

ADSA

Revision

# Integer Arithmetic

week 1

School method of Addition.

School method of Multiplication

Representation :

Assume that integers are represented as digit strings.

Base B number system,  $B \geq 1$

Digits  $0, 1, \dots, B-1$

$a_{n-1} a_{n-2} \dots a_1 a_0$

Store in an array  $a$  of size  $n$  represents  $\sum_{i=0}^{n-1} a_i B^i$  containing  $n$  digits.

## School Method for addition

Input: Two integers  $a = (a_{n-1} \dots a_0)$  and  $b = (b_{n-1} \dots b_0)$

Compute:  $s = a + b$ ,

where  $s = (s_n \dots s_0)$  is an  $n+1$  digit integer.

$a_{n-1} \dots a_1 a_0$  first operand

$b_{n-1} \dots b_1 b_0$  second operand

$c_n c_{n-1} \dots c_1 0$  carries

$s_n s_{n-1} \dots s_1 s_0$  sum

$c_n \dots c_0$  is sequence of carries

$$c_0 = 0, c_{i+1} \cdot B + s_i = a_i + b_i + c_i, (0 \leq i < n), s_n = c_n$$

## Pseudo-code:

$c = 0;$

for ( $i = 0, i < n, i++$ )

{

    add  $a_i, b_i$  and  $c$  to form  $s_i$  and a new carry  $c$ ;

}

$s_n := c$

## Theorem:

The addition of two  $n$ -digit integers requires exactly  $n$  primitive operations. The result is an  $n+1$  digit integer.

# Example for Multiplication

$$a = 342, b = 26, B = 10$$

1. Compute partial products

$$a \cdot b_1 = 342 \times 2$$

$$a \cdot b_0 = 342 \times 6$$

$$P_1 = 6840$$

$$P_2 = 2052$$

2. Sum up aligned partial products

$$\begin{array}{r} 6840 \\ + 2052 \\ \hline 8892 \end{array}$$

## School Method for Multiplication

Input: Two integers  $a = (a_{n-1} \dots a_0)$  and  $b = (b_{n-1} \dots b_0)$

Compute:  $p = a \cdot b$ , where  $p = (p_{2n-1} \dots p_0)$  is a  $2n$  digit-long number.

number of primitive operations:

$n$  multiplications,  $n+1$  additions,  $2n+1$  in total.

Example:  $a = 342, b_j = 6, B = 10$

computing  $c$ 's and  $d$ 's

$$a_0 \cdot b_j = 2 \cdot 6 = 12$$

summing up:

$$c_0 = 1, d_0 = 2$$

Base.

#

$$a_1 \cdot b_j = 4 \cdot 6 = 24$$

$i = 1, 0$

$$c_1 = 2, d_1 = 4$$

$$\begin{array}{r} + 842 \\ \hline 2052 \end{array}$$

$$a_2 \cdot b_j = 3 \cdot 6 = 18$$

$$c_2 = 1, d_2 = 8$$

## Pseudo - code

```
P = 0;  
for (j = 0; j < n; j++)  
{  
    P = P + a · b_j · B^j;  
}
```

## Theorem

The school method multiplies two  $n$ -digit integers with  $3n^2 + 2n = \Theta(n^2)$  primitive Operations

## Weeks 2

Recursive Version of the School Method (Divide and conquer)  
Karatsuba Multiplication.



$$a \cdot b = a_1 \cdot b_1 \cdot B^{2k} + (a_1 \cdot b_0 + a_0 \cdot b_1) \cdot B^k + a_0 \cdot b_0$$

- Divide the problem into sub-problems.
- Solve sub problems using same approach.
- Obtain solution for original problem from solution to sub solutions.

### Algorithm:

1. Split  $a$  and  $b$  to obtain  $a_1, a_0, b_1$  and  $b_0$ .
2. Compute  $a_1 \cdot b_1, a_1 \cdot b_0, a_0 \cdot b_1$ , and  $a_0 \cdot b_0$ .
3. Add the aligned products to obtain  $p = a \cdot b$ .

