

Graph Visualization:

Topology-Shape-Metrics

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Topology-Shape-Metrics

1. Background

- Graphs and Graph Drawings
- Planarity
- Topology

2. The topology-shape-metrics approach

1. Topology
2. Shape
3. Metrics

3. Remarks

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Background: **Graphs and Graph Drawings**

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A **graph** as an adjacency matrix

	0	1	2	3	4	5
0	0	1	0	1	0	1
1	1	0	1	0	1	0
2	0	1	0	1	0	1
3	1	0	1	0	1	0
4	0	1	0	1	0	1
5	1	0	1	0	1	0

The same graph as an adjacency list

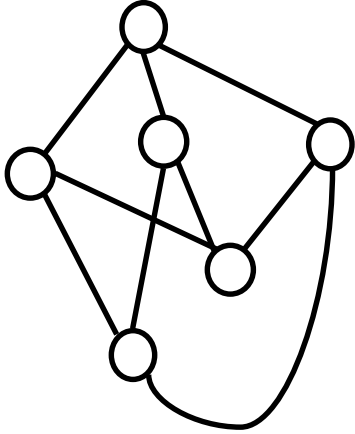
vertex	Adjacent vertices
0	1, 3, 5
1	0, 2, 4
2	1, 3, 5
3	0, 2, 4
4	1, 3, 5
5	0, 2, 4

A *graph* has

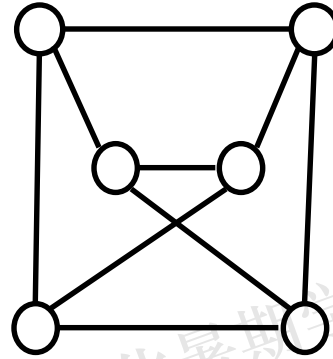
- no geometry
- no graphics

It is *purely combinatorial* information.

A graph drawing



Another drawing of the same graph



A graph drawing has layout

- A position for each vertex
- A route for each edge

It is combinatorial plus geometric information.

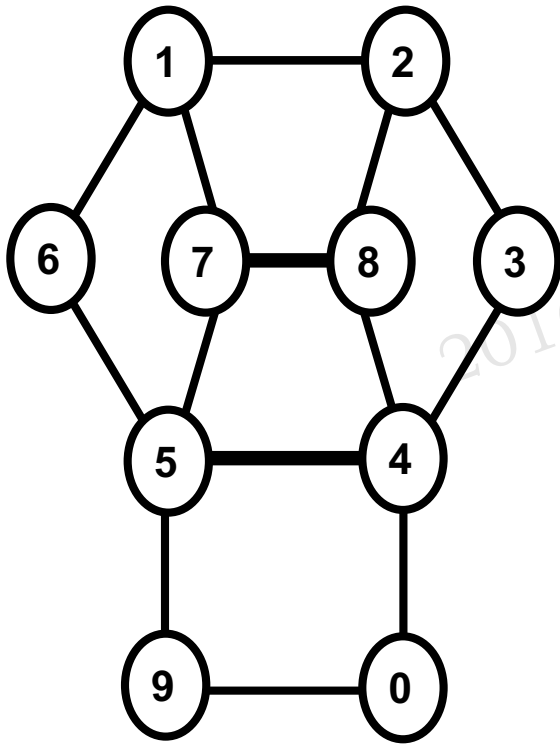
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Background: **Planarity**

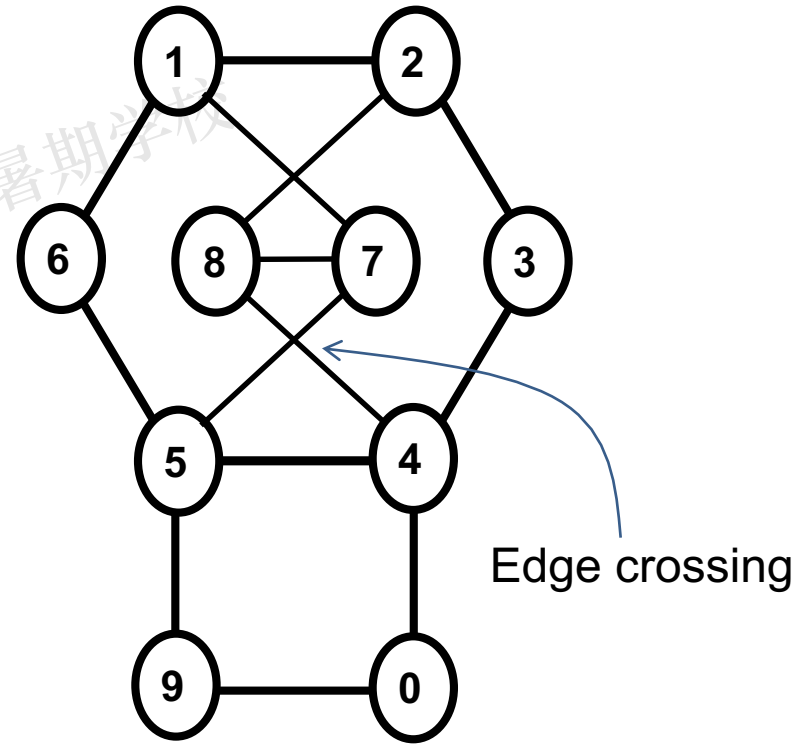
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A graph *drawing* is *planar* if it has no edge crossings.

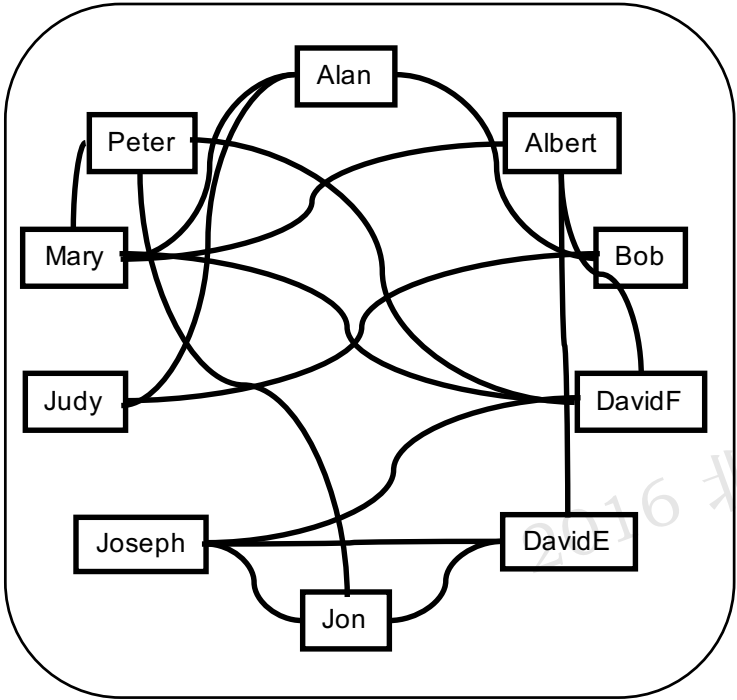
A planar graph drawing



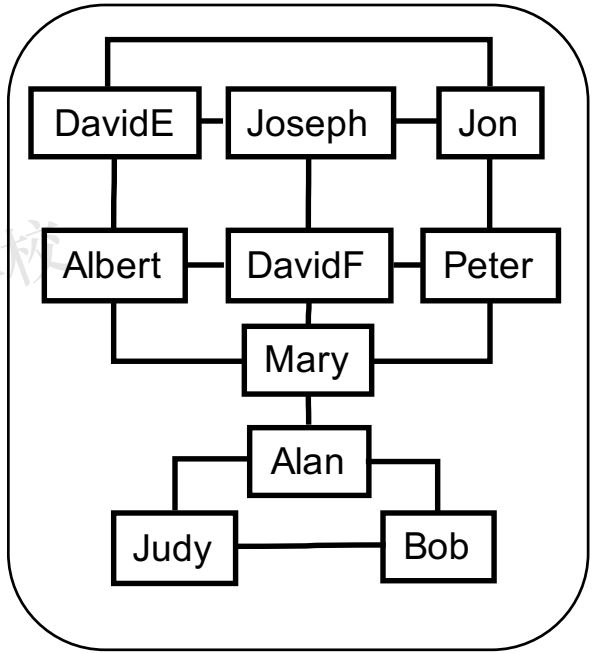
A non-planar graph drawing



Note: Planar drawings make beautiful pictures



Bad picture



Good picture

A graph

is planar

if it can be drawn without edge crossings.

A planar graph G_1

	0	1	2	3	4	5	6	7	8	9
0					1					1
1			1				1	1		
2		1		1					1	
3			1		1					
4	1			1		1			1	
5					1		1	1		1
6		1				1				
7		1				1			1	
8			1		1			1		
9	1					1				

A non-planar graph G_2

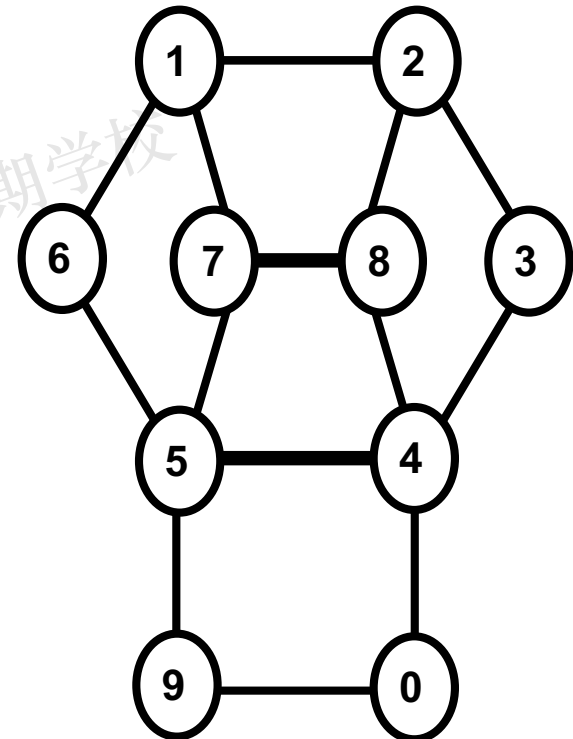
	0	1	2	3	4	5
0		1		1		1
1	1		1		1	
2		1		1		1
3	1		1		1	
4		1		1		1
5	1		1		1	

Lemma:

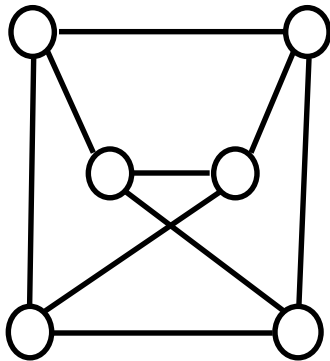
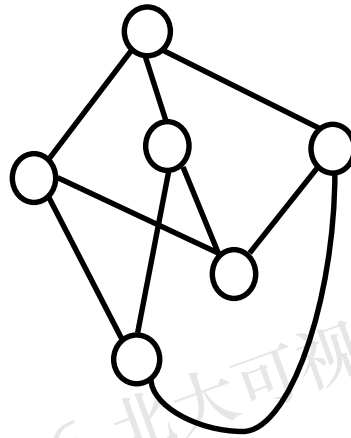
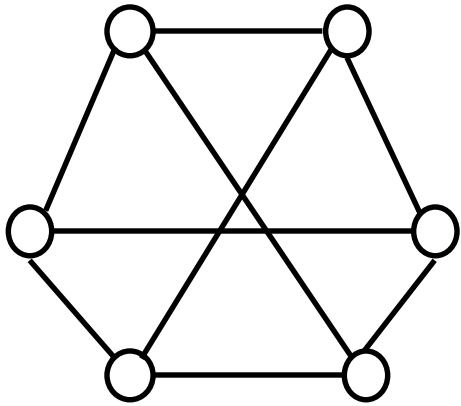
G_1 is planar.

	0	1	2	3	4	5	6	7	8	9
0					1					1
1			1				1	1		
2		1		1					1	
3			1		1					
4	1			1		1			1	
5					1		1	1		1
6		1				1				
7		1				1			1	
8			1		1			1		
9	1					1				

Proof



A graph G is non-planar
 if every drawing of G has edge crossings.



A non-planar graph G_2

	0	1	2	3	4	5
0		1		1		1
1	1		1		1	
2		1		1		1
3	1		1		1	
4		1		1		1
5	1		1		1	

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Note: There are many beautiful theorems about planar graphs

Theorem: (from Euler, 1700s) For a planar graph $G = (V, E)$, $|E| \leq 3|V| - 6$. If $|E| = 3|V| - 6$, then G is triconnected and each face of the embedding of G has 3 edges.

Theorem (Kuratowski, 1930) A graph is planar if and only if it does not contain a subgraph that is a subdivision of K_5 or $K_{3,3}$.

Theorem: (Steinitz, 1930s) For every triconnected planar graph G , there is a convex polyhedron P in 3D such that the vertex-edge graph of P is isomorphic to G .

Theorem: (Appel-Haken, 1970s) A planar graph can be colored in 4 colors.

Theorem: (Lipton-Tarjan, 1980s) For every planar graph G with n vertices, there is set of $O(\sqrt{n})$ vertices whose removal divides the graph into components of size at most $\frac{2n}{3}$.

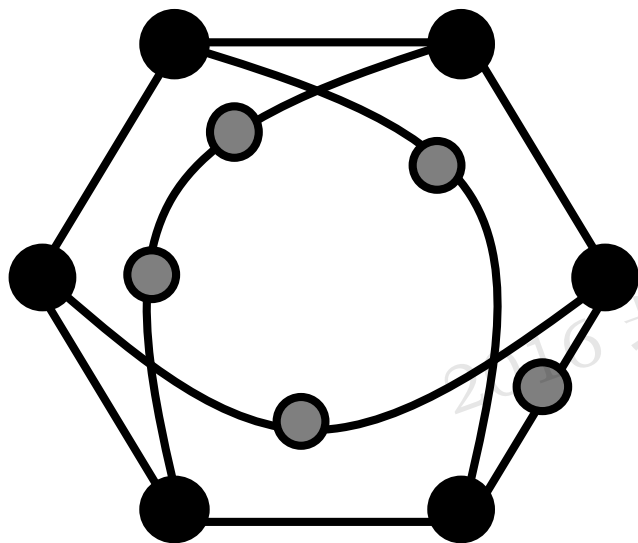
Theorem: (Trivial) A graph is planar if and only if each of its triconnected components is planar.

Graph theorists love planar graphs

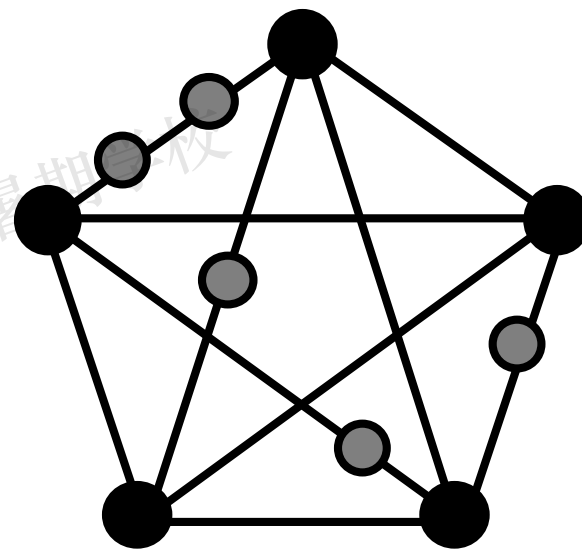
Note: There are many beautiful theorems about planar graphs

Theorem (Kuratowski, 1930)

A graph is planar if and only if it does not contain a subgraph that is a subdivision of K_5 or $K_{3,3}$.



$K_{3,3}$ (complete bipartite graph on 6 vertices) subdivision



K_5 (the complete graph on 5 vertices) subdivision

Theorem (Kuratowski, 1930)

A graph is planar if and only if it does not contain a subgraph that is a subdivision of K_5 or $K_{3,3}$.

Corollary

G_2 is non-planar

Proof:

G_2 is $K_{3,3}$.

G_2

	0	1	2	3	4	5
0		1		1		1
1	1		1		1	
2		1		1		1
3	1		1		1	
4		1		1		1
5	1		1		1	

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Remarks: **Planar** graphs and **real-world** graphs

- Most real-world graphs are not planar
- But most are “nearly” planar in some sense:
 - deletion of $o(n)$ edges gives a planar graph
 - scale-free networks are locally dense and globally sparse

Planarity testing algorithms

Hopcroft-Tarjan planarity testing algorithm (1974)

- Tests whether a graph is planar or not, in linear time
- Very complicated algorithm; implementation difficult
 - First published version incorrect; corrected by Deo (1976)
 - Most implementations incorrect
 - First correct implementation (I believe) 1994.

Many subsequent planarity testing algorithms

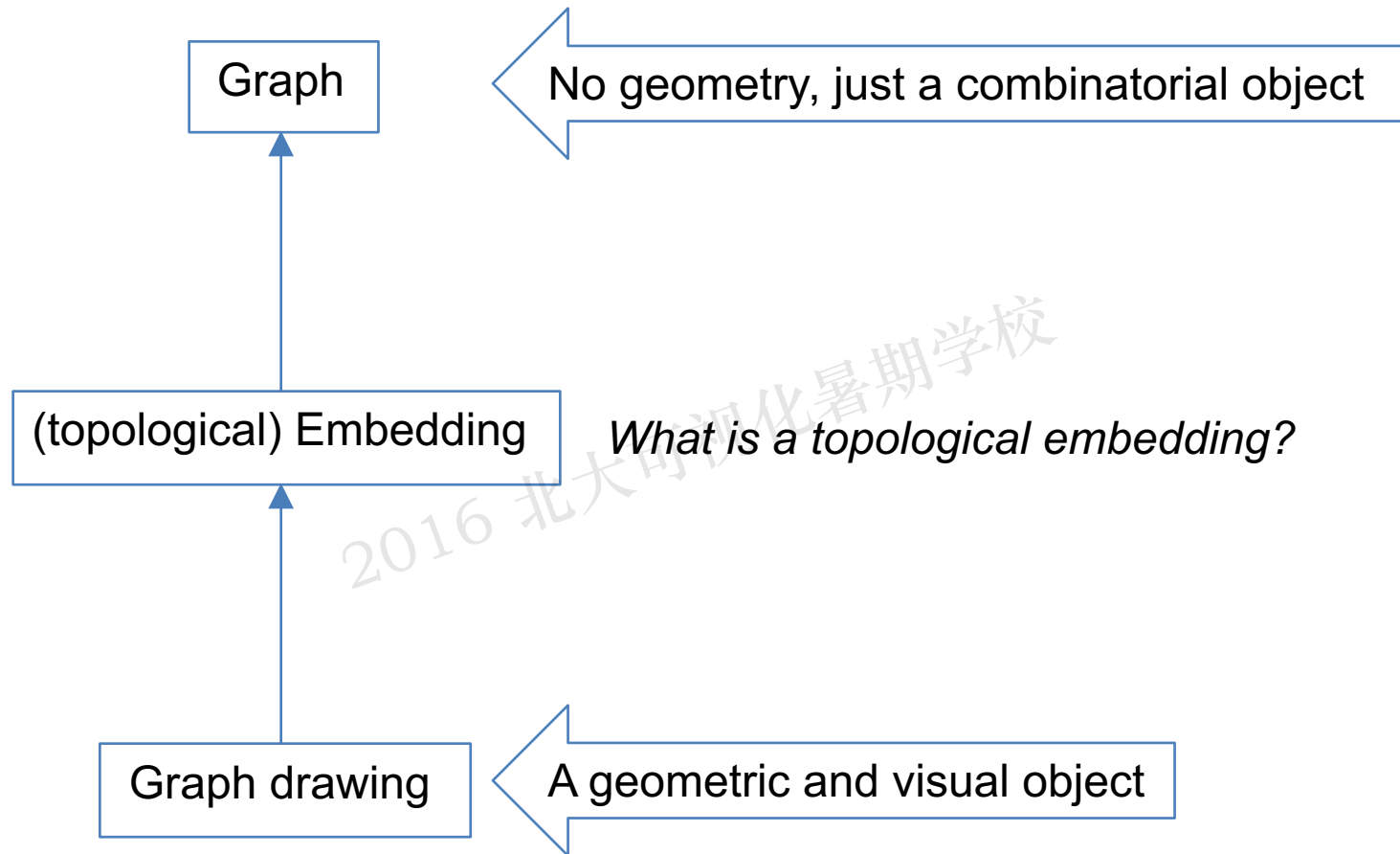
- Lempel-Even-Cederbaum 1966
- Booth-Lueker 1976
- Rosenstiel-de Frayssieux 1990
- Hsu/Boyer-Myvold 2000

