

算法正确性证明 Correctness

1.1 Loop Invariant: LI

LI: At the start of each iteration of `for loop`, keys originally in $A[1 \dots j-1]$ in sorted order.

```

Insertion-Sort( $A$ ) // sorts  $A[1 \dots n]$ 
1   for  $j = 2$  to  $A.length$ 
2      $key = A[j]$ 
3     // Insert  $A[j]$  into the sorted sequence  $A[1 \dots j-1]$ 
4      $i = j - 1$ 
5     while  $i > 0$  and  $A[i] > key$ 
6        $A[i + 1] = A[i]$ 
7        $i = i - 1$ 
8      $A[i + 1] = key$ 

```

(Basis) Initialization: LI True Initially

(Inductive) Maintenance: If LI True ^(IH) before an arbitrary iteration of loop
it remains true before next one

Termination: LI True when loop ends.

对本题: Init: $j=2$ holds

Maintain: $A[1 \dots j-1]$ sorted : insert j -th item after k -th position if $A[k] \leq A[j] < A[k+1]$
 \therefore sorted order is maintained

Analysis's Running Time

$T(n)$ = worst-case running time on input of size n .

$$\text{Ansatz: } \sum_{j=2}^n (j-1) = \frac{n(n-1)}{2} = \Theta(n^2)$$

1.2 Asymptotic Notation

Input Size: Time gets larger as inputs do

Parameterize input: size n

O -notation: O -notation: There exist positive constant c and n_0 such that

$$O(g(n)) = 0 \leq f(n) \leq c g(n) \text{ for all } n \geq n_0 \quad \text{upper bound}$$

Ω -notation: Ω -notation: There exist positive constant c and n_0 such that

$$\Omega(g(n)) = 0 \leq c(g(n)) \leq f(n) \text{ for all } n \geq n_0 \quad \text{lower bound}$$

Θ -notation: Θ -notation $T(n) = \Theta(g(n)) \iff T(n) = O(g(n)) = \Omega(g(n))$

i.e. $0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n)$ (e.g. $T(n) = 4n^2 + 22n + 12 = \Theta(n^2)$) Tight bound

Properties: Const: $\Theta(1)$

$$\text{Addition: } O(f(n)) + O(g(n)) = O(\max(f(n), g(n)))$$

$$\text{Multiplication: } O(f(n)) \cdot O(g(n)) = O(f(n) \cdot g(n))$$

$$T(n) \leq 4 \cdot T\left(\frac{n}{2}\right) + O(n) = O(n^2)$$

$$\text{分析: } X \cdot Y = (x_1 + x_0)(y_1 + y_0) \stackrel{\cong}{=} \underline{x_1 y_1} + \underline{x_0 y_1} + \underline{x_1 y_0} + \underline{x_0 y_0}$$

$$(x_1 + x_0)(y_1 + y_0) - x_1 y_1 - x_0 y_0$$

$$\Rightarrow 3 \cdot T\left(\frac{n}{2}\right) + O(n) = O(n^{1.585})$$

```

FastMultiply(x, y) // x, y n-digit integers, assume n = 2^k
1  if n == 1
2    use multiplication table to find z = x * y
3  else divide x, y in half:
4    x = 10^{n/2}x_1 + x_0
5    y = 10^{n/2}y_1 + y_0
6    // x_1, y_1, x_0, y_0 each have n/2 digits
7    P_1 = FastMultiply(x_1, y_1)  $T\left(\frac{n}{2}\right)$ 
8    P_0 = FastMultiply(x_0, y_0)  $T\left(\frac{n}{2}\right)$ 
9    P_2 = FastMultiply(x_1 - x_0, y_1 - y_0)  $T\left(\frac{n}{2}\right)$ 
10   T = P_1 + P_0 - P_2
11   z = 10^n P_1 + 10^{n/2}T + P_0
12  return z

```

Week 2.

2-1 Quick Sort

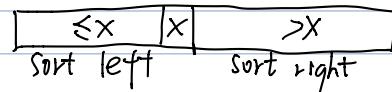


Moving data while dividing

Quick Sort : $\left\{ \begin{array}{l} \text{Divide and conquer} \\ \text{Sorts in place} \end{array} \right.$

1. Divide and conquer

1. Divide : Partition input array into 2 subarrays with pivot x
such that elements in left $\leq x$
elements in right $> x$



2. conquer : recursively sort left, sort right

3. combine : None : B/c sort in place

Partition pseudocode

```
Partition (A, p, r)          // A[p...r]      p=1  r=n
01  X = A[r]                // pivot = A[r]
02  i = p - 1               // i = last element of  $\leq x$  subarray so far
03  for j = p to r-1
04    if A[j]  $\leq x$ 
05      i = i + 1
06    exchange A[i] and A[j]
07  exchange A[i+1] and A[r]
08  return i+1
```

Invariant

Initially :

The diagram shows a horizontal array from index p to r . The element at index r is enclosed in a box and labeled ' X '. An arrow points from index $p-1$ to the element at index r , and another arrow points from index $p-1$ to index i .

Later :

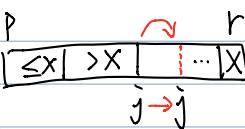
The diagram shows the array after partitioning. The subarray $A[p:i]$ contains elements $\leq x$. The subarray $A[i+1:r]$ contains elements $> x$. The pivot x is at index $i+1$. Arrows point from index p to the start of the $\leq x$ subarray, and from index $i+1$ to the end of the $\leq x$ subarray.

End :

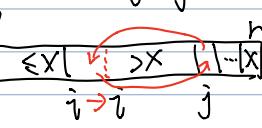
The diagram shows the final state of the array. The subarray $A[p:i]$ contains elements $\leq x$. The element at index $i+1$ is the pivot x . The subarray $A[i+1:r]$ contains elements $> x$. Arrows point from index p to the start of the $\leq x$ subarray, and from index $i+1$ to the end of the $\leq x$ subarray.

is $A[j] \leq x$?

if $A[j] > x$:



if $A[j] \leq x$:



Initialization: Before loop begins $A[r] = \text{pivot}$ and sub-arrays $A[p..i]$, $A[i+1..j-1]$ are empty. Conditions 1–3 hold.

Maintenance: During loop execution, if $A[j] \leq \text{pivot}$, $A[j]$ and $A[i+1]$ are exchanged and i and j are incremented. If $A[j] > \text{pivot}$, then only j is incremented. Conditions 1–3 hold in both cases.

Termination: When loop terminates, $j = r$, so all entries in A are partitioned into one of the 3 cases: $A[p..i] \leq \text{pivot}$, $A[i+1..j-1] > \text{pivot}$, $A[r] = \text{pivot}$.

\therefore Conditions 1–3 always hold.

Quick-Sort Pseudocode

```
Quicksort (p,q,r) // A [p...r] p=1 r=n
01. if p < r
02.   q = Partition (A, p, r)
03.   Quicksort (A, p, q-1)
04.   Quicksort (A, q+1, r)
```

Initial call
Quicksort (A, 1, n)

Correctness : 1st: pivot element is put in correct (sorted) position \exists in final
2nd: after partition no element move to the other side \exists sorted order

Time analysis

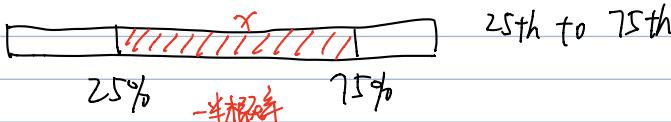
```
Quicksort(A, p, r) // T(n)
1 if p < r // O(1)
2   q = Partition(A, p, r) // Θ(n)
3   Quicksort(A, p, q - 1) // T(q - 1)
4   Quicksort(A, q + 1, r) // T(n - q)
```

Recurrence:

$$T(n) = \underbrace{T(q-1) + T(n-q)}_{\text{cost of subproblems}} + \underbrace{\Theta(n)}_{\text{Partition cost}}$$

Average Running time : close to best

\Rightarrow call a pivot \times good if it lies between



$$T(n) \leq T\left(\frac{n}{4}\right) + T\left(\frac{3}{4}n\right) + \Theta(n)$$

$$T(n) = \Theta(n \lg n)$$

Best case : pivot always split array into 2 subarrays of size $\frac{n}{2}$

$$T(n) \leq 2T\left(\frac{n}{2}\right) + \Theta(n) = \Theta(n \lg n)$$

(same as merge sort)

Worst case : require $\Omega(n^2)$ comparisons $q=n$ or $q=1$

$$A[1..n] = [1, 2, \dots, n] \quad \text{或} \quad [n, \dots, 1]$$

1st pivot: n partition: $n-1$

$$[\underbrace{1, \dots, n-1}_{L}, \underbrace{n}_{R}] \xrightarrow{O(n)}$$

2nd pivot: $n-1$ partition: $n-2$

$$\text{Total} = 1 + \dots + n-1 = \frac{n(n-1)}{2}$$

$$T(n) = \Theta(n^2)$$

CLRS 4.4-9 p.93

定理 for $a < b < c$ 且 $a+b=c$ $T(n) = T(an) + T(bn) + \Theta(n)$ solves to $\Theta(n \lg n)$

2-2 Randomized - Quicksort

Partition around random element

Randomized-partition (A, p, r)

- 01 $i = \text{random}(p, r)$ // pick i at random
- 02 exchange ($A[r], A[i]$)
- 03 return partition (A, p, r)

CLRS P.17

Analysis: Compute a bound on expected # of comparisons

Lemma: for randomized quicksort $E(\# \text{comparison}) \in [1, 2n \ln n]$

Proof: input $A[1] \dots A[n]$ let s_1, \dots, s_n be sorted (A)

a pair s_i and s_j are compared at most once during algo.

