

# Algorithm Analysis

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## Attribute

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- Input
- Output
- Definiteness
- Finiteness
- Effectiveness

## Notation

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$$\begin{aligned}T(N) = O(f(N)) &: \text{for } N \geq n_0, T(N) \leq c \cdot f(N) \\T(N) = \Omega(f(N)) &: \text{for } N \geq n_0, T(N) \geq c \cdot f(N) \\T(N) = \Theta(f(N)) &: T(N) = O(f(N)) \& T(N) = \Omega(f(N)) \\T(N) = o(f(N)) &: T(N) = O(f(N)) \& T(N) \neq \Theta(f(N))\end{aligned}$$

## Checking

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1. function, like:

$$T(N) = O(N^2), \frac{T(2N)}{T(N)} \approx 4$$

2. limit:

$$T(N) = O(f(N)), \lim_{N \rightarrow \infty} \frac{T(N)}{f(N)} \approx \text{Constant}$$

## List

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## Data Type

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$$\{Objects\} \cup \{Operations\}$$

## Implementation

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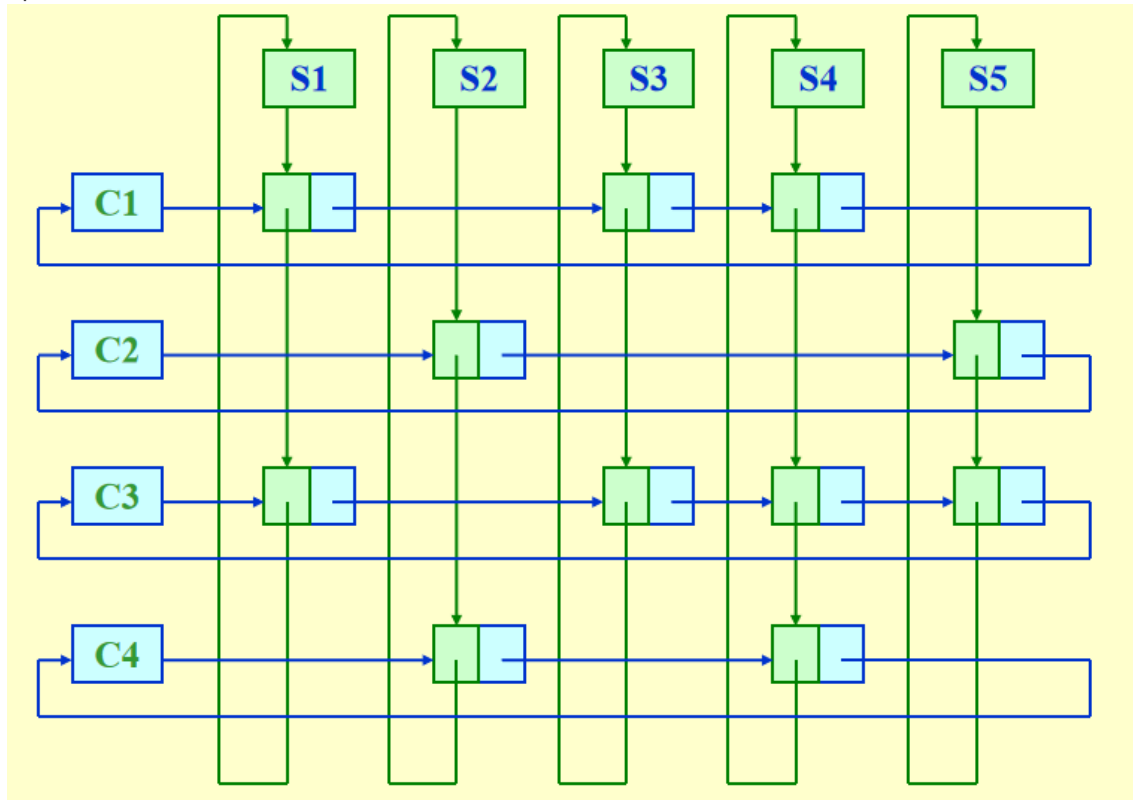
1. Simple array using **sequential mapping**
  - easy for random access
  - **limited size**
  - **inconvenient for insertion and deletion**
2. Linked lists
  - easy to insert and delete (with a dummy head)

- flexible size
- **linear search to find the  $k^{\text{th}}$  element**

### 3. Doubly linked circular list

- a list that is more convenient for random access

### 4. Sparse matrix

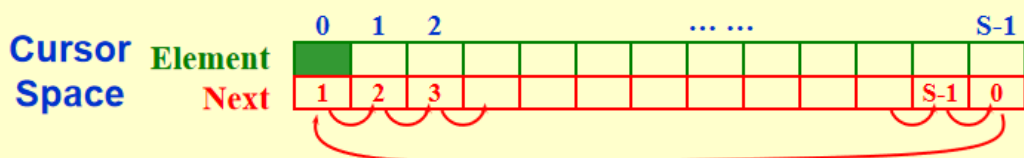


### 5. Cursor implementation

In some languages not supporting pointers, cursor implementation is adopted.

#### 👉 Features that a linked list must have:

- The data are stored in a collection of structures. Each structure contains **data** and a pointer to the **next** structure.
- A new structure can be obtained from the system's global memory by a call to **malloc** and released by a call to **free**.



Usually, the implementation depends on the **desired operations**.

## Stack

### FILO

- The stack model must be well encapsulated. That is, no part of your code, except for the stack routines, can attempt to access the Array or Top variable.
- Error check must be done before Push or Pop (Top).

## Infix to Postfix

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- Add parentheses if necessary
- If a letter is read, output the letter
- If an operand is read:
  - If the stack is empty, push the operand
  - If the last operand  $\geq$  current operand, output the stack and then push the current
  - If a right parenthesis is read, output all the operands between the parentheses
  - If the end is reached, output all the operands
- Observe that when  $($  is not in the stack, its precedence is the highest; but when it is in the stack, its precedence is the lowest. Define in-stack precedence and incoming precedence for symbols, and each time use the corresponding precedence for comparison.

## Queue

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FIFO

1. Linked list
2. Array implementation
  - To differentiate FULL and EMPTY:
    - one space is abandoned, that is, EMPTY means  $\text{front} == \text{rear} + 1$  and no elements; FULL means  $\text{front} == \text{rear} + 2$  and  $N - 1$  elements
    - add a SIZE field field to signal the size
    - add a FULL boolean variable

