

1 Definitions

✍ **G(V, E)** where **G ::= graph**, **V = V(G) ::= finite nonempty set of vertices**, and **E = E(G) ::= finite set of edges**.

✍ **Undirected graph:** $(v_i, v_j) = (v_j, v_i) ::= \text{the same edge}$.

✍ **Directed graph (digraph):** $\langle v_i, v_j \rangle ::= \begin{matrix} v_i \\ \rightarrow \\ v_j \end{matrix} \neq \langle v_j, v_i \rangle$

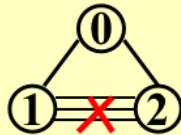


✍ **Restrictions :**

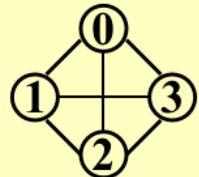
(1) **Self loop** is illegal.



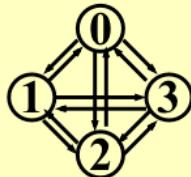
(2) **Multigraph** is not considered



✍ **Complete graph:** a graph that has the maximum number of edges

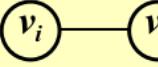


$$\begin{array}{l} \# \text{ of } V = n \Rightarrow \\ \# \text{ of } E = C_n^2 = \frac{n(n-1)}{2} \end{array}$$



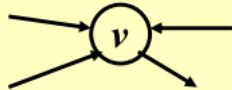
$$\begin{array}{l} \# \text{ of } V = n \Rightarrow \\ \# \text{ of } E = P_n^2 = n(n-1) \end{array}$$

- 自己指向自己是不合法的
- 多重指向也是不考虑的
- 无向图 / 有向图

- ✍  **v_i and v_j are adjacent ;
 (v_i, v_j) is incident on v_i and v_j**
- ✍  **v_i is adjacent to v_j ; v_j is adjacent from v_i ;
 $<v_i, v_j>$ is incident on v_i and v_j**
- ✍ **Subgraph $G' \subset G ::= V(G') \subseteq V(G) \ \&\& \ E(G') \subseteq E(G)$**
- ✍ **Path ($\subset G$) from v_p to v_q ::= $\{v_p, v_{i1}, v_{i2}, \dots, v_{in}, v_q\}$ such that (v_p, v_{i1}) ,
 $(v_{i1}, v_{i2}), \dots, (v_{in}, v_q)$ or $<v_p, v_{i1}>, \dots, <v_{in}, v_q>$ belong to $E(G)$**
- ✍ **Length of a path ::= number of edges on the path**
- ✍ **Simple path ::= $v_{i1}, v_{i2}, \dots, v_{in}$ are distinct**
- ✍ **Cycle ::= simple path with $v_p = v_q$**
- ✍ **v_i and v_j in an undirected G are connected if there is a path from v_i to v_j
(and hence there is also a path from v_j to v_i)**
- ✍ **An undirected graph G is connected if every pair of distinct v_i and v_j are connected**

- 无向图只考虑相连，而有向图还考虑方向
- 联通图：任意两个节点都能相通（无向图中）

- ✍ (Connected) Component of an undirected $G ::=$ the maximal connected subgraph
- ✍ A tree ::= a graph that is connected and *acyclic*
- ✍ A DAG ::= a directed acyclic graph
- ✍ Strongly connected directed graph $G ::=$ for every pair of v_i and v_j in $V(G)$, there exist directed paths from v_i to v_j and from v_j to v_i . If the graph is connected without direction to the edges, then it is said to be **weakly connected**
- ✍ Strongly connected component ::= the maximal subgraph that is strongly connected
- ✍ $\text{Degree}(v) ::=$ number of edges incident to v . For a directed G , we have **in-degree** and **out-degree**. For example:



$\text{in-degree}(v) = 3$; $\text{out-degree}(v) = 1$; $\text{degree}(v) = 4$

- ✍ Given G with n vertices and e edges, then

$$e = \left(\sum_{i=0}^{n-1} d_i \right) / 2 \quad \text{where } d_i = \text{degree}(v_i)$$

- 最大联通分量：部分联通的最多的子图
- 树其实就是没有cycle的graph
- DAG: 有向、非周期
- 强联通图：任意两个点都存在互相的路径（有向图）
- 强联通分量：最大联通的子图
- indegree: 需要所有节点; outdegree: 当前节点就可
- 边的个数 = (点赞 + 收赞) / 2

1.1 Representation of Graphs

1.1.1 Adjacency Matrix

adj_mat [n] [n] is defined for $G(V, E)$ with n vertices, $n \geq 1$:

$$\text{adj_mat}[i][j] = \begin{cases} 1 & \text{if } (v_i, v_j) \text{ or } (v_j, v_i) \in E(G) \\ 0 & \text{otherwise} \end{cases}$$

Note: If G is undirected, then $\text{adj_mat}[][]$ is symmetric.

Thus we can save space by storing only half of the matrix.

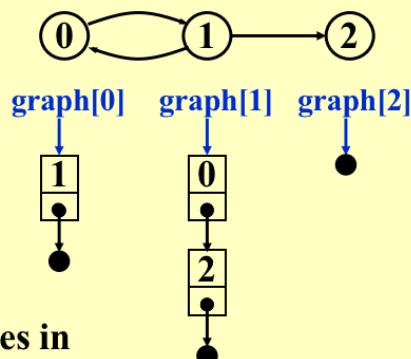
$$\begin{aligned} \text{degree}(i) &= \sum_{j=0}^{n-1} \text{adj_mat}[i][j] \quad (\text{if } G \text{ is undirected}) \\ &\quad + \sum_{j=0}^{n-1} \text{adj_mat}[j][i] \quad (\text{if } G \text{ is directed}) \end{aligned}$$

- 如果是无向的，那么只用存一半就行了

1.1.2 Adjacency Lists

【Example】

$$\text{adj_mat}[3][3] = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$



Note: The order of nodes in each list does not matter.

For undirected G :

$$S = n \text{ heads} + 2e \text{ nodes} = (n+2e) \text{ ptrs} + 2e \text{ ints}$$

- replace each row by a linked list

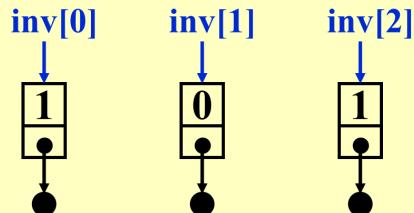
Degree(i) = number of nodes in graph[i] (if G is undirected).

T of examine $E(G) = O(n + e)$

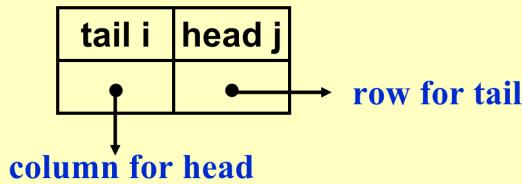
If G is **directed**, we need to find $\text{in-degree}(v)$ as well.

Method 1 Add inverse adjacency lists.

