, \dots, e_n, v_n),$$

such that every its edge e_i has the ends v_{i-1}, v_i .

Such a sequence really is a “*walk*” through the graph, see for instance how an IP packet is routed through the internet – it often repeats vertices.

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Lemma 2.1. *Let \sim be a binary relation on the vertex set $V(G)$ of a graph G , such that $u \sim v$ if, and only if, there exists a walk in G starting in u and ending in v . Then \sim is an *equivalence* relation.*

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Proof. The relation \sim is *reflexive* since every vertex itself forms a walk of length 0. It is also *symmetric* since any undirected walk can be easily “reversed”, and *transitive* since two walks can be concatenated at the common endvertex. \square

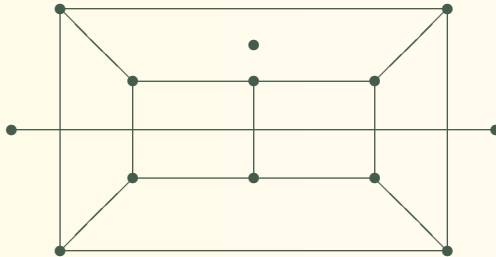
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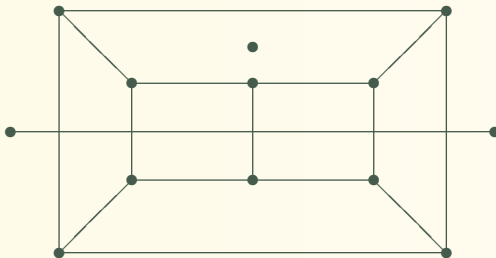
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Can you see all the *three* components in this picture?

Recall a *path* in a graph — it is a walk without repetition of vertices.

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- Assume we have built a starting fragment $(w_0, e_1, w_1, \dots, w_i) = W$ (inductively from $i = 0$, i.e. w_0) where $w_i = v_j$ for some $j \in \{0, 1, \dots, n\}$.

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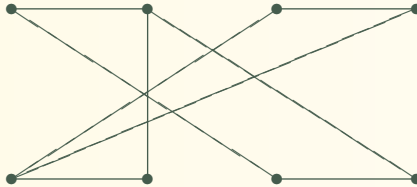
Proof; a shorter, but *nonconstructive* alternative.

Among all the walks between u and v in G , we choose the (one of) *shortest one* as W . It is clear that if the same vertex repeated in W , then W could be shortened further, a contradiction. Hence W is a path in G .

□

Basic connectivity

Definition 2.3. **Graph G is connected** if G consists of at most one connected component. By Theorem 2.2, this means if every two vertices of G are connected by a **path**.



2.2 Exploring (Searching) a Graph

For an illustration, we present a very general scheme of **searching through a graph**. This meta-algorithm works with the following data states and structures:

- **A vertex**: having one of the states ...
 - *initial* – assigned at the beginning,
 - *discovered* – after we have find it along an edge,
 - *finished* – assigned after exploring all incident edges.
 - (Can also be post-processed, after finishing all its descendants.)
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 - *initial* – assigned at the beginning,
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- **Stack (depository):** is a supplementary data structure (a set) which
 - keeps all the discovered vertices until they have been finished.

Graph search has several variants mostly defined by the way vertices are picked from the depository. For greater generality, we actually record vertices together with their access edges. **Specific programming tasks** can be (are) performed at each vertex or edge of G while processing them.

Algorithm 2.4. Searching through a connected component G

This algorithm visits and processes every vertex and edge of a *connected* graph G .

```
input < graph  $G$ ;  
state(all vertices and edges of  $G$ ) < initial;  
stack  $U = \{(\emptyset, v_0)\}$ , for any vertex  $v_0$  of  $G$ ;  
search tree  $T = \emptyset$ ;  
while ( $U$  nonempty) {  
    choose  $(e, v) \in U$ ;  
     $U = U \setminus \{(e, v)\}$ ;  
    if  $(e \neq \emptyset)$  PROCESS( $e$ );  
    if (state( $v$ )  $\neq$  finished) {  
        foreach ( $f$  incident with  $v$ ) {  
             $w =$  the opposite vertex of  $f = vw$ ;  
            if (state( $w$ )  $\neq$  finished)  
                 $U = U \cup \{(f, w)\}$ ;  
        }  
        PROCESS( $v$ );  
        state( $v$ ) = finished;  
         $T = T \cup \{e, v\}$ ;  
    }  
}  
 $G$  is finished;
```

Implementation variants of graph searching

- *DFS* depth-first search – the depository U is a “LIFO” stack.