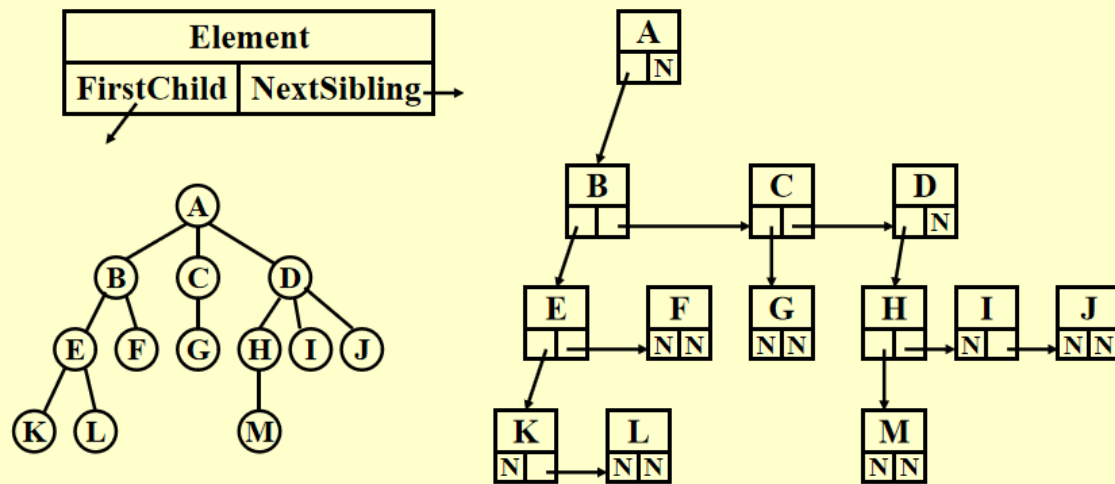
## Tree

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- degree of a node: **number of subtrees** of the node. For example,  $\text{degree}(A) = 3$ ,  $\text{degree}(F) = 0$
- degree of a node: max degree of all the nodes
- depth of  $n_i$ : length of the unique path from the root to  $n_i$ .  $\text{Depth}(\text{root}) = 0$ .
- height of  $n_i$ : length of the longest path from  $n_i$  to a leaf.  $\text{Height}(\text{leaf}) = 0$
- level:  $\text{level}(\text{root}) = 1$
- **Right** and **Left** matters!!

## FirstChild-NextSibling Representation

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## Traversal

### Preorder

- Use stack implicitly. (System stack)
- visit the node before a push
- depth-first search

### Postorder

- Use stack implicitly. (System stack)
- visit the node after a push

### Inorder

- visit the node right after the first push(if left child exists)

### Level order

- use a queue instead of a stack
- visit a node and then push its children until the traversal ends
- breadth-first search

## Threaded Binary Tree

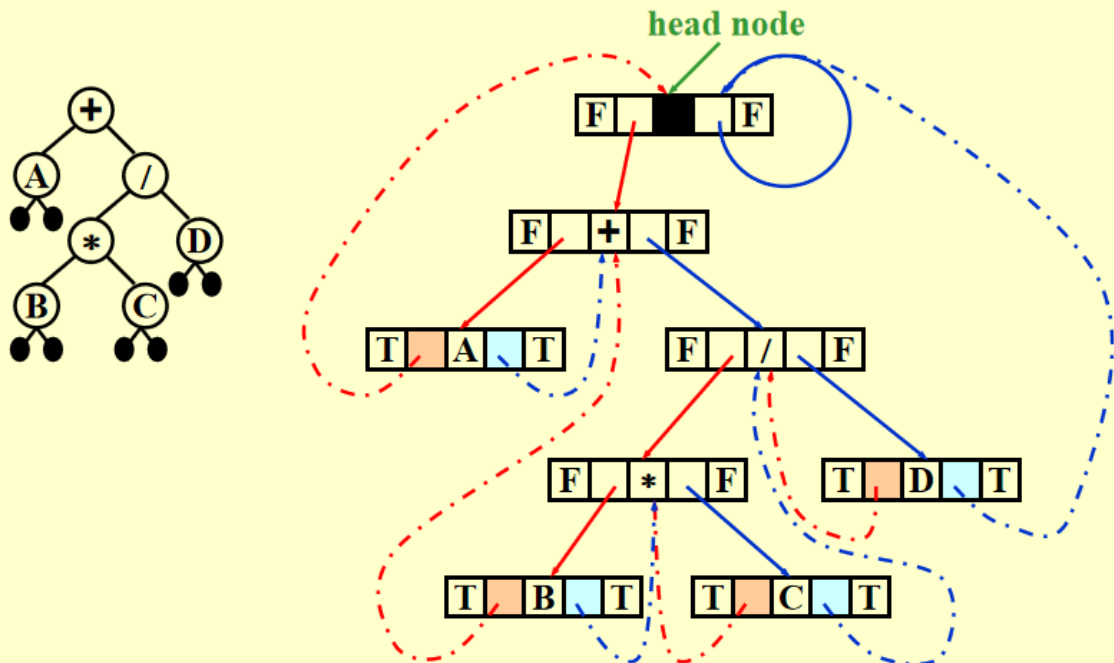
Rule 1: If Tree->Left is null, replace it with a pointer to the **inorder** predecessor of Tree.

Rule 2: If Tree->Right is null, replace it with a pointer to the **inorder** successor of Tree.

Rule 3: **There must not be any loose threads.** Therefore a threaded binary tree must have a head node of which the left child points to the first node.

[[Example]] Given the syntax tree of an expression (infix)

$$A + B * C / D$$



## Binary Search Tree

- ☑ The maximum number of nodes on level  $i$  is  $2^{i-1}$ ,  $i \geq 1$ .  
The maximum number of nodes in a binary tree of depth  $k$  is  $2^k - 1$ ,  $k \geq 1$ .
- ☑ For any nonempty binary tree,  $n_0 = n_2 + 1$  where  $n_0$  is the number of leaf nodes and  $n_2$  the number of nodes of degree 2.

A binary search tree is a binary tree. It may be empty. If it is not empty, it satisfies the following properties:

1. Every node has a key which is an integer, and the keys are **distinct**.
2. The keys in a nonempty left subtree must be **smaller** than the key in the root of the subtree.
3. The keys in a nonempty right subtree must be **larger** than the key in the root of the subtree.
4. The left and right **subtrees** are also binary search trees.
5. From the above, the leftmost element is the smallest; the rightmost element is the largest.

```
Position Find(ElementType X, Tree T)
{
    while(T)
    {
        if(X == T->Element)
        {
            return T;
        }
    }
}
```

```

    }
    else if(X < T->Element)
    {
        T = T->Left;
    }
    else
    {
        T = T->Right;
    }
}
return NULL;
}

```

```

Tree Insert(ElementType X, Tree T)
{
    if(T == NULL)
    {
        T = malloc(sizeof(struct treenode));
        if(T == NULL)
        {
            Error();
        }
        else
        {
            T->Element = X;
            T->Left = T->Right = NULL;
        }
    }
    else
    {
        if(X < T->Element)
        {
            T->Left = Insert(X, T->Left);
        }
        else if(X > T->Element)
        {
            T->Right = Insert(X, T->Right);
        }
        // else X == T->Element, do nothing
    }
    return T;
}

```

```

SearchTree Delete( ElementType X, SearchTree T )
{
    Position TmpCell;
    if ( T == NULL )    Error();
    else if ( X < T->Element ) /* Go left */
        T->Left = Delete( X, T->Left );
    else if ( X > T->Element ) /* Go right */
        T->Right = Delete( X, T->Right );
    else /* Found element to be deleted */
        if ( T->Left && T->Right ) { /* Two children */

```

```

        /* Replace with smallest in right subtree */
        TmpCell = FindMin( T->Right );
        T->Element = TmpCell->Element;
        T->Right = Delete( T->Element, T->Right );
    } /* End if */
else { /* One or zero child */
    TmpCell = T;
    if ( T->Left == NULL ) /* Also handles 0 child */
        T = T->Right;
    else if ( T->Right == NULL )
        T = T->Left;
    free( TmpCell );
} /* End else 1 or 0 child */
return T;
}

```

## Heap (Priority Queue)

### Perfect Binary Tree:

A tree of height  $h$  has  $2^{h+1} - 1$  nodes.