【Example】



Method 2 Multilist (Ch 3.2) representation for adj_mat[i] [j]



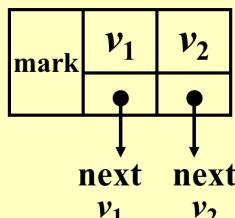
- 其实被指向就是指向矩阵的转置

1.1.3 Adjacency Multilists

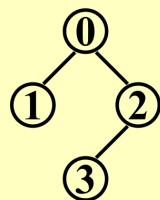
Adjacency Multilists

In adjacency list, for each (i, j) we have two nodes:

$\text{graph}[i] \rightarrow \boxed{j}$ $\bullet \rightarrow \dots \dots$ Now let's combine the two nodes
 $\text{graph}[j] \rightarrow \boxed{i}$ $\bullet \rightarrow \dots \dots$ into one: $\text{graph}[i] \rightarrow \boxed{\text{node}}$ $\leftarrow \text{graph}[j]$

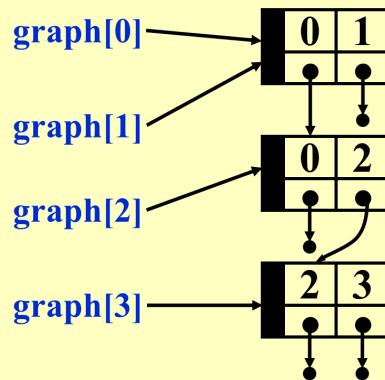


【Example】



Weighted Edges

- $\text{adj_mat}[i][j] = \text{weight}$
- adjacency lists \ multilists : add a **weight** field to the node.



- 每个节点实际上是一条边

2 Topological Sort 拓扑排序

- 其实不是一个排序算法，是一个调度算法

AOV Network ::= digraph G in which V(G) represents activities (e.g. the courses) and E(G) represents precedence relations (e.g. $C_1 \rightarrow C_3$ means that C_1 is a prerequisite course of C_3).

i is a predecessor of j ::= there is a path from i to j
 i is an immediate predecessor of j ::= $\langle i, j \rangle \in E(G)$
 Then j is called a successor (immediate successor) of i

Partial order ::= a precedence relation which is both transitive ($i \rightarrow k, k \rightarrow j \Rightarrow i \rightarrow j$) and irreflexive ($i \rightarrow i$ is impossible).

Note: If the precedence relation is reflexive, then there must be an i such that i is a predecessor of i . That is, i must be done before i is started. Therefore if a project is feasible, it must be irreflexive.

Feasible AOV network must be a dag (directed acyclic graph).

A topological order is a linear ordering of the vertices of a graph such that, for any two vertices, i, j , if i is a predecessor of j in the network then i precedes j in the linear ordering.

Note: The topological orders may not be unique for a network. For example, there are several ways (topological orders) to meet the degree requirements in computer science.

```

1 void Topsort( Graph G )
2 {   int Counter;
3   Vertex V, W;
4   for ( Counter = 0; Counter < NumVertex; Counter ++ ) {
5     V = FindNewVertexOfDegreeZero( ); // 每次要扫描中间优先级空的
6     if ( V == NotAVertex ) {
7       Error ( "Graph has a cycle" );
8       break;
9     }
10    TopNum[ V ] = Counter; /* or output V */
11    for ( each W adjacent to V )
12      // 和V这个节点有关系的其要求都会-1, 一直减到0, 那么下次循环就会被塞到
13      Indegree[ W ] -- ;
14  }
15 }
```

$$T = O(|V|^2) \quad (6)$$

Improvement: Keep all the unassigned vertices of degree 0 in a special box (queue or stack).

```

1 | void Topsort( Graph G )
2 | { Queue Q;
3 |   int Counter = 0;
4 |   Vertex V, W;
5 |   Q = CreateQueue( NumVertex ); MakeEmpty( Q );
6 |   // 入度为0的, 说明不需要准备的, 直接纳入queue
7 |   for ( each vertex V )
8 |     if ( Indegree[ V ] == 0 ) Enqueue( V, Q );
9 |
10|   while ( !IsEmpty( Q ) ) {
11|     V = Dequeue( Q );
12|     TopNum[ V ] = ++ Counter; /* assign next */
13|     for ( each W adjacent to V )
14|       if ( -- Indegree[ W ] == 0 ) Enqueue( W, Q );
15|   } /* end-while */
16|
17|   if ( Counter != NumVertex )
18|     Error( "Graph has a cycle" );
19|   DisposeQueue( Q ); /* free memory */
20| }
```

$$T = O(|V|) \quad (7)$$

- 这个算法会要求，上面那个较差的算法是不要求的
- 就是每次不去找已经为0的了，就是队列当前节点对其所有关系点做工作，做完工作顺便判断

3 Shortest Path Algorithm

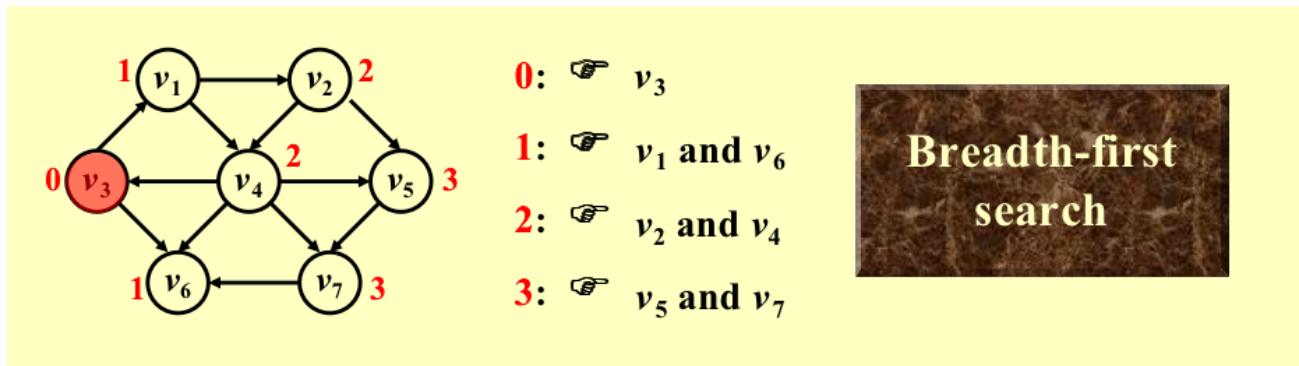
- Given a digraph $G = (V, E)$, and a cost function $c(e)$ for $e \in E(G)$. The length of a path P from source to destination is $\sum_{e_i \in P} c(e_i)$ (also called weighted path length).
- 不带权的最短路就是权都等于1

3.1 Single-Source Shortest-Path Problem

Given as input a weighted graph, $G = (V, E)$, and a distinguished vertex, s , find the shortest weighted path from s to every other vertex in G .