Note:

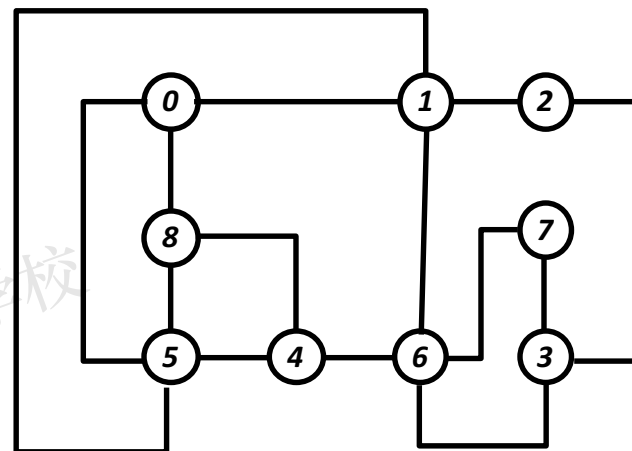
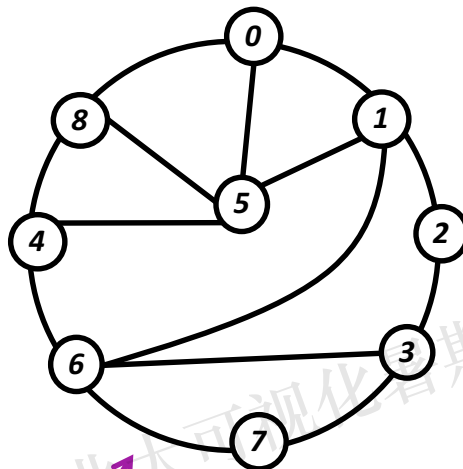
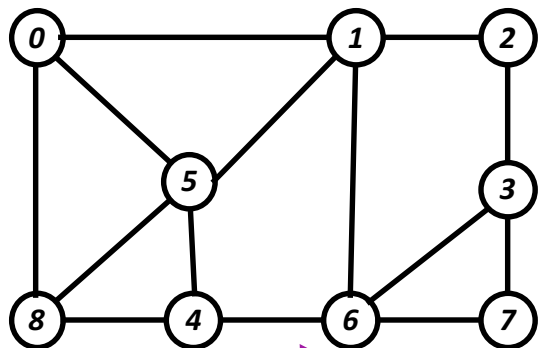
- All these planarity testing algorithms are efficient and effective, but none is elegant.
- Finding an elegant linear time planarity testing algorithm is still an unsolved problem.

Background: **Topology**

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Three different drawings of the same graph

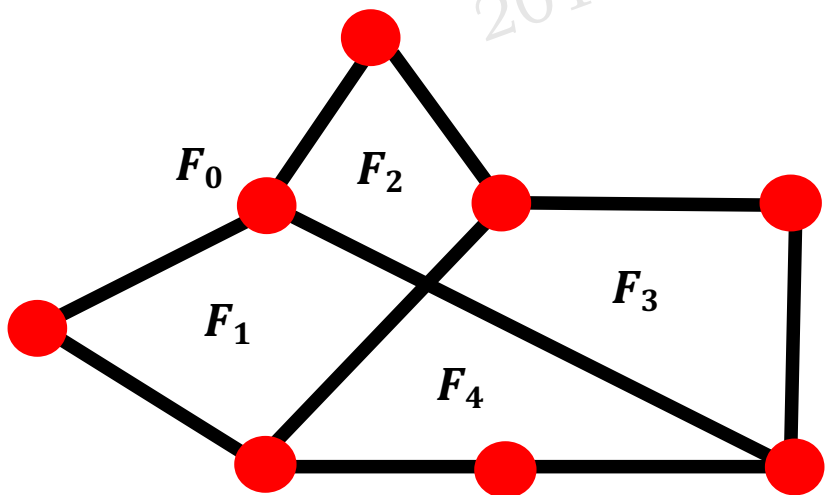
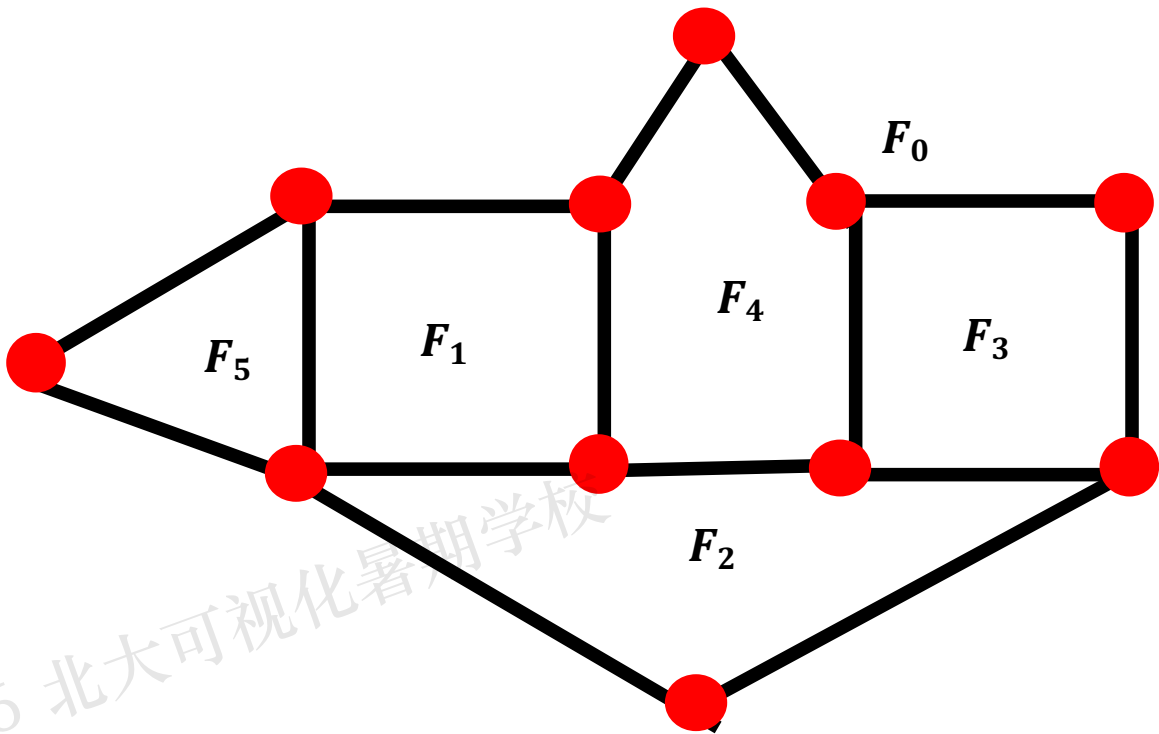


The same topological embedding

Different topological embeddings

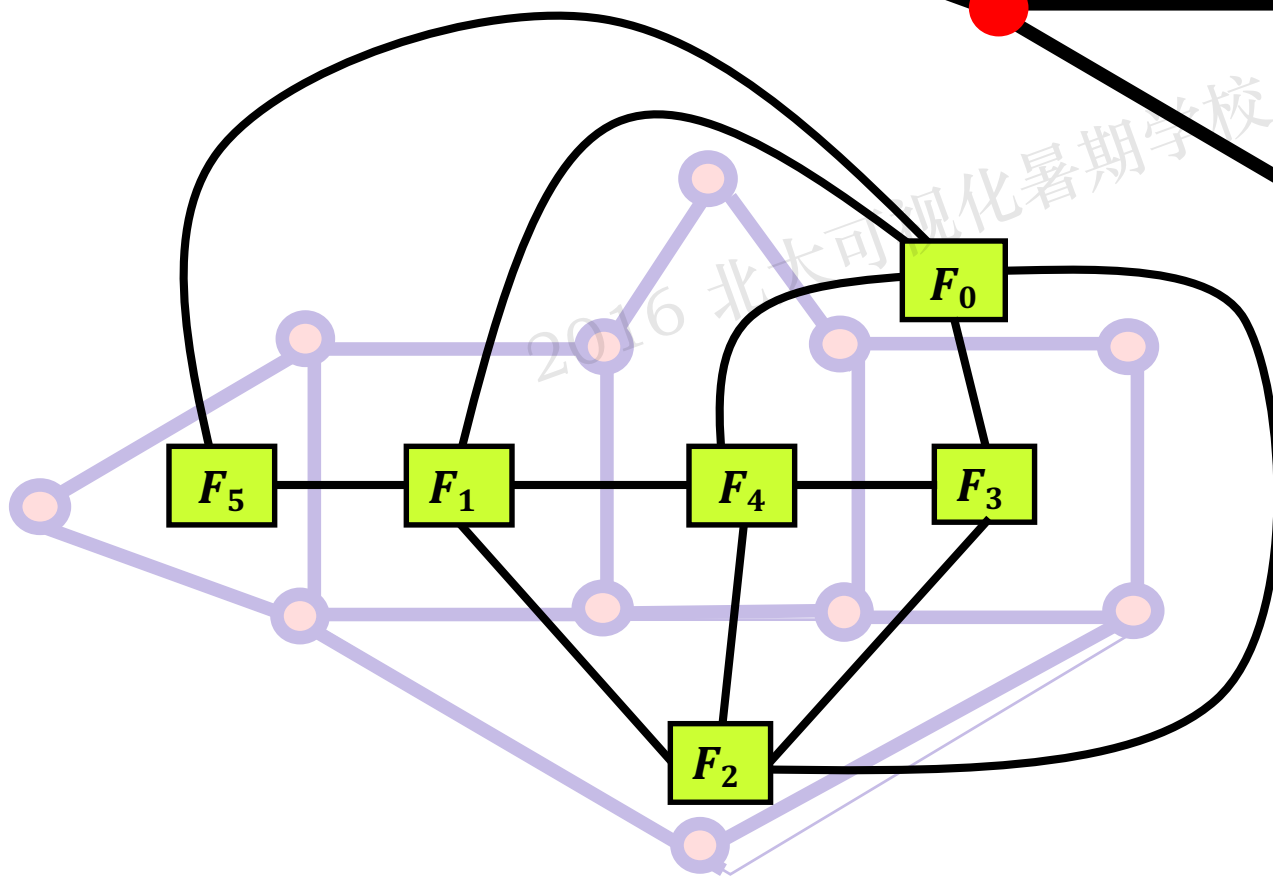
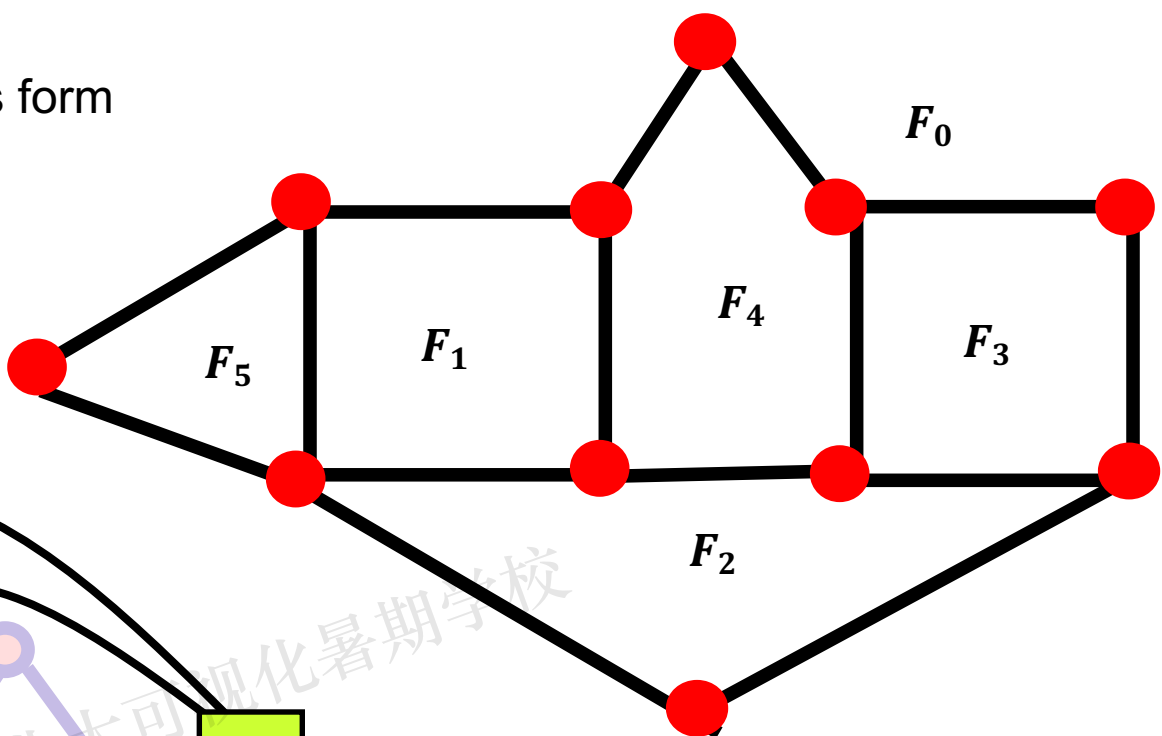
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A graph drawing divides the plane into regions called faces.



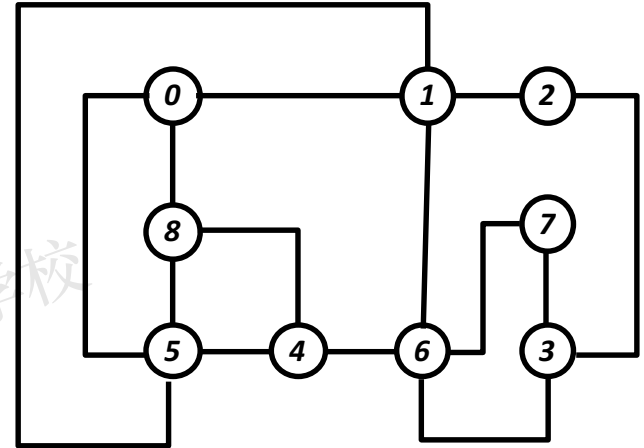
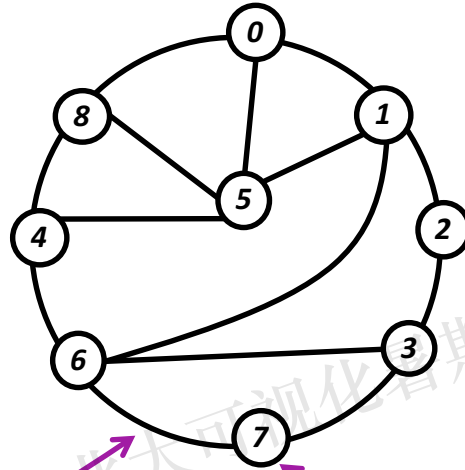
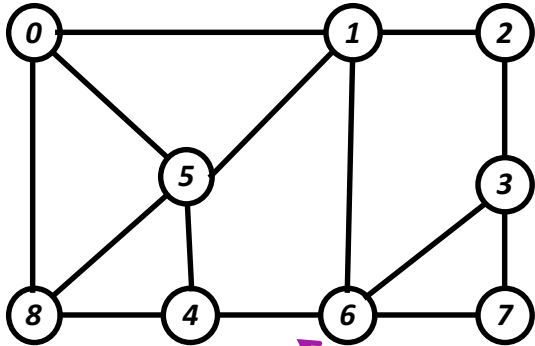
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The faces plus their adjacencies form the dual graph of the drawing.



Definition:

Two drawings of a graph are *topologically equivalent* if there is a homeomorphism of the plane/sphere that maps one to the other.



Topologically equivalent

Not topologically equivalent

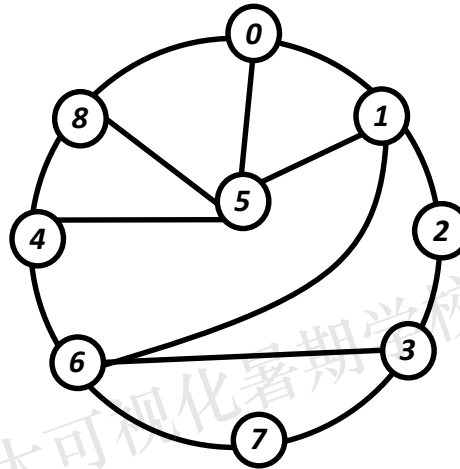
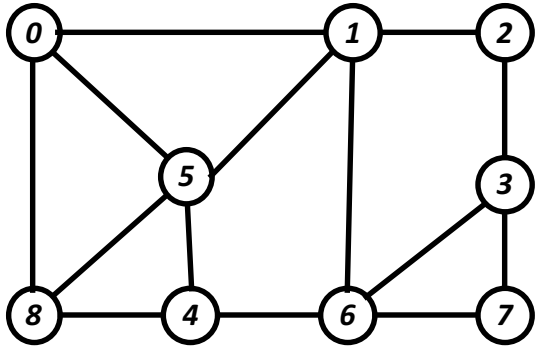
Alternative definitions:

Two drawings of a graph are topologically equivalent if

1. they have the “same” dual graph, or alternatively
2. they have the same clockwise circular ordering of edges around each vertex.

Definition:

Two drawings of a graph are *topologically equivalent* if there is a homeomorphism of the plane/sphere that maps one to the other.



0	8,1,5
1	5,0,2,6
2	1,3
3	6,2,7
4	8,5,6
5	8,0,1,4
6	4,1,3,7
7	6,3
8	0,5,4

Same
clockwise
circular
orderings of
edges around
each vertex

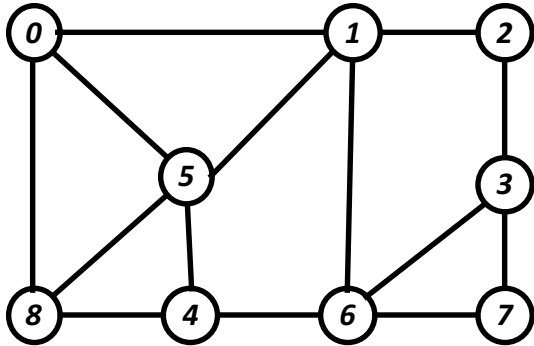


Same
topological
embedding

0	5,8,1
1	6,5,0,2
2	3,1
3	7,6,2
4	8,5,6
5	4,8,0,1
6	4,1,3,7
7	6,3
8	4,0,5

Definition:

A (topological) embedding of a graph G is an equivalence class of drawings of G under topological equivalence.



0	8,1,5
1	5,0,2,6
2	1,3
3	6,2,7
4	8,5,6
5	8,0,1,4
6	4,1,3,7
7	6,3
8	0,5,4

Alternative definitions:

1. An embedding consists of a graph, plus the dual graph.
2. An embedding consists of a graph, plus a clockwise circular ordering of edges around each vertex.