### Complete Binary Tree:

A binary tree with  $n$  nodes and height  $h$  is complete *iff* its nodes correspond to the nodes numbered from 1 to  $n$  in the perfect binary tree of height  $h$ .

Then, in an array representing the complete binary tree sequentially, for a node with index  $i$ :

$$\begin{aligned}
 \text{parent}(i) &= \begin{cases} \lfloor \frac{i}{2} \rfloor & i \neq 1 \\ \text{none} & i = 1 \end{cases} \\
 \text{leftchild}(i) &= \begin{cases} 2i & 2i \leq n \\ \text{none} & 2i > n \end{cases} \\
 \text{rightchild}(i) &= \begin{cases} 2i + 1 & 2i + 1 \leq n \\ \text{none} & 2i + 1 > n \end{cases}
 \end{aligned}$$

For the perfect binary tree of height  $h$  containing  $2^{h+1} - 1$  nodes, the sum of the heights of the nodes is  $2^{h+1} - 1 - (h + 1)$ .

## Order property

A min tree is a tree in which the key value in each node is **no larger than the key values in its children (if any)**. A min heap is a **complete binary tree** that is also a min tree.

height of the heap:  $h = \log \lfloor N \rfloor$

```

void Insert( ElementType X, PriorityQueue H )
{
    int i;

```

```

if ( IsFull( H ) ) {
    Error( "Priority queue is full" );
    return;
}

// percolate up
for ( i = ++H->Size; H->Elements[ i / 2 ] > x; i /= 2 )
    H->Elements[ i ] = H->Elements[ i / 2 ];

// faster than swap in the loop
H->Elements[ i ] = x;
}

```

```

ElementType DeleteMin( PriorityQueue H )
{
    int i, Child;
    ElementType MinElement, LastElement;
    if ( IsEmpty( H ) ) {
        Error( "Priority queue is empty" );
        return H->Elements[ 0 ];
    }
    /* save the min element */
    MinElement = H->Elements[ 1 ];
    /* take last and reset size */
    LastElement = H->Elements[ H->Size-- ];

    for ( i = 1; i * 2 <= H->Size; i = Child ) { /* Find smaller child */
        Child = i * 2;
        if ( Child != H->Size && H->Elements[Child+1] < H->Elements[Child] )
            Child++;
        if ( LastElement > H->Elements[ Child ] ) /* Percolate one level */
            H->Elements[ i ] = H->Elements[ Child ];
        else break; /* find the proper position */
    }
    H->Elements[ i ] = LastElement;
    return MinElement;
}

```

```

Heap *Build(Heap *H, Size N)
{
    // H[1]~H[N] contains the original sequence
    if(N!=1){
        int i=N/2; // the first subroot that needs adjust
        for( ; i>=1; i--){
            PercolateDown(i);
        }
    }
    return H;
}

```



faster than to build the heap anew

## Union and Find

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### Equivalence Relation:

- reflexive
- symmetric
- transitive

### Partial Order:

- reflexive
- anti-symmetric
- transitive

```
void SetUnion ( DisjSet S, SetType Rt1, SetType Rt2 )
{   S [ Rt2 ] = Rt1 ;   }
```

```
SetType Find ( ElementType X, DisjSet S )
{   for ( ; S[X] > 0; X = S[X] )   ;
    return X ;
}
```

## Smart Union

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**Union by size:** `S[root]=size(initialized as -1)`

Always change the smaller tree

$$height(T) \leq \lfloor \log_2 N \rfloor + 1$$

$N$  Union and  $M$  Find:  $O(N + M \log_2 N)$

### Union by height:

Always change the shallow tree

## Path Compression

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```

SetType Find ( ElementType X, DisjSet S )
{
    if ( S[ X ] <= 0 )    return X;
    else    return S[ X ] = Find( S[ X ], S );
}

```

```

SetType Find ( ElementType X, DisjSet S )
{
    ElementType root, trail, lead;
    for ( root = X; S[ root ] > 0; root = S[ root ] )
        ; /* find the root */
    for ( trail = X; trail != root; trail = lead ) {
        lead = S[ trail ] ;
        S[ trail ] = root ;
    } /* collapsing */
    return root ;
}

```

## Graph

Simple path ::=  $v_{i1}, v_{i2}, \dots, v_{in}$  are distinct

Cycle ::= simple path with  $v_p = v_q$

## Representation

- Adjacency Matrix
- Adjacency List

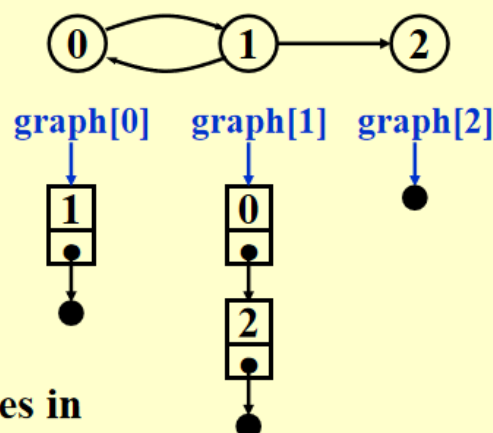
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### Adjacency Lists

Replace each row by a linked list

[[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.

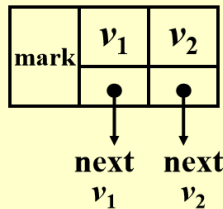
**For undirected G:**

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

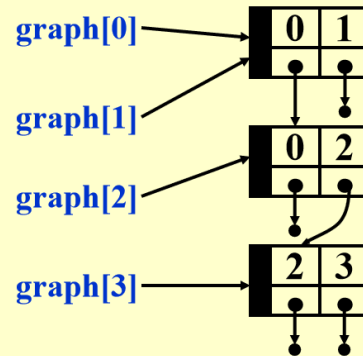
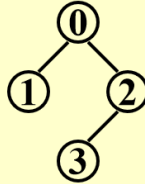
## Adjacency Multilists

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

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



[[ Example ]]



## Weighted Edges

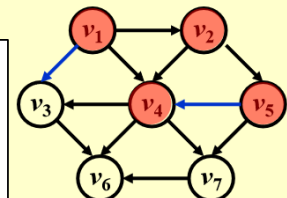
- $\text{adj\_mat}[i][j] = \text{weight}$
- adjacency lists \ multilists : add a **weight** field to the node.

When the program needs to mark the edge

## Topological Sort

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