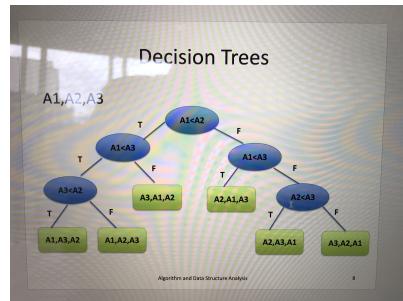
# Linear-Time Sorting Algorithm (Counting sort)

	Insertion Sort	Merge Sort	Heap Sort	Quick Sort
worst case:	$O(n^2)$	$O(n \log n)$	$O(n \log n)$	$O(n^2)$
average case:	$O(n^2)$			
best case:	$O(n)$	$O(n \log n)$		$O(n \log n)$

Theorem: all comparison sorts are  $\Omega(n \log n)$

Decision Tree : provide an abstraction of comparison sorts

e.g.



Any decision tree that sorts  $n$  elements has height  $\Omega(n \log n)$ .

# Skip List

Lecture 12 week 7

Runtime of search:  $O(\log n)$

Height of skip list:  $O(\log n)$

# of pointers  $O(2n + \lceil \log n \rceil + 3)$

关键操作:

Search()

Split L) 分割  $\rightarrow$  skip list

Concatenate (L) 聚合两个 skip list.

Delete(L)

Insert()

Space Usage: The expected space usage of a skip list with  $n$  items is  $O(n)$ .

Height: a skip list with  $n$  entries has height at most  $O(\log n)$  with probability at least  $1 - \frac{1}{n^{C-1}}$  that is asymptotically 1 for large constant  $C$ .

# Hashing (1)

Lecture 13 week 7

Worst case analysis of data structures:

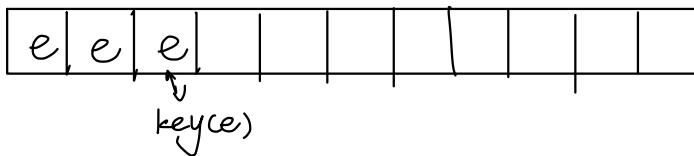
Name:	Insert( $x$ )	Remove( $x$ )	Find( $x$ )
-------	---------------	---------------	-------------

Linked List:	$O(n)$	$O(1)$	$\Theta(1)$
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AVL Trees	$O(\log n)$	$O(\log n)$	$O(\log n)$
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# Associative Arrays

关联数组

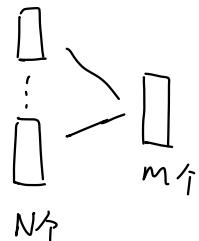


## Hash table

Idea: use hash function  $h$  to map potential  $N$  keys to  $m$  values, where  $m < N$ .

Let  $t$  be a hash table of size  $m$ .

Store element  $e$  in index  $h(\text{key}(e))$  of  $t$ .



worst case: hash function of elements returns the same value. e.g. insert zebra, boa, ABBA, cobra.

通常来说:

$$n < \frac{N}{m}$$

number of potential keys  
number of elements      number of possible hash function values

但: it is possible to have all  $n$  elements in one table entry.

## Insert( $e$ )

1. Get index  $h(\text{key}(e))$

2. Add  $e$  to the end of the list at  $t[h(\text{key}(e))]$

$O(1)$

∴ worst case insert of linked list is  $O(n)$

∴ Insert( $e$ : Element) is  $O(1)$

$O(1)$

## Find ( $k$ : Key)

∴ Worst case find of linked list is

$\Theta(n)$

∴ Find ( $k$ : key) is  $\Theta(n)$

1. Get index  $h(k)$
2. Search through list at  $t[h(k)]$
3. If element  $e$  with unique key  $k$  is in list, return  $e$ .  
Else return null.

$\Theta(n)$

$O(1)$

## Remove ( $k$ : Key)

∴ worst case find of linked list is  $\Theta(n)$

worst case remove of linked list is  $\Theta(n)$

∴ remove ( $k$  : key) is  $\Theta(n)$

1. Get index  $h(k)$
2. Search through list at  $t[h(k)]$
3. If element  $e$  with unique key  $k$  is in list, remove  $e$ .