Why? 1st time compare either s_i or s_j pivot, have compare again

for $i < j$ let $R_{ij} = \begin{cases} 1 & s_i, s_j \text{ compared} \\ 0 & \text{o/w} \end{cases}$

Let $R = \text{total } \# \text{ of comparisons}$

$$R = \sum_{i=1}^{n-1} \sum_{j=i+1}^n R_{ij}$$

$$E(R) = E\left[\sum \sum R_{ij}\right] = \sum \sum E(R_{ij}) = 1 \cdot \Pr(R_{ij}=1) + 0 \cdot \Pr(R_{ij}=0)$$

$$\hookrightarrow = \Pr(R_{ij}=1)$$

Note: 1st time a pivot is chosen from $s_i \dots s_j$ it's equally likely to be any of them

$$\Pr(R_{ij}=1) = \frac{2}{j-i+1}$$

$$\sum_{i=1}^{n-1} \sum_{j=i+1}^n E(R_{ij}) = 2 \cdot \sum_{i=1}^{n-1} \left(\frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n-i+1} \right) \leq 2 \sum_{i=1}^{n-1} \underbrace{\left(\frac{1}{2} + \dots + \frac{1}{n} \right)}_{\text{constant}} = \int_2^n (n-x) dx \leq 2 \cdot n \cdot \ln n$$

2-3 Hashing

Hashing Basic

Chained Hash Tables - Universal Hashing

Open Addressing

- Linear Probing
- Double Hashing

Hashing :

Dictionary Problem : Maintain a dynamic set of items, each with a key, in a table.

3 operations:

- insert (item) add item
- delete (item) remove item
- search (key) return item

$O(1)$
per operation
in expectation

(Assume items have distinct keys)

①

Hash table:

Give short name, # between 1~250 to each of 2^{32} IP Addresses.

Store records of 250 IP's in table of size 250 . indexed by short names.

What if more than one record associated with same name? \rightarrow collision

Two issues:
① How assign "short names"
② How resolve "collision"

① General Set Up:
• Universe: $U = \{1, \dots, 2^w\} \rightarrow$ Table $T[0 \dots m-1]$
 2^{32} IPs
 250 Entries ($m=250$)

• Subset $S \subseteq U$ $|S|=n$

• Hash Func $h: U \rightarrow \{0, 1, \dots, m-1\}$ store $x \in U$ at $T[h(x)]$

when $\begin{cases} x \neq y \\ h(x) = h(y) \end{cases} \Rightarrow$ collision
 $\text{goal} \Rightarrow$ rare
cheap

②

Chain hash table

Chaining { table $T[0 \dots m-1]$

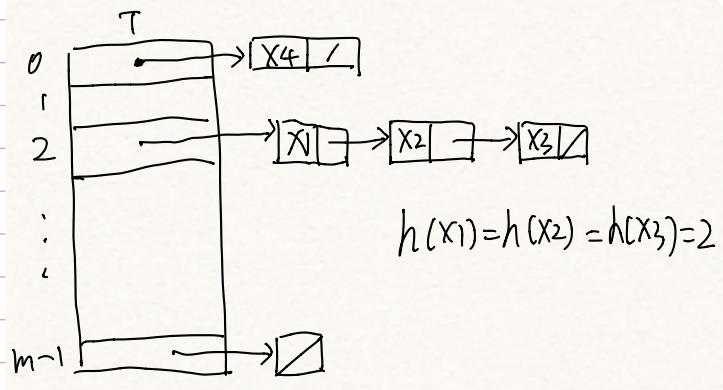
$T[i]$ is linked list of keys with $h(x)=i$

■ Insert

key
Hash-insert (T, x)

insert at head of list $T[h(x)]$ $O(1)$

if allow duplicates $O(\text{length of list})$



■ search

Hash-search (T, x)

search for element with key x in $T[h(x)]$

Time: $O(\text{length of list})$

Running time:

proportional to the length
of list of element in slot $h(x)$

Time for search

$O(l + \ell(x))$ $\ell(x)$: # of keys
such that
compute $h(x)$ such that
 $h(x) = h(c)$

■ Time analysis (hash n items into table of size m)

Worst case : all hash to same position : search $\Theta(n)$

Expected Time : • Assumption: each key is equally likely to hash to any slot of table independent of where other key hash to

• Define : load factor : $\alpha = \frac{n}{m} = \underbrace{\text{avg}}_{\text{Avg length}} (\# \text{keys per slot})$

• expected performance given assumption

$$O(1 + E[\ell(x)]) \quad E[\ell(x)] = \frac{n}{m} = \alpha$$

$$\begin{aligned} \text{Time} &= O(1 + \alpha) &= O(1) &\text{ iff } \alpha = O(1) \text{ i.e. } m = \Omega(n) \\ &\uparrow \quad \leftarrow \text{search list} && \text{clearly like } m = \Omega(n) \text{ and } m = O(n) \\ \text{compute } i = h(x) &&& \Rightarrow m = \Theta(n) \\ &&& \text{e.g. } \frac{1}{4} \leq m \leq 4n \text{ very nice!} \end{aligned}$$

■ How compute $h(x)$?

keys as natural values. e.g. string \rightarrow ASCII

$$\begin{aligned} \text{"CLRS"} &= 67 \times 128^3 \\ &+ 78 \times 128^2 \\ &+ 82 \times 128^1 \\ &+ 83 \times 128^0 \end{aligned}$$

- Division method : $h(x) = x \bmod m$

advantage: easy $O(1)$

disadvantage: must avoid certain values of m power of 2

BAD: if $m = 2^p$ for integer p . then $h(x)$ is

just p least significant bits of x

$h(x)$ 只是 x 的 p 个最低有效位

e.g. CLRS = $\dots + 83 \times 128^0$

↓ divide

128: 1010011

↑ divide

ABCS = $\dots + 83 \times 128^0$

- $x_1 \equiv x_2 \pmod{m} \Rightarrow \text{prob(collision)} = \frac{1}{m}$

- But if keys: $k, 2k, 3k, \dots$, if $k|m$ have common factor of d
 \Rightarrow we only use $\frac{d}{d}$ of table

→ Good Practice: Choose m prime, not too close to power of 2

- universal hashing: randomize hash function

Fix in advance: A set of hash function H

H : - easy
- cheap

At run time choose $h \in H$ at random

Statistical Property: universal = for all keys $x, y \in U$, $x \neq y$

$$\Pr_{h \in H} [h(x) = h(y)] \leq \frac{1}{m} \quad m: \text{table size}$$

example: Multiply-add hash func

Run Time {

- choose prime number $p > |U|$
- choose integer $a \in \{1, \dots, p-1\}$ randomly
- choose integer $b \in \{0, 1, \dots, p-1\}$ randomly

$$\Rightarrow h_{a,b}(x) = ((ax + b) \bmod p) \bmod m$$

$\in \{0, 1, \dots, p-1\}$

$\in \{0, 1, \dots, m-1\}$ ∈ table index

$$\text{Worst case : } x_1 \neq x_2 \quad \underset{a,b}{\text{Prob}} \{ h(x_1) = h(x_2) \} = \frac{1}{m}$$

$$E(\text{collision}) = \frac{h}{m} = d$$

Week 3. Dynamic Programming

Fibonacci Members

0. 1. 1. 2. 3. 5. 8. 13. 21

转移方程

$$F(n) = F(n-1) + F(n-2) \quad \text{if } n > 1 \quad F(0) = 0 \quad F(1) = 1$$

$$\text{Exponential in } n : F(n) = 2^{0.694n} = (1.618)^n = \infty \quad \text{指数级}$$

$F(n)$ // 问题 N

```

01 if n=0 return 0    O(1)
02 if n=1 return 1    O(1)
03 return F(n-1) + F(n-2) O(1)
  }
```

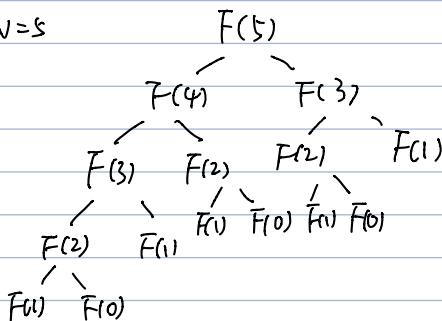
$T(n) = \# \text{ computer steps to compute } F(n)$

$T(n) \leq 2 \text{ for } n \leq 1$

$$T(n) = T(n-1) + T(n-2) + 3$$

e.g. $N=5$

$F(5)$



=>

Many computations repeated

↓

这样子做的话如果 $n=0$ return 0

02 create $f[0..n]$

03 $f[0] = 0 \quad f[1] = 1$

04 for $i = 2$ to n

05 $f[i] = f[i-1] + f[i-2]$

06 return $f[n]$ $O(n) \rightarrow \text{linear}$

Dynamic Programming (DP)

Knapsack Problem

CLRS PP425-427

N items

Weight: w_1, w_2, \dots, w_n

Value: v_1, v_2, \dots, v_n

W : weight limit

Optimization Problem:

find $S \subseteq \{1, \dots, n\}$ subject to constraint: $\sum_{k \in S} w_k \leq W$

↑
subject

object: $\text{Max}(\sum_{k \in S} v_k)$

(possible choices): $2^n = C(n, 0) + C(n, 1) + \dots + C(n, n)$

△ exponentially

↓ ?