```
void Topsort( Graph G )
{
    Queue Q;
    int Counter = 0;
    Vertex V, W;
    Q = CreateQueue( NumVertex ); MakeEmpty( Q );
    for ( each vertex V )
        if ( Indegree[ V ] == 0 ) Enqueue( V, Q );
    while ( !IsEmpty( Q ) ) {
        V = Dequeue( Q );
        TopNum[ V ] = ++ Counter; /* assign next */
        for ( each W adjacent to V )
            if ( -- Indegree[ W ] == 0 ) Enqueue( W, Q );
    } /* end-while */
    if ( Counter != NumVertex )
        Error( "Graph has a cycle" );
    DisposeQueue( Q ); /* free memory */
}
```



Indegree	
v1	0
v2	0
v3	0
v4	0
v5	0
v6	3
v7	1

## Shortest Path Problem

# Unweighted Shortest Path

Breadth-first search

## ❖ Implementation

**Table[ i ].Dist** ::= distance from  $s$  to  $v_i$  /\* initialized to be  $\infty$  except for  $s$  \*/

**Table[ i ].Known** ::= 1 if  $v_i$  is checked; or 0 if not

**Table[ i ].Path** ::= for tracking the path /\* initialized to be 0 \*/

```
void Unweighted( Table T )
{
    /* T is initialized with the source vertex S given */
    Queue Q;
    Vertex V, W;
    Q = CreateQueue (NumVertex ); MakeEmpty( Q );
    Enqueue( S, Q ); /* Enqueue the source vertex */
    while ( !IsEmpty( Q ) ) {
        V = Dequeue( Q );
        T[ V ].Known = true; /* not really necessary */
        for ( each w adjacent to V )
            if ( T[ W ].Dist == Infinity ) {
                T[ W ].Dist = T[ V ].Dist + 1;
                T[ W ].Path = V;
                Enqueue( W, Q );
            } /* end-if Dist == Infinity */
    } /* end-while */
    DisposeQueue( Q ); /* free memory */
}
```

# Weighted Shortest Path

```
void Dijkstra( Table T )
{
    /* T is initialized by Figure 9.30 on p.303 */
    Vertex V, W;
    for ( ; ; ) {
        V = smallest unknown distance vertex;
        if ( V == NotAVertex )
            break;
        T[ V ].Known = true;
        for ( each w adjacent to V )
            if ( !T[ W ].Known )
                if ( T[ V ].Dist + Cvw < T[ W ].Dist ) {
                    Decrease( T[ W ].Dist to T[ V ].Dist + Cvw );
                    T[ W ].Path = V;
                } /* end-if update W */
    } /* end-for( ; ; ) */
}
/* not work for edge with negative cost */
```

## ❖ Implementation 1

**V** = smallest unknown distance vertex;

*/\* simply scan the table –  $O(|V|)$  \*/*

$T = O(|V|^2 + |E|)$   Good if the graph is dense

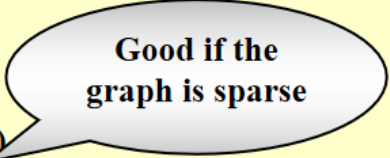
## ❖ Implementation 2

**V** = smallest unknown distance vertex;

*/\* keep distances in a priority queue and call DeleteMin –  $O(\log|V|)$  \*/*

Decrease(  $T[W].Dist$  to  $T[V].Dist + C_{vw}$  );

*/\* 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)

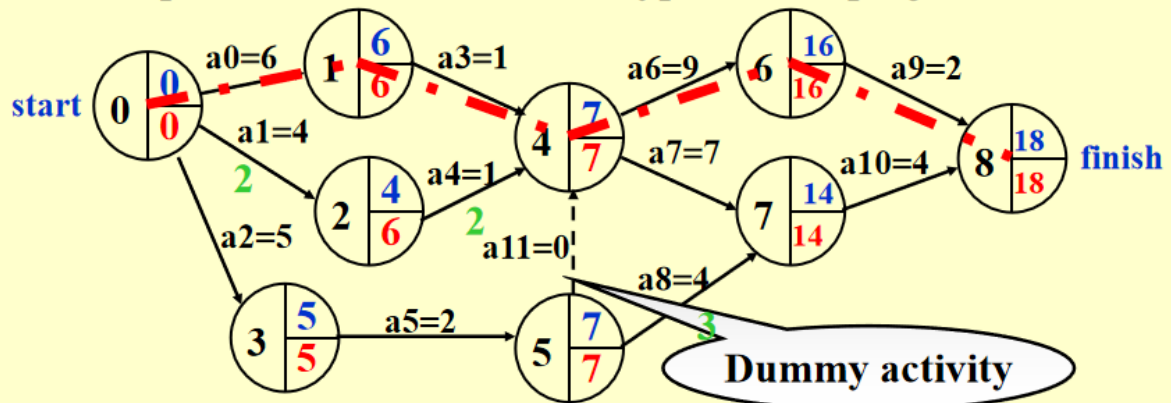
## Negative Cost

```
void WeightedNegative( Table T )
{
    /* T is initialized by Figure 9.30 on p.303 */
    Queue Q;
    Vertex V, W;
    Q = CreateQueue( NumVertex ); MakeEmpty( Q );
    Enqueue( S, Q ); /* Enqueue the source vertex */
    while ( !IsEmpty( Q ) ) {
        V = Dequeue( Q );
        for ( each w adjacent to V )
            if ( T[V].Dist + Cvw < T[W].Dist ) {
                T[W].Dist = T[V].Dist + Cvw;
                T[W].Path = V;
                if ( w is not already in Q )
                    Enqueue( W, Q );
            } /* end-if update */
        } /* end-while */
    DisposeQueue( Q ); /* free memory */
}
/* negative-cost cycle will cause indefinite loop */
```

each vertex is visited more than once

## Acyclic Graph

## [[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.

Those nodes with zero slack are critical, and called critical path. If their values change, the aggregate time changes.

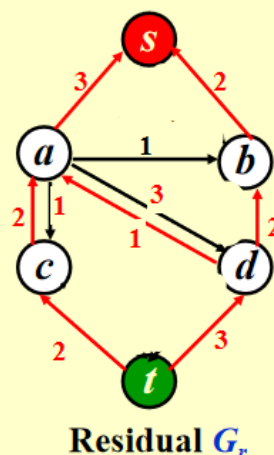
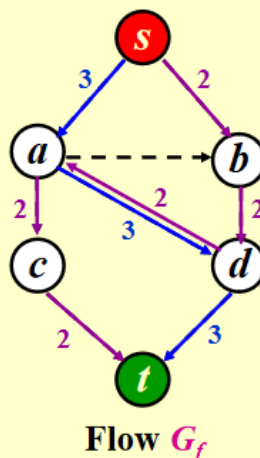
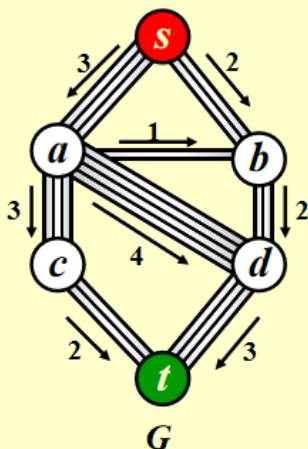
$$T = O(|E| + |V|)$$

## Network Flow

### 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  $G_r$ .



[[Proposition]] If the edge capabilities are **rational numbers**, this algorithm always terminate with a maximum flow.

**Note:** The algorithm works for  $G$  with ***cycles*** as well.

**最大流最小割 (Max-flow min-cut) 定理:** 在任何网络中, 最大流 $flow$ 的值=最小割的 $capacity$

<https://www.cnblogs.com/ecjtuacm-873284962/p/7113635.html>

$$T = O(f \cdot |E|)$$

## Minimum Spanning Tree

**Prim's Algorithm:**

- Start from a vertex,  $S = \{s_0\}$
- for  $s_i \in S$ , find  $w$  such that  $\min \langle s_i, w \rangle$ , and add  $w$  to  $S$

**Kruskal's Algorithm:**