$O(1)$

## Average Case Analysis

### Theorem 4.1

If  $n$  elements are stored in a hash table  $t$  with  $m$  entries and a random hash function is used, the expected execution time of remove or find is  $O(1+n/m)$ .

### Theorem 4.2

Let  $c$  be a positive constant. A family  $H$  of functions from Key to  $0..m-1$  is called  $c$ -universal if any two distinct keys collide with a probability of at most  $c/m$ .

### Theorem 4.3

If  $n$  elements are stored in a hash table with  $m$  entries using hashing with chaining and a random hash function from a  $c$ -universal family is used, the expected execution time of remove or find is  $O(1+cn/m)$ .

# Hashing (2)

week 8

I universal family 是什么东西?

## Alternative Approach to Hashing

Hashing with chaining is a closed hashing approach.

- closed hashing: handles collision by storing all elements with the same hashed key in one table entry.
- open hashing: handles collision by storing subsequent elements with the same hashed key in different table entries.

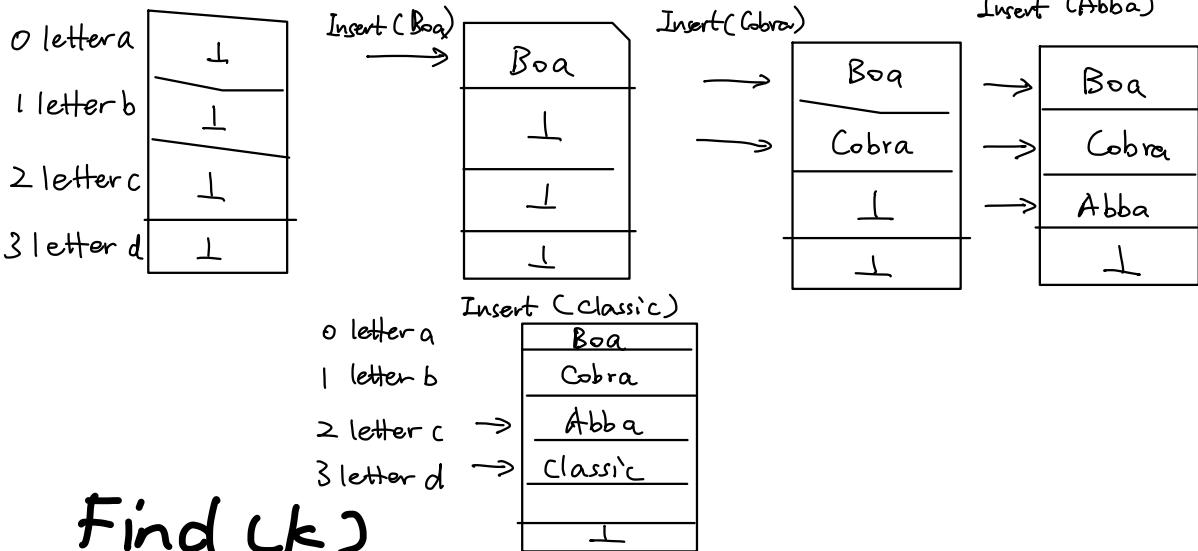
## Hashing with linear Probing (open hashing)

All unused entries in  $t$  are set to  $\perp$

When inserting on a collision, insert the element to the next free entry.

### Insert( $e$ : Element)

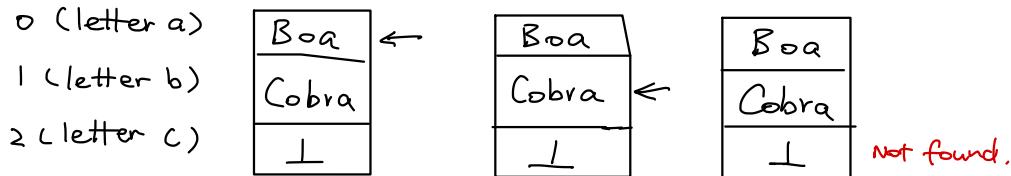
1. get index  $i = h(\text{key}(e))$
2. if  $t[i] == \perp$ , store  $e$  at  $t[i]$  只有在 $t[i]$ 是空的时候才可以 insert.
3. if  $t[i]$  is not empty, increase  $i$  by 1 and go to step 2.