A data structure for planar embeddings:

- a) List of vertices.
- b) For each vertex u , a circular list of vertices v adjacent to u .

Planar embedding algorithms

- Most planarity testing algorithms “can be adjusted” to output a planar embedding of a planar graph in linear time.
- All these algorithms are efficient and effective; none is elegant.

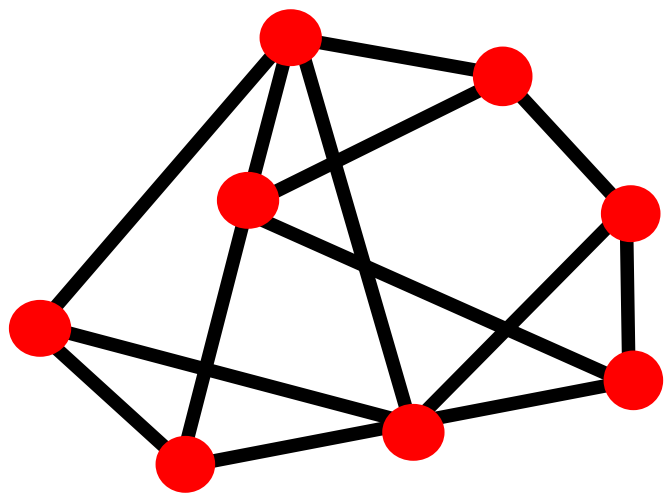
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Background: **Planarization**

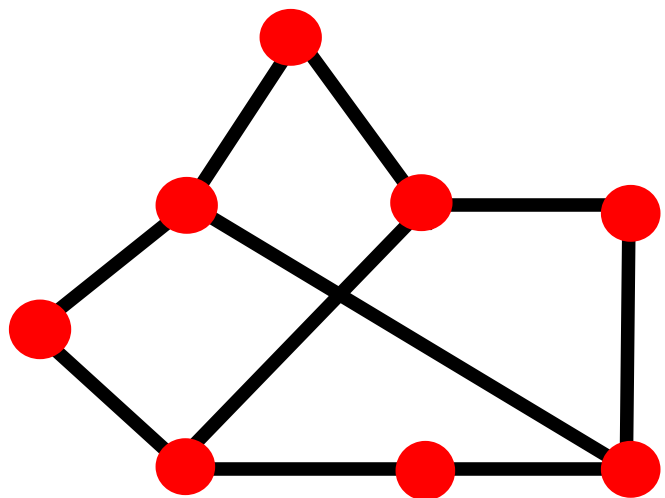
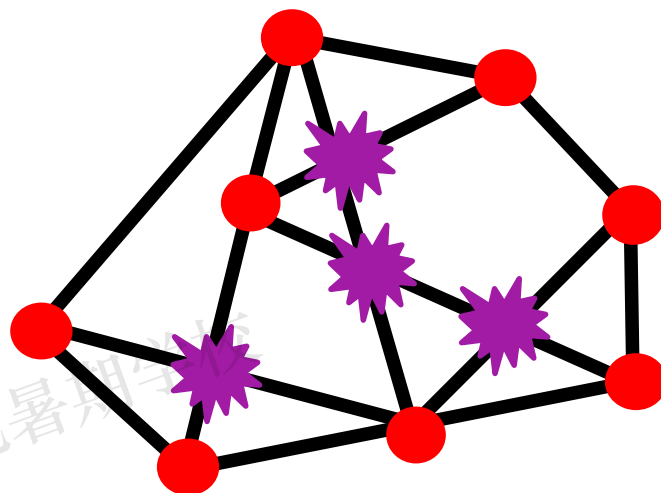
- a) Planarize a topological embedding
- b) Planarize a graph

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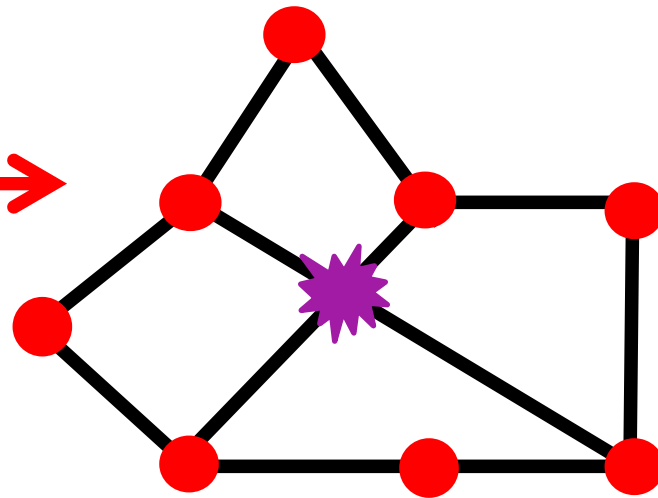
A non-planar topological embedding can be planarized by placing dummy vertices at the edge crossings.



planarize
→

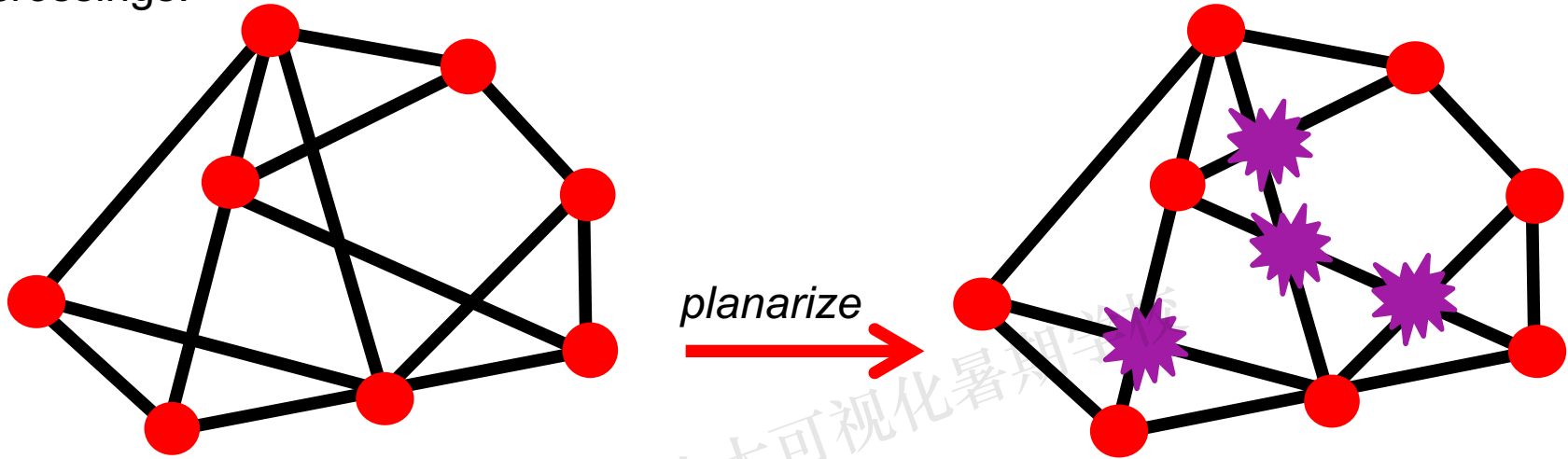


planarize
→



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A non-planar embedding can be planarized by placing dummy vertices at the edge crossings.



Note:

- Planarization of a topological embedding can be done in linear time.
- The concept of planarization allows us to apply terminology, definitions, and data structures about planar embeddings to non-planar embeddings.
- If the number of crossing points is small, then planarization might not increase asymptotic time complexity of algorithms.
- Normally, we use planarization to give a data structure for a non-planar embedding.

A non-planar graph can be planarized by placing gluing independent edges together.

A non-planar graph

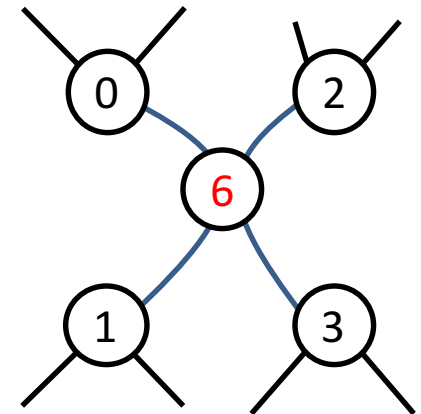
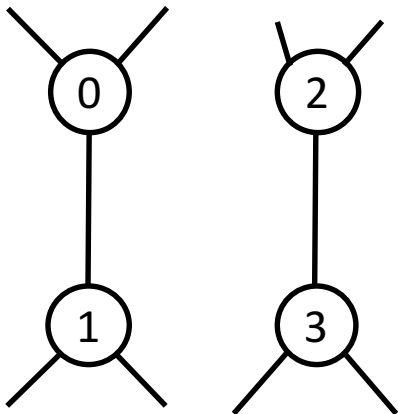
vertex	Adjacent vertices
0	1, 3, 5
1	0, 2, 4
2	1, 3, 5
3	0, 2, 4
4	1, 3, 5
5	0, 2, 4

Glue edges
(0,1) and (2,3)



vertex	Adjacent vertices
0	6, 3, 5
1	6, 2, 4
2	1, 6, 5
3	0, 6, 4
4	1, 3, 5
5	0, 2, 4
6	0, 1, 2, 3

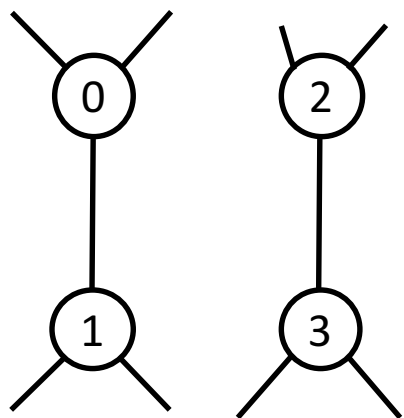
A planar graph



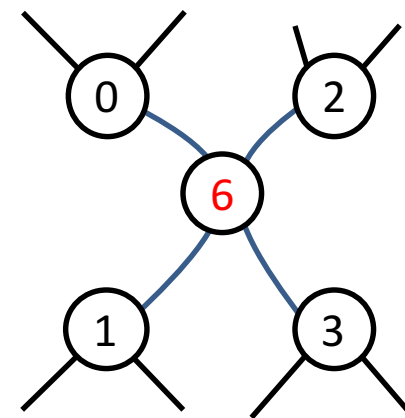
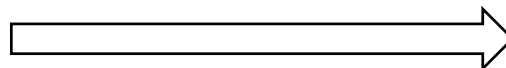
A non-planar graph can be planarized by placing gluing independent edges together.

Theorem: For every graph G , there is a sequence of edge-gluing that makes G planar.

Theorem: Finding a minimum sequence of edge-gluing that makes a graph planar is an NP-complete problem.



Glue edges
(0,1) and (2,3)



3. Graph drawings

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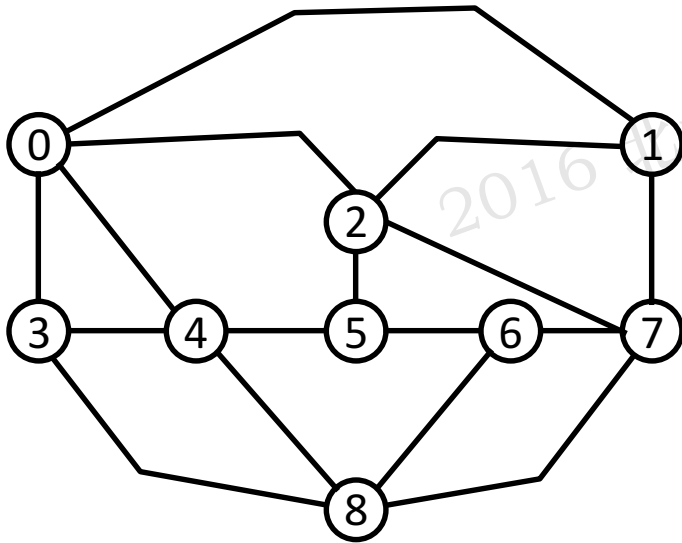
A drawing of a graph $G = (V, E)$ consists of

- a location $p(u)$ for each vertex u , and
- a Jordan arc $c(u, v)$ for each edge (u, v) such that the endpoints of $c(u, v)$ are $p(u)$ and $p(v)$.
- (plus a lot of non-degeneracy conditions)

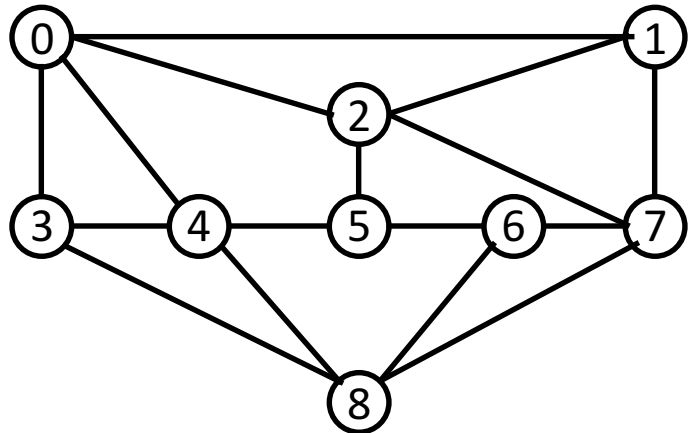
There are many kinds of graph drawings

- *Grid drawing*: vertices (and edge bends?) are located at integer grid points
- *Polyline drawing*: edges are polylines
- *Straight-line drawing*: edges are straight line segments
- *Orthogonal drawing*: edges are polylines made up of vertical and horizontal line segments
-
-

Polyline drawing

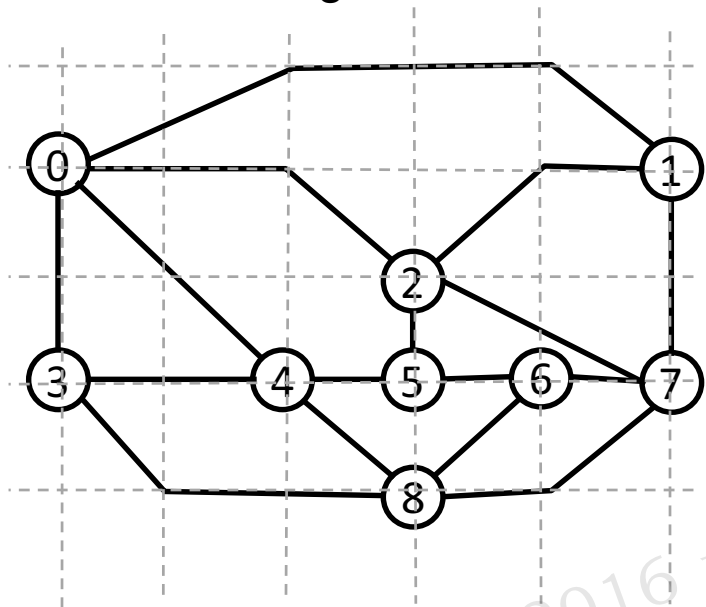


Straight-line drawing

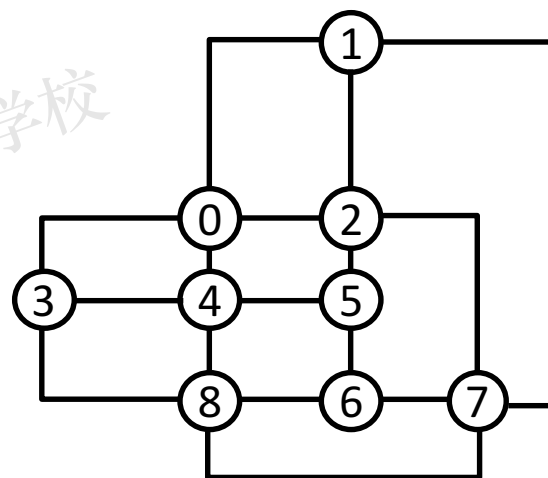


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Grid drawing

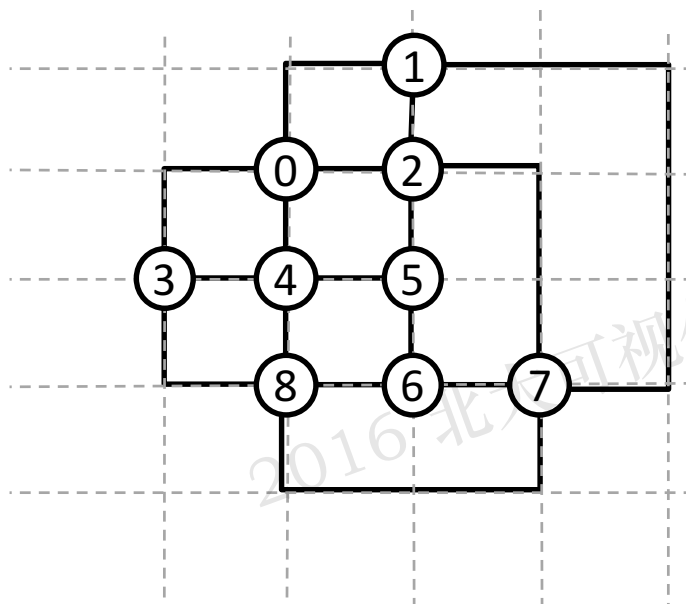


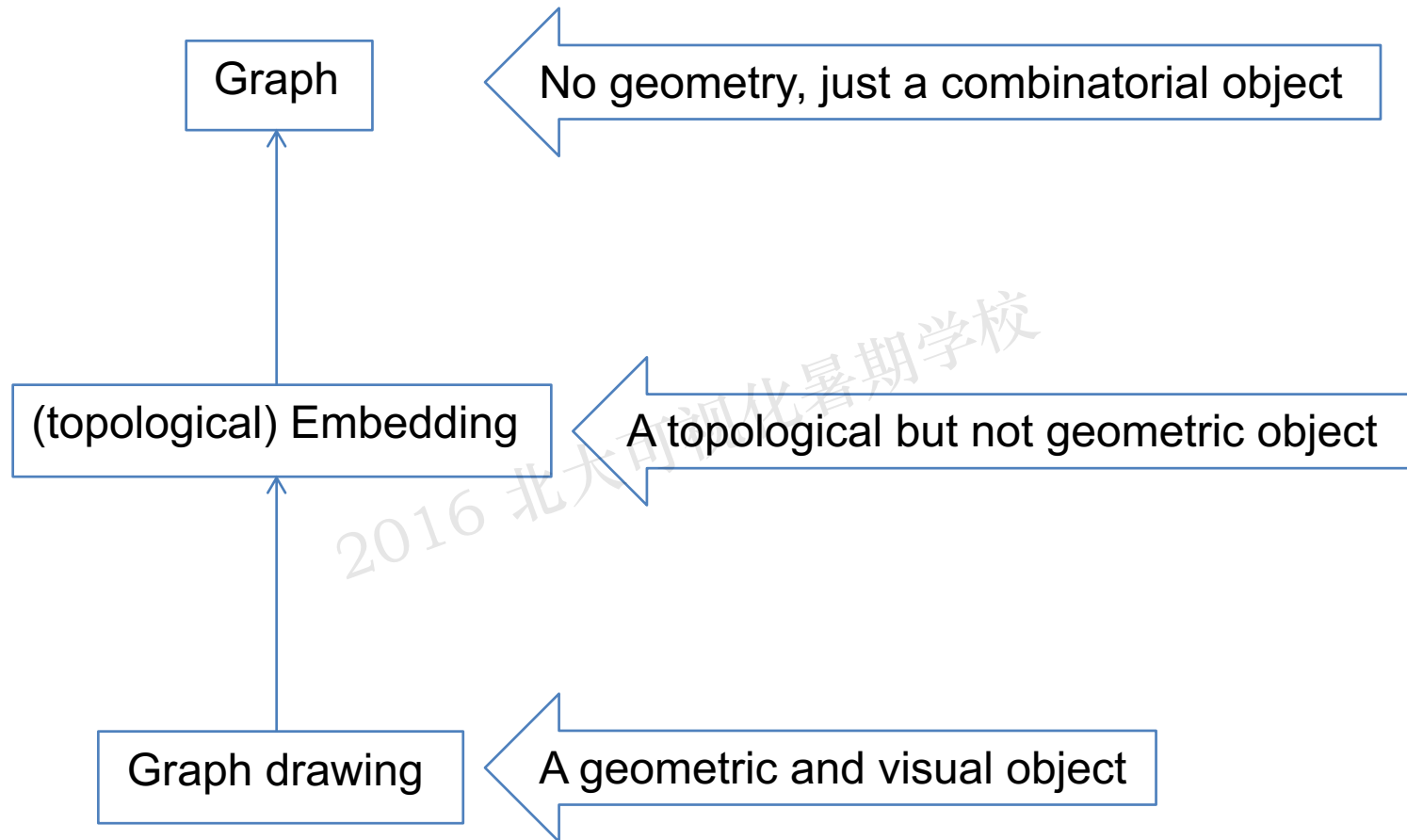
Orthogonal drawing



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Orthogonal grid drawing





Topology-shape-metrics approach

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Topology-shape-metrics method:

Input: a graph G

Algorithm:

1. Topology: Compute a good topological embedding of G
2. Shape: Compute a good orthogonal shape for this topological embedding
3. Metrics: Compute a good orthogonal grid drawing of G

Output: an *orthogonal grid* drawing of G

Aim: produce a topological embedding with few edge crossings

1. Compute a good topological *embedding* of G

Input: a graph $G = (V, E)$

- a) Compute a planar subgraph $G' = (V, E')$, where E' is a subset of E , such that $|E'|$ is as large as possible.
- b) Compute a planar embedding G'' of G' .
- c) Insert the edges of $E - E'$ into G'' , creating as few crossings as possible, to create an embedding G''' of G .