```
void Kruskal ( Graph G )
{   T = { } ;
    while ( T contains less than |V| -1 edges && E is not empty ) {
        choose a least cost edge (v, w) from E ;
        delete (v, w) from E ;
        if ( (v, w) does not create a cycle in T )
            add (v, w) to T ;
        else
            discard (v, w) ;
    }
    if ( T contains fewer than |V|-1 edges )
        Error ( "No spanning tree" ) ;
}
```

- Start from an edge,  $E = \{e_0\}$
- find  $e \notin E$  such that  $\min \langle e \rangle \&not\ cycle$ , and add  $e$  to  $E$

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

## DFS

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

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

```

void ListComponents ( Graph G )
{
    for ( each v in G )
        if ( !visited[ v ] ) {
            DFS( v );
            printf("\n");
        }
}

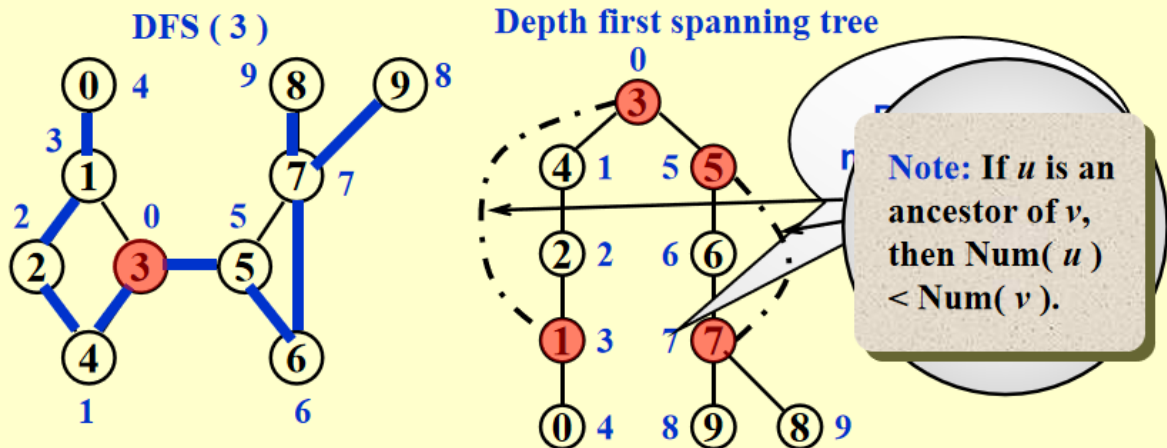
```

## Biconnectivity

- $v$  is an articulation point if  $G' = DeleteVertex(G, v)$  has at least 2 connected components.
- $G$  is a biconnected graph if  $G$  is connected and has no articulation points.
- A biconnected component is a maximal biconnected subgraph.
- $E(G)$  is partitioned by the biconnected components of  $G$ .

### Finding the **biconnected components** of a connected undirected $G$

➤ Use **depth first search** to obtain a spanning tree of  $G$



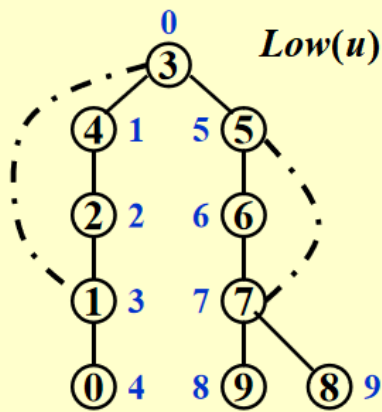
➤ Find the **articulation points** in  $G$

- ⊕ The **root** is an articulation point iff 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**.

*Num*: the order produced when a spanning tree is generated. The smaller, the more superior

If  $u$  is missed, the superior and the subordinate can not get in touch.





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

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  $Low(child) \geq Num(u)$ .

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

$Low$  represents the highest level that  $u$  can get in touch, with an exception(back edge)

If  $Low(child) \geq Num(u)$ , at least one child must get in touch with the superior by  $u$ , so  $u$  is critical

## Euler Circuit



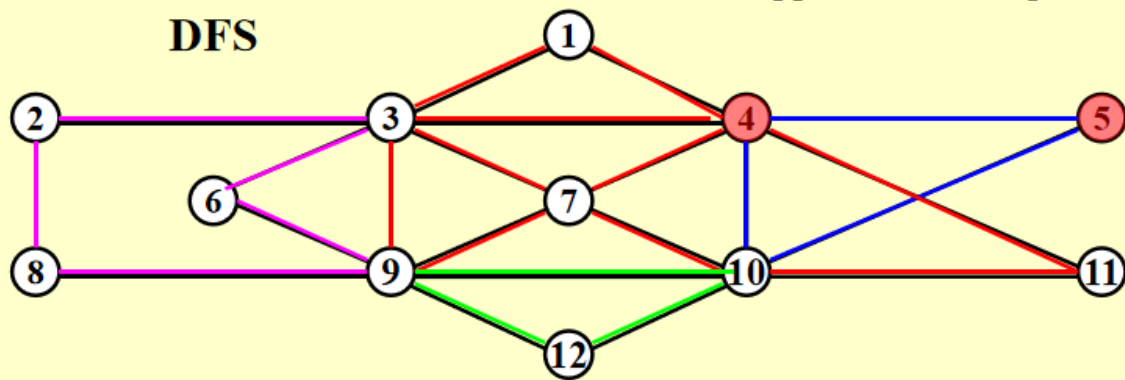
Draw each line exactly once without lifting your pen from the paper – *Euler tour*



Draw each line exactly once without lifting your pen from the paper, AND finish at the starting point – *Euler circuit*

[[**Proposition**]] An Euler circuit is possible only if the graph is connected and each vertex has an **even** degree.

[[**Proposition**]] An Euler tour is possible if there are exactly **two** vertices having odd degree. One must start at one of the odd-degree vertices.



**Note:**

- The path should be maintained as a linked list.
- For each adjacency list, maintain a pointer to the last edge scanned.
- $T = O(|E| + |V|)$



Find a simple cycle in an undirected graph that visits every vertex – *Hamilton cycle*

## Sort

### Shellsort

#### § 4 Shellsort ---- by Donald Shell

[[Example]] Sort:

	81	94	11	96	12	35	17	95	28	58	41	75	15
5-sort	35	17	11	28	12	41	75	15	96	58	81	94	95
3-sort	28	12	11	35	15	41	58	17	94	75	81	96	95
1-sort	11	12	15	17	28	35	41	58	75	81	94	95	96

✎ Define an *increment sequence*  $h_1 < h_2 < \dots < h_t$  ( $h_1 = 1$ )

✎ Define an  $h_k$ -sort at each phase for  $k = t, t-1, \dots, 1$

**Note:** An  $h_k$ -sorted file that is then  $h_{k-1}$ -sorted **remains**  $h_k$ -sorted.

make use of insertion sort