Find (k)

1. Get index  $i = h(k)$
  2. If  $t[i] == \perp$ , return not found
  3. If element  $e$  at  $t[i]$  has key ( $e$ ) ==  $k$ , return found.  
Else increase  $i$  by 1 and go to step 2.

e.g. Find (ABBA)



# Remove (k)

Cannot remove the element with key(e)==k and replace it with NULL.

If we replace element e1 at t[i] with NULL, how do we find an element e2 with the same h(k)

Instead, first remove the element with key(e) ==k and then fix the invariant.

## Sudo Code:

1. Get index i=h(k) search (k)
2. If t[i] == NULL, return
3. If element e at t[i] has key(e) != k, increase i by 1 and go to step 2.
4. Set t[i] = NULL
5. Set index j=i+1
6. If t[j] ==NULL, return
7. If h(t[j]) > i, increase j by 1
8. Else set t[i] = t[j] and t[j] = NULL
9. Set i = j and go to step 5

repair

## Remove(Cobra)

0 letter a	Boa
1 letter b	Cobra
2 letter c	ABBA
3 letter d	Classic
4 letter e	⊥

Step 1, 3

Boa
Cobra
ABBA
Classic
⊥

Step 4

Boa
⊥
ABBA
Classic
⊥

Step 5

Boa
⊥
ABBA
Classic
⊥

Step 8 9 5

0 letter a	Boa
1 letter b	ABBA
2 letter c	⊥
3 letter d	Classic
4 letter e	⊥

Step 8 9 5

Boa
ABBA
Classic
⊥
⊥

Step 6

Boa
ABBA
Classic
⊥
⊥

	Chaining	VS.	Linear Probing
pro	referential integrity		use of contiguous memory
con	waste of space		gets slower as table fills up

# Graphs - I

week 8

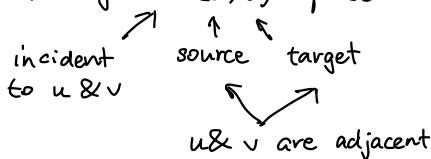
A directed graph (digraph)  $G=(V,E)$  is a pair consisting of a node set (vertex set)  $V$  and An edge set (arc set)  $E \subseteq V \times V$

$$n = |V| \quad \text{the \# of vertices} \qquad m = |E| \quad \text{the \# of edges}$$

$C: E \rightarrow \mathbb{R}$     edge weights / costs

## Terminology 术语

An edge  $e = (u, v)$  represents a connection from  $u$  to  $v$ .



\* Edge  $(v, v)$  is called a self-loop.



- The number of outgoing edges of a vertex  $v$  is called the out-degree of  $v$ :  $\text{outdegree}(v) = |\{(v, u) \in E\}|$
- The number of incoming edges of a vertex  $v$  is called the indegree of  $v$ :  $\text{indegree}(v) = |\{(u, v) \in E\}|$

**Bidirected graph** :  $(u,v) \in E \Rightarrow (v,u) \in E$

**Undirected graph** : streamlined representation of a  
bidirected graph.  
we write  $(u,v)$  and  $(v,u)$   
as a two element set  $\{u,v\}$

## Subgraph

Paths :

## Simple Graph Algorithm

Observation: node with outdegree zero can not appear in a cycle .

An undirected graph is called a tree if there is exactly one path  
between any pair of nodes.

An undirected graph is called a forest if there is at most one path  
between any pair of nodes.

0 1

Note that each component of a forest is a tree.

## Two choices for representation:

### 1. Adjacency Lists

use for each node  $v$  a double-linked list  
that stores its outgoing neighbours (alternatively  
we can also use the incoming neighbours or lists for  
both).

### 2. Adjacency Matrices

Represent a graph consisting of  $n$  nodes by  
an  $n \times n$  matrix  $A$ . Set  
 $A_{ij} = 1$  if  $(i,j) \in E$   
 $A_{ij} = 0$  otherwise.

### Advantage

- Insertion of edges goes in constant time
- well suited for sparse graphs

Insertion, removal , edge queries work in constant time .

$O(n)$  to obtain an edge entering or leaving a node.

### Disadvantage

storage requirement  $\Theta(n^2)$  even for sparse graphs.