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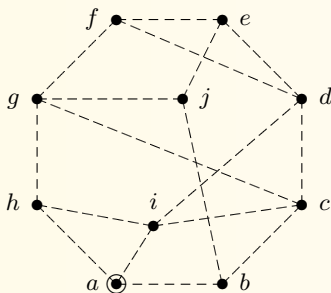
Implementation variants of graph searching

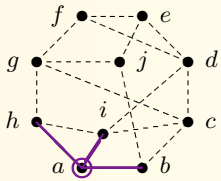
- *DFS* depth-first search – the depository U is a “LIFO” stack.
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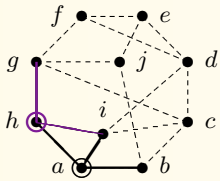
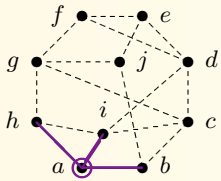
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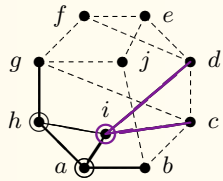
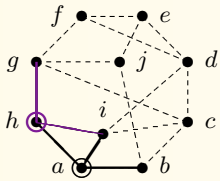
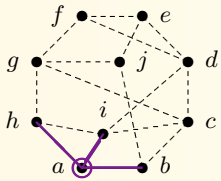
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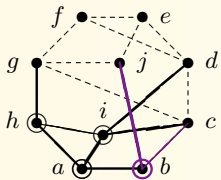
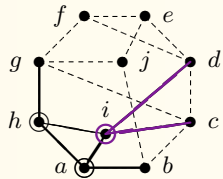
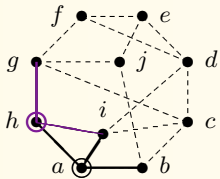
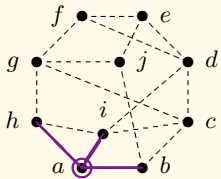
Example 2.15. An example of a *breadth-first* search run from a vertex a .

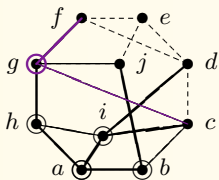
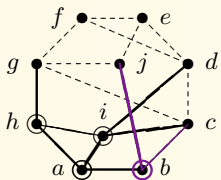
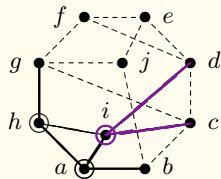
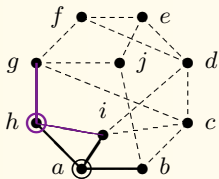
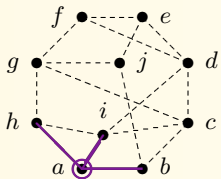


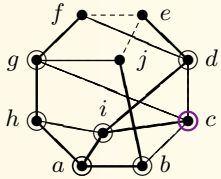
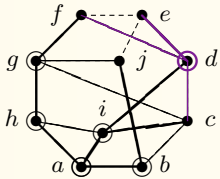
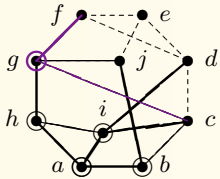
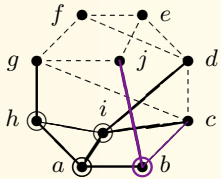
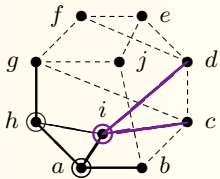
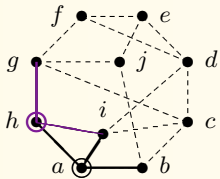
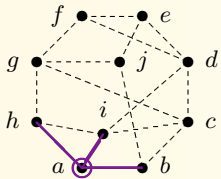