3.1.1 Unweighted Shortest Paths BFS

3.1.1.1 Sketch of the idea



3.1.1.2 Implementation

```

1 Table[ i ].Dist ::= distance from s to vi /* initialized to be infinite
except for s */
2 Table[ i ].Known ::= 1 if vi is checked; or 0 if not
3 Table[ i ].Path ::= for tracking the path /* initialized to be 0 */

```

3.1.1.2.1 basis

```

1 void Unweighted( Table T )
2 {   int CurrDist;
3   Vertex V, W;
4   for ( CurrDist = 0; CurrDist < NumVertex; CurrDist ++ ) {
5     for ( each vertex V )
6       // 这里很冗余, 我做了一次后, 全部再遍历一遍很花时间
7       // 可以直接用queue来处理
8       if ( !T[ V ].Known && T[ V ].Dist == CurrDist ) {
9         T[ V ].Known = true;
10        for ( each W adjacent to V )
11          if ( T[ W ].Dist == Infinity ) {
12            T[ W ].Dist = CurrDist + 1;
13            T[ W ].Path = V;
14          } /* end-if Dist == Infinity */
15        } /* end-if !Known && Dist == CurrDist */

```

```

16 } /* end-for CurrDist */
17 }
18

```

- Worst case: 一条线的时候，时间复杂度到 $T(N) = O(N^2)$

3.1.1.2.2 Improvement (要求)

```

1 void Unweighted( Table T )
2 { /* T is initialized with the source vertex S given */
3     Queue Q;
4     Vertex V, W;
5     Q = CreateQueue( NumVertex ); MakeEmpty( Q );
6     Enqueue( S, Q ); /* Enqueue the source vertex */
7     while ( !IsEmpty( Q ) ) {
8         V = Dequeue( Q );
9         T[ V ].Known = true; /* not really necessary */
10        for ( each W adjacent to V )
11            if ( T[ W ].Dist == Infinity ) {
12                T[ W ].Dist = T[ V ].Dist + 1;
13                T[ W ].Path = V;
14                Enqueue( W, Q );
15            } /* end-if Dist == Infinity */
16        } /* end-while */
17        DisposeQueue( Q ); /* free memory */
18    }

```

时间复杂度：本质上就是每个点每条边都会算一次，那么复杂度就是根据节点个数和边数来定

$$T(N) = O(|V| + |E|) \quad (8)$$

3.1.2 Dijkstra's Algorithm (for weighted shortest paths)

- Let $S = \{ s \text{ and } vi \text{'s whose shortest paths have been found} \}$ 定义一个known set，确定已知的最短
- For any $u \notin S$, define $\text{distance}[u] = \text{minimal length of path } s \rightarrow (vi \in S) \rightarrow u$. If the paths are generated in non-decreasing order, then
 - the shortest path must go through ONLY $v_i \in S$;
 - u is chosen so that $\text{distance}[u] = \min_w \notin S \text{ | } \text{distance}[w]$ (If u is not unique, then we may select any of them); /* Greedy 贪婪 Method */
 - if $\text{distance}[u_1] < \text{distance}[u_2]$ and we add u_1 into S , then $\text{distance}[u_2]$ may change. If so,

a shorter path from s to u_2 must go through u_1 and distance' [u_2] = distance [u_1] + length($< u_1, u_2 >$).

```

1 void Dijkstra( Table T )
2 { /* T is initialized by Figure 9.30 on p.303 */
3     Vertex V, W;
4     for ( ; ; ) {
5         V = smallest unknown distance vertex;
6         // current smallest is known as the global smallest
7         // 可以用 Heap 实现
8         if ( V == NotAVertex )
9             break;
10        T[ V ].Known = true;
11        for ( each W adjacent to V )
12            if ( !T[ W ].Known )
13                if ( T[ V ].Dist + Cvw < T[ W ].Dist ) {
14                    Decrease( T[ W ].Dist to T[ V ].Dist + Cvw );
15                    T[ W ].Path = V;
16                } /* end-if update W */
17            } /* end-for( ; ; ) */
18 /* not work for edge with negative cost!!! */

```

❖ Implementation 1

§ 3. Shortest Path Algorithms

$V = \text{smallest unknown distance vertex};$
/ simply scan the table – $O(|V|)$ */*

$T = O(|V|^2 + |E|)$

Good if the graph is dense

❖ Implementation 2

$V = \text{smallest unknown distance vertex};$
/ keep distances in a priority queue and call DeleteMin – $O(\log|V|)$ */*

$\text{Decrease}(T[W].Dist \text{ to } T[V].Dist + Cvw);$

/ Method 1: DecreaseKey – $O(\log|V|)$ */*

$T = O(|V| \log|V| + |E| \log|V|) = O(|E| \log|V|)$

Good if the
graph is sparse

/ Method 2: insert W with updated Dist into the priority queue */*

/ Must keep doing DeleteMin until an unknown vertex emerges */*

$T = O(|E| \log|V|)$ but requires $|E|$ DeleteMin with $|E|$ space

❖ Other improvements: Pairing heap (Ch.12) and Fibonacci heap (Ch. 11)

3.1.3 Graphs with Negative Edge Costs

```

1 void WeightedNegative( Table T )
2 { /* T is initialized by Figure 9.30 on p.303 */
3     Queue Q;
4     Vertex V, W;
5     Q = CreateQueue (NumVertex); MakeEmpty( Q );
6     Enqueue( S, Q ); /* Enqueue the source vertex */
7     while ( !IsEmpty( Q ) ) {
8         V = Dequeue( Q );
9         for ( each W adjacent to V )
10            if ( T[ V ].Dist + Cvw < T[ W ].Dist ) {
11                T[ W ].Dist = T[ V ].Dist + Cvw;
12                T[ W ].Path = V;
13                if ( W is not already in Q )
14                    Enqueue( W, Q ); // 可以在程序内设一个Inqueue[], enqueue的时候设为1, dequeue的时候设为0
15            } /* end-if update */ // 进入if中的都是需要update的, 把其放入queue中
16        } /* end-while */
17        DisposeQueue( Q ); /* free memory */
18    }
19 /* negative-cost cycle will cause indefinite loop */
20 // 为解决这种问题可以限制dequeue的次数

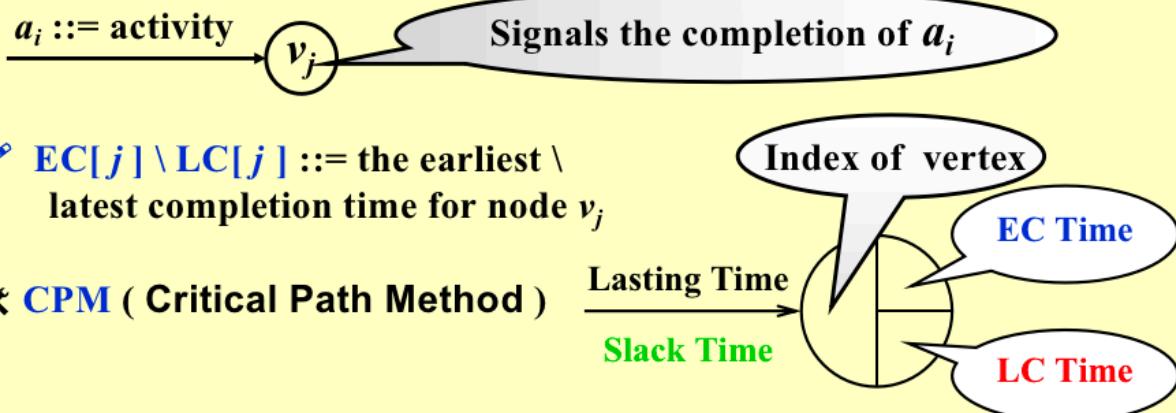
```