## Graph Traversal 遍历

### breath - first - search 广度优先

use Adjacency array and Priority Queues.

Total Runtime :  $O(n+m)$

### Pseudo - code

Cheat Sheet イ- タシ!

# Depth first Search 深度优先搜索

Idea for DFS

Whenever you visit a vertex, explore in the next step one of its non-visited neighbors.

Implementation

When visiting a node, mark it as visited and recursively call DFS for one of its non-visited neighbors.

If there is no non-visited neighbor end recursive call.

Order in which nodes are finished.



# DFS Cheat Sheet

-x- +j!

Runtine:  $O(m+n)$

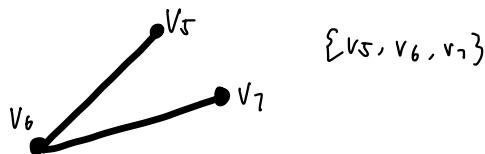
# Strongly connected components

Two nodes  $u$  and  $v$  belong to the same strongly connected component if there is a path from  $u$  to  $v$  and a path from  $v$  to  $u$ .

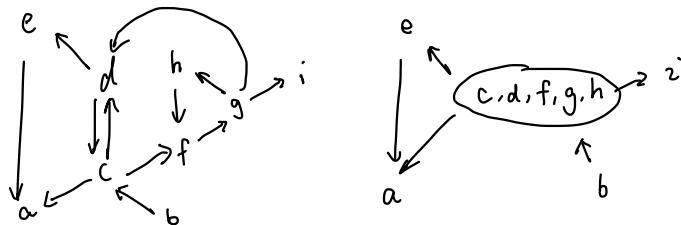
↑ undirected graph always say "yes" !

有从  $U$  到  $V$  的路也有从  $V$  到  $U$  的路.

## Undirected Graph



## Directed Graph



## Runtime

- Use adjacency lists to represent the directed graph.
- We use DFS twice (time  $O(m+n)$ )
- Have to compute the transpose graph (time  $O(m+n)$ )
- Total runtime :  $O(m+n)$



# Shortest Path

week 8

single Source Shortest Paths Problem  
单源最短路径问题

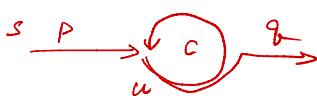
{ ① Dijkstra 单源最短路径算法，时间复杂度为  $O(E + V \log V)$ ，要求权值非负。  
② Bellman - Ford 单源最短路径算法，时间复杂度为  $O(VE)$ ，适用于带负权值情况。

All-pairs Shortest Paths Problem : 即分别以每个顶点作为源-顶点，并求其至其它顶点的最短距离。  
全源最短路径问题。

Single source shortest path problem:

Compute for a given node  $s$  of  $V$  a shortest path to any other node in  $V$ .

If a path from  $s$  to  $v$  contains a negative cycles then a shortest path does not exist ( $c$  is not defined)



有 negative cycle 存在那么就不存在最短路径

命题：最短路径中的子路径也一定是最短的。 ✓

Dijkstra's Algorithm  $\nabla$

works for acyclic graphs and for non-negative edge cost.

<https://youtu.be/pVfj6mxhdMw>

## Runtime:

Original:  
Insert and decreaseKey take time  $O(c)$   
Delete Min takes time  $O(n)$   
Total Runtime:  $O(cm+n^2)$

Improvement:  
Binary Heaps:  $O((m+n)\log n)$   
Fibonacci Heaps:  $O(m+n\log n)$

# Bellman - Ford Algorithm

Solves the problem for arbitrary edge costs.  
任意一边长度.