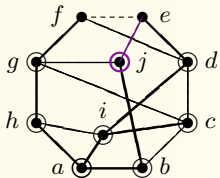
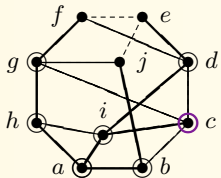
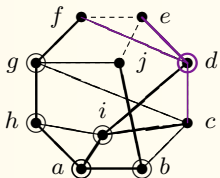
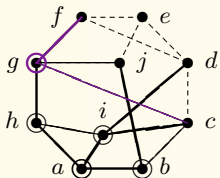
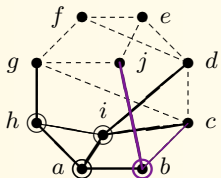
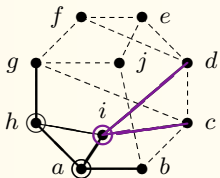
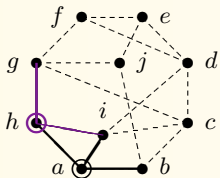
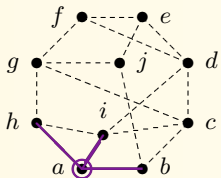


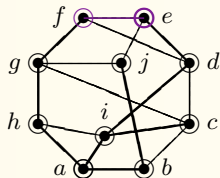
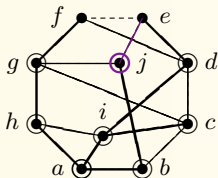
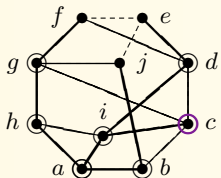
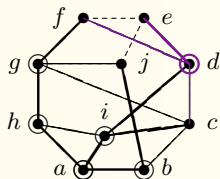
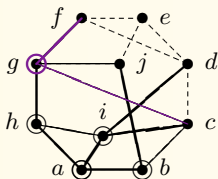
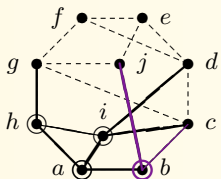
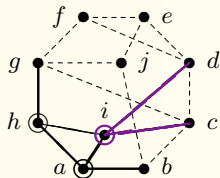
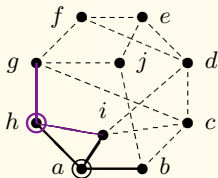
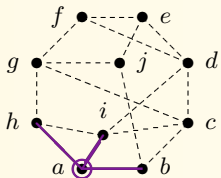






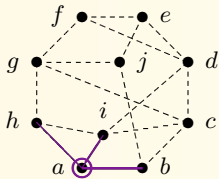




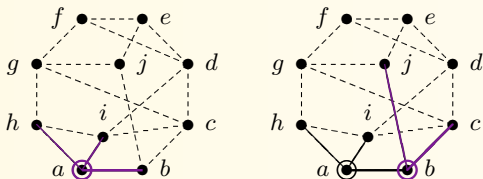


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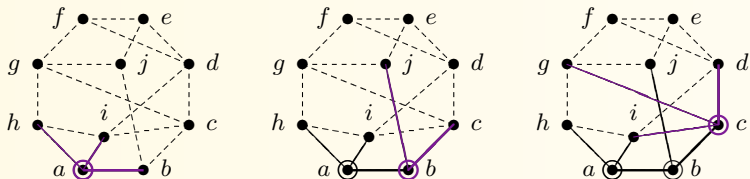
Example 2.16. An example of a *depth-first* search run from a vertex a .