```
void Shellsort( ElementType A[ ], int N )
{
    int i, j, Increment;
    ElementType Tmp;
    for ( Increment = N / 2; Increment > 0; Increment /= 2 )
        /*h sequence */
        for ( i = Increment; i < N; i++ ) { /* insertion sort */
            Tmp = A[ i ];
            for ( j = i; j >= Increment; j -= Increment )
                if( Tmp < A[ j - Increment ] )
                    A[ j ] = A[ j - Increment ];
            else
                break;
            A[ j ] = Tmp;
        } /* end for-I and for-Increment loops */
}
```

The sequence is critical:


### Hibbard's Increment Sequence:

$h_k = 2^k - 1$  ---- consecutive increments have no common factors.


**【Theorem】** The worst-case running time of Shellsort, using Hibbard's increments, is  $\Theta(N^{3/2})$ .



### Conjectures:

  $T_{\text{avg-Hibbard}}(N) = O(N^{5/4})$

Shellsort is a very simple algorithm, yet with an extremely complex analysis. It is good for sorting up to moderately large input (tens of thousands).

 Sedgewick's best sequence is  $\{1, 5, 19, 41, 109, \dots\}$  in which the terms are either of the form  $9 \times 4^i - 9 \times 2^i + 1$  or  $4^i - 3 \times 2^i + 1$ .  $T_{\text{avg}}(N) = O(N^{7/6})$  and  $T_{\text{worst}}(N) = O(N^{4/3})$ .

## Heap Sort

```

void Heapsort( ElementType A[ ], int N )
{
    int i;
    for ( i = N / 2; i >= 0; i -- ) /* BuildHeap */
        PercDown( A, i, N );
    for ( i = N - 1; i > 0; i -- ) {
        Swap( &A[ 0 ], &A[ i ] ); /* DeleteMax */
        PercDown( A, 0, i );
    }
}

```

maintain a max heap

Heapsort data start from position 0.

## Mergesort

```

void MSort( ElementType A[ ], ElementType TmpArray[ ],
            int Left, int Right )
{
    int Center;
    if ( Left < Right ) { /* if there are elements to be sorted */
        Center = ( Left + Right ) / 2;
        MSort( A, TmpArray, Left, Center ); /* T( N / 2 ) */
        MSort( A, TmpArray, Center + 1, Right ); /* T( N / 2 ) */
        Merge( A, TmpArray, Left, Center + 1, Right ); /* O( N ) */
    }
}

void Mergesort( ElementType A[ ], int N )
{
    ElementType *TmpArray; /* need O(N) extra space */
    TmpArray = malloc( N * sizeof( ElementType ) );
    if ( TmpArray != NULL ) {
        MSort( A, TmpArray, 0, N - 1 );
        free( TmpArray );
    }
    else FatalError( "No space for tmp array!!!" );
}

void Merge( ElementType A[ ], ElementType TmpArray[ ],
            int Lpos, int Rpos, int RightEnd )
{
    int i, LeftEnd, NumElements, TmpPos;
    LeftEnd = Rpos - 1;
    TmpPos = Lpos;
    NumElements = RightEnd - Lpos + 1;
    while( Lpos <= LeftEnd && Rpos <= RightEnd ) /* main loop */
        if ( A[ Lpos ] <= A[ Rpos ] )
            TmpArray[ TmpPos++ ] = A[ Lpos++ ];
        else
            TmpArray[ TmpPos++ ] = A[ Rpos++ ];
    while( Lpos <= LeftEnd ) /* Copy rest of first half */
        TmpArray[ TmpPos++ ] = A[ Lpos++ ];
    while( Rpos <= RightEnd ) /* Copy rest of second half */
        TmpArray[ TmpPos++ ] = A[ Rpos++ ];
    for( i = 0; i < NumElements; i++, RightEnd -- )
        /* Copy TmpArray back */
        A[ RightEnd ] = TmpArray[ RightEnd ];
}

```

It is hardly ever used for internal sorting, but is quite useful for external sorting, since Mergesort requires linear **extra memory**, and **copying an array is slow**.

$$T = O(N \log N)$$

## Quicksort

```
ElementType Median3( ElementType A[ ], int Left, int Right )
{
    int Center = ( Left + Right ) / 2;
    if ( A[ Left ] > A[ Center ] )
        Swap( &A[ Left ], &A[ Center ] );
    if ( A[ Left ] > A[ Right ] )
        Swap( &A[ Left ], &A[ Right ] );
    if ( A[ Center ] > A[ Right ] )
        Swap( &A[ Center ], &A[ Right ] );
    /* Invariant: A[ Left ] <= A[ Center ] <= A[ Right ] */
    Swap( &A[ Center ], &A[ Right - 1 ] ); /* Hide pivot */
    /* only need to sort A[ Left + 1 ] ... A[ Right - 2 ] */
    return A[ Right - 1 ]; /* Return pivot */
}

void Qsort( ElementType A[ ], int Left, int Right )
{
    int i, j;
    ElementType Pivot;
    if ( Left + Cutoff <= Right ) { /* if the sequence is not too short */
        Pivot = Median3( A, Left, Right ); /* select pivot */
        i = Left;    j = Right - 1; /* why not set Left+1 and Right-2? */
        for( ; ; ) {
            while ( A[ ++i ] < Pivot ) { } /* scan from left */
            while ( A[ --j ] > Pivot ) { } /* scan from right */
            if ( i < j )
                Swap( &A[ i ], &A[ j ] ); /* adjust partition */
            else break; /* partition done */
        }
        Swap( &A[ i ], &A[ Right - 1 ] ); /* restore pivot */
        Qsort( A, Left, i - 1 ); /* recursively sort left part */
        Qsort( A, i + 1, Right ); /* recursively sort right part */
    } /* end if - the sequence is long */
    else /* do an insertion sort on the short subarray */
        InsertionSort( A + Left, Right - Left + 1 );
}
```

`i=Left; j=Right-1;` because of the loop uses `++i` and `--j`. And that is because the sequent swap and sequent recursion.

If equal, make the swap happen to make two balanced arrays.

$$T_{AVG} = O(N \log N)$$

$$T_{WORST} = O(N^2)$$

## Sorting Large Structures

## § 8 Sorting Large Structures

**Problem:** Swapping large structures can be very much expensive.

**Solution:** Add a pointer field to the structure and swap pointers instead – **indirect sorting**. Physically rearrange the structures at last if it is really necessary.

[[Example]] **Table Sort**

The sorted list is

list	[0]	[1]	[2]	[3]	[4]	[5]
key	d	b	f	c	a	e
table	4	1	3	0	5	2

list [ table[0] ], list [ table[1] ], ....., list [ table[n-1] ]

**Note:** Every permutation is made up of disjoint cycles.

list	[0]	[1]	[2]	[3]	[4]	[5]
key	a	b	c	d	e	f
table	0	1	2	3	4	5

temp = d  
current = 3  
next = 3

In the worst case there are  $\lfloor N/2 \rfloor$  cycles and requires  $\lfloor 3N/2 \rfloor$  record moves.

$T = O(mN)$  where  $m$  is the size of a structure.

$table[i]$  records the current position of the element that is supposed to be here

like,  $a, b, c, \dots$  is mapped to  $1, 2, 3, \dots$ .  $e$  is supposed to mapped to 4 and its **current position** is 5, so  $table[4] = 5$

## A General Lower Bound for Sorting

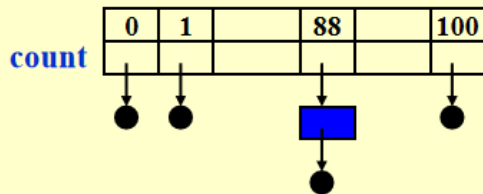
Any algorithm that sorts by comparisons only must have a worst case computing time of  $\Omega(N \log N)$ .

## Bucket Sort and Radix Sort

## § 10 Bucket Sort and Radix Sort

### ☞ Bucket Sort

[[Example]] Suppose that we have  $N$  students, each has a grade record in the range 0 to 100 (thus there are  $M = 101$  possible distinct grades). How to sort them according to their grades in **linear** time?



What if  
 $M \gg N$ ?

#### Algorithm