3.1.4 Acyclic Graphs

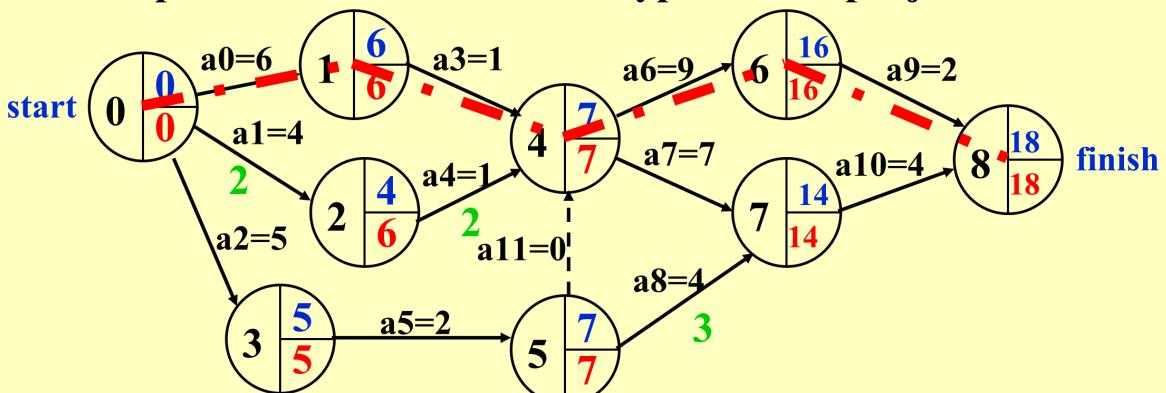
- If the graph is acyclic, vertices may be selected in topological order since when a vertex is selected, its distance can no longer be lowered without any incoming edges from unknown nodes.

$T = O(|E| + |V|)$ and no priority queue is needed.

❖ Application: AOE (Activity On Edge) Networks
—— scheduling a project



【Example】 AOE network of a hypothetical project



➤ Calculation of EC: Start from v_0 , for any $a_i = \langle v, w \rangle$, we have

$$EC[w] = \max_{(v,w) \in E} \{EC[v] + C_{v,w}\}$$

➤ Calculation of LC: Start from the last vertex v_8 , for any $a_i = \langle v, w \rangle$, we have $LC[v] = \min_{(v,w) \in E} \{LC[w] - C_{v,w}\}$

➤ Slack Time of $\langle v, w \rangle = LC[w] - EC[v] - C_{v,w}$

➤ Critical Path ::= path consisting entirely of zero-slack edges.

- EC: 最迟完工时间
- LC: 最早完工时间
- Slack Time: 可摸鱼时间
- A、从开始顶点 v_1 出发, 令 $ve(1)=0$, 按拓扑有序序列求其余各顶点的可能最早发生时间。

- $V_e(k) = \max\{v_e(j) + dut(j, k)\}$, $j \in T$ 。其中 T 是以顶点 v_k 为尾的所有弧的头顶点的集合 ($2 \leq k \leq n$)。如果得到的拓扑有序序列中顶点的个数小于网中顶点个数 n , 则说明网中有环, 不能求出关键路径, 算法结束。
- B、从完成顶点 v_n 出发, 令 $v_l(n) = v_e(n)$, 按逆拓扑有序求其余各顶点的允许的最晚发生时间: $v_l(j) = \min\{v_l(k) - dut(j, k)\}$, $k \in S$ 。其中 S 是以顶点 v_j 是头的所有弧的尾顶点集合 ($1 \leq j \leq n-1$)。
- C、==求每一项活动 a_i ($1 \leq i \leq m$) 的最早开始时间 $e(i) = v_e(j)$, 最晚开始时间 $l(i) = v_l(k) - dut(j, k)$ ==。

若某条弧满足 $e(i) = l(i)$, 则它是关键活动。

3.1.4.1 All-Pairs Shortest Path Problem

For all pairs of v_i and v_j ($i \neq j$), find the shortest path between.

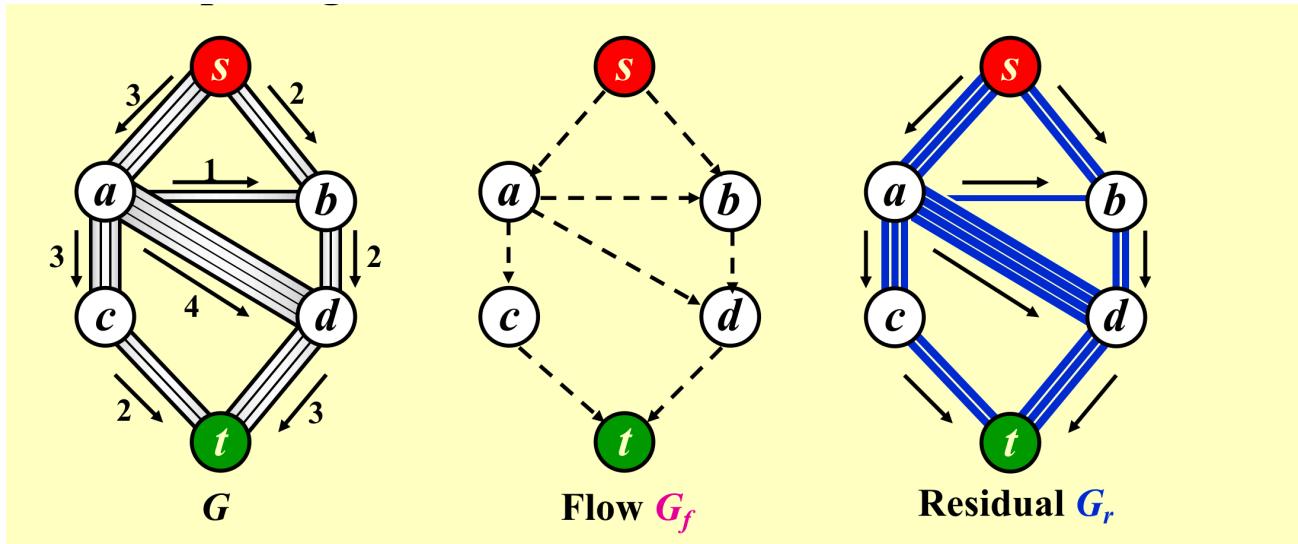
Method 1 Use **single-source algorithm** for $|V|$ times.