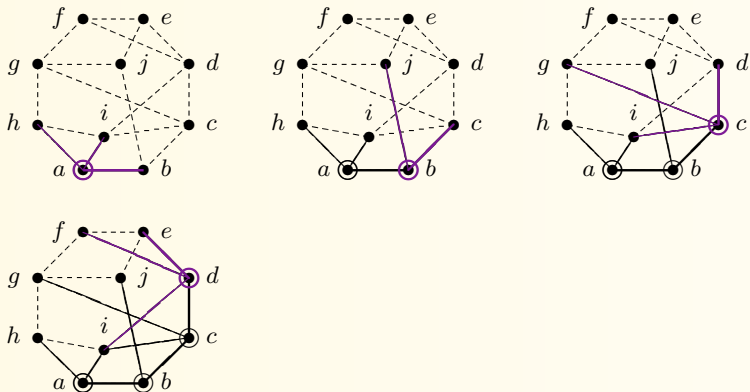
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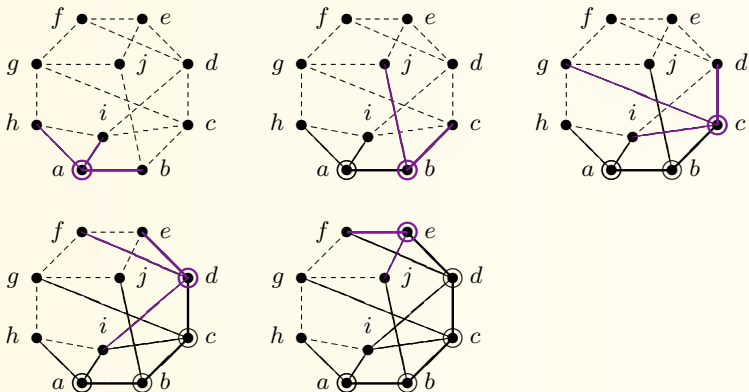
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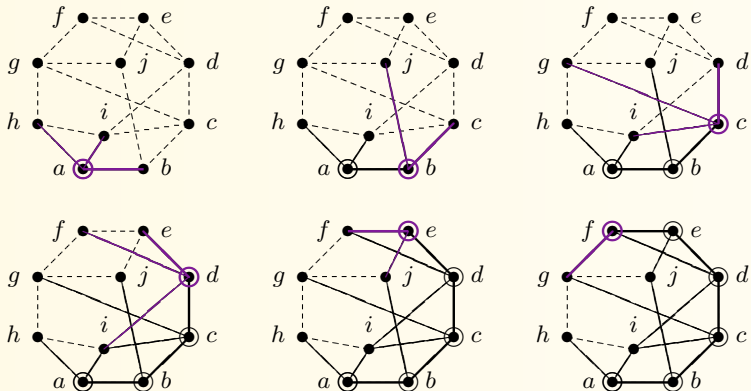
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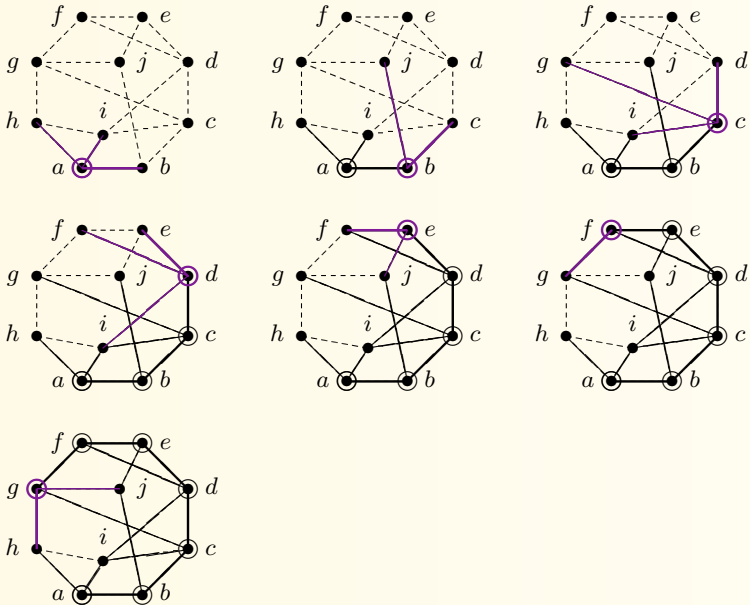
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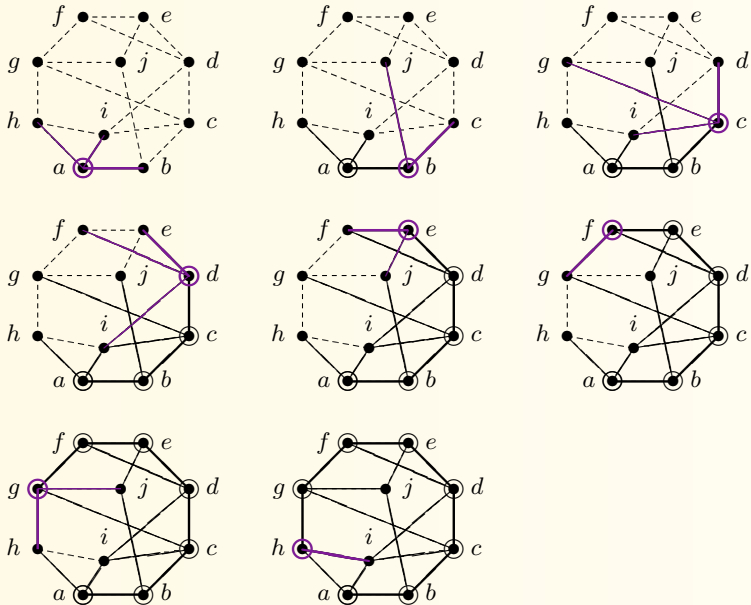
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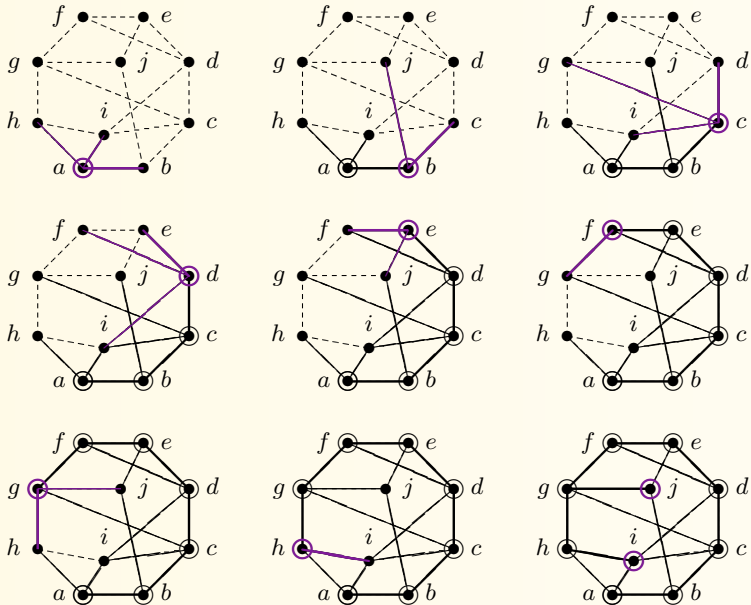
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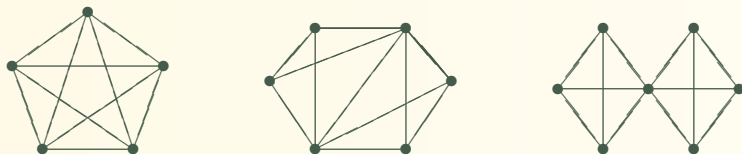
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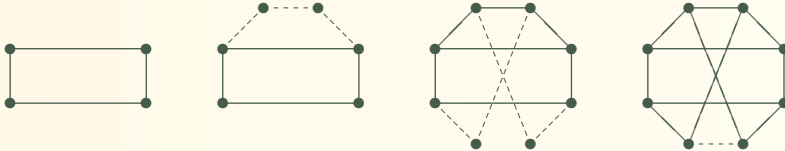
Graphs that are “vertex / edge-1-connected” are simply connected.