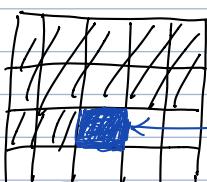
polynomial

⇒ 指数限制 ⇒ 有界背包 ⇒ w_i 和 W ⇒ 正整数

Theorem: can be solved in $O(nW)$ time

pseudo polynomial

DP Idea



Subproblem:

a knapsack problem
with different inputs

↑
should be easy based on before result

状态转移表

For this case:

	0	$j \rightarrow$	W
1	0 0 0	0 0 0	
	0 [i,j-w]		
\downarrow	0	0	
i	0	0	
j	0	0	
n	0	0	

i : item
 j : weights

$M[i, j] = \max \underline{\text{value}} \text{ for } [i, j]$

$[i, j]$: in knapsack restricted to { items 1 ... i sum(weights) $< j$

$$\Rightarrow \left\{ M[i, j] = \max \left\{ \sum_{k \in S} V_k : \sum_{k \in S} W_k \leq j \right\} \right. \\ \left. S \subseteq \{1, 2, \dots, i\} \right\}$$

Recurrence

Base cases:
 $M[0, j] = 0$
 $M[i, 0] = 0$

knapsack ($n, W[1..n], V[1..n], W$)

01 for $i = 0$ to n // item } $O(h)$
 02 $M[i, 0] = 0$
 03 for $j = 1$ to W // weight } $O(n)$
 04 $M[0, j] = 0$

$M[i, j] = \max \begin{cases} M[i-1, j] & \text{if do not take item } i \\ M[i-1, j-W_i] + V_i & \text{if take item } i \end{cases}$

positive

保证能放下 item i

$O(nW) \Rightarrow \text{space / time}$

05 for $i = 1$ to n } $O(nW)$
 06 for $j = 1$ to W
 07 if $W_i > j$:
 08 $M[i, j] = M[i-1, j]$ // too heavy
 09 else
 10 $M[i, j] = \max(M[i-1, j], M[i-1, j-W_i] + V_i)$
 11
 12 return $M[n, W]$

我们不知道包里放了什么?

EX:	ITEM	WEIGHT	VALUE	$W=5$
	1	1	6	
	2	2	10	
	3	3	12	
	i	0 1 2 3 4 5		
L	0	0 0 0 0 0 0		
	1	0 6 6 6 6 6		
	2	0 6 10 16 16 16		
	3	0 6 10 16 18 22	拿没拿	
		#1 #2 #1,2 #1,3 #2,3		

Optional knapsack ($M[0..n, 0..W], V[1..n], W$)

01 create array $L[0..n]$
 02 $k = 0$ // # items in optional knapsacks
 03 $j = W$ // unused capability
 04 for $i = n$ down to 1
 05 if $M[i, j] > M[i-1, j]$ // take i
 06 $k = k+1$
 07 $L[k] = i$
 08 $j = j - W_i$
 09 return L

↑ 算了不少没用的.

可以真有用吗?

Memorized-Knapsack (i, j) // input $w[1..n], v[1..n], W, M[0..n, 0..w]$

- 01 create $M[0..n, 0..w]$
- 02 for $i = 0$ to n $M[i, 0] = 0$
- 03 for $j = 0$ to W $M[0, j] = 0$
- 04 if $M[i, j] < 0$ // not compute
- 05 if $j < w_i$ // Item i too heavy
- 06 value = Memorized-Knapsack ($i-1, j$)
- 07 else
- 08 value = max { Memorized-Knapsack ($i-1, j$)
Memorized-Knapsack ($i-1, j-w_i$) + v_i }
- 09 $M[i, j] = \text{value}$ ← 这一步很重要，保证不会重复计算
- 10 return $M[i, j]$

Initially call $M[n, w]$ ⇒ 也就是利用例查只算有用的部分

IV Longest Common Subsequence (LCS)

15.4 in CLRS

Problem: Give 2 sequences $X = [1..m]$ $Y = [1..n]$

Object: Find a longest subsequence common to both

↳ element must be in order but not need to be consecutive

Ex: $X: A B C B D A B$
 $Y: \begin{matrix} B \\ B \\ D \\ C \\ A \\ B \\ A \end{matrix}$

↙ B CBA
 ↙ BCAB
 ↙ BDA B

length = 4

但结果不止一个

Brute Force Algorithm

$m \leq n$ for every subsequence of X , check if it is a subsequence of Y , keep track of length

IV worst case time: ?

- ① 2^m subsequence of X to check
- ② each check need $O(n)$
 \Rightarrow worst $O(n2^m)$ exponential !! $\xrightarrow{\text{DP?}} \star$ Polynomial

Key Idea: LCS of X and Y can be expressed as smaller sequences

- ① if $x_m = y_n$, set last symbol of LCS to x_m

find LCS of $X[1..m-1] Y[1..n-1]$

- ② if $x_m \neq y_n$, then LCS of $x_i y_j$ is

LCS $X[1..m] Y[1..n-1]$

↓ Recursive & DP

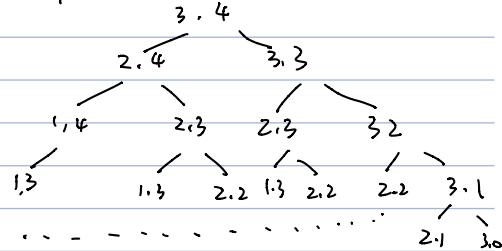
IV Recursive Method

LCS - length (x[1..m] y[1..n])

worst case : totally different

Ex: X: AAA

Y: B B B B



height: $O(m+n)$

time : $O(2^{O(m+n)})$

distinct problems: ($m \times n$)

DP method

Table of LCS Length

④ subproblems : each recursive call of form : $LCS\text{-length } (\mathbf{x}[1..i] \mathbf{y}[1..j]) \quad 1 \leq i \leq m \quad 1 \leq j \leq n$

④ Table: m rows n columns $c[i,j] = \text{LCS-length } (x[1..i], y[1..j])$

j →

 Base Case : $c[i, 0] = 0$

$$C[i, j] = \begin{cases} 1 & \text{if } x_i = y_j \\ 0 & \text{if } x_i \neq y_j \end{cases} = \max(C[i-1, j], C[i, j-1])$$

LCS-length (x, y, m, n)

01 Create new table C[0..m, 0..n]

02 for i=1 to m

] ocm)

$$[i, 0] = 0$$

] $O(n)$

Time / space $O(n \ln n)$

for $j = 1$ to n
65 $C[0, j] = 0$

2 (m+n)

b] $j = 1 \text{ to } n$

2

$j = 1 \text{ to } n$
if $x[i] = Y[i]$

$$c[i, j] = 1 + c[i-1, j-1]$$

else

$$C[i:j] = \max(C[i-1:j], C[i:j-1])$$

12 return C

Matrix Chain Multiplication

Parenthesization Problem: Optimal Evaluation
Associative ExpressionGiven "chain" (sequence) of n matrices

$$A_1 \cdot A_2 \cdots A_n$$

where for $i=1 \dots n$ matrix A_i has dimension $p_{i-1} \times p_i$

Object: To find the most efficient way to multiply these matrices

Given: $A_1 \ A_2 \ A_3 \dots A_n$ Rows: $r_1 \ r_2 \ r_3 \dots r_n$ Cols: $c_1 \ c_2 \ c_3 \dots c_n$ Where: $c_i = r_{i-1}$ $1 \leq i \leq n$

不然没法乘

两矩阵相乘 cost: $r_i \cdot c_1 \cdot c_2 \leftarrow$ $\rightarrow \min (\# \text{ of scalar multiplication})$ EX $A_1 \ A_2 \ A_3 \ A_4$

1. $(A_1 A_2) (A_3 A_4)$

2. $(A_1 (A_2 (A_3 A_4)))$

3. $A_1 ((A_2 A_3) A_4)$

∴ 第 3 种

Ex: $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix} = \begin{pmatrix} 14 & 32 \\ 32 & 77 \end{pmatrix}$

$$\begin{matrix} 2 \times 3 & & 3 \times 2 \\ & \diagdown & \\ & 2 \times 2 & \end{matrix} \quad \begin{matrix} & \downarrow \\ 12 & \end{matrix}$$

12 scalar multiplication

Fact: $12 = r_1 \times c_1 \times c_2$

Fact: $i \times j$ matrix with $j \times k$ matrix $\Rightarrow \# \text{ scalar multiplication} : i \cdot j \cdot k$

$$\textcircled{1} \quad (A_1 \times A_2) \times A_3 \quad \# \text{ multi} = 1 \times 10^3 \times 1 + 1 \times 1 \times 10^3 \\ \begin{matrix} & \checkmark \\ 1 \times 10^3 & 10^3 \times 1 & 1 \times 10^3 \end{matrix} \quad = \boxed{2 \cdot 10^3}$$

$$\textcircled{2} \quad A_1 \times (A_2 \times A_3) \quad \# \text{ multi} = 10^3 \times 1 \times 10^3 + 1 \times 10^3 \times 10^3 \\ \begin{matrix} & \checkmark \\ 10^3 \times 10^3 & \end{matrix} \quad = \boxed{2 \cdot 10^6}$$

Week 4 Data Structures : Heap, Binary Search Tree (BST)

Data structures

e.g. {
Linked-list : good for insert ; bad for search
Array : good for search ; bad for insert

Heap

Def: Realization of the abstract data type priority queue

priority queue: maintains a set S of elements X , each associated with a numeric key K supports operations:

- insert (S, x)
- Min (S) : return element of S with smallest key
- Extract-Min (S) : return element of S with smallest key and delete it
- decrease-key (S, x, k) : decreases value of x 's key to new value k

Implementation of PQ:

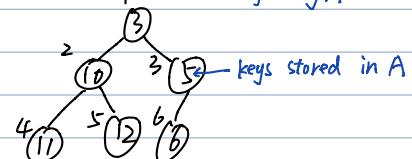
Binary Tree: An array, represented as a binary tree
(nearly complete)

arbitrary size arrays

e.g.