$T = O(|V|^3)$ – works fast on sparse graph.

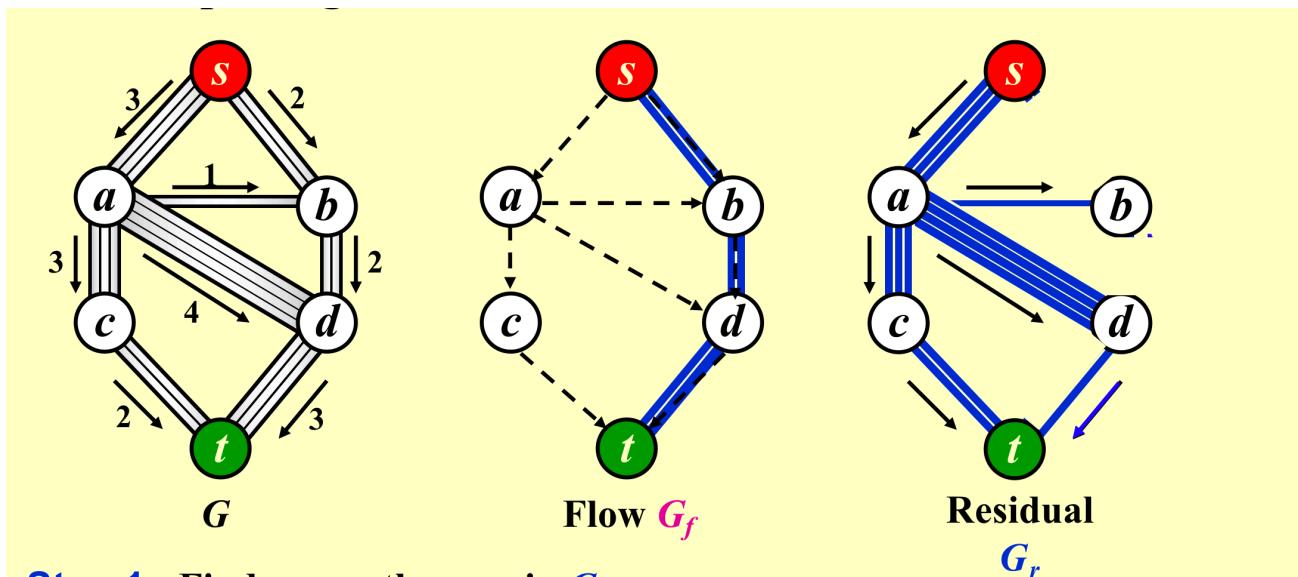
Method 2 $O(|V|^3)$ algorithm given in Ch.10, works faster on dense graphs.

4 Network Flow Problems

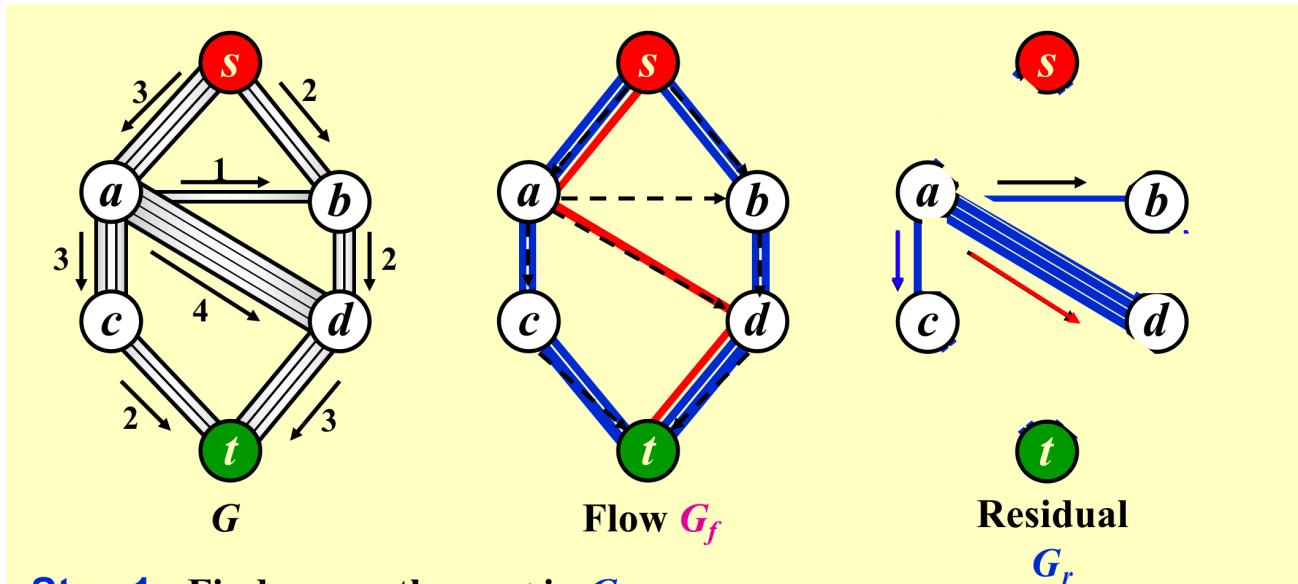
4.1 A Simple Algorithm



- Step 1: Find any path $s \in t$ in G_r ;



- Step 2: Take the minimum edge on this path as the amount of flow and add to G_f ;
- Step 3: Update G_r and remove the 0 flow edges; 加一条路径，就会在残差图里减掉一个



- Step 4:

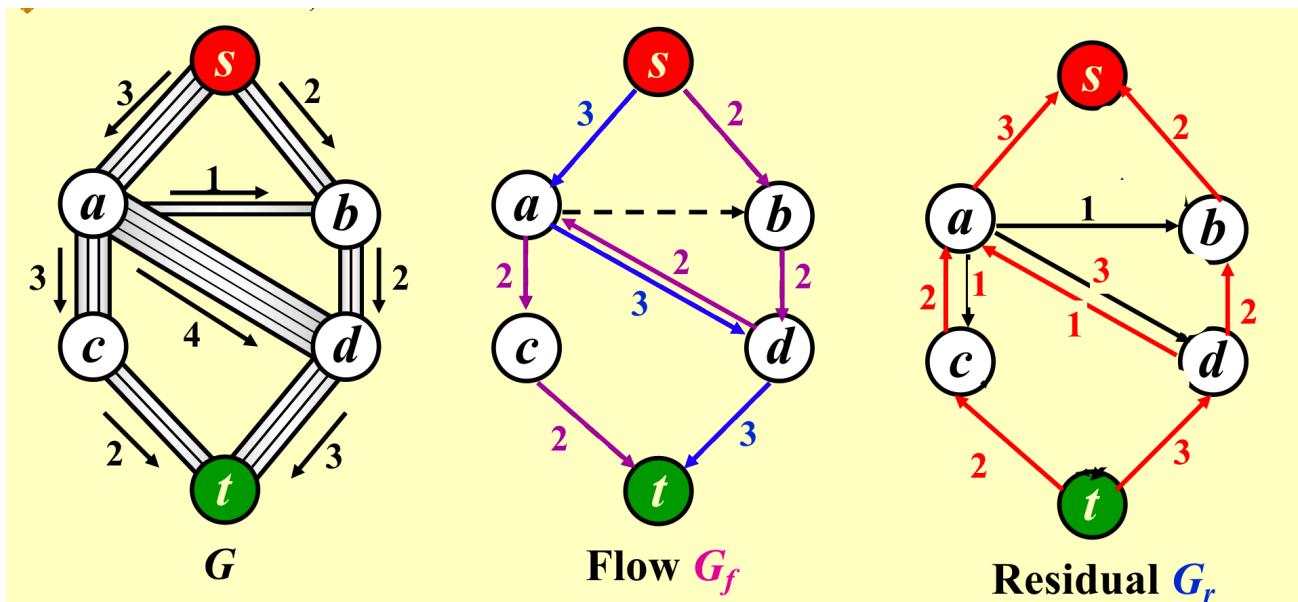
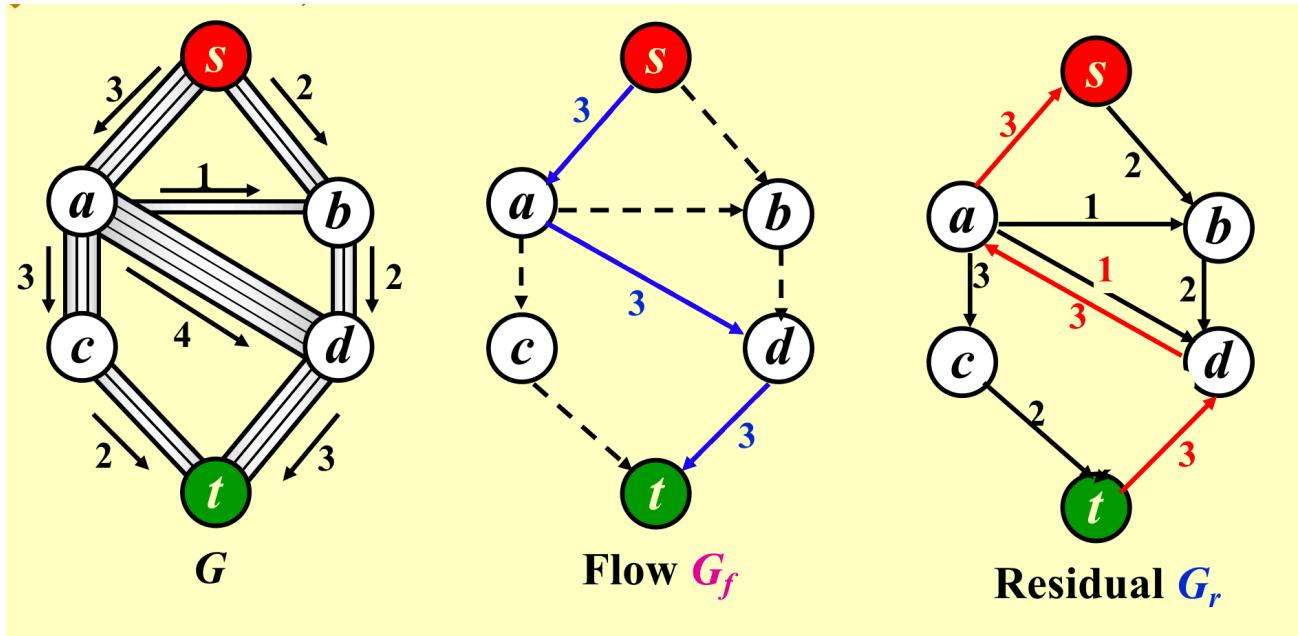
```

1 if (there is a path s \in t in Gr )
2   Goto Step 1;
3 else
4   End.

```

- Problem: 如果一开始的路不好，就会把后面的路都给影响到

4.2 A Solution – allow the algorithm to undo its decisions

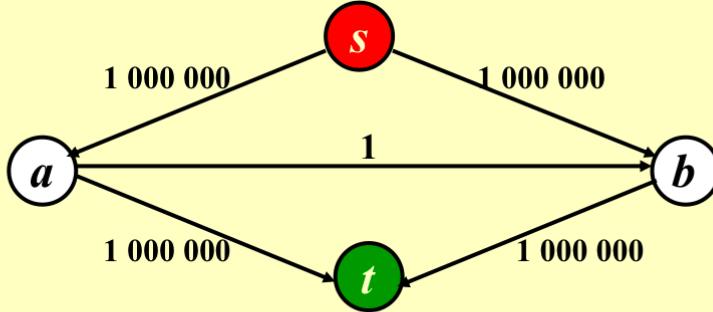


- For each edge (v, w) with flow $f_{v,w}$ in G_f , add an edge (w, v) with flow $f_{v,w}$ in G_r .
- 就是说设两个图，一个是flow的图，一个是残差图。之前的残差图对于选好的路径才用删除的做法，但是现在的残差图才用增加反向流来判断残差，是可修改的。实现了自取消的能力。
- 还是很巧妙的一种算法

4.3 Analysis

- ☞ An augmenting path can be found by an unweighted shortest path algorithm.

$T = O(f \cdot |E|)$ where f is the maximum flow.



- ☞ Always choose the augmenting path that allows the largest increase in flow. /* modify Dijkstra's algorithm */

$$\begin{aligned} T &= T_{\text{augmentation}} * T_{\text{find a path}} \\ &= O(|E| \log \text{cap}_{\max}) * O(|E| \log |V|) \\ &= O(|E|^2 \log |V|) \text{ if } \text{cap}_{\max} \text{ is a small integer.} \end{aligned}$$

- ☞ Always choose the augmenting path that has the least number of edges.

$$\begin{aligned} T &= T_{\text{augmentation}} * T_{\text{find a path}} \\ &= O(|E|) * O(|E| \cdot |V|) \text{ /* unweighted shortest path algorithm */} \\ &= O(|E|^2 |V|) \end{aligned}$$

Note:

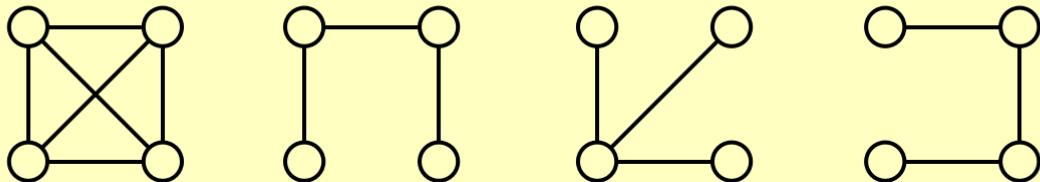
- If every $v \notin \{s, t\}$ has either a single incoming edge of capacity 1 or a single outgoing edge of capacity 1, then time bound is reduced to $O(|E||V|^{1/2})$.
- The **min-cost flow** problem is to find, among all maximum flows, the one flow of minimum cost provided that each edge has a cost per unit of flow.