Sometimes we speak about a k -connected graph, and then we usually mean it to be **vertex- k -connected**. High vertex connectivity is a (much) stronger requirement than edge connectivity...

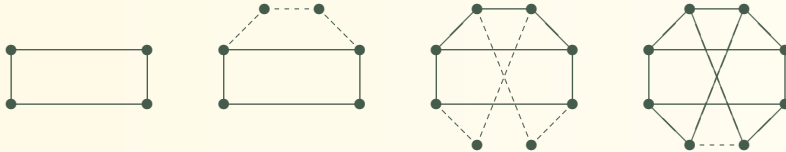
About 2-connected graphs

Theorem 2.5. *A simple graph is 2-connected if, and only if, it can be constructed from a cycle by “adding ears”; i.e. by iterating the operation which adds a new path (of arbitrary length, even an edge, but not a parallel edge) between two existing vertices of a graph.*



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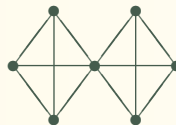
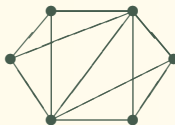
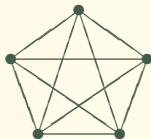


Theorem 2.6. *Assume G is a 2-connected graph. Then every two edges in G lie on a common cycle.*

Menger's theorem and related

Theorem 2.7. A graph G is edge- k -connected if, and only if, there exist (at least) k edge-disjoint paths between any pair of vertices (the paths may share vertices).