1	2	3	4	5	6
3	10	5	11	12	6

1 ← indices of array A



Heap as tree:

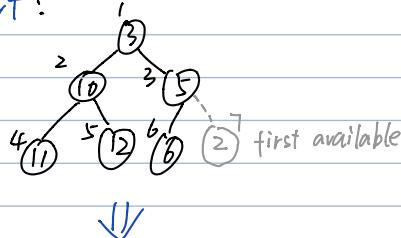
- Root: 1st element in array ($i=1$)
- parent [i] = $\lfloor \frac{i}{2} \rfloor$ (index of parent)
- left [i] = $2i$ (index of left child)
- Right [i] = $2i+1$ (index of right child)

Min-heap Property (ordering constraint)

key (parent(i)) \leq key (i) // such as example A

Heap Operations:

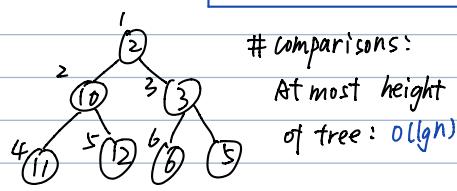
- insert :



Min-heap-insert (A, key)

```
01 heap_size[A] = heap_size[A] + 1
02 i = heap_size[A]
03 while i > 1 and A[parent[i]] > key
04     A[i] = A[parent[i]]
05     i = parent[i]
06 A[i] = key
```

但 $2 < 5$. Why??
 $2 < \text{parent}[7] = 5$
 $\text{key}[7] \leftarrow \text{key}[5]$
直到 $2 \geq 5$ 才停止



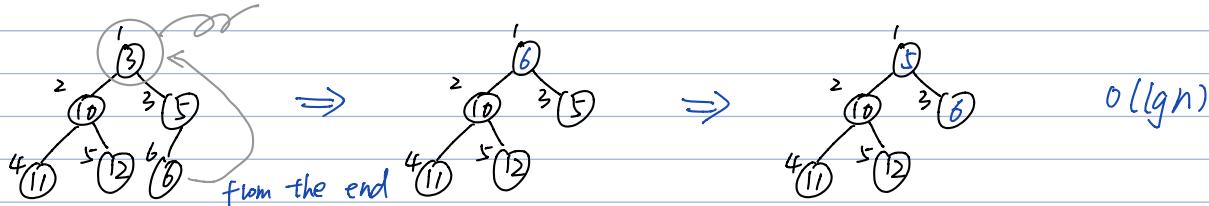
• decrease-key

similar to insert, except element already in tree $O(lgn)$

decrease-key (A, key)

- 01 if $A[i] \leq key$
- 02 return
- 03 while $i > 1$ and $A[\text{parent}[i]] > key$
- 04 $A[i] = A[\text{parent}[i]]$
- 05 $i = \text{parent}[i]$
- 06 $A[i] = key$

• extract-min



Heap-extract-min (A)

- 01 $min = A[1]$
- 02 $A[1] = A[\text{heapsize}[A]]$
- 03 $\text{heapsize}[A] = \text{heapsize}[A] - 1$
- 04 min-heapfy ($A, 1, \text{heapsize}[A]$)
- 05 return $[A]$

min-heapfy ($A, i, \text{heapsize}[A]$)

// left, right subtrees of i are min-heap
// goal: make i 's subtree a min-heap

- 01 if $\text{left}(i) \leq \text{heapsize}[A]$ and $A[\text{left}[i]] < A[i]$
- 02 then $\text{smallest} = \text{left}[i]$ //if violation
- 03 else
- 04 $\text{smallest} = i$ //if no violation
- 05 if $\text{right}(i) \leq \text{heapsize}[A]$ and $A[\text{right}[i]] < A[i]$
- 06 then $\text{smallest} = \text{right}[i]$
- 07 if $\text{smallest} \neq i$
- 08 then exchange ($A[i], A[\text{smallest}]$)
- 09 min-heap ($A, \text{smallest}, \text{heapsize}[A]$)

IV Binary Search Trees (BSTs)

① Def: Maintains complete ordering of data, rooted binary tree

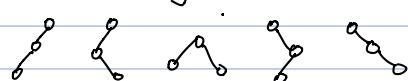
• Rooted binary tree defined on a finite set of nodes

either (1) empty

or (2) consists of a node called the root

• Together with 2 rooted binary trees: left subtree, right subtree

• Order matters



② each node X contains:

- key $[x]$ (possibly other data)
 ↳ data field

- three pointers: $\text{left}[x]$: point to left child


```

05   if key[z] < key[x]
06     x = left[x]
07   else
08     x = right[x]
09   p[z] = y
10   if y == nil
11     root[T] = z
12   else if key[z] < key[y]
13     left[y] = z
14   else
15     right[y] = z

```

$O(n)$

① Successor Assume all keys distinct

def: successor of node x is node y such that $\text{key}[y]$ is the smallest key that $> \text{key}[x]$

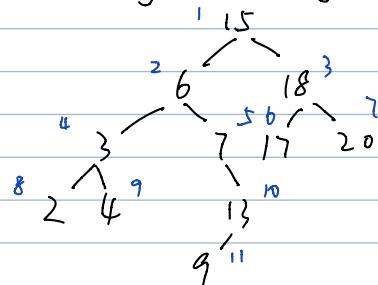
Tree-Successor(x)

```

01  if right[x] ≠ nil
02    return tree-min(right(x))
03  y = p(x)
04  while y ≠ nil and x = right[y]:
05    x = y
06    y = p[y]
07  return y

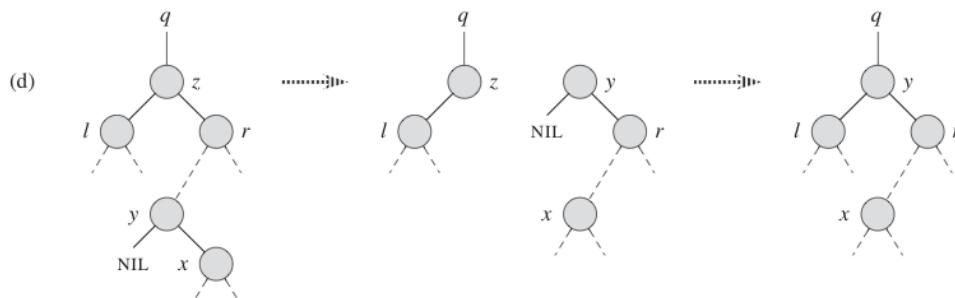
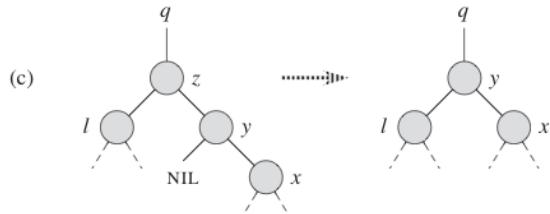
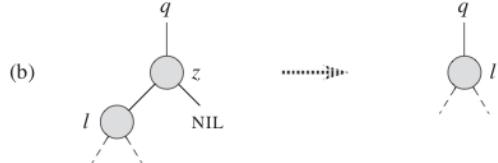
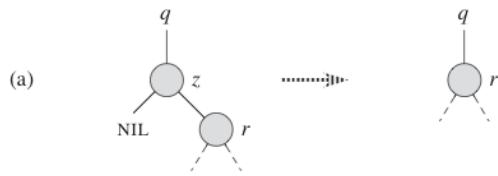
```

(有右子樹的根)



Successor[Node(value=13)] =
Node(value=15)
Successor[11] = 1

② Delete z (fix successor)



★ part one:

Transplant: replace one subtree as child of its parents by another subtree

```

Transplant (T, u, v)           // u: replaced   v: new
01 if p[u] == nil             // u is root of T
02   root[T] = v
03 else if u = left[p[u]]     // u is left child
04   left[p[u]] = v           // v is left child
05 else                         // u is right child
06   right[p[u]] = v          // v is right child
07 if v != nil
08   v.p = u.p
  
```

★ part two:

Tree-delete-node

Tree-delete (T, z)

```

01 if left[z] == nil
02   Transplant (T, z, right[z])
03 else if right[z] == nil
04   Transplant (T, z, left[z])
05 else y = Tree-min (right[z])           // z has 2 children
06   if p[y] != z                         // y is the successor but it not connect
07     Transplant (T, y, right[y])         // z & y (condition d)
08     right[y] = right[z]
09     p[right[y]] = y
10   Transplant (T, z, y) → 换枝
11   left[y] = left[z]
12   p[left[y]] = y → 换左子
  
```

↑ y & z (condition d)
换枝
换左子

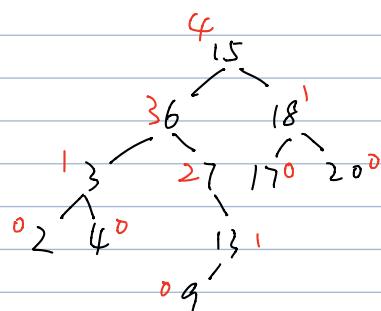
⑦ Insert (T, z) => 加入有关 height 的信息

Tree-insert (T, z) // to insert key v into BST T input node z (new node) with key[z] = v
 // left[z] = nil right[z] = nil

```

01 y=nil // parent of root
02 x=root[T]
03 while x != nil:
04   y=x
05   if key[z] < key[x]
06     x=left[x]
07   else
08     x=right[x]
09   p[z]=y
10   height[z]=0
11   if y == nil
12     root[T]=z // z is the only node in the tree
13   else if key[z] < key[y]
14     left[y]=z
15   else
        right[y]=z
  
```

height
O(n)



16 if $\text{left}[y] == \text{nil}$ or $\text{right}[y] == \text{nil}$
 17 $x = p[z]$ // z is the only child of y
 18 while $x \neq \text{nil}$
 19 $\text{height}[x] = \text{height}[x] + 1$
 20 $x = p[x]$

z is the only child of y

② Delete (T, z) // 调整高度版 T augmented

Tree-delete (T, z)

01 if $\text{left}[z] == \text{nil}$
 02 Transplant ($T, z, \text{right}[z]$)
 03 $x = p[z]$
 04 else if $\text{right}[z] == \text{nil}$
 05 Transplant ($T, z, \text{left}[z]$)
 06 $x = p[z]$
 07 else $y = \text{Tree-min}(\text{right}[z])$ // z has 2 children
 08 if $p[y] \neq z$ // y is the successor but it not connect
 09 Transplant ($T, y, \text{right}[y]$)
 10 $\text{right}[y] = \text{right}[z]$
 11 $p[\text{right}[y]] = y$
 12 Transplant (T, z, y)
 13 $\text{left}[y] = \text{left}[z]$
 14 $p[\text{left}[y]] = y$
 15 $y = \text{height}[z] - 1$
 16 $x = p[y]$
 17 while $x \neq \text{nil}$
 18 $\text{height}[x] = \max\{\text{height}[\text{left}[x]]\} + 1$
 19 $x = p[x]$

Week 5 Giraph

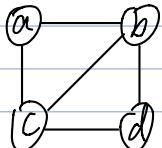
Graph: $G = (V, E)$

V : finite set of vertices ($|V| \neq 0$)

E : set of edges (vertex pairs)

Directed / undirected

Undirected Graph:

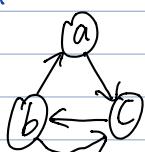


$$V = \{a, b, c, d\}$$

$$E = \{(a,b), (a,c), (b,d), (c,d), (b,c)\}$$

Note: $(u,v) = (v,u)$

Directed Graph



$$V = \{a, b, c\}$$

$$E = \{(b,a), (a,c), (c,b), (b,c)\}$$

Note: $(u,v) \neq (v,u)$

loop:



$|V|$: number of vertices

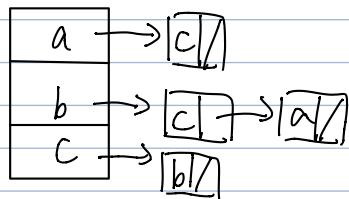
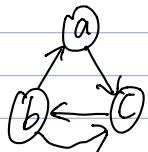
$|E|$: number of edges

$$0 \leq |E| < |V|^2$$

Graph Representations:

Adjacency List

- V = set of vertices array of $|V|$ linked list
- for each vertex $u \in V$ $\text{Adj}[u] =$ list of vertices adjacent to u .



Good for sparse graph

where $|E| \ll |V|^2$

space required: $O(|V|+|E|)$

Adjacency Matrix

Assume $V = \{1, 2, \dots, |V|\}$

Let $A = (a_{ij}) = |V| \times |V|$ matrix $\Rightarrow a_{ij} \left\{ \begin{array}{ll} 1 & \text{if } (i,j) \in E \\ 0 & \text{o/w} \end{array} \right.$



$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \end{matrix} & \end{matrix}^T$$

Good for dense graphs
where $|E| \approx |V|^2$
space = $O(V^2)$

Graph Search

Given: Graph $G = (V, E)$ Start vertex: $s \in V$

Explore: visit every vertex reachable from s

(1) s is reachable

(2) if u is reachable from s and $v \in \text{adj}[u]$ then v is reachable from s

(3) if u reachable from s then there is a path from s to u

$s \rightarrow v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k \rightarrow u$ of length $k+1$ leading from s to u

\Rightarrow find path algorithm

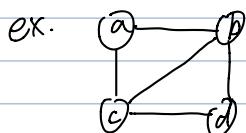
- if v discovered from u , then u parent of v

- keep track of parents $\pi[s] = \text{nil}$ s has no parent
 $\pi[v] = u$ $u \rightarrow v$

- length of path: #(Edges) in the path

\hookrightarrow simple: if all its vertices are distinct

circle: a closed path $(v_0, v_1, \dots, v_k, v_0)$



Ex. distance : $\text{dist}(u, v) = \delta(u, v)$ is the length of shortest path from u to v

Single-source shortest paths problem

- select source vertex $s \in V$

- return: list of distances $\text{dist}(s, v)$ for all $v \in V$ $O(V+E)$

- tree: $\pi[v]$ $\pi[\pi[v]]$

Breadth-First Search (BFS)

Input: $G = (V, E)$ $s \in V$ ADJ List format

Output: $d[v] = \min \#$ of edges in path from s to v

$\pi[v] =$ parent / predecessor of v in tree

Time: $O(V+E)$

Label: status of vertex

white:	not yet discovered / scanned
gray:	discovered / scanned
black:	finished

Maintain 3 variables: $\text{color}[v]$: status
 $d[v]$: distance
 $\pi[v]$: parent

Maintain a list of gray vertices in order of discovery

when discover a vertex, add it to end of list

when finish a vertex, remove it from beginning of list

\hookrightarrow FIFO \rightarrow queue

append to list $O(1)$ enqueue (Q, key)

remove from list $O(1)$ dequeue (α, key)

vertices discovered but not finished (only gray)

Ex: G $A \rightarrow C \rightarrow E$
 $\downarrow \quad \downarrow \quad \uparrow \downarrow$
 $B \rightarrow D \rightarrow F$
 $C \rightarrow D \rightarrow E$
 $D \rightarrow F$
 $E \rightarrow F$
 $F \rightarrow E$

Source: A
 BFS tree:

A	/	C	/	1
B		E	2	distan ^{ce}
D		F	3	Tree is not unique, but
C	\	B	1	the distance stay the same.

$D \rightarrow E \rightarrow F$
 2
 3

BFS (G, S)

```

01 for each u in V
02   color[u] = white
03   d[u] = ∞
04   π[u] = nil
05 color[S] = gray
06 d[S] = 0
07 Q = ∅ // queue
08 enqueue(Q, S)
09 while Q ≠ ∅:
10   u = dequeue(Q)
11   for every vertex v in Adj[u]
12     if color[v] = white
13       color[v] = gray
14       d[v] = d[u] + 1
15       π[v] = u
16       enqueue(Q, v)
17   color[u] = black

```

$O(V+E)$

Printed-Path (G, S, V)

```

01 if V == S
02 print "S"
03 else if π[V] == nil
04   print "no path"
05 else
06   print path(G, S, π[V])
07   print V

```

$O(V)$

Depth-first-search (DFS)

Input: $G(V,E)$ directed or undirected , in Adj List format

Goal: what parts of graph reachable from a given vertex

Output: $d[v]$: discovery time

$f[v]$: finish time

$π[v]$: parent of v on dfs tree

Time: $O(V+E)$

DFS (G) // Adj List Format

```

01 for each vertex u in V
02   color[u] = white
03   π[u] = nil
04   time = 0 // global variable
05 for each vertex u in V
06   if color[u] = white
07     DFS-visit(G, u)