5 Minimum Spanning Tree

§ 5 Minimum Spanning Tree

【Definition】 A *spanning tree* of a graph G is a **tree** which consists of $V(G)$ and a subset of $E(G)$

【Example】 A complete graph and three of its spanning trees



Note:

- The minimum spanning tree is a **tree** since it is acyclic -- the number of edges is $|V| - 1$.
- It is **minimum** for the total cost of edges is minimized.
- It is *spanning* because it covers every vertex.
- A minimum spanning tree exists iff G is **connected**.
- Adding a non-tree edge to a spanning tree, we obtain a *cycle*.

- 首先要是个生成树，到了每个顶点；
- 生成树的边是顶点个数减1
- 边的总和的最小的认为是最小生成树

5.1 Greedy Method - Prim's Algorithm

Make the best decision for each stage, under the following constraints :

- (1) we must use only edges within the graph;
- (2) we must use exactly $|V| - 1$ edges;
- (3) we may not use edges that would produce a cycle.

- 看ppt的图，和Dijkstra算法很像
- 复杂度和Dijkstra一样？？？

5.2 Kruskal's Algorithm – maintain a forest

```

1 void Kruskal ( Graph G )
2 {   T = { } ;
3     while ( T contains less than |V| - 1 edges && E is not empty ) {
4         choose a least cost edge (v, w) from E ; /* DeleteMin 可以用
5             Heap */
6         delete (v, w) from E ;
7         if ( (v, w) does not create a cycle in T ) // 可以验证两个端点不
8             是一个集合里的, 用并查集
9             add (v, w) to T ; /* Union / Find */
10        else
11            discard (v, w) ;
12    }
13    if ( T contains fewer than |V| - 1 edges )
14        Error ( "No spanning tree" ) ;
15 }
```

$$T = O(|E| \log |E|) \quad (9)$$

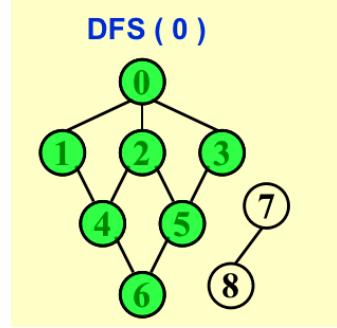
6 Applications of DFS

```

1 void DFS ( Vertex V ) /* this is only a template */
2 {   visited[ V ] = true; /* mark this vertex to avoid cycles */
3     for ( each W adjacent to V )
4       if ( !visited[ W ] )
5         DFS( W );
6 } /* T = O( |E| + |V| ) as long as adjacency lists are used */
```

- 就是先找到最深的节点。最多是每条边和每个节点都访问一次

6.1 Undirected Graph



```

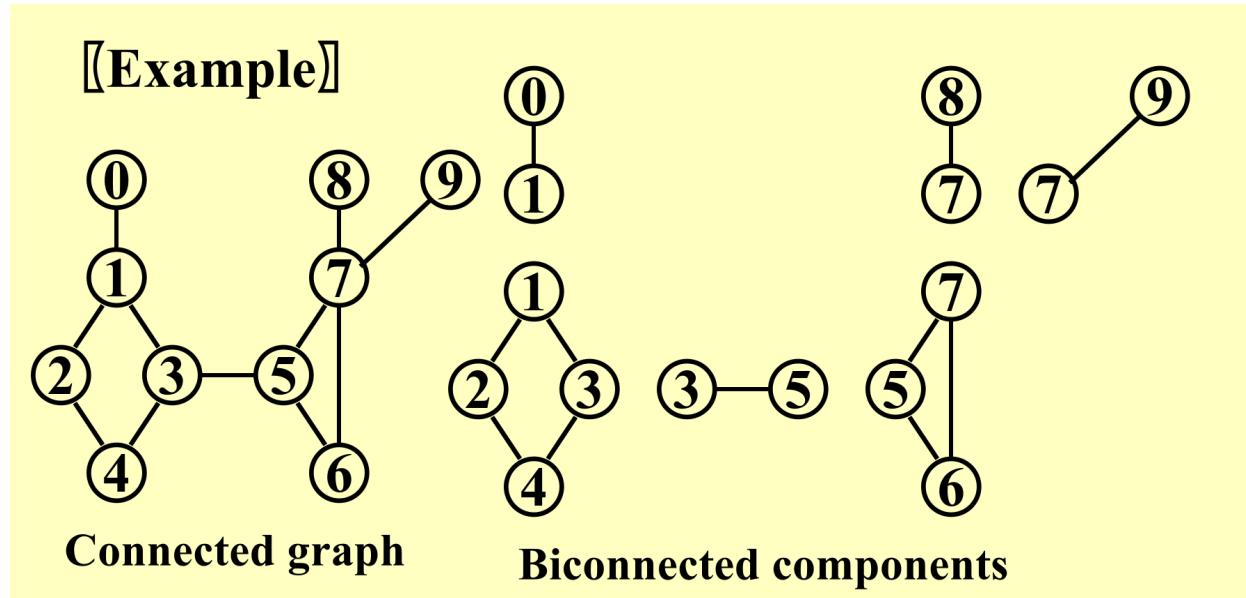
1 void ListComponents ( Graph G )
2 {   for ( each V in G )
3     if ( !visited[ V ] ) {
4       DFS( V ); // 一个DFS相当于把联通了的所有节点都访问一次
5       printf("\n");
6     }
7 }
8 // 0 1 4 6 5 2 3
9 // 7 8

```

- 其实有点像中序遍历
- 有几个回车符，就有几个联通元素

6.2 Biconnectivity

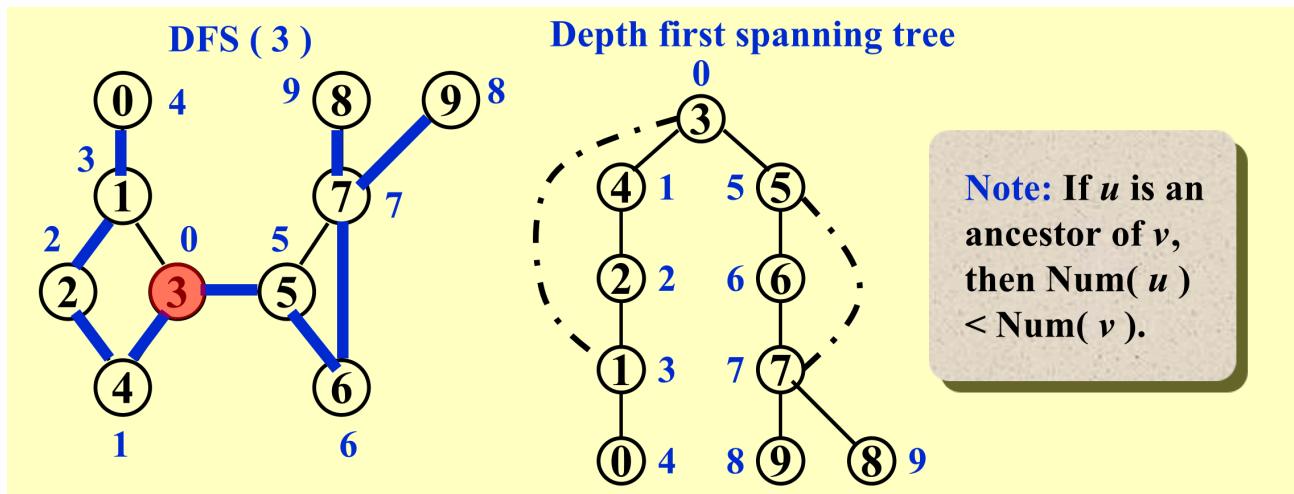
- v is an articulation point (关节点) if $G' = \text{DeleteVertex}(G, v)$ has at least 2 connected components. 关节点（就是关节）去除后，会变成至少两个联通分量。
- G is a biconnected graph if G is connected and has no articulation points.
- A biconnected component is a maximal biconnected subgraph.