```
{
  initialize count[ ];
  while (read in a student's record)
    insert to list count[stdnt.grade];
  for (i=0; i<M; i++) {
    if (count[i])
      output list count[i];
  }
}
```

$$T(N, M) = O(M + N)$$

[[Example]] Given  $N = 10$  integers in the range 0 to 999 ( $M = 1000$ )  
Is it possible to sort them in **linear** time?

### ☞ Radix Sort

**Input:** 64, 8, 216, 512, 27, 729, 0, 1, 343, 125

Sort according to the **Least Significant Digit** first.

Bucket	0	1	2	3	4	5	6	7	8	9
Pass 1	0	1	512	343	64	125	216	27	8	729
Pass 2	0	512	125		343		64			
	1	216	27							
	8		729							
Pass 3	0	125	216	343		512		729		
	1									
	8									
	27									
	64									

**Output:** 0, 1, 8, 27, 64, 125, 216, 343, 512, 729

What if we sort  
according to the **Most  
Significant Digit** first?

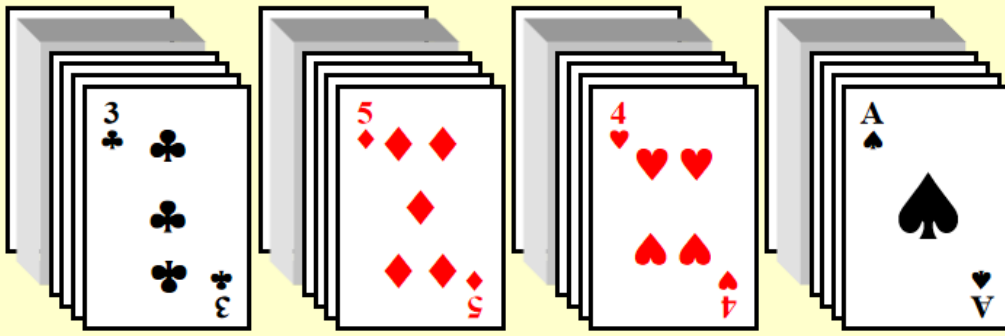
$T = O(P(N+B))$   
where  $P$  is the  
number of  
passes,  $N$  is the  
number of  
elements to sort,  
and  $B$  is the  
number of  
buckets.

Sample:

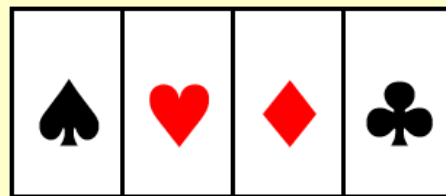


## ☞ MSD ( Most Significant Digit ) Sort

① Sort on  $K^0$ : for example, create 4 buckets for the suits

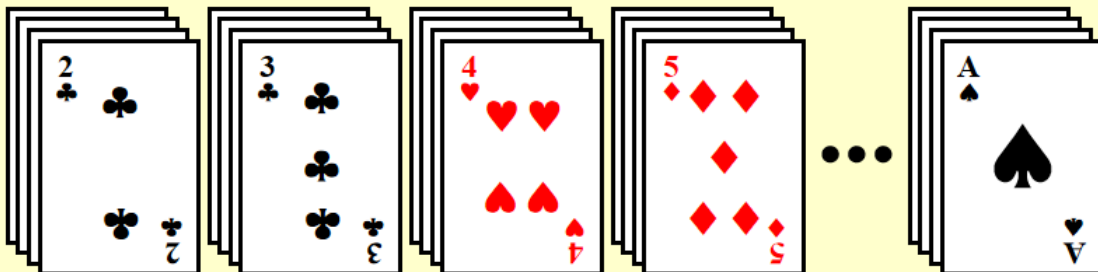


② Sort each bucket independently (using any sorting technique)



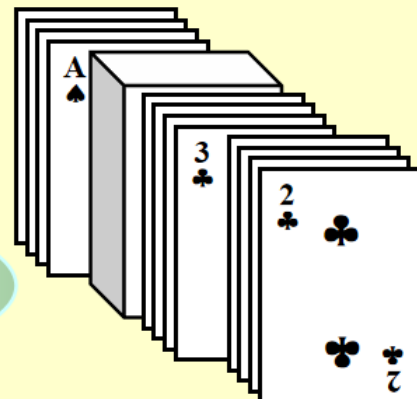
## ☞ LSD ( Least Significant Digit ) Sort

① Sort on  $K^1$ : for example, create 13 buckets for the face values



② Reform them into a single pile

③ Create 4 buckets and resort

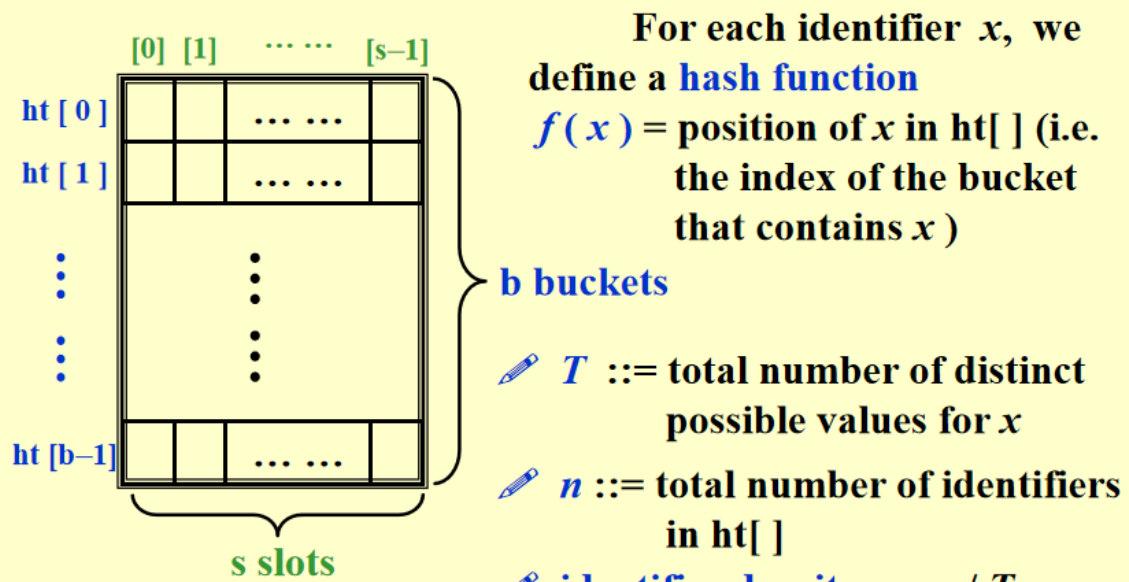


**Question:**  
Is LSD always faster than MSD?

## Hashing



## Hash Tables



- identifier density  $::= n/T$
- loading density  $\lambda ::= n/(sb)$

*Table size* should be a prime number!

Without overflow:

$$T_{search} = T_{insert} = T_{delete} = O(1)$$

**a number**

---

$$f(x) = x \% \text{Table size}$$

**a string**

---

$$f(x) = (\sum x[i]) \% TableSize ; \quad /* \text{ if } x \text{ is a string } */$$

[[Example]]  $TableSize = 10,007$  and string length of  $x \leq 8$ .

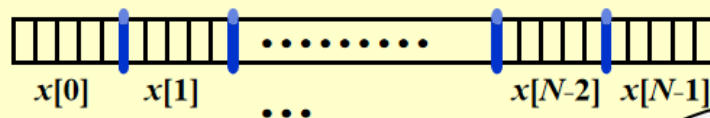
If  $x[i] \in [0, 127]$ , then  $f(x) \in [0, 1016]$  ☹

$$f(x) = (x[0] + x[1]*27 + x[2]*27^2) \% TableSize ;$$

Total number of combinations =  $26^3 = 17,576$

☹ Actual number of combinations < 3000

$$f(x) = (\sum x[N-i-1] * 32^i) \% TableSize ;$$