```

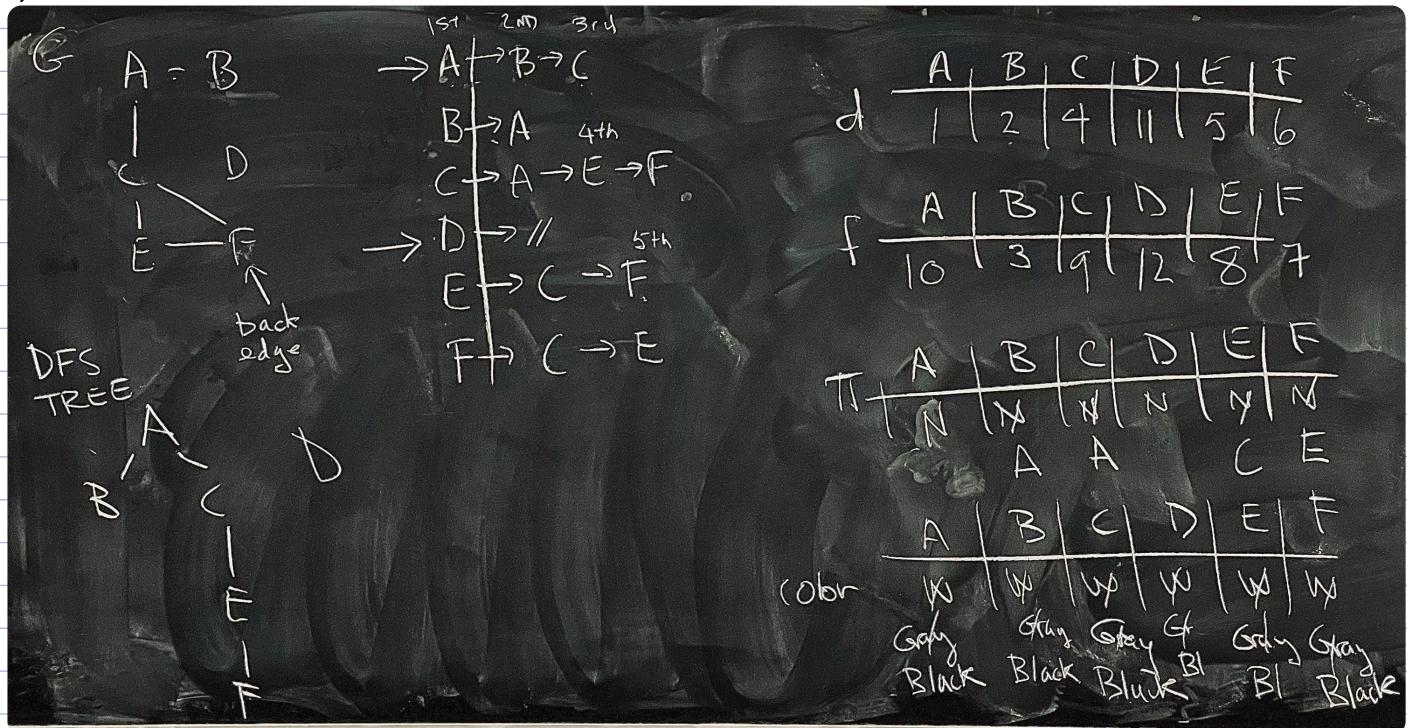
DFS-visit (G, u)

```

01 time = time + 1
02 d[u] = time
03 color[u] = gray
04 for each v in Adj[u]
05   if color[v] == white
06     π[v] = u
07     DFS-visit(G, v)
08   color[u] = black
09   time = time + 1
10   f[u] = time

```

Ex for DFS process



- Not all edges in G are in DFS tree

Tree Edges Back-Edges (None tree)

AB, AC, CE, EF

FC

Properties

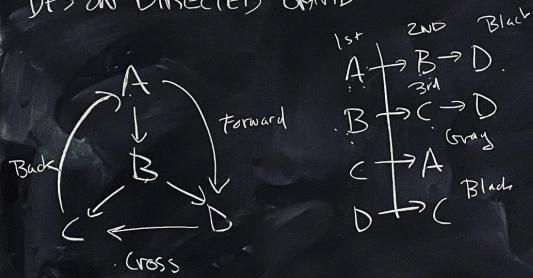
- Tree Edge (u,v) : vertex v white when edge explored 1st time

$$\begin{array}{cccccc} u=A & \overset{A}{\text{---}} & \overset{B}{\text{---}} & \overset{B}{\text{---}} & \overset{A}{\text{---}} \\ v=B & 1 & 2 & 3 & 10 \end{array}$$

- Back Edge (u,v) : vertex v gray when edge explored 1st time

$$\begin{array}{cccccc} u=F & \overset{C}{\text{---}} & \overset{F}{\text{---}} & \overset{F}{\text{---}} & \overset{C}{\text{---}} \\ v=C & 4 & 6 & 7 & 9 \end{array}$$

DFS on DIRECTED GRAPHS



	A	B	C	D
Color	W	W	W	W
$d[u]$	1	2	3	4
$f[v]$				

	A	B	C	D
π	N	N	N	N
α	A	B	C	D
b				

	A	B	C	D
d	1	2	3	5

	A	B	C	D
f	1	8	7	6

- TYPES OF EDGES (GRAY, WHITE)
- ① TREE EDGES (u,v) : $(A,B), (B,C), (B,D)$ most important
 $d[u] < d[v] < f[v] < f[u]$
 - ② BACK EDGES (u,v) : (C,A)
 $d[v] < d[u] < f[u] < f[v]$
 - ③ FORWARD EDGES (u,v) : (A,D)
 $d[u] < d[v] < f[v] < f[u]$
 - ④ CROSS EDGES (u,v) : (D,C)
 $d[v] < f[v] < d[u] < f[u]$

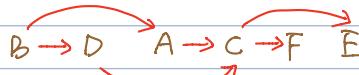
(*) tree : forward, cross $f[v] < f[u]$

(*) back $f[u] < f[v]$

Directed Acyclic Graphs (DAG)

Ex: $A \rightarrow C \rightarrow E$ \leftarrow discover/finish
 $B \rightarrow D$ $F \rightarrow E$
 $9/12 \quad 10/11 \quad 5/6$
list: E F C A D B
output by finish time
reverse \Rightarrow

$A \rightarrow C$
 $B \rightarrow A \rightarrow D$
 $C \rightarrow E \rightarrow F$
 $D \rightarrow C$
 $E \rightarrow //$
 $F \rightarrow //$



topological sort :

Goal: a list of vertices such that all edges go from left to right

topological-sort(G)

01 $L = \emptyset$

02 DFS(G) to compute $f[v]$ $O(v+E)$

03 when vertex finish append to L

04 reverse L $O(v)$

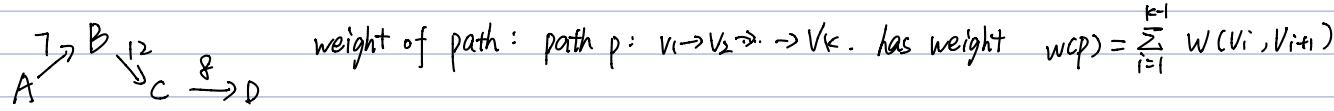
$O(v+E)$

Ex.

Week 6 Graph

Shortest path

weighted directed graph $G = (V, E, W)$ where $W: E \rightarrow \mathbb{R}$ associated with a real number to each edge



shortest path: from u to v is a path of minimum weight from u to v

shortest path weight: notation

$$s(u, v) = \min \{w(p) : p \text{ path from } u \text{ to } v\}$$

when do shortest path not exist?

negative edge weights \Rightarrow some shortest path may not exist

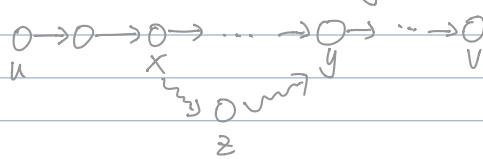
e.g.



No path exist: $s(u, v) = \infty$

Claim: A subpath of a shortest path is a shortest path

Proof: suppose we have a subpath of a shortest path



R: there is another path from $x \rightarrow z \rightarrow y$ shorter than $x \rightarrow y$
 $\dots \Rightarrow u \rightarrow v$ longer than $u \rightarrow z \rightarrow v \Rightarrow$ contradiction R is false

Dijkstra's Algorithm

input: (G, W, S) $G = (V, E)$ directed, adj list format

$W: \text{weight function } W: E \rightarrow \mathbb{R}$

$S \in V$: source vertex

output: $d[v]$: distance = $s(S, v)$

$\pi[v]$: parent of v on shortest path to v from S

The unique path from S to each vertex in shortest paths tree is a shortest path

Maintains a current cost to every vertex

$d[v]$: current min cost of reaching vertex V .

$\pi[v]$: parent of V

Dijkstra (G, W, s)

```

01 Initialize-single-source ( $\Theta, s$ )
02  $T = \emptyset$  // Store the node has already find path
03  $Q = V$  // keyed by  $d[v]$   $Q$ : min-heap
04 while  $Q \neq \emptyset$ :
05    $u = \text{extract-min } (Q)$ 
06    $T = T \cup \{u\}$ 
07   for each  $v$  in  $\text{adj}[u]$ 
08     Relax ( $u, v, w$ )
  
```

Initialize-single-source (G, S)

```

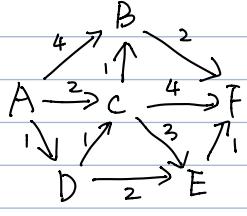
01 for each vertex  $v$  in  $V$ 
02  $d[v] = \infty$ 
03  $\pi[v] = \text{nil}$ 
04  $d[s] = 0$ 
  
```

Relax (u, v, w)

```

01 if  $d[v] > d[u] + w(u, v)$ 
02  $d[v] = d[u] + w(u, v)$ 
03  $\pi[v] = u$ 
  
```

Ex:



$A \rightarrow B \rightarrow C \rightarrow D$
 $B \rightarrow F$
 $C \rightarrow B \rightarrow E \rightarrow F$
 $D \rightarrow C \rightarrow E$
 $E \rightarrow F$
 $F \rightarrow \emptyset$

$Q : A B C D E F$

$T : A D C B E F$

	A	B	C	D	E	F
d	0	∞	∞	∞	∞	∞
π		4	2	1	3	5
		3			5	
				4		
	A	B	C	D	E	F
π	N	N	N	N	N	N
	A	A	A	D	C	
	C	B		B		F

Correctness: NTS: $d[v] = \delta(s, v)$ when Dijkstra terminate

LEMMA: when u is deleted from Q and added to T , $d[u] = \delta(s, u)$

Proof: • Basis: at 1st iteration $u=s$ $d[u]=d[s]=0$ and $0=\delta(s, s)$

• Induction: Suppose u just deleted from Q .