Output: an embedding of G with few crossings

Use a maximum-planar-subgraph method

We need solutions for a difficult problem:

Maximum Planar Subgraph (MPS)

Input: a graph G

Output: a planar subgraph of G with a maximum number of edges.

Note

- ◆ The Maximum Planar Subgraph problem is NP-complete
- ◆ Many heuristic approaches have been investigated, implemented, and tested over at least the last 30 years

One successful approach to MPS so far is *integer linear programming*

Integer Linear Program for the Maximum Planar Subgraph problem

Given a graph $G = (V, E)$:

Variables

x_e for each edge $e \in E$

Objective

Maximize $\sum_{e \in E} x_e$

Constraints

a) $x_e \in \{0, 1\}$

b) For each Kuratowski subgraph $K = (V_K, E_K)$ of G :

$$\sum_{e \in E_K} x_e < |E_K|$$

Interpretation:

$$x_e = \begin{cases} 1 & \text{if } e \in E' \\ 0 & \text{otherwise} \end{cases}$$

Algorithm:

- Use a traditional “branch&cut” approach, with cutting planes from the many theorems on planar graphs.

1. Compute a good topological *embedding* of G

Input: a graph $G = (V, E)$

- a) Compute a planar subgraph $G' = (V, E')$, where E' is a subset of E , such that $|E'|$ is as large as possible.
- b) Compute a planar embedding G'' of G' .
- c) Insert the edges of $E - E'$ into G'' , creating as few crossings as possible, to create an embedding G''' of G .

Output: an embedding of G with few crossings

Use planar embedding algorithms

- For example, a variation on the Hopcroft-Tarjan planarity algorithm

1. Compute a good topological *embedding* of G

Input: a graph $G = (V, E)$

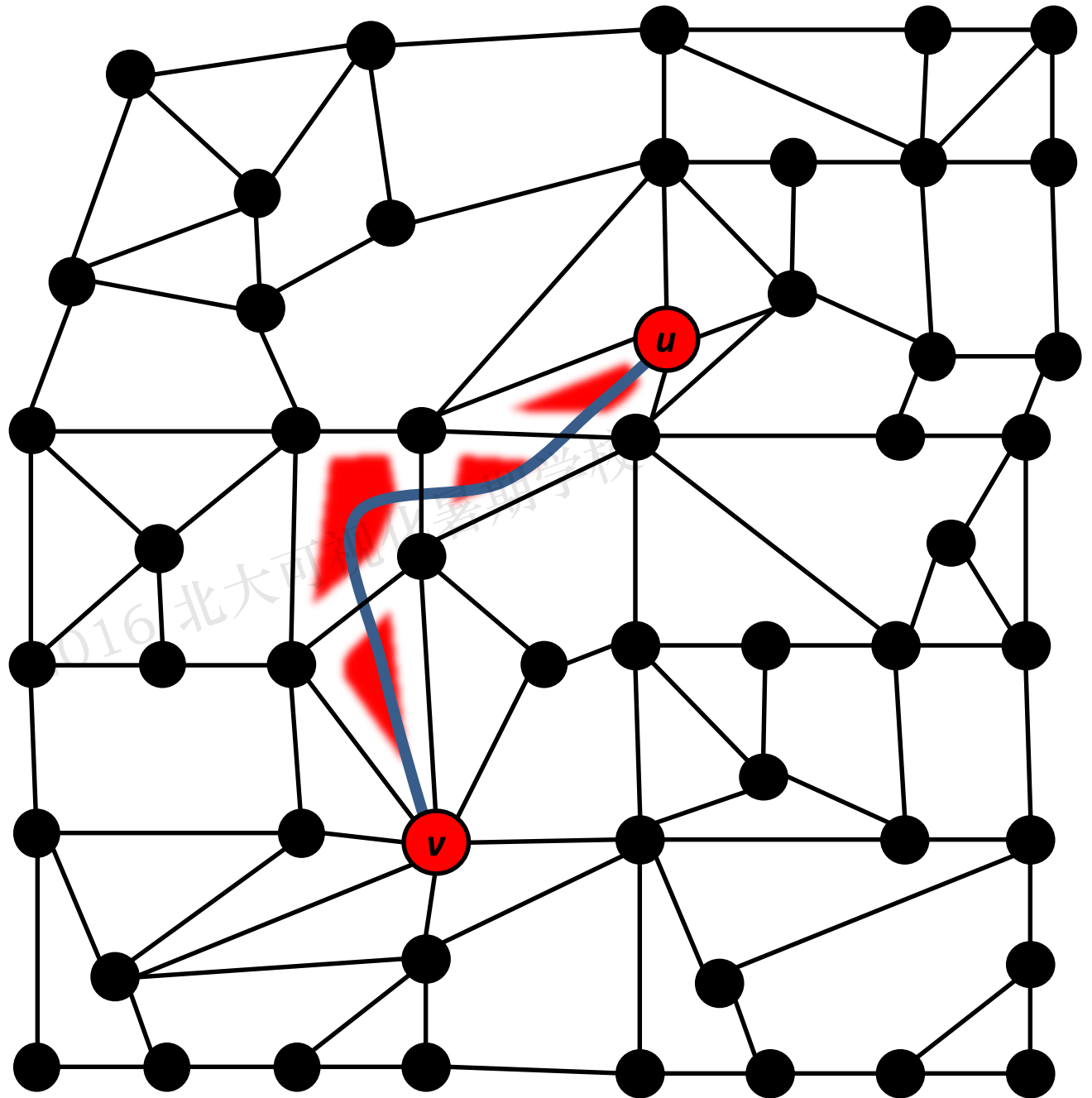
- a) Compute a planar subgraph $G' = (V, E')$, where E' is a subset of E , such that $|E'|$ is as large as possible.
- b) Compute a planar embedding G'' of G' .
- c) Insert the edges of $E - E'$ into G'' , creating as few crossings as possible, to create an embedding G''' of G .

Output: an embedding of G with few crossings

Use planar shortest path in the dual

To insert an edge $(u, v) \in E - E'$ into G'' :

- Construct the dual graph of G''
- Let f_u be the set of faces containing u
- Let f_v be the set of faces containing v
- Route (u, v) via a shortest path from f_u to f_v .



Topology-shape-metrics approach:

Input: a graph G

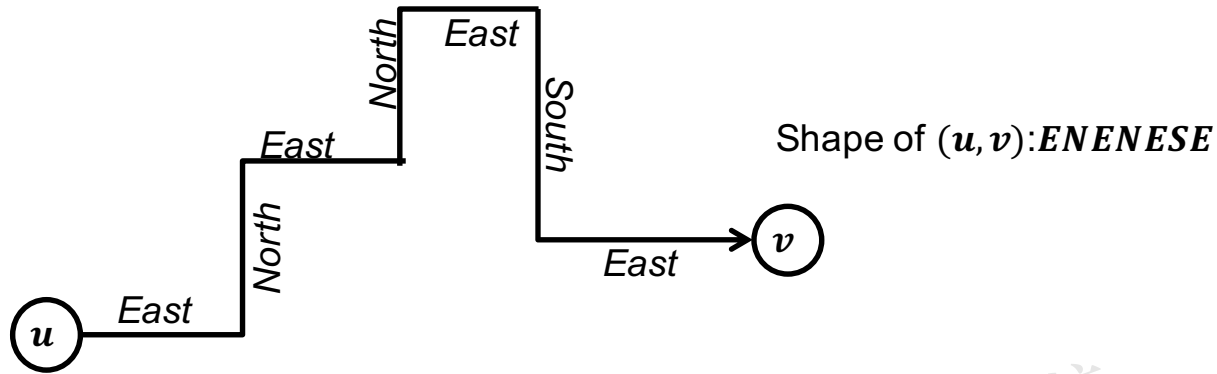
Algorithm:

1. Topology: Compute a good topological embedding of G
2. Shape: Compute a good orthogonal *shape* for this topological embedding
3. Metrics: Compute a good orthogonal grid drawing of G

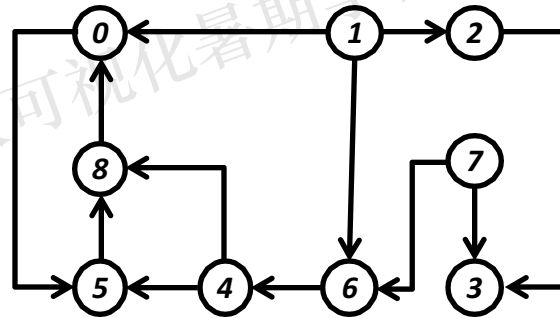
Output: an *orthogonal grid* drawing of G

Aim: give shape with a small number of edge bends.

The shape of a directed orthogonal edge is the sequence of North/South/East/West turns.



The shape of an orthogonal drawing consists of the shape of each edge (after directing edges arbitrarily)



0→5	WSE
1→0	W
1→2	E
1→6	S
2→3	ESW
4→5	W
4→8	NW
5→8	N
6→4	W
7→3	S
7→6	WSW
8→0	N

2. Shape:

- Compute a good orthogonal shape for the topological embedding output from the topology step.
- We want a small number of bends

Minimum Bends Problem

Input: An embedding G

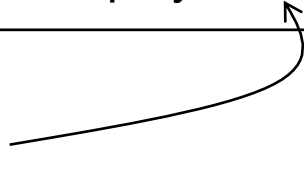
Output: A shape for G with a minimum number of bends.

Surprising result

Theorem (Tamassia, ~1987)

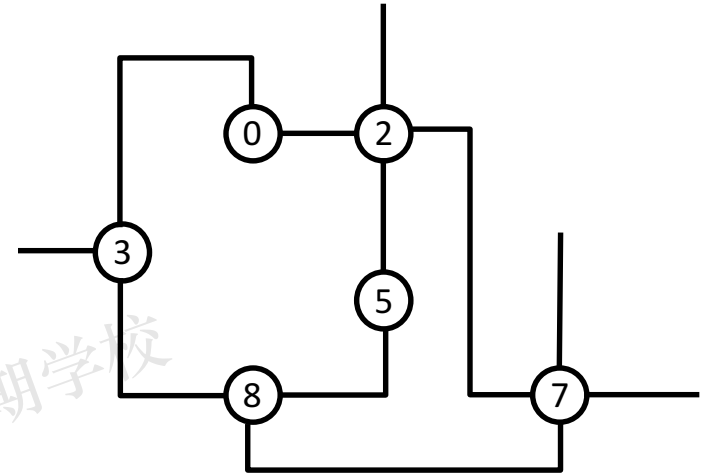
The Minimum Bends Problem can be solved in polynomial time.

$O(n^{1.75} \log n)$



Tamassia's algorithm to give a shape with a minimum total number of bends

- Note that $\frac{\pi}{2}$ angles in a drawing satisfy some linear constraints
 - The sum of angles around a face is $2(a + b - 4)\frac{\pi}{2}$, where a is the number of vertices and b is the number of bends in the face.
 - The sum of angles around a vertex is $4\frac{\pi}{2}$.
 - ... plus other constraints from theorems on planar graphs
 - ...
 - ...



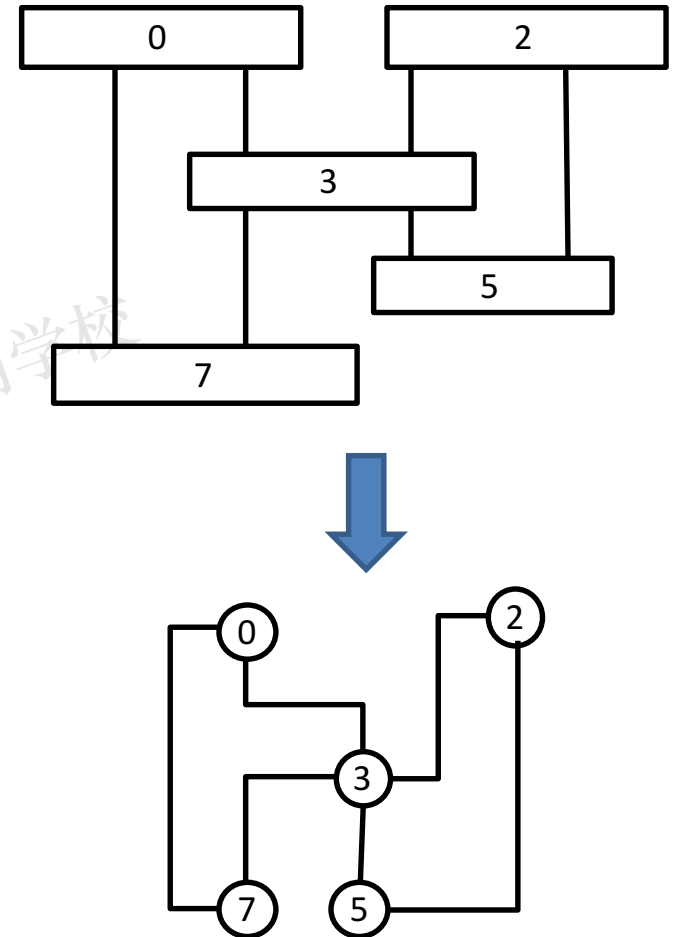
- These constraints can be modelled as a minimum flow problem
- We can solve the minimum flow problem in polynomial time.

Visibility Algorithm

Alternative method to give a shape with a small number of bends

1. Create a visibility representation of the input embedding
2. Adjust the visibility representation to give an orthogonal shape

- Runs in linear time
- Relatively elegant
- Does not give a minimum total number of bends
- But guarantees that the number of bends on an edge is at most 4.

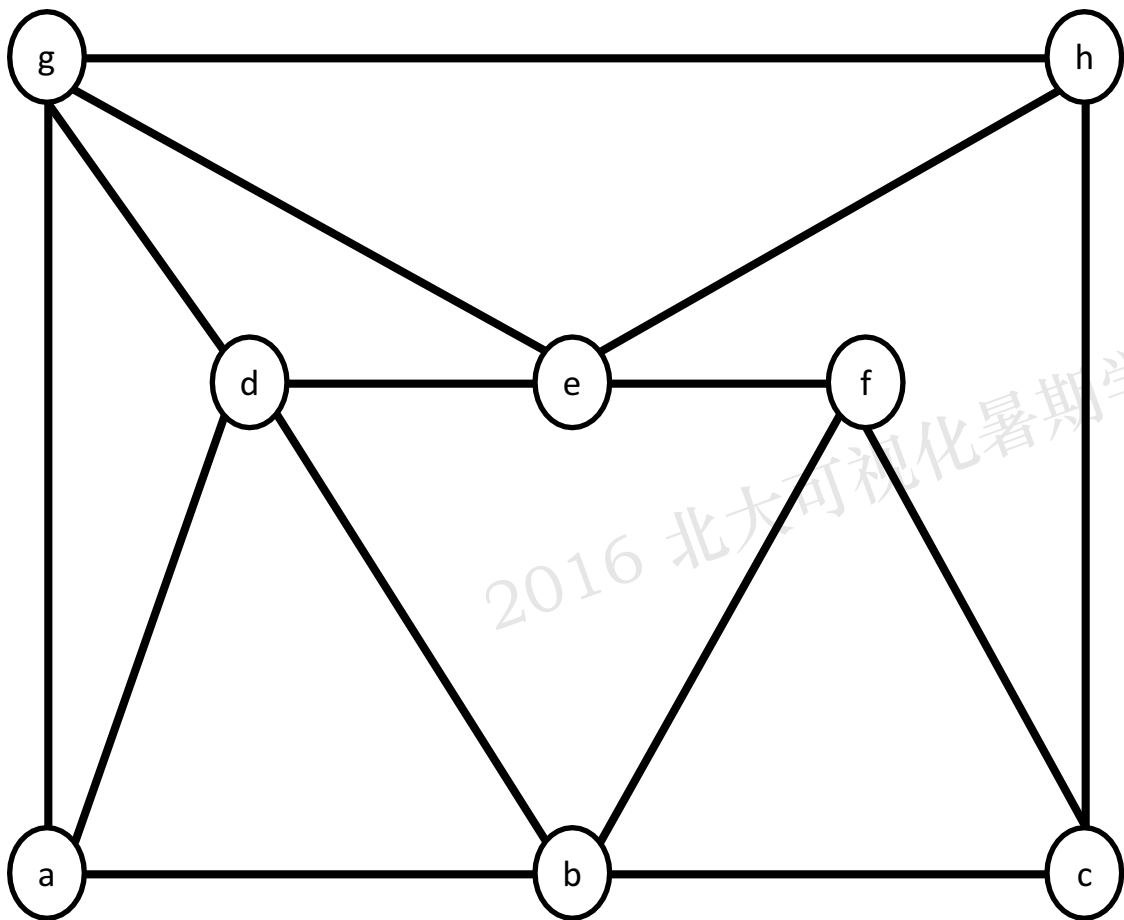


Visibility algorithm

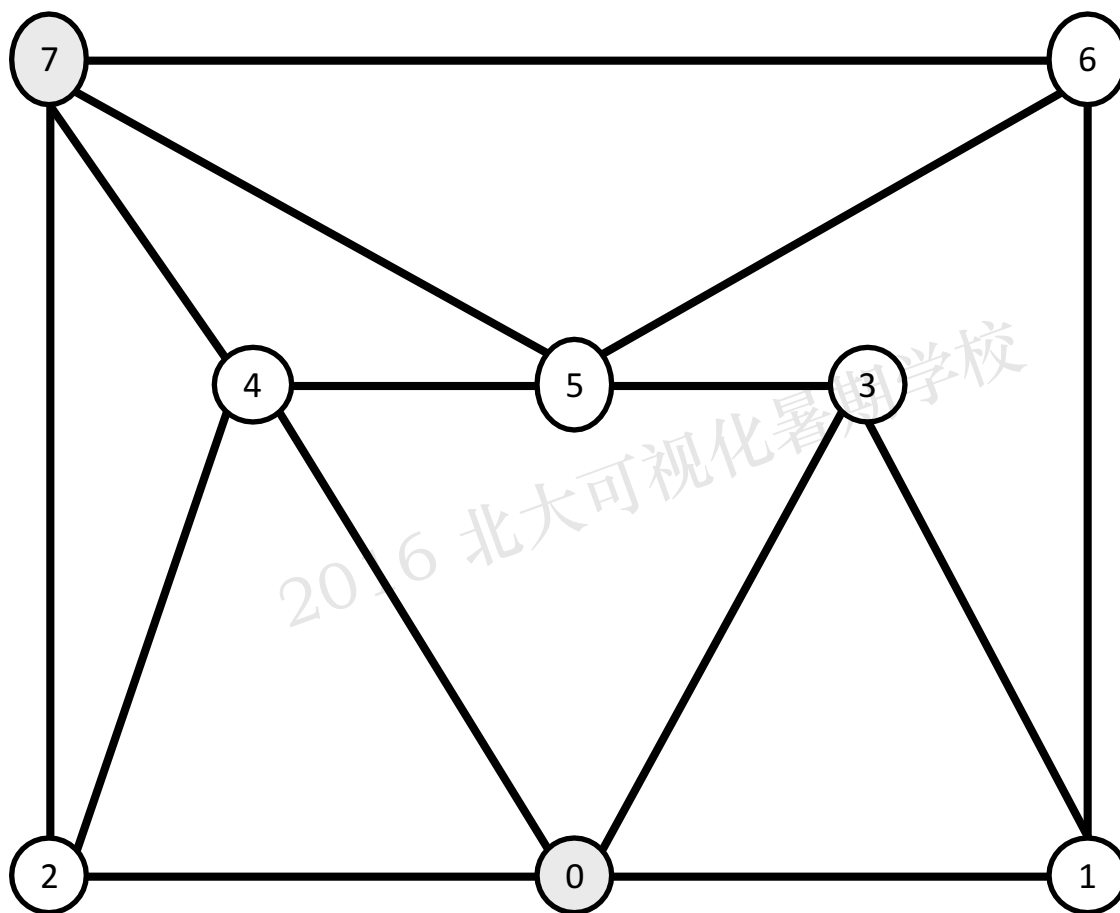
Input: 2-connected topological embedding $G = (V, E)$