- Note: No edges can be shared by two or more biconnected components. Hence $E(G)$ is partitioned by the biconnected components of G .

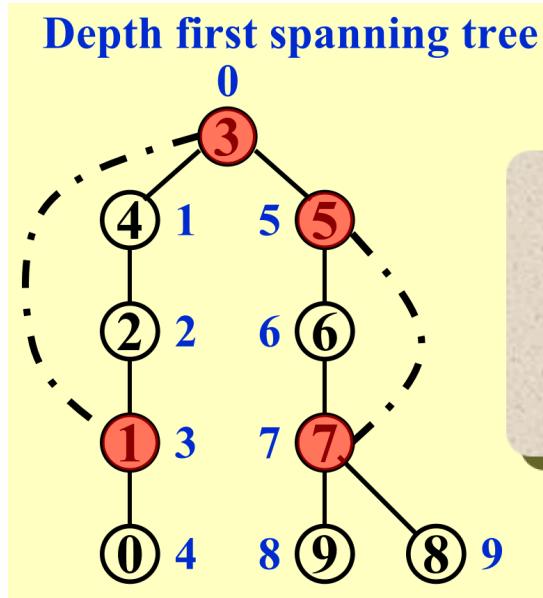
6.2.1 How to find articulation point

6.2.1.1 Use depth first search to obtain a spanning tree of G

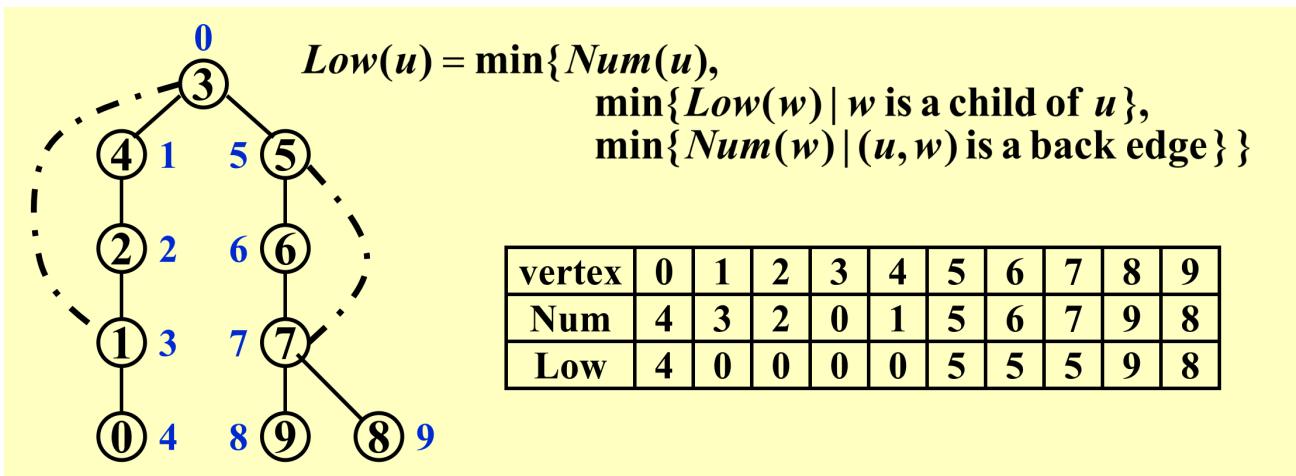


- 先用DFS找出spanning tree，根据DFS顺序标号
- 把缺失的边补齐，记作backedge

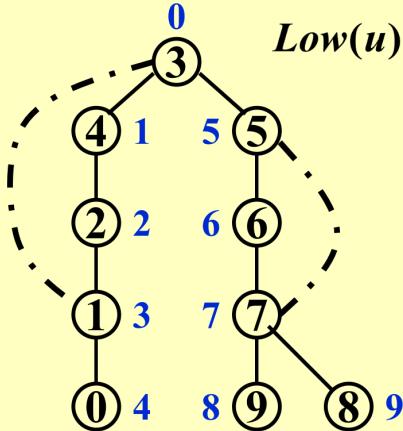
6.2.1.2 Find the articulation points in G



- The **root** is an articulation point if it has **at least 2 children**
- Any **other vertex** u is an articulation point iff u has **at least 1 child**, and it is **impossible to move down at least 1 step and then jump up to u 's ancestor**. 判断是不是在环路上, 不是环路关键节点。



- 从根节点开始
- 1、7节点有back edge
- 2、4、6就是用 $Low(w)$ child
- 0、9、8用的是 $Num(u)$



$Low(u) = \min\{Num(u), \min\{Low(w) | w \text{ is a child of } u\}, \min\{Num(w) | (u, w) \text{ is a back edge}\}\}$

vertex	0	1	2	3	4	5	6	7	8	9
Num	4	3	2	0	1	5	6	7	9	8
Low	4	0	0	0	0	5	5	5	9	8

Therefore, u is an **articulation point** iff

(1) u is the **root** and has **at least 2 children**; or

(2) u is not the root, and has **at least 1 child** such that

$$Low(\text{child}) \geq Num(u).$$

Please read the pseudocodes on p.327 and p.329 for more details.

6.3 Euler Circuits

- Draw each line exactly once without lifting your pen from the paper – Euler tour (一笔画)
 - An Euler tour is possible if there are exactly **two vertices having odd degree**. One must start at one of the odd-degree vertices.
- Draw each line exactly once without lifting your pen from the paper, AND finish at the starting point – Euler circuit (欧拉环, 一笔画+回到起点)
 - An Euler circuit is possible **only if the graph is connected and each vertex has an even degree**.
- 具体做法
- Note:
 - The path should be maintained as a linked list.
 - For each adjacency list, maintain a pointer to the last edge scanned.

$$T = O(|E| + |V|) \tag{10}$$