• IH: For all vertices z previously deleted from Q $d[z] = \delta(s, z)$ NTS: $d[u] = \delta(s, u)$

• Suppose for contradiction: $d[u] \neq \delta(s, u)$, we know that for any v .

when relax sets $d[v]$ to a finite value, there is always evidence of a path of that weight. $\therefore d[v] \geq \delta(s, v)$
 so if $d[v] \neq \delta(s, v)$ then $d[v] > \delta(s, v)$. thus there is a shorter path from s to v with weight $< d[v]$. call this path p .

Let y be 1st vertex on P that is not in T . (y might be u), let $x = \pi[y]$ on P . Then $x = \pi(y) \in T$,
 so by IH: $d[x] = \delta(s, x)$

$\therefore d[y] \leq \delta(s, x) + w(x, y) \leq \text{weight of } P < d[u]$

why \leq ? since (x, y) was relaxed weight of optional
 when x added to T . path from s to u

$d[y] < d[u]$ means u is not min, contradiction!

$\therefore d[u] = \delta(s, u)$

Time analysis: Q : Data Struct

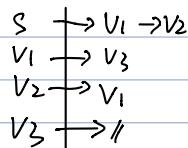
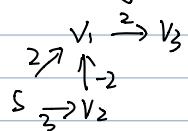
insert (Q, v)
 extract-min (Q)
 decrease-key (Q, v, new, key)

Times	min-heap	array
$ V $	$O(\lg V)$	$O(1)$
$ V $	$O(\lg V)$	$O(V)$
$O(E)$	$O(\lg V)$	$O(1)$

$\} O(V^2)$ better for dense

with heap: $|V|O(|V|) + |V|O(|V|) + O(E) \cdot O(|V|) = O((V+E)|V|)$ better for sparse

Ex. where DIJKSTRA fails.



d	S	V ₁	V ₂	V ₃
0	0	∞	∞	∞

d	S	V ₁	V ₂	V ₃
2	0	2	4	∞

这个时候 V_1 已经加入了 T。但 $d[V_1] = 2 \neq s(V_1)$

Bellman-Ford Algorithm

Maintains: $d[v]$: current best estimate of shortest path weight from s to v

Initially: $d[S] = 0$ $d[v] = \infty \forall v \neq s$

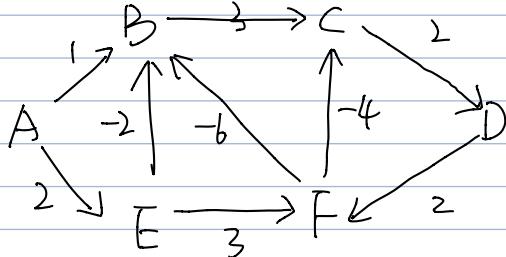
IDEA: Build a tree of shortest paths by relaxing all edges in arbitrary ordering then repeat, until edges relaxed $|V|-1$ times.

Bellman-Ford (G, W, S)

- 01 Initialize-single-source (G, s)
- 02 for $i = 1$ to $|V|-1$
- 03 for each edge (u, v) in E
- 04 Relax (u, v, w)
- 05 for each edge (u, v) in E
- 06 if $d[v] > d[u] + w(u, v)$
- 07 return False
- 08 return True

$$\begin{array}{l} \text{part 1} \quad O(1) \quad O(V) \\ \text{part 2} \quad O(2-4) \quad O(VE) \\ \text{part 3} \quad O(5-8) \quad O(E) \end{array} \} \Rightarrow O(VE) \quad \text{if dense } O(V^3)$$

Ex



A \rightarrow B \rightarrow E

B \rightarrow C

C \rightarrow D

D \rightarrow F

E \rightarrow B \rightarrow F

F \rightarrow B \rightarrow C

A B C D E F

d 0 ∞ ∞ ∞ ∞ ∞

1 ∞ ∞ ∞ ∞ ∞

2 ∞ 1 ∞ ∞ ∞

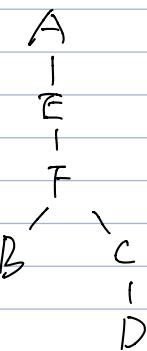
3 ∞ ∞ 1 ∞ ∞

4 ∞ ∞ ∞ 1 ∞

5 ∞ ∞ ∞ ∞ 1

-1 ∞ ∞ ∞ ∞ 1

second loop



A B C D E F

d ∞ ∞ ∞ ∞ ∞ ∞

A B C D E F

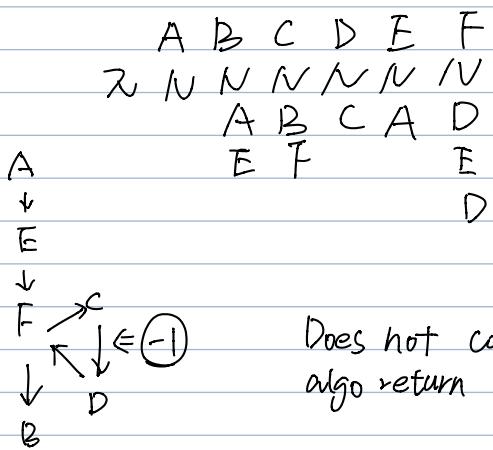
E F

F

发现 DF 树成环了。我们有负圈 C \rightarrow D \rightarrow F 现在会怎样？期末

	A	B	C	D	E	F
d	0	∞	∞	∞	∞	∞
	1	4	6	2	7	
	0	1		5		
	-1					

Second loop	-2	0	3	4		
third loop	-3	-1	2	3		
	:	:	:	:		
	:	:	:	:		
	:	:	:	1		



Does not converge. So the algo return false

Why this algo works?

- Invariant: in absence of neg-weight cycle d-values always either overestimate or exactly correct
- start at ∞ : only max change is by relaxing along an edge

Relax (u, v, w) : $d[v] = \min \{ d[v], d[u] + w(u, v) \}$ D.P.

- it gives correct distance to v when u is the 2nd-to-last vertex in shortest path to v and $d[u]$ correctly set.
- if never make $d[v]$ too small :: extra relax don't hurt

shortest path from s to t $s \rightarrow u_1 \rightarrow u_2 \rightarrow \dots \rightarrow u_k \rightarrow t$

Week 7 Minimum Spanning Tree

MST

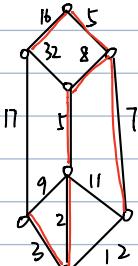
Input: Connected, Undirected, $G = (V, E)$ with weight function $w: E \rightarrow \mathbb{R}$

No constraint on weight

Output: A Spanning tree $T = (V, E')$, $E' \subseteq E$, of minimum weight $w(T) = \sum_{(u,v) \in E'} w(u,v)$

Kruskal's Algorithm

Ex.



- take cheapest edge one by one

but cannot create cycle

2, 3, 5, 5, 7, 8, 16

weight = 46 Greedy Algorithm

Greedy Choice Property:

Locally optimal choices

lead to Global optimal Solution

Kruskal's Algorithm

1. sort edges by weight
2. choose any edge with smallest weight put it in spanning tree constructing
3. successively consider remaining edges in order of increasing weight. Add edge to tree if it does not create a cycle with edges already in tree.

Kruskal (G, m) // $m = |E|$

① 用 DFS

① sort edges by weight: $w(e_1) \leq w(e_2) \leq \dots \leq w(e_m)$

② 最简单的办法

② $X := \emptyset$ // edge in MST

树有 $n-1$ 条边

③ for $i=1$ to m

若边 $> n-1$, 则有环

④ if e_i does not create a cycle with edges in X

⑤ then add e_i to X

Time: $O(|E| \lg |E|)$

$$\Leftarrow |E| \leq |V|^2 \Rightarrow \lg |E| \leq 2 \lg |V|$$

$$\lg |E| = O(\lg |V|)$$

$$\underbrace{O(E \lg |E|)}_{\text{sort}} + \underbrace{O(|V|)}_{\text{union}} + \underbrace{O(|V|E|)}_{\text{using DFS}} = O(|V|E)$$

Prim's Algorithm (Greedy)

Start from S 每次找与当前树相连所有边中的最小边

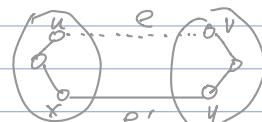
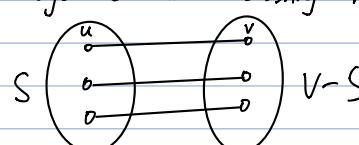
- start at a "root"

Then grow a tree from the root by selecting the least-weight edge with one vertex already in the tree

- cut property

Def: a cut in $G = (V, E)$ is any partition of vertices V into non-empty sets, S and $V-S$

Theorem: (cut property): For any cut $(S, V-S)$ in a connected, weighted, undirected Graph $G = (V, E, w)$ any least-weight edge $e = (u, v)$ crossing the cut, i.e. $u \in S$, $v \in V-S$, is in some MST of G .