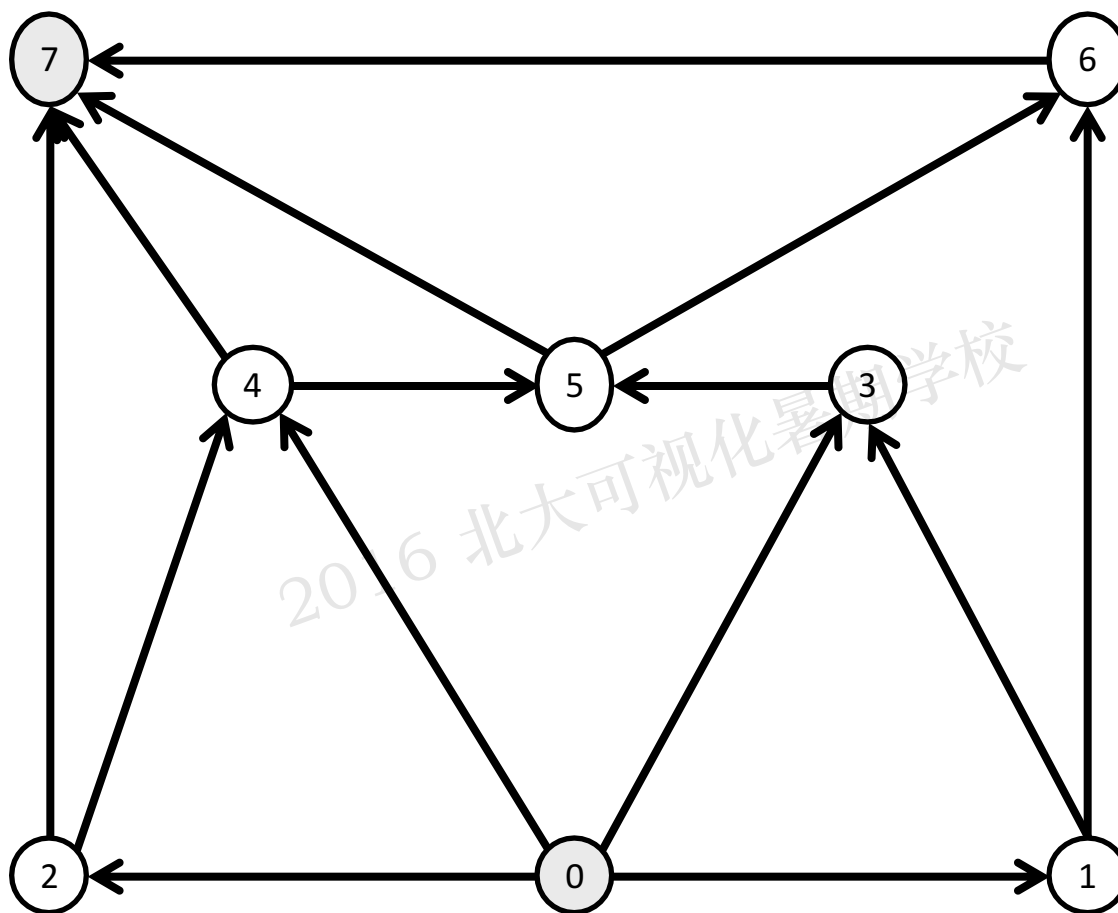
Output: Visibility drawing of G

1. Construct an st-numbering y of G , and direct G according to y .
2. Construct the directed dual D , and topologically sort the nodes of D , to give an x coordinate $x(f)$ for each face f of G .
3. For each edge $e = (u, v) \in E$:
Let f_e be the face to the left of e
Draw e as a vertical line segment from $(x(f_e), y(u))$ to $(x(f_e), y(v))$.
4. For each vertex u in V :
Let $x_{min}(u) = \min_e(x(f_e))$ over all edges e incident to u
Let $x_{max}(u) = \max_e(x(f_e))$ over all edges e incident to u
Draw u as a horizontal line segment from $(x_{min}(u), y(u))$ to $(x_{max}(u), y(u))$.
5. Convert the visibility drawing to an orthogonal drawing.

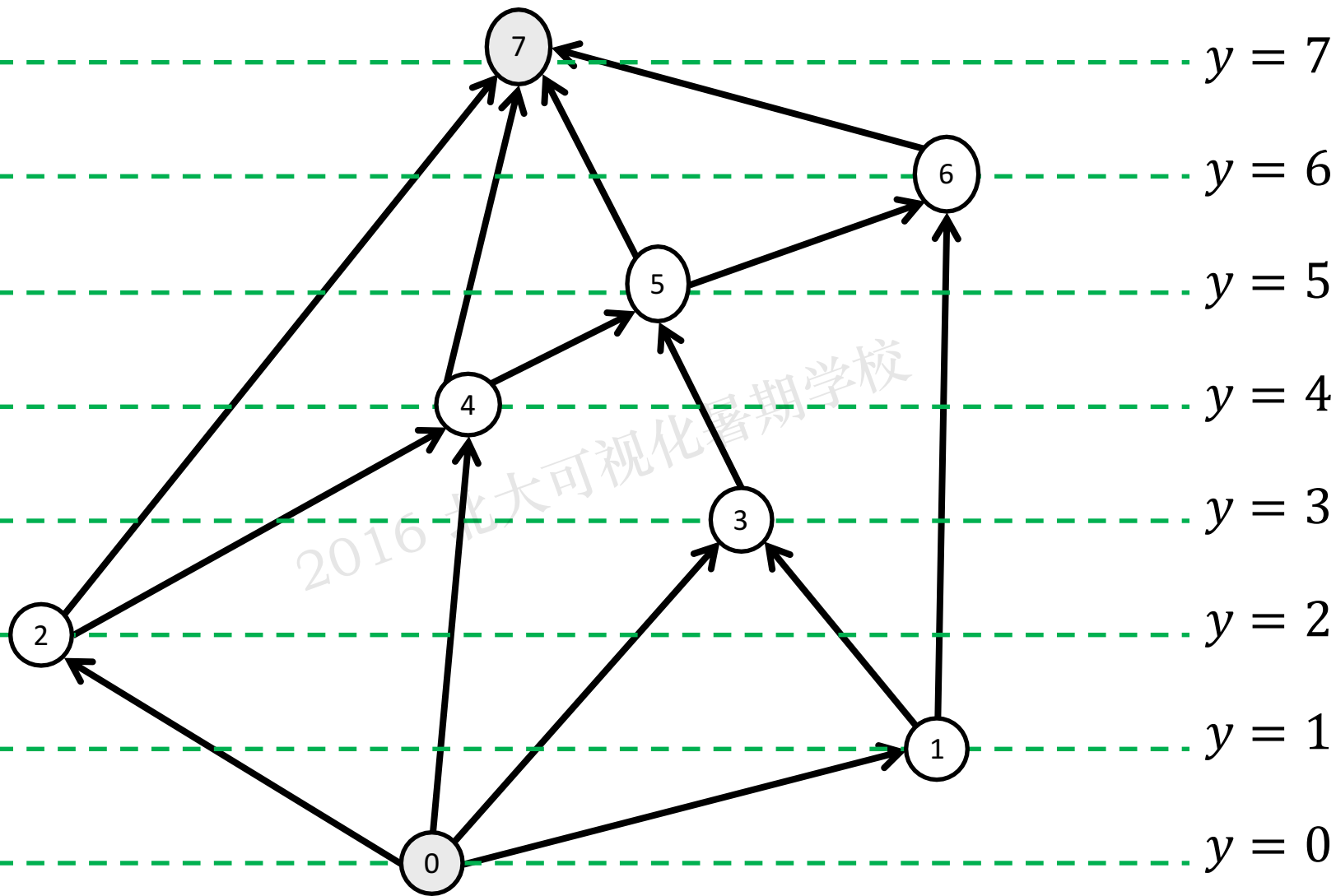


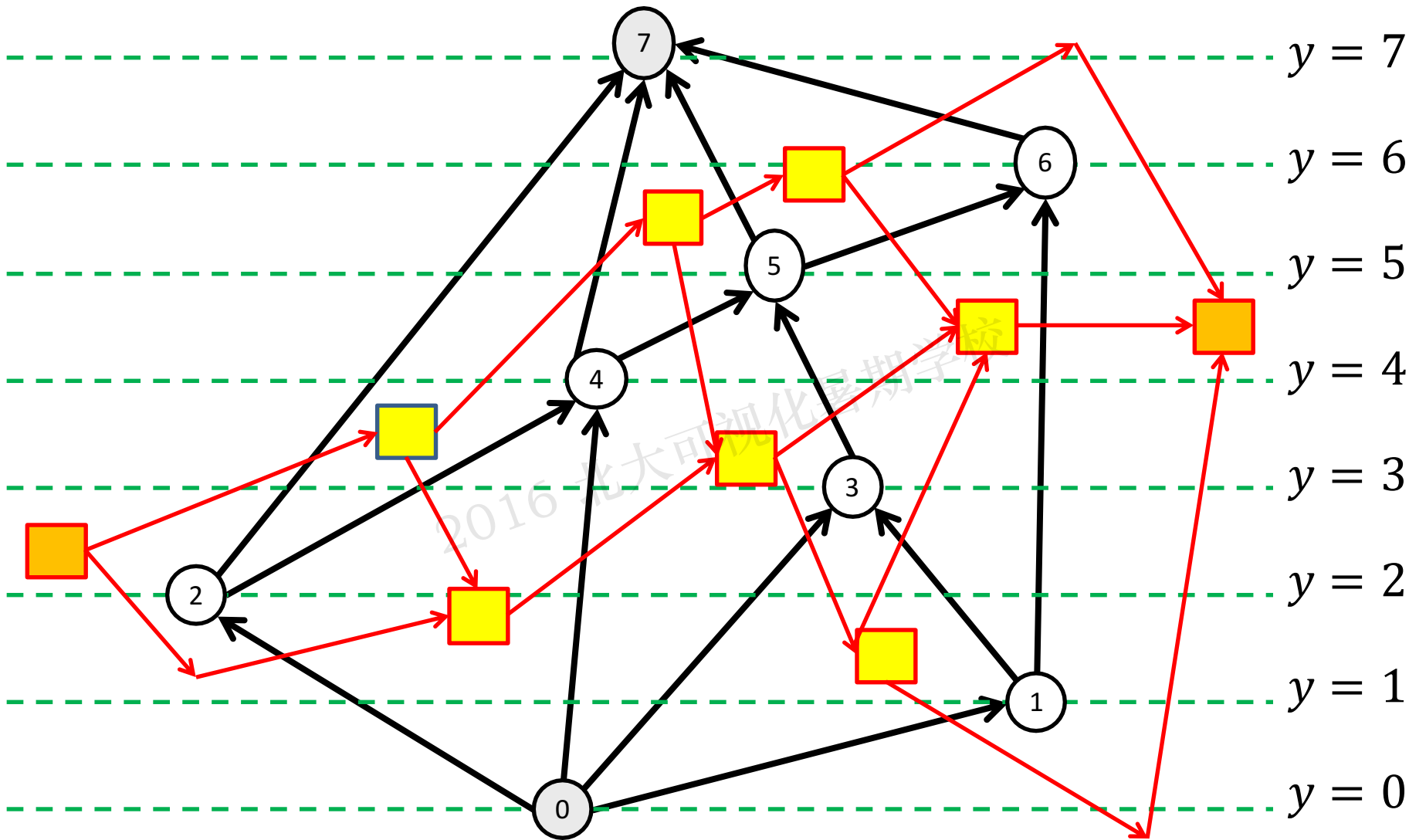
a	b,g,d
b	a,d,f,c
c	b,f,h
d	a,g,e,b
e	d,g,h,f
f	c,b,e
g	a,h,e,d
h	c,e,g