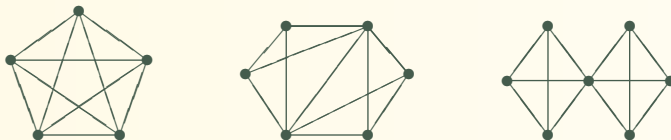
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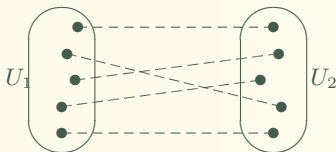
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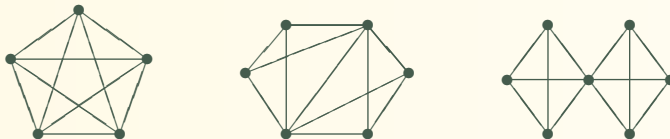
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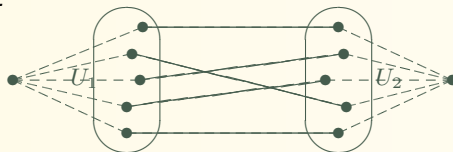
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2.4 Connectivity in Directed Graphs

At the beginning we proceed analogically to the undirected case...

Definition: A *directed walk* of length n in a graph D is a sequence of alternating vertices and directed edges

$$(v_0, e_1, v_1, e_2, v_2, \dots, e_n, v_n),$$

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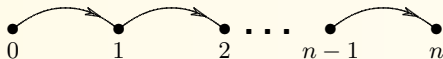
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Theorem 2.9. *If there exists a directed walk from u to v in a digraph D , then there also exists a *directed path* from u to v in this D .*



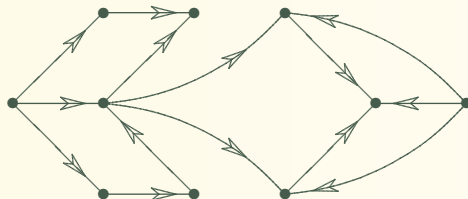
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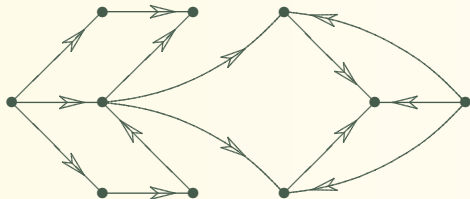
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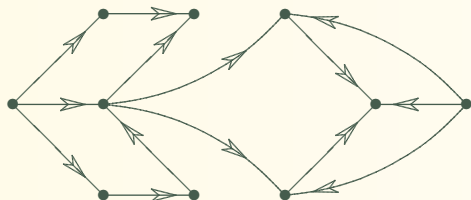


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Strong connectivity

Lemma 2.10. *Let \approx be a binary relation on the vertex set $V(D)$ of a **directed** graph D such that $u \approx v$ if, and only if, there exist two directed walks in D – one starting in u and ending in v and the other starting in v and ending in u .*

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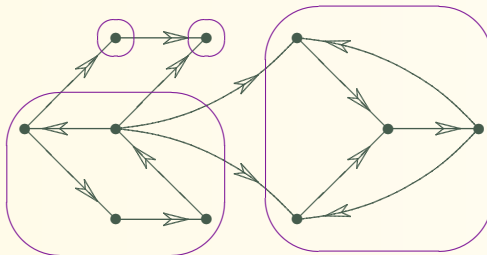
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Definition 2.11. The **strong components** of a digraph D are formed by the equivalence classes of the above relation \approx (Lemma 2.10) on $V(D)$. A digraph is *strongly connected* if it has at most one strong component.

See the four strong components in this illustration picture:



Condensation of a digraph

Definition: A digraph Z whose vertices are the strong components of D , and the arcs of Z exist exactly between those pairs of distinct components of D such that D contains an arc between them, is called a *condensation* of D .