可以有负权边，不能有负权环。 Bellman - Ford 算法可判断图中是否有负权环。

如果一个图没有负权环，从一点到另外一点的最短路径，最多经过所有的V个顶点，有V-1条边。  
否则，存在顶点经过了两次，既存在负权环。

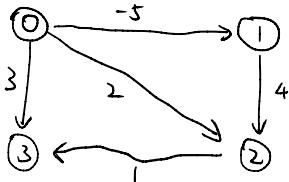
松弛操作：尝试经由一个点，再回到原来的点，得到更短的路径，多一条边，权值更小。

对所有的点进行  $V-1$  次松弛操作，理论上就找到了从源点到其它所有点的最短路径。如果还可以继续松弛，  
所说原图中有负权环。

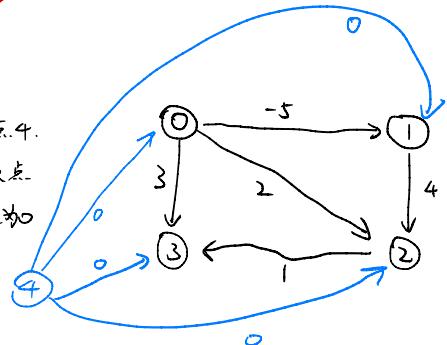


# All-pairs Shortest Paths Problem

week 10



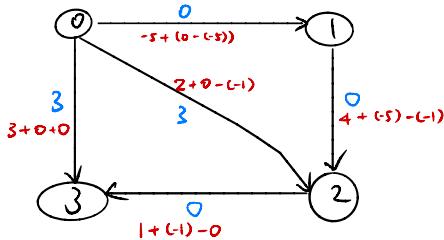
首先，新增一个源顶点4.  
使其与所有顶点连通，新边赋权值为0



使用 Bellman-Ford 算法计算新的顶点到其它顶点的最短路径，则从4至{0,1,2,3}的最短路径分别是{0, -5, -1, 0}。当得到这个h[u]信息后，将新增的顶点4移除，然后使用如下公式对所有边的权值进行 re-weight.

$$w[u, v] = w[u, v] + (h[u] - h[v])$$

此时所有权值都为非负，就可以用 Dijkstra 算法对每个顶点进行最短路径计算了。



Runtime:  $O(V^2 \log V + VE)$

# Floyd-Warshall Algorithm

针对稠密图的 Floyd-Warshall 算法 时间复杂度为  $O(V^3)$

2 种可能结果：

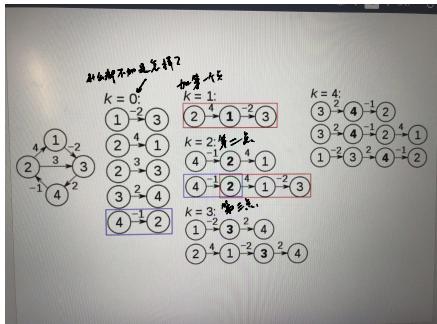
- ① The node  $k+1$  does not improve the shortest path from  $i$  to  $j$ .
- ② we get a shorter path by going from  $i$  to  $k+1$  and from  $k+1$  to  $j$ .

Take the best option, we get:

$$d_{k+1}(i, j) = \min \{ d_k(i, j), d_k(i, k+1) + d_k(k+1, j) \} \quad \checkmark$$

The length of a shortest path from  $i$  to  $j$  in given Graph G is  $d_{n-1}(i, j)$

e.g.



The distance matrix at each iteration of  $k$ , with the updated distances in bold, will be:

$k=0$	$j$	1	2	3	4
i	1 0 $\infty$ -2 $\infty$				
1	2 4 0 3 $\infty$				
2	$\infty$ 0 2 0				
3	$\infty$ -1 $\infty$ 0				

$k=1$	$j$	1	2	3	4
i	1 0 $\infty$ -2 $\infty$				
1	2 4 0 2 $\infty$				
2	$\infty$ 0 2 0				
3	$\infty$ -1 $\infty$ 0				

$k=2$	$j$	1	2	3	4
i	1 0 $\infty$ -2 $\infty$				
1	2 4 0 2 $\infty$				
2	$\infty$ 0 2 0				
3	$\infty$ -1 $\infty$ 0				

$k=3$	$j$	1	2	3	4
i	1 0 $\infty$ -2 0				
1	2 4 0 2 4				
2	$\infty$ 0 2 0				
3	$\infty$ -1 0 2				
4	3 -1 1 0				

$k=4$	$j$	1	2	3	4
i	1 0 -1 -2 0				
1	2 4 0 2 4				
2	5 1 0 2				
3	3 -1 1 0				

		1	2	3	4
	j	1	2	3	4
i	1	0 -1 -2 0			
2	4 0 2 4				
3	5 1 0 2				
4	3 -1 1 0				

Runtime:

$$\mathcal{O}(n^3)$$

# Minimum Spanning Trees

week 10

最小生成树.