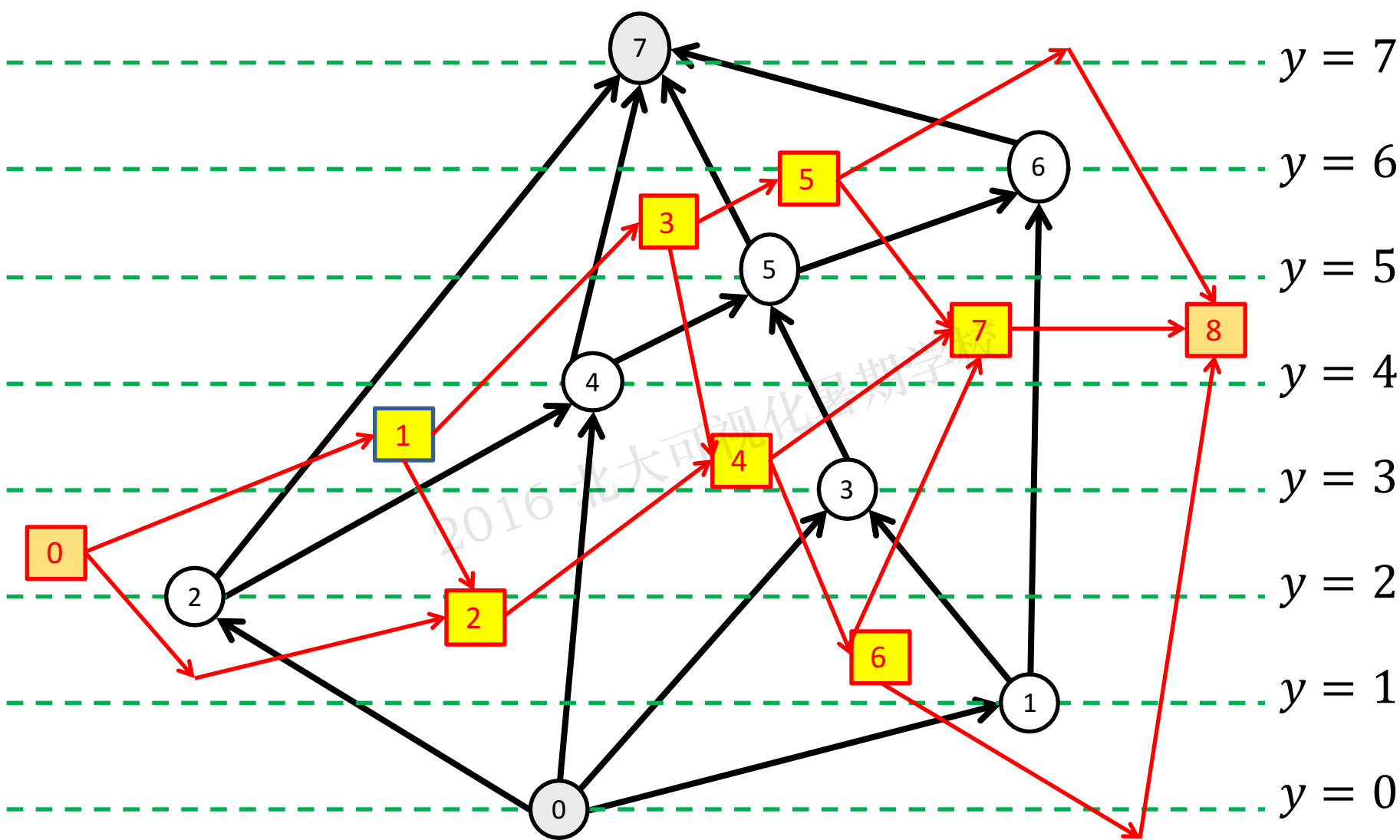


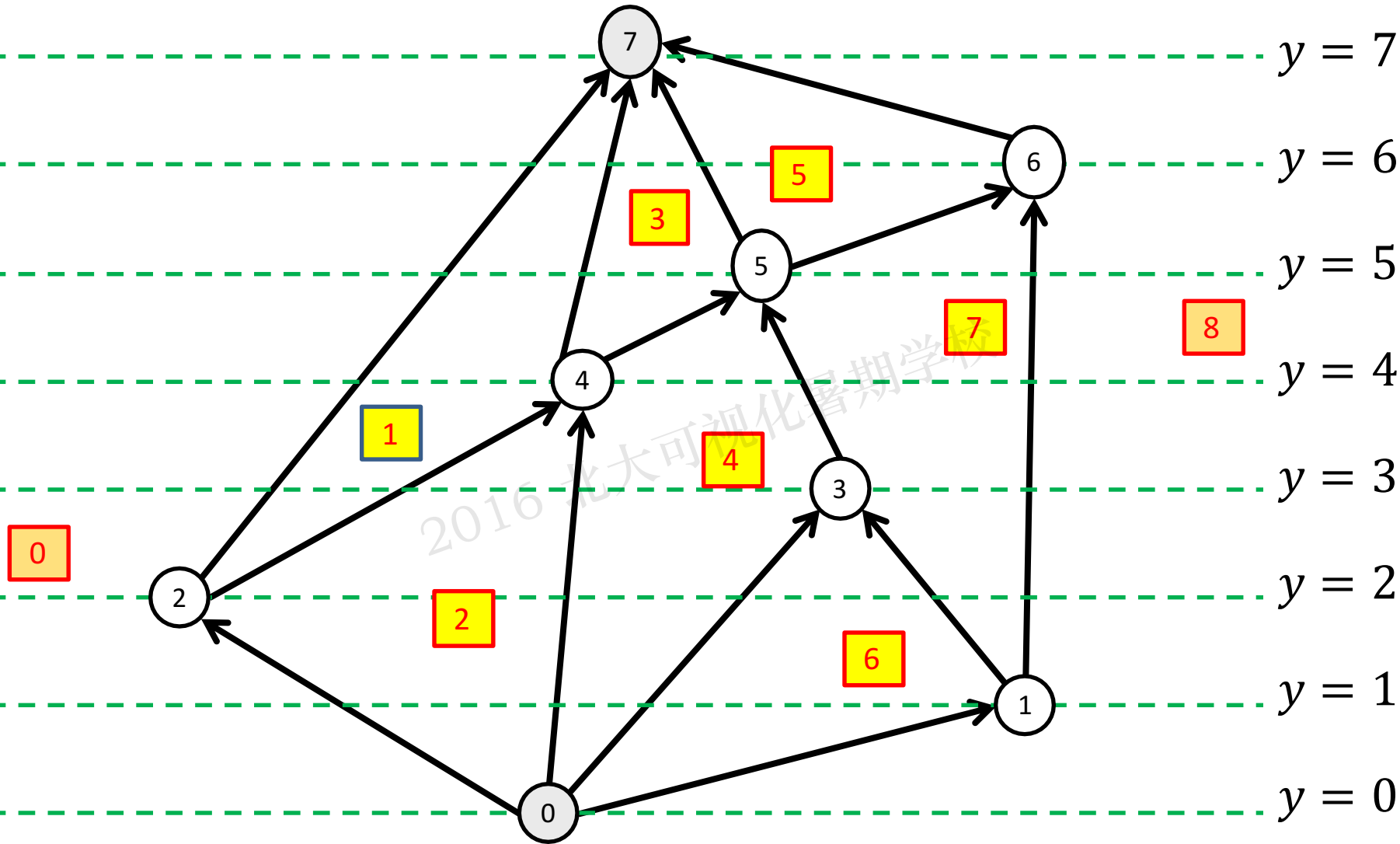


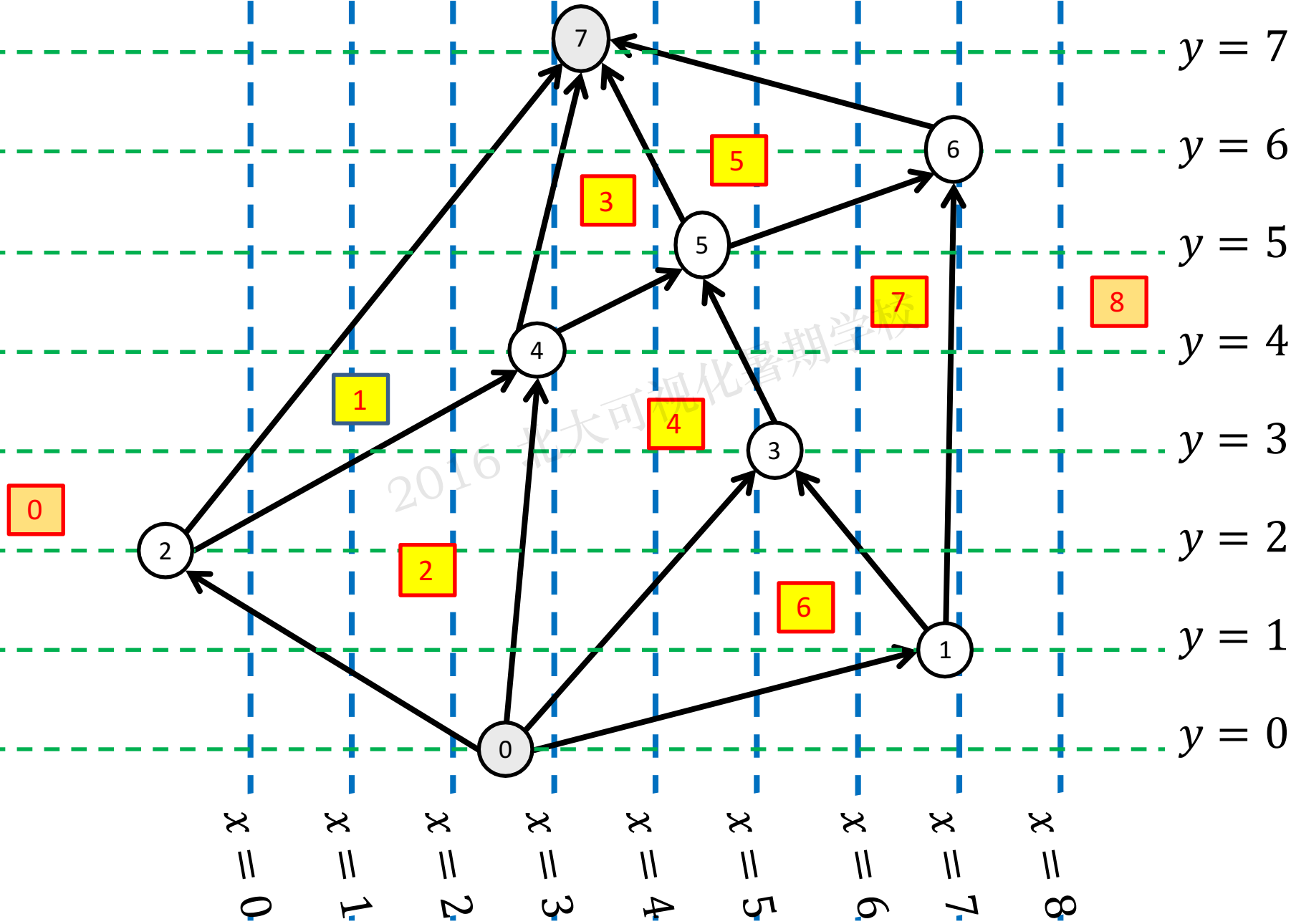
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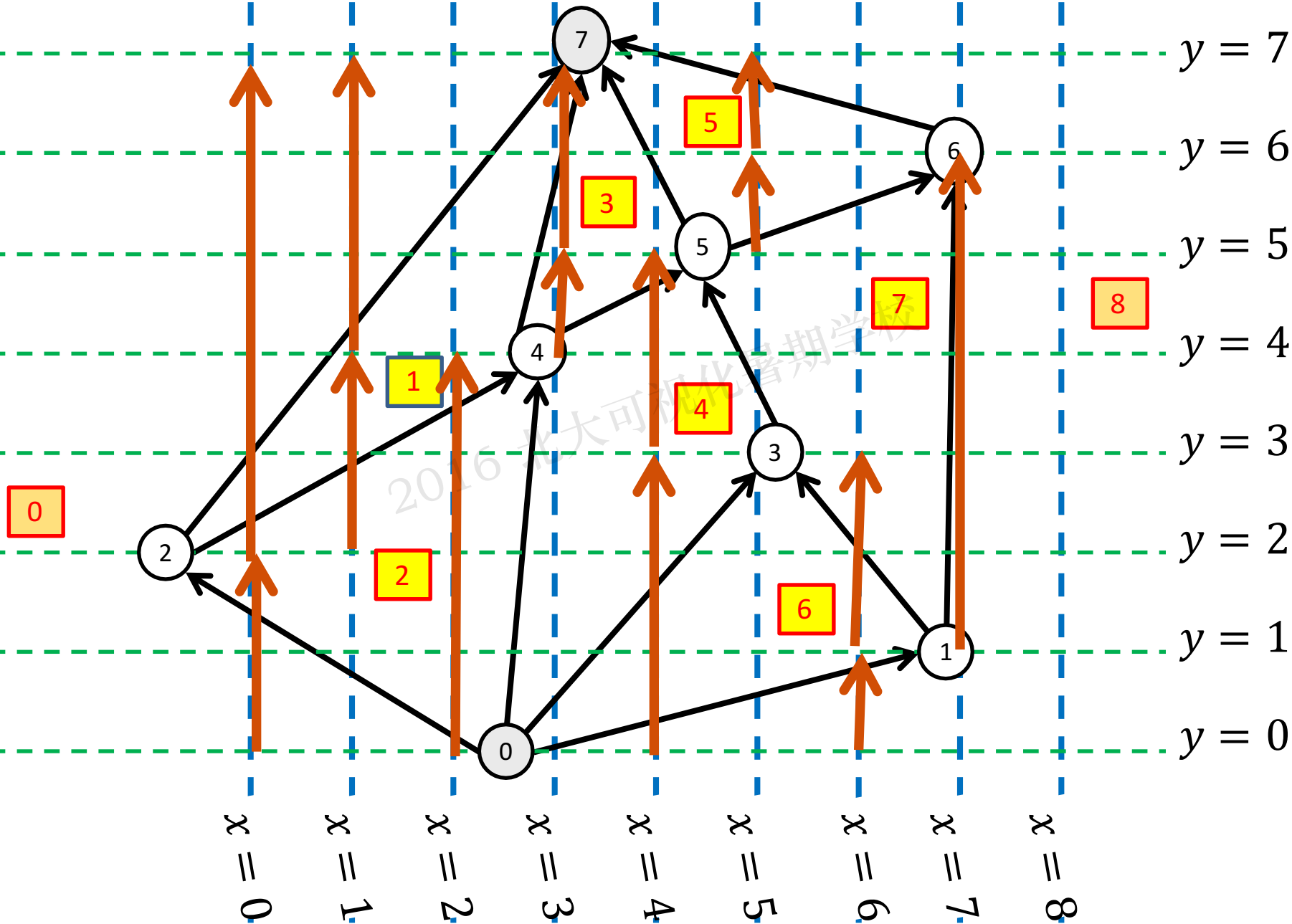