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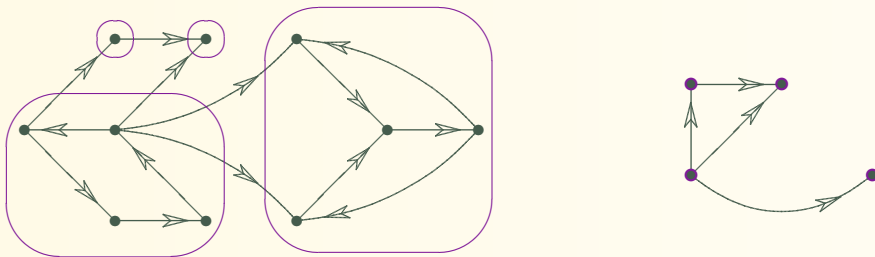
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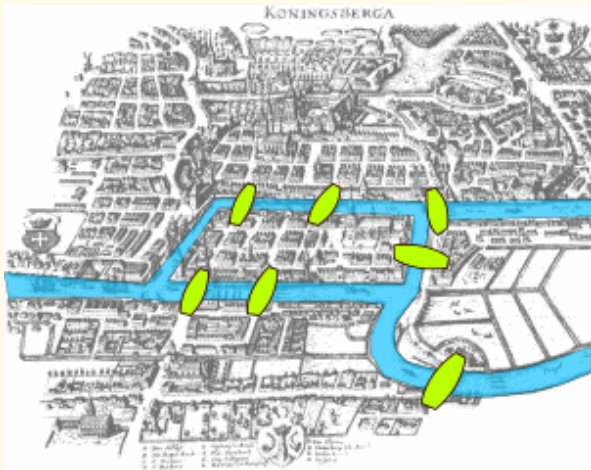
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Proposition 2.12. *The condensation of any digraph is an acyclic digraph.*



2.5 Eulerian Trails

Perhaps the **oldest recorded result** of graph theory comes from famous Leonardo Euler—it is the “**7 bridges of Königsberg**” (Královec, now Kaliningrad) problem.

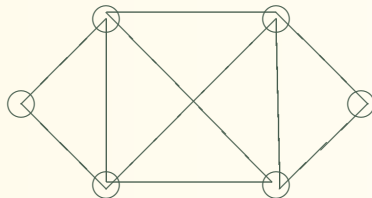
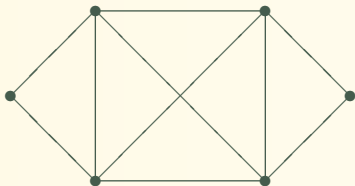


So what was the problem? The city majors that time wanted to walk through the city while crossing each of the 7 bridges exactly once. . .

This problem led Euler to introduce the following:

Definition: A *trail* in a graph is a walk which does not repeat edges.

A *closed trail (tour)* is such a trail that ends in the same vertex it started with. The opposite is an *open trail*.

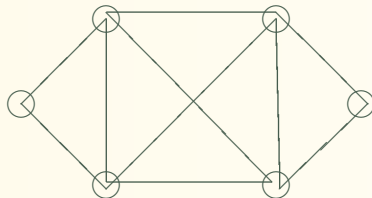
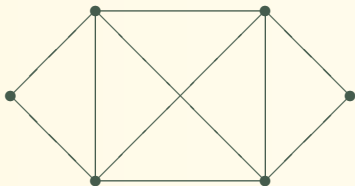


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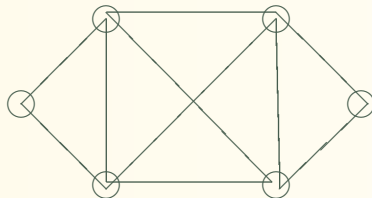
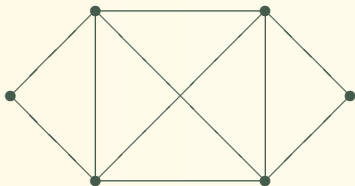
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Corollary 2.14. A (multi)graph G consists of one open trail if, and only if, G is connected and all the vertex degrees in G **but two** are even.

Analogous results hold true also for digraphs (the proofs are the same)...