定义: Given a connected undirected graph  $G = (V, E)$  with positive edge costs  $c: E \rightarrow \mathbb{R}$ .  
Find a connected subgraph  $T$  of  $G$  that has minimal cost.

例子: Nodes are the computers of the network  
Edges are possible connections between computers  
Costs are the cost of establishing a particular connection.

### Minimum Spanning Trees

- Edge costs are positive.
- Implies that the connected subgraph of minimal cost does not contain a cycle.
- It is a tree spanning all nodes of the graph (called a spanning tree).
- A spanning tree of minimal cost is called a minimum spanning tree (MST).

An MST of a given graph  $G$  can be constructed by **Greedy Algorithm**.

贪心算法。

### Crucial Properties:

**cut property:** Let  $e$  be an edge of minimum cost in a cut  $C$ . Then there is an MST that contains  $e$ .

**cycle property:** an edge of maximal cost in any cycle does not need to be considered for computing an MST.

### Kruskal's algorithm

Sort in the edge of the graph with respect to increasing weights.

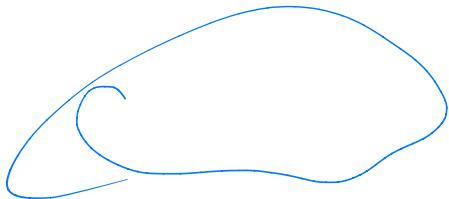
Introduce the edges in ascending order that do not create cycles.

**Runtime:** Sorting of the edges takes time  $O(n \log m)$

We need to

- test whether an edge connects two different components
- join the two components when an edge is introduced.

20

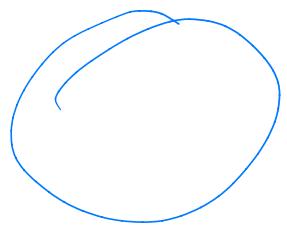


Union Find Data Structure 联合查找数据结构 -

List-oriented data structure

Jarnik-Prim Algorithm

—  
—  
—



Prim Algorithm	Kruskal Algorithm
$O(m + n \log n)$	$O(m \log m)$
more efficient for dense graphs. i.e. where $m = \Theta(n^2)$ holds	

# P & NP

week 11

## Examples

Problems that can be solved in polynomial time:

- Integer Addition -  $O(n)$
- Integer Multiplication –  $O(n^2)$
- Test whether a graph is acyclic
- Shortest paths – Dijkstra  $O(m+n^2)$
- Minimal spanning trees –Kruskal  $O(m \log m)$ .
- Almost all problems that we consider in this course.

## Difficult Problems

There are many problems for which no efficient algorithm is known.

Examples (see Mehlhorn/Sanders page 54):

- Hamiltonian cycle problem
- Traveling Salesman Problem
- Boolean Satisfiability Problem
- Clique Problem
- Graph Coloring Problem

## Class NP

Guess an answer

Verify the answer true/false

If we can use polynomial time to verify the answer is true.

The the problem is NP

### Class NP

- How can we prove a problem is NP?
  - A problem is not in NP if its solution cannot be verified in polynomial time.
- Any example?
  - All optimisation problems, whose answers cannot be checked in polynomial time.
  - TSP
  - Bin Packing
  - Timetabling

Data Structures and Algorithms

### NP-Complete Problems

- We don't know whether polynomial time algorithms exists for the mentioned problems.
- It is very likely (and almost all people in computer science believe) that there are no polynomial time algorithms for these problems.
- They belong to a class of equivalent problems known as **NP-complete problems**. (NP stands for “nondeterministic polynomial time”)

Data Structures and Algorithms

## Class P

- A decision problem is **polynomial solvable** iff its characteristic function is polynomial-time computable.
- We use **P** to denote the class of polynomial-time-solvable decision problems.

能在 polynomial 的时间 内 解决 什么 问题

MST problem 是 Class P ?

- Hint: Upper bound on Kruskal

$$O(m \log n) < O(m^2)$$