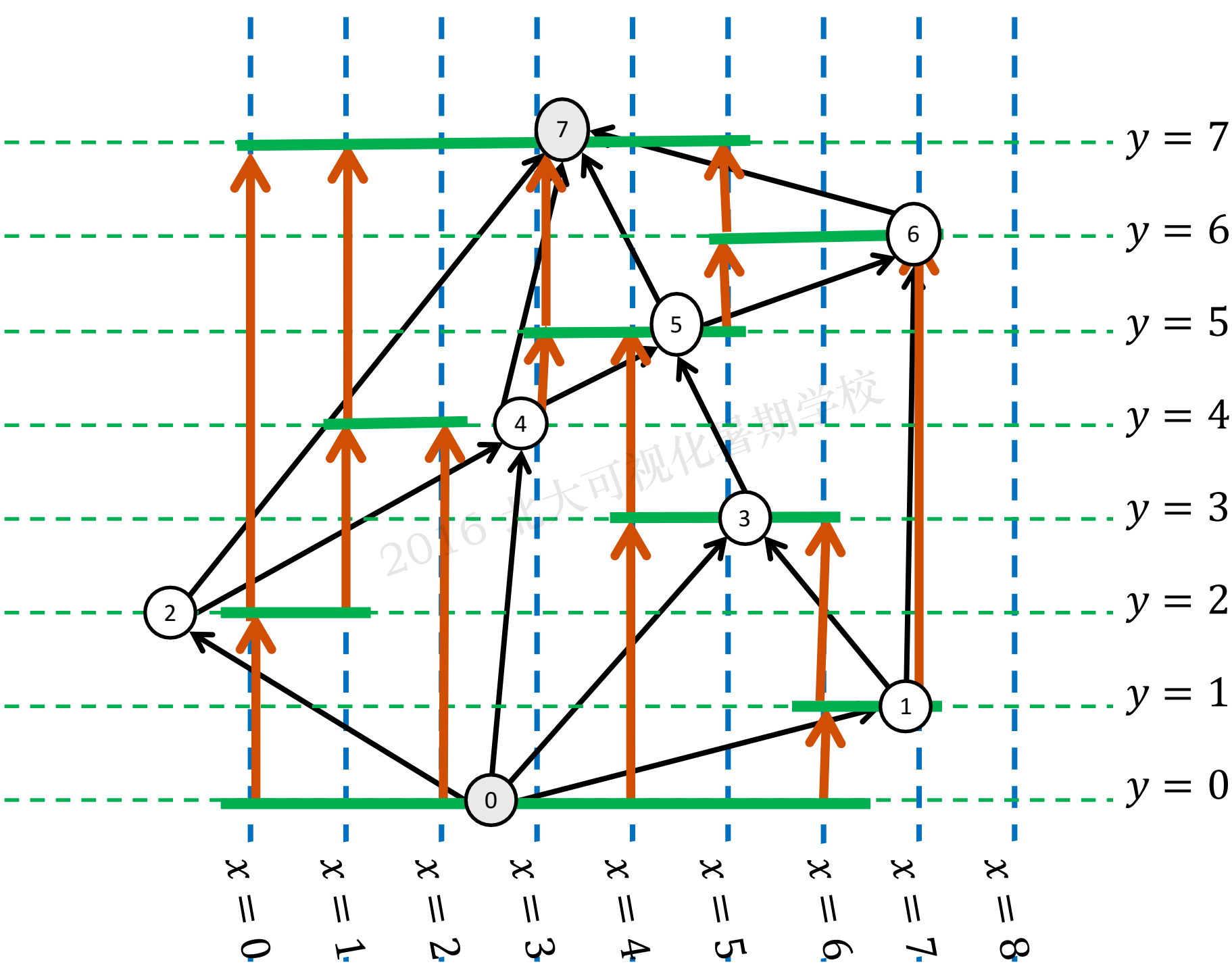


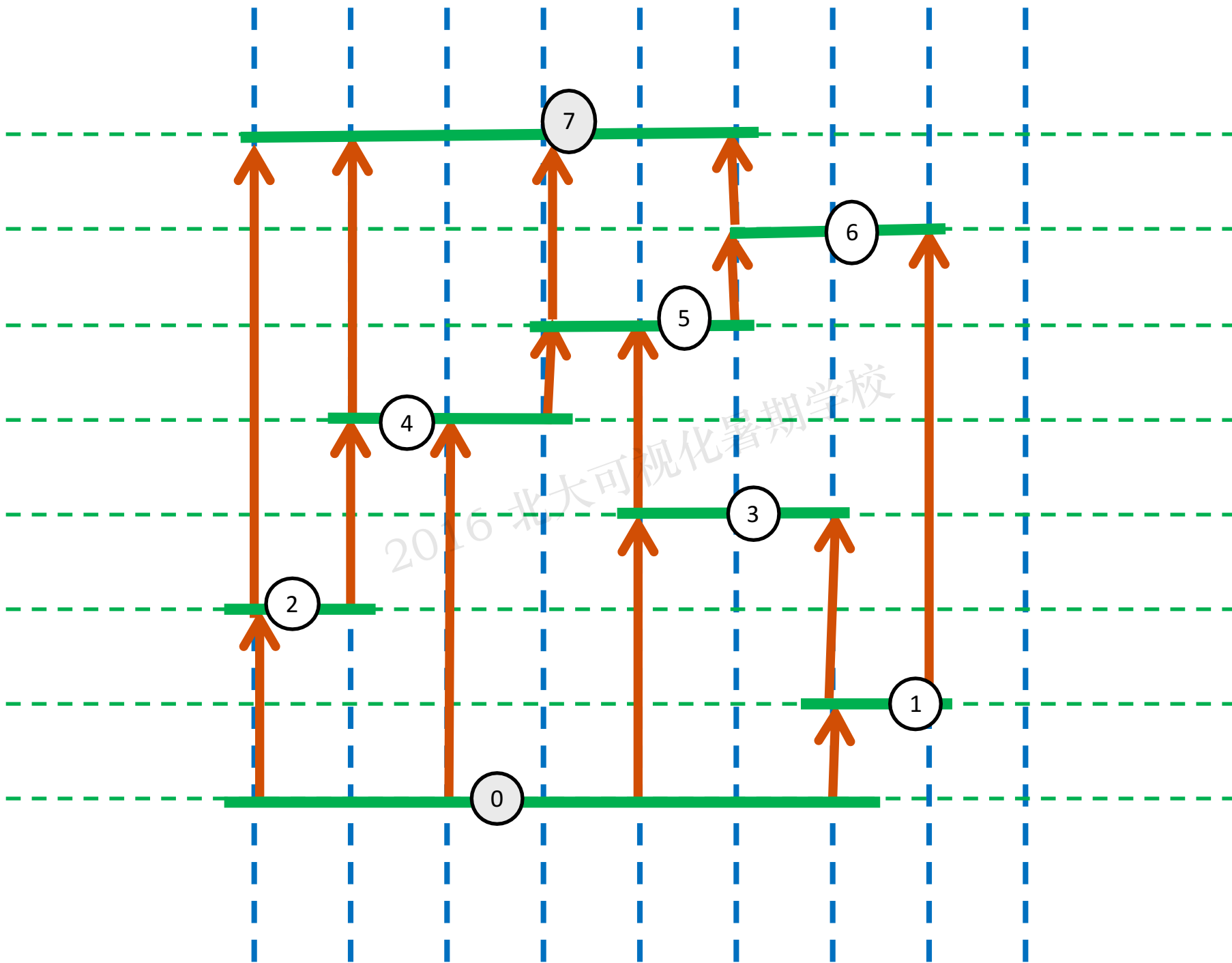


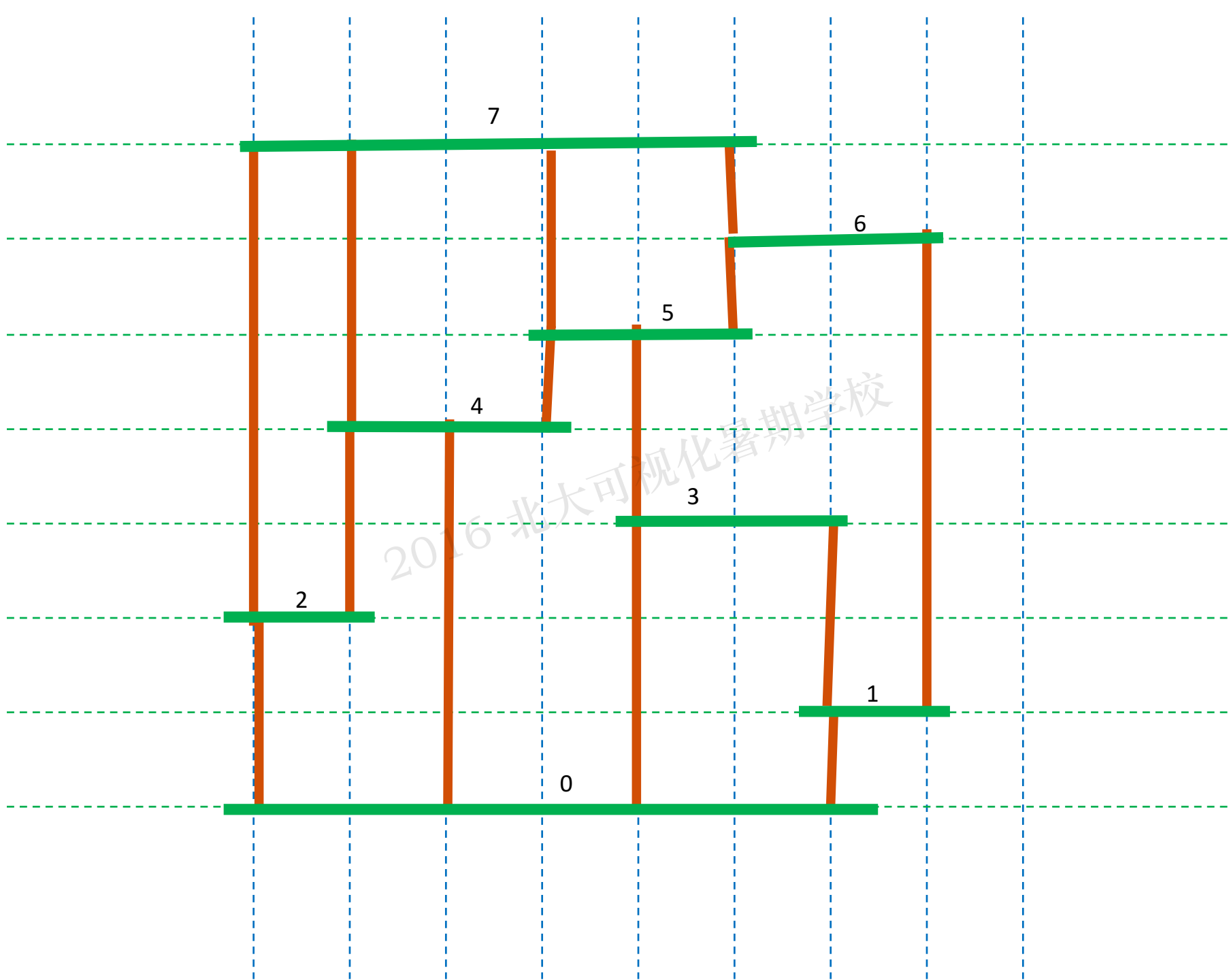


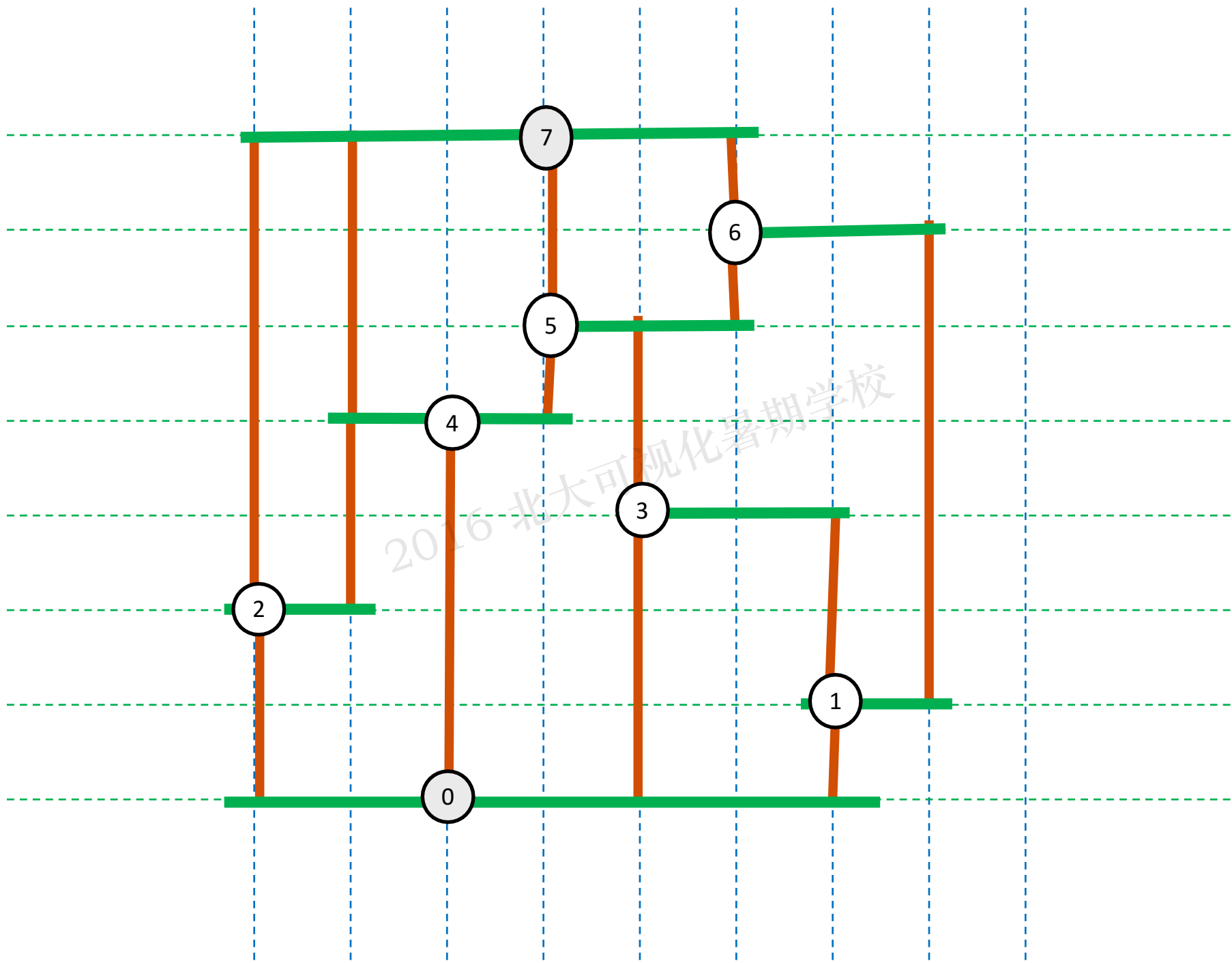


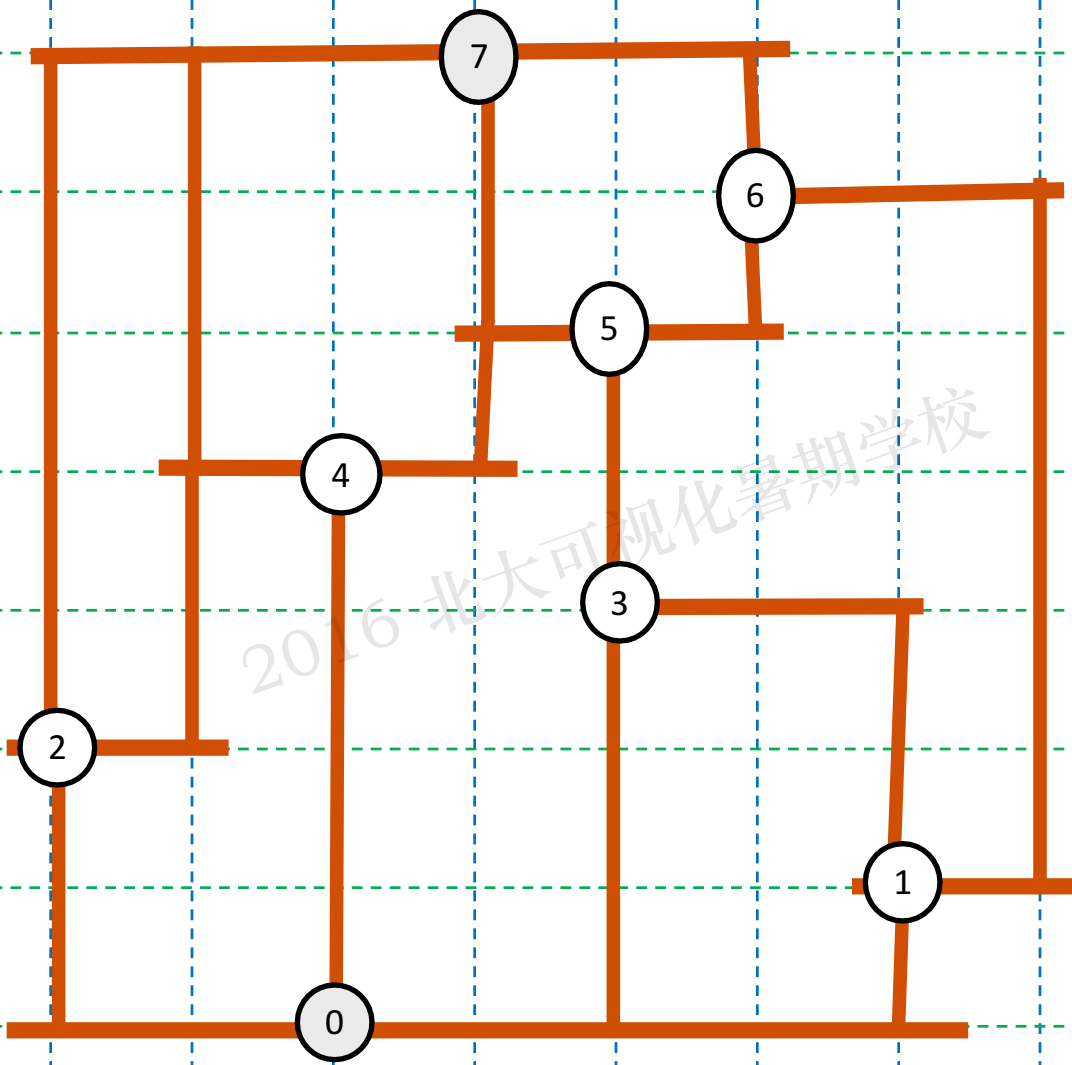


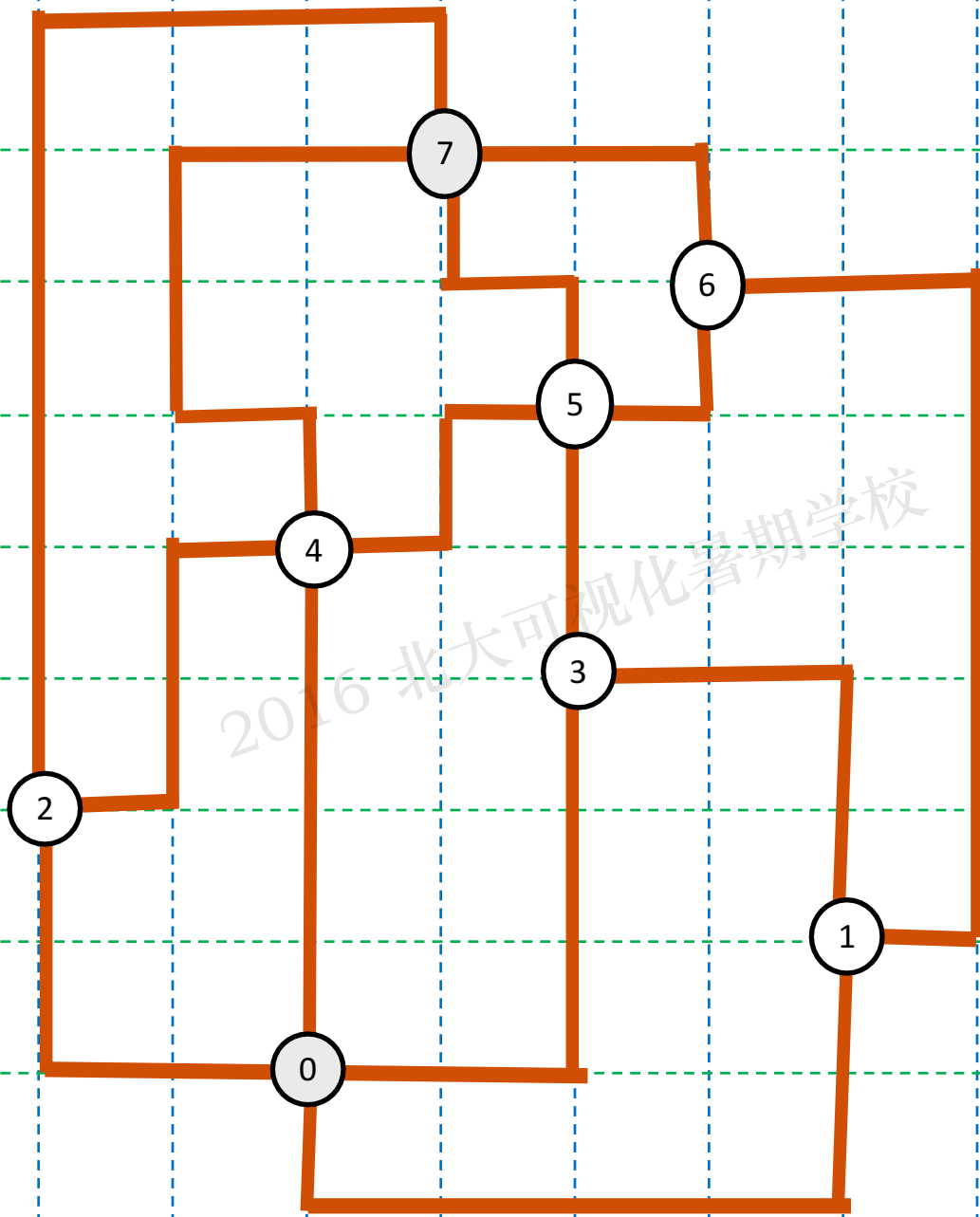




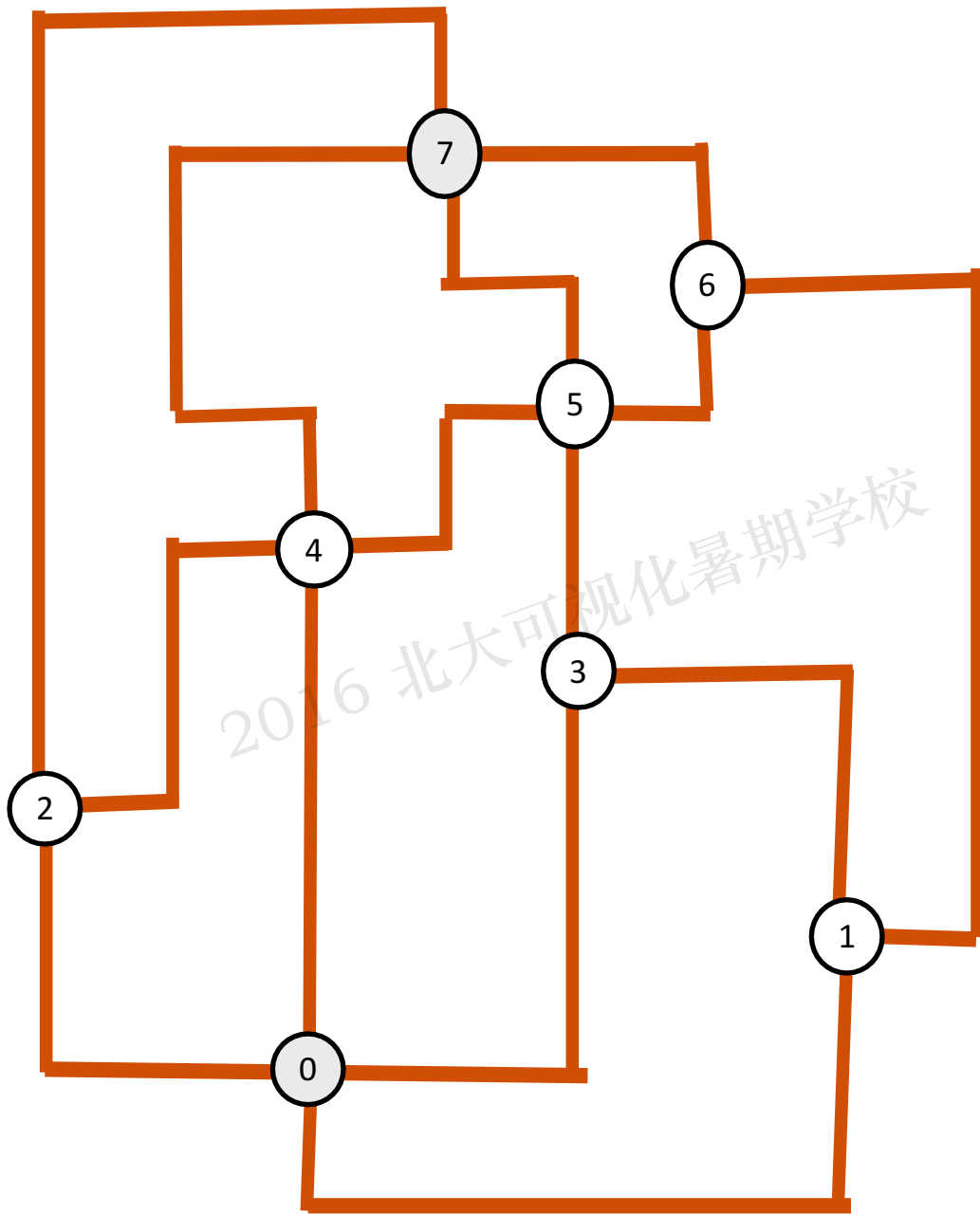








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Topology-shape-metrics approach:

Input: a graph G

Algorithm:

1. Topology: Compute a good topological *embedding* of G
2. Shape: Compute a good orthogonal *shape* for this topological embedding
3. Metrics: Compute a good orthogonal grid drawing of G

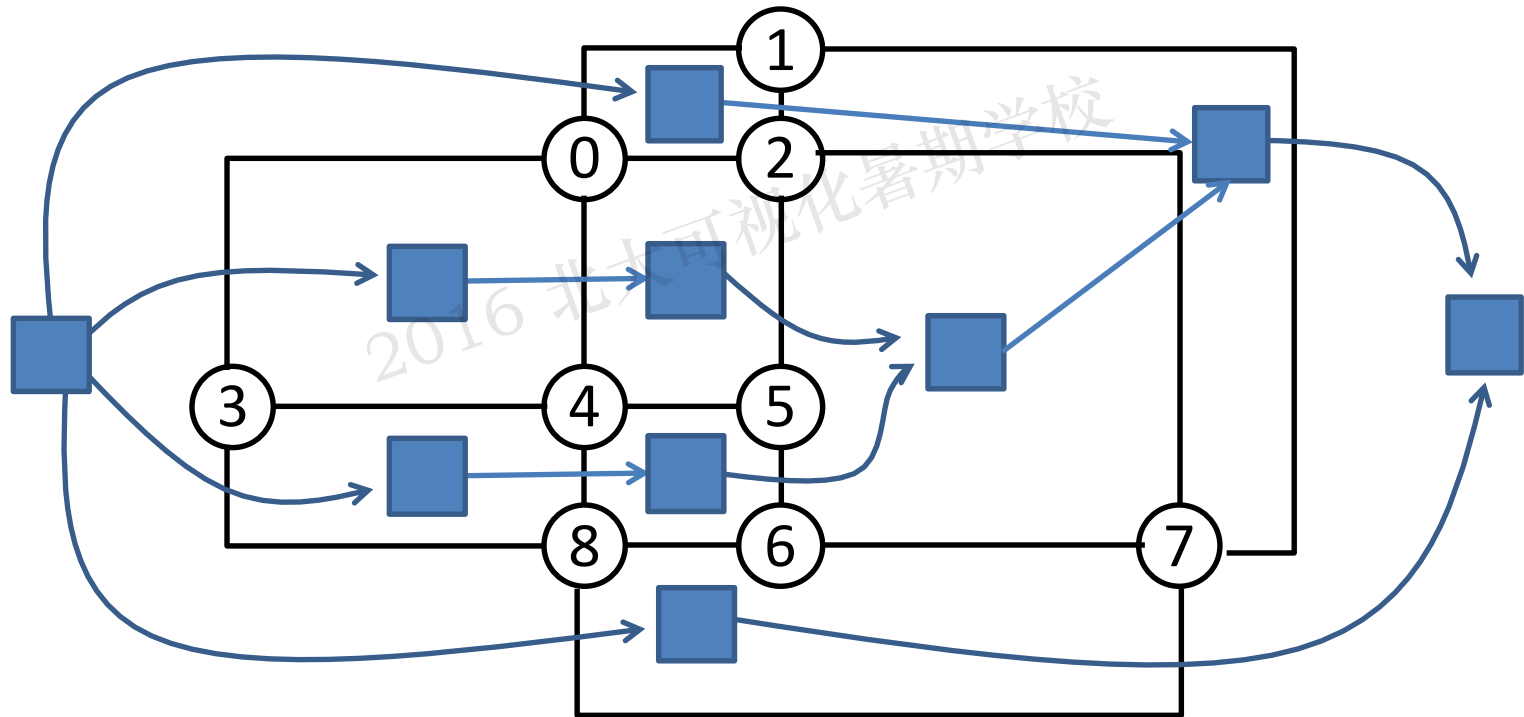
Output: an *orthogonal grid* drawing of G

Aim: give a drawing with good vertex resolution

Metrics step: Use VLSI-inspired compaction methods to get a drawing on a small grid

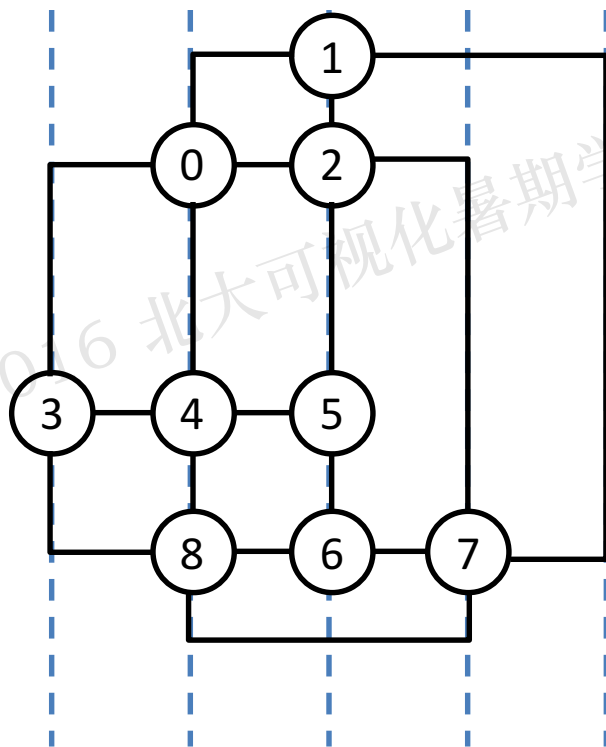
Compaction in the x direction

1. Construct a directed visibility graph H on the dual with source at the left and sink at the right.
2. For each vertex u in H , find a longest path in H from the source to u .
3. Assign y -coordinates using path-length from the source.