```
Index Hash3( const char *x, int TableSize )
{
    unsigned int HashVal = 0;
    /* 1*/ while( *x != '\0' )
    /* 2*/     HashVal = ( HashVal << 5 ) + *x++;
    /* 3*/ return HashVal \% TableSize;
}
```

Can you make it faster?  
Carefully select some characters from  $x$ .

☹ If  $x$  is too long (e.g. street address), the early characters will be left-shifted out of place.

## For collision

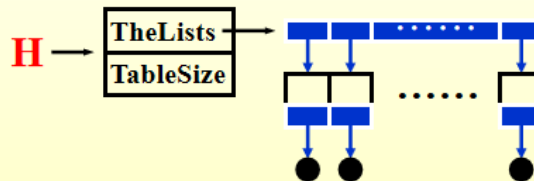
### Separate Chaining

keep a list of all keys that hash to the same value

```

HashTable InitializeTable( int TableSize )
{
    HashTable H;
    int i;
    if ( TableSize < MinTableSize ) {
        Error( "Table size too small" ); return NULL;
    }
    H = malloc( sizeof( struct HashTbl ) ); /* Allocate table */
    if ( H == NULL ) FatalError( "Out of space!!!" );
    H->TableSize = NextPrime( TableSize ); /* Better be prime */
    H->TheLists = malloc( sizeof( List ) * H->TableSize ); /* Array of lists */
    if ( H->TheLists == NULL ) FatalError( "Out of space!!!" );
    for( i = 0; i < H->TableSize; i++ ) { /* Allocate list headers */
        H->TheLists[ i ] = malloc( sizeof( struct ListNode ) ); /* Slow! */
        if ( H->TheLists[ i ] == NULL ) FatalError( "Out of space!!!" );
        else H->TheLists[ i ]->Next = NULL;
    }
    return H;
}

```



```

Position Find ( ElementType Key, HashTable H )
{
    Position P;
    List L;

    L = H->TheLists[ Hash( Key, H->TableSize ) ];

    P = L->Next;
    while( P != NULL && P->Element != Key ) /* Probably need strcmp */
        P = P->Next;
    return P;
}

```

```

void Insert ( ElementType Key, HashTable H )
{
    Position Pos, NewCell;
    List L;
    Pos = Find( Key, H );
    if ( Pos == NULL ) { /* Key is not found, then insert */
        NewCell = malloc( sizeof( struct ListNode ) );
        if ( NewCell == NULL ) FatalError( "Out of space!!!" );
        else {
            L = H->TheLists[ Hash( Key, H->TableSize ) ];
            NewCell->Next = L->Next;
            NewCell->Element = Key; /* Probably need strcpy! */
            L->Next = NewCell;
        }
    }
}

```

```
}
```

Insertion happens right after the hear node.

## Open Addressing

find another empty cell to solve collision (avoiding pointers)

### Algorithm: insert key into an array of hash table

```
{  
  index = hash(key);  
  initialize i = 0 ----- the counter of probing;  
  while ( collision at index ) {  
    index = ( hash(key) + f(i) ) % TableSize;  
    if ( table is full ) break;  
    else i ++;  
  }  
  if ( table is full )  
    ERROR ("No space left");  
  else  
    insert key at index;  
}
```

Collision  
resolving  
function.  
 $f(0) = 0.$

## Linear Probing

$$f(i) = i$$

Cause primary clustering:

any key that hashes into the cluster will add to the cluster after several attempts to resolve the collision.

## Quadratic Probing

$$f(i) = i^2$$

If quadratic probing is used, and the table size is prime, then a new element can always be inserted if the table is at least half empty.

If the table size is a prime of the form  $4k + 3$ , then the quadratic probing  $f(i) = \pm i^2$  can probe the **entire** table

For each collision, try from  $i = 1$

```

Position Find ( ElementType Key, HashTable H )
{
    Position CurrentPos;
    int CollisionNum;
    CollisionNum = 0;
    CurrentPos = Hash( Key, H->TableSize );
    while( H->TheCells[ CurrentPos ].Info != Empty &&
        H->TheCells[ CurrentPos ].Element != Key ) {
        CurrentPos += 2 * ++CollisionNum - 1;
        if ( CurrentPos >= H->TableSize ) CurrentPos -= H->TableSize; //faster
    }
    return CurrentPos;
}

```

```

void Insert ( ElementType Key, HashTable H )
{
    Position Pos;
    Pos = Find( Key, H );
    if ( H->TheCells[ Pos ].Info != Legitimate ) { /* OK to insert here */
        H->TheCells[ Pos ].Info = Legitimate;
        H->TheCells[ Pos ].Element = Key; /* Probably need strcpy */
    }
}

```

Although primary clustering is solved, secondary clustering occurs – that is, keys that hash to the same position will probe the same alternative cells.

## Double Hashing

$$f(i) = i * hash_2(x)$$

$hash_2(x) = R - (x \% R)$  with  $R$  a prime smaller than  $TableSize$ , will work well.

## Rehashing

- Build another table that is **about twice** as big
- Scan down the entire original hash table for non-deleted elements
- Use a new function to hash those elements into the new table

When to rehash?

1. As soon as the table is **half full**
2. When an insertion fails
3. When the table reaches a certain load factor

**The new position of an element is decided by the new hash function!**

0	1	2	3
null	21 and 9	null	7

after rehash:

0	1	2	3	4	5	6	7
null	9	null	null	null	21	null	7