Compaction in the x direction

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Similarly compact in the y -direction.

Remarks

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Topology-shape-metrics method:

Input: a graph G

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2. Shape: Compute a good orthogonal shape for this topological embedding
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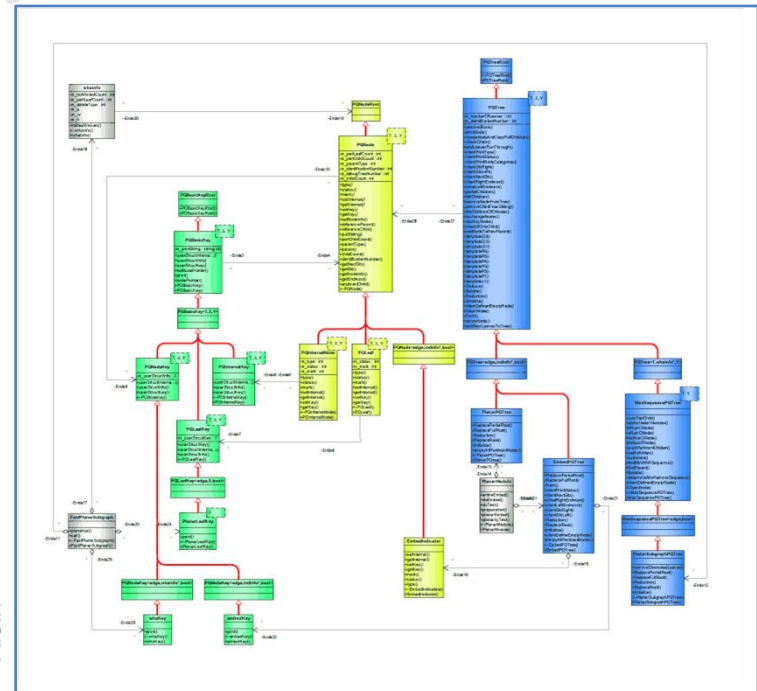
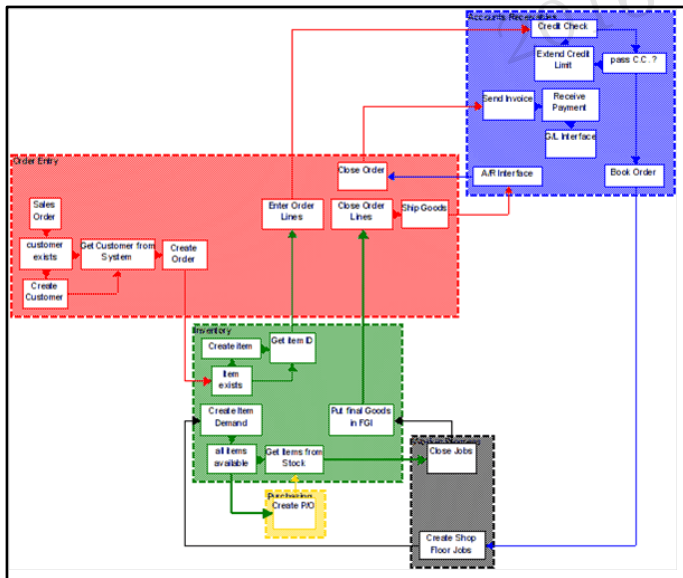
Output: an *orthogonal grid* drawing of G

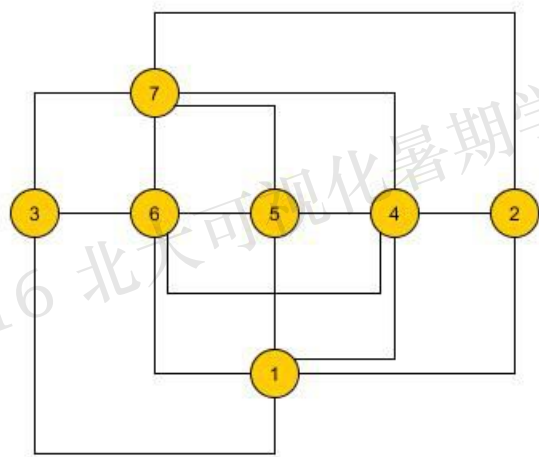
Is this method any good?

Topology-shape-metrics approach

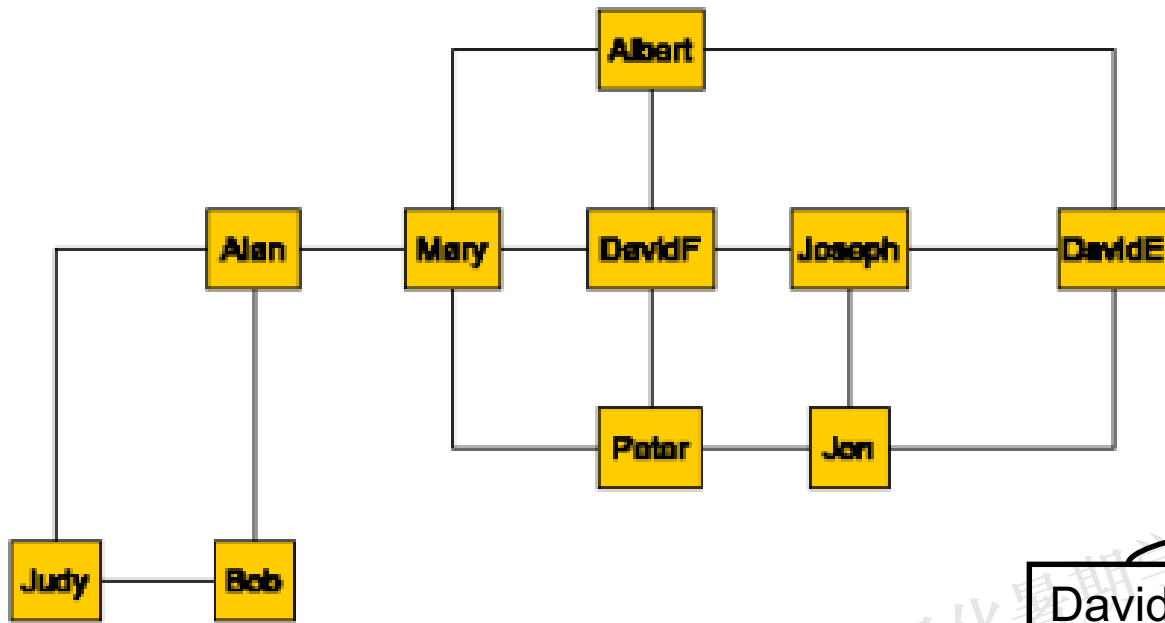
Good things

- Works well on small graphs
- Relatively fast (varies from $O(n)$ to $O(n^2 \log n)$)
- Validated readability
- Can be adjusted to handle vertices of large degree and large size
- Can be adjusted for some constraints

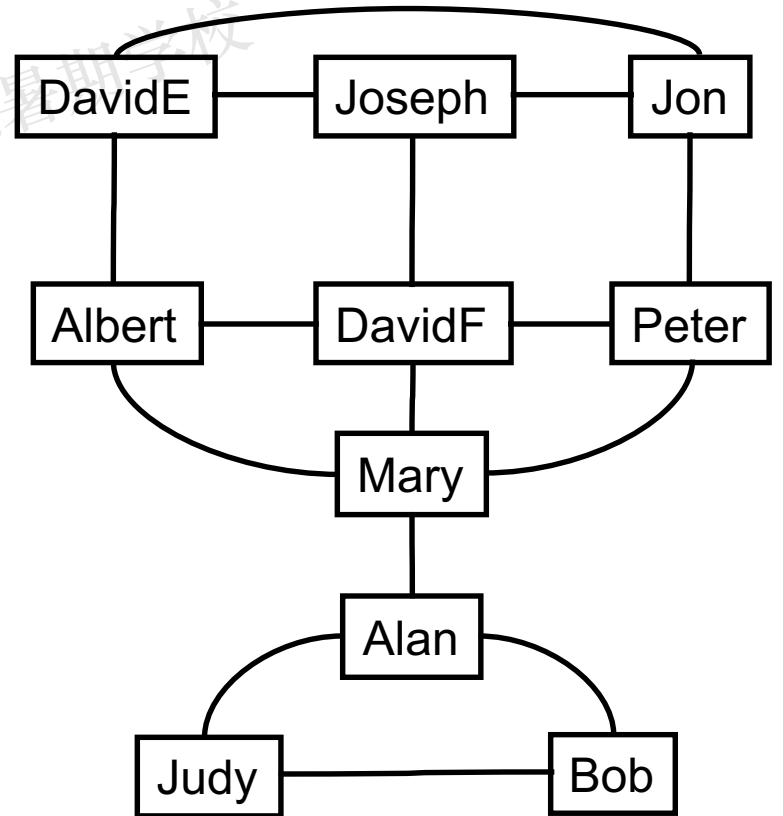


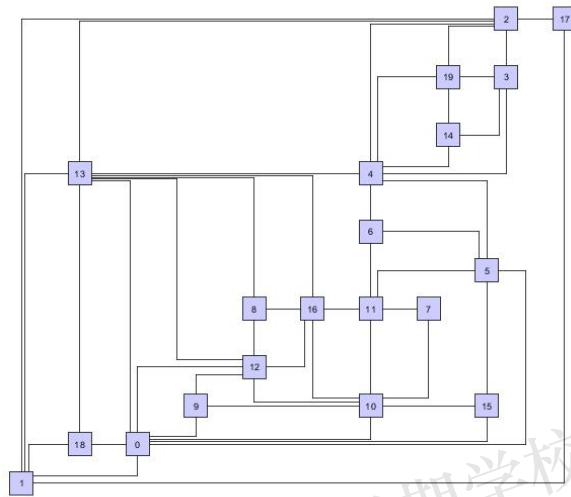
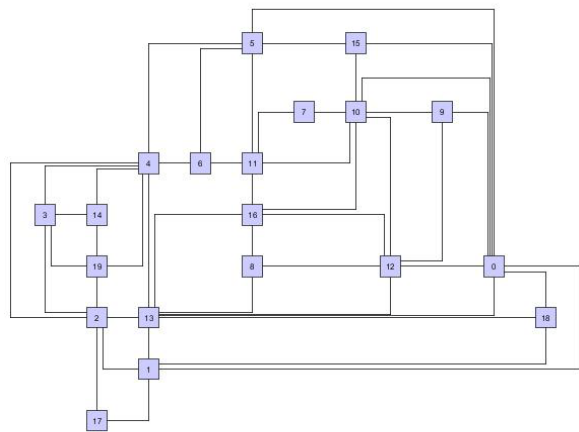


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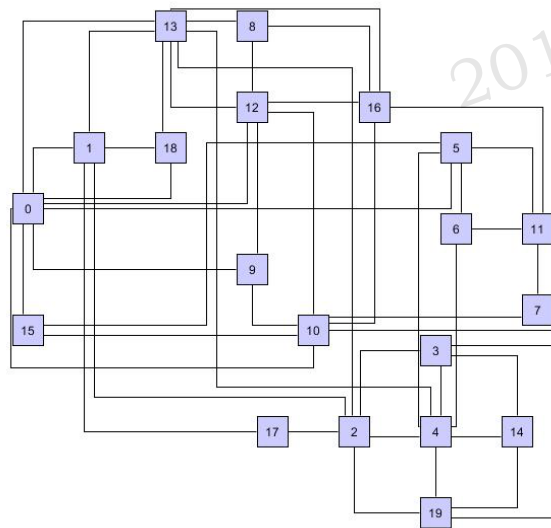


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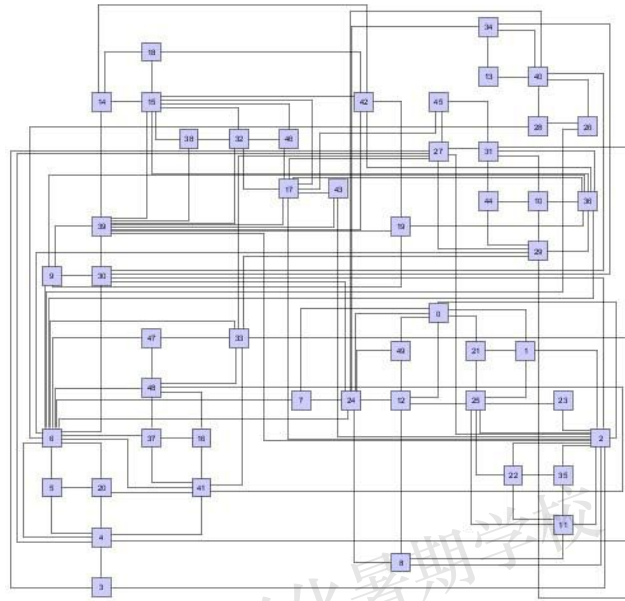
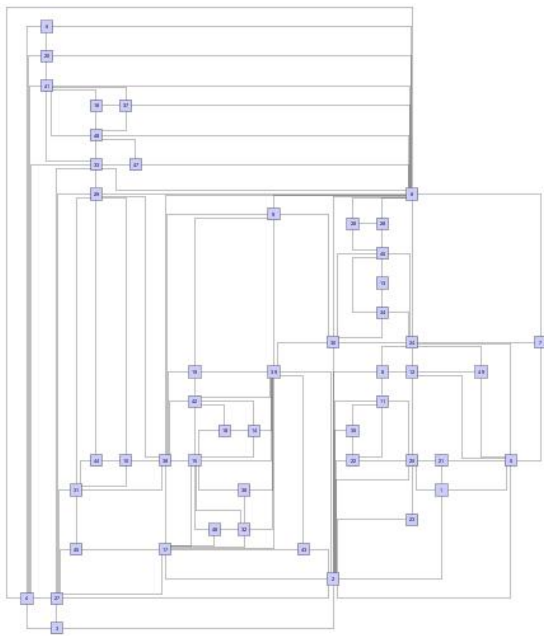




$$n = 20$$

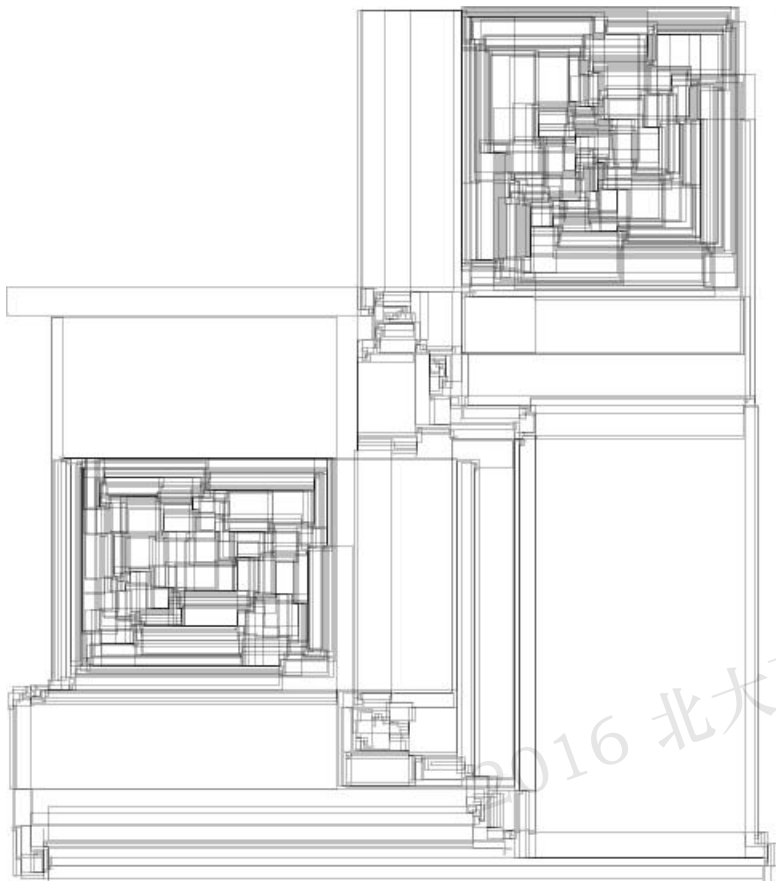


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$$n = 50$$

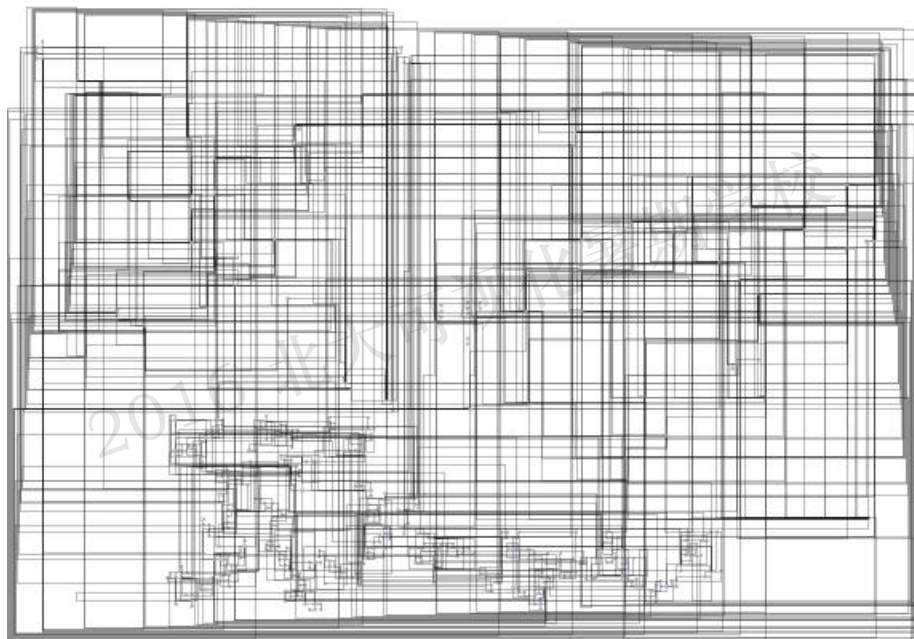
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$n = 500$

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$n = 500$



Topology-shape-metrics approach

Bad things

- Large drawings often look bad (poor faithfulness?)
- Very difficult to code