Proof: • let T be an MST of G , if T contains e . Done.

• If T not contain $e = (u, v)$, since T is a tree, there is a unique path between u, v .

Since $u \in S, v \in V-S$, this path must contain $e' = (x, y)$ crossing the cut. Add e , we have a cycle $\Rightarrow T \cup e - \{e'\} = T'$. T' is the ST contain e .

$w(u,v) < w(x,y)$ i.e. T' is a MST

- Cycle property

Theorem (cycle property) let $G = (V, E)$ be a connected, undirected, weighted graph. let C be any cycle in G . let $e = (uv)$ be a max-weight edge on C then e does not belong to some MST of G

- Prim's Algorithm (what is relevant cut $(S, V-S)$)?

Let S be the subset of vertices in current tree T . (start with one vertex, r (root))

Add cheapest edge e with exactly one endpoint in S .

Cut property asserts that e is in MST

- Question: How to find cheapest edge with one endpoint in S ?

- Answer: Maintain such edges in a priority queue, delete min to determine next edge to add to T .

binary-min-heap

Prim(G, w, r) // $G = (V, E)$ adj list format, $r \in V$ start vertex

$O(V)$	01 for each $u \in V$ 02 $\text{key}[u] = \infty$ // $\text{key}[u]$ = weight of an edge connected u to a vertex in S 03 $\pi[u] = \text{nil}$
--------	--

$O(1)$	04 $\text{key}[r] = 0$	$O(V) + O(\lg V) (O(V) + O(V) + O(E))$
--------	------------------------	--

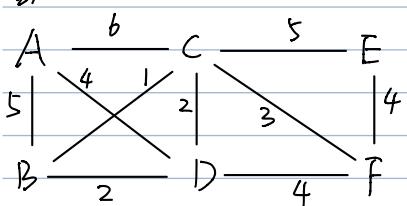
$ V \text{ insert}$ $O(V)$	05 $Q = V$ // ordered by $\text{key}[u]$	$= O(V) + O((V+E)\lg V)$
--------------------------------	--	--------------------------

$ V \text{ extract-min}$ $O(V)$	07 $u = \text{extract-min}(Q)$ 08 $\text{color}[u] = \text{Black}$ // add to S 09 for each vertex v in $\text{adj}[u]$:	$= O((V+E)\lg V) = O(E\lg V)$ $ E \geq V $
-------------------------------------	--	---

$O(E)$	10 if $\text{color}[v] = \text{white}$ and $w(u,v) < \text{key}[v]$ $\pi[v] = u$
--------	---

decrease-key	11 $\text{key}[v] = w(u,v)$
-----------------------	-----------------------------

Ex:



$A \rightarrow B \rightarrow C \rightarrow D$
 $B \rightarrow A \rightarrow C \rightarrow D$
 $C \rightarrow A \rightarrow B \rightarrow D \rightarrow E \rightarrow F$
 $D \rightarrow A \rightarrow B \rightarrow C \rightarrow F$
 $E \rightarrow C \rightarrow F$
 $F \rightarrow C \rightarrow D \rightarrow E$

$(V-S)$: Q: A B R D E F

S: A D B C F E

key	A	B	C	D	E	F
0	∞	∞	∞	∞	∞	∞
X		b	4			
2		2		4		
1						
	X		3			
			4			

π A B C D E F
NNNNNNN

A A A

D D D

B B B

R C C

F F F

Color A B C D E F
W W W W W W

BL

BL

BL

BL

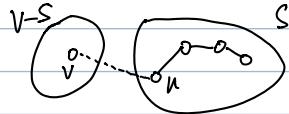
E - F - C - B - D - A

Kruskal Algorithm

What is the relevant cut

- edges sorted in increasing order
- if adding $e = (u,v)$ to T does not create a cycle, then e is the min-weight with exactly one endpoint in S , so cut property asserts e in MST.

Question: How to check if adding an edge to T creates a cycle?



\Rightarrow Answer: Union find data structure

CLRS Union by rank p571

kruskal (G, w)

- 01 $X = \emptyset$ // maintain connected component (edges)
- 02 for each vertex u in V
 - 03 Makeset(u)
 - 04 Sort(E)
 - 05 for each edge (u,v) in E
 - 06 if Find(u) \neq Find(v) // if u, v in different CC's
 - 07 $X = X \cup \{(u,v)\}$ // add to X
 - 08 Union(u, v)
 - 09 return X

// each vertex has rank, height of subtree hanging from x

Makeset(x)

01 $\pi(x) = x$

02 $\text{rank}(x) = 0$

Find(x) // find root

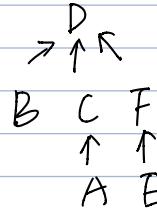
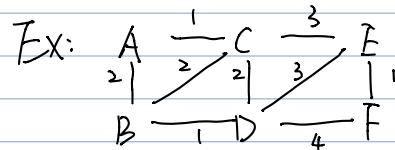
01 while $x \neq \pi(x)$

02 $x = \pi(x)$

03 return x

Union(x, y)

01 Link(Find(x), Find(y))



② sorted edge

- (A, C, 1) (B, D, 1) (E, F, 1)
 (A, B, 2) (B, C, 2) (C, D, 2)
 (C, E, 3) (D, E, 3) (D, F, 4)

Link(x, y)

01 if $\text{rank}(x) > \text{rank}(y)$

02 $\pi(y) = x$

03 else $\pi(x) = y$

04 if $\text{rank}(x) = \text{rank}(y)$

05 $\text{rank}(y) = \text{rank}(y) + 1$

	A	B	C	D	E	F	Rank	A	B	C	D	E	F
0	A	B	C	D	E	F	0	0	0	0	0	0	0
1	C						1						
2	D						2						
3		F					3						
4		D					4						
5				D			5						
6					D		6						
7						D	7						
8							8						
9							9						

$X = \{\textcircled{1}, \textcircled{2}, \textcircled{3}, \textcircled{4}, \textcircled{7}\}$
 $\textcircled{1} (A, C) \text{ Link}(A, C) \pi(A) = C \text{ rank}(C) = 1$
 $\textcircled{2} (B, D) \text{ Link}(B, D) \pi(B) = D \text{ rank}(D) = 1$
 $\textcircled{3} (E, F) \text{ Link}(E, F) \pi(E) = F \text{ rank}(F) = 1$
 $\textcircled{4} (A, B) \text{ Link}(\text{Find}(A), \text{Find}(B)) = \text{Link}(C, D) \pi(C) = D \text{ rank}(D) = 2$
 $\textcircled{5} (B, C) \text{ find}(B) = \text{find}(C) \text{ do nothing}$
 $\textcircled{6} (C, D) \text{ find}(C) = \text{find}(D) \text{ do nothing}$
 $\textcircled{7} (C, E) \text{ Link}(\text{find}(C), \text{find}(E)) = \text{Link}(D, F) \pi(F) = D$
 $\textcircled{8} (D, E) \text{ Find}(D) = \text{Find}(E) = F \text{ do nothing}$
 $\textcircled{9} (E, F) \text{ Find}(E) = \text{Find}(F) = F \text{ do nothing}$

Complexity	Times	Cost
Kruskal	$ V $ makeset's	$O(1)$
	$2 E $ find's	proportional to height $O(gv)$
	$N-1$ union	$O(1)$
	$O(Egv)$	

Partial cost $O(v) + O(Egv) + O(Vlgv) + O(Egv)$
 $= O(Egv)$

- Cycle property

Theorem (cycle property) let $G = (V, E)$ be a connected, undirected, weighted graph. let C be any cycle in G . let $e = (u, v)$ be a max-weight edge on C then e does not belong to any MST of G

Proof for cycle property : Let T be a MST of G . Suppose $e = (u, v) \in T$.

Assume $e \in C$, e' is on path from u to v in C

Remove e from C . consider : $T - \{e\}$. now we have 2 set of vertex : X and $V - X$. Because C is a cycle. $\exists e' \in C$ $e \neq e'$, and e' crosses the cut $(X, V - X)$, and is on the path from u to v . Since e is max-weight edge in C , $w(e') < w(e)$, the new MST T' has less weight than T
 $\therefore w(T') < w(T) \Rightarrow$ Contradiction

Week 8 Network Flow

■ Flow network : Directed graph with non-negative edge weights called **capacities**
edges: carry some measurable quantity
each edge has a given **capacity** : Limit amount to quantity edge carry

■ DEF : Flow network: a flow network : directed graph $G = (V, E)$ with a source vertex $s \in V$. a sink vertex $t \in V$ and edge capacity $c: V \times V \rightarrow \mathbb{R}_{\geq 0}$ Define $c(u,v) = 0$ for any pair of vertices $(u,v) \notin E$

■ DEF : Flow: a flow on G is a function $f: V \times V \rightarrow \mathbb{R}$ satisfying:

capacity: $0 \leq f(u,v) \leq c(u,v) \quad \forall u,v \in V$

conservation: for every vertex $v \in V$, $v \neq s, t$, $\sum_{u \in V} f(u,v) = \sum_{u \in V} f(v,u)$
 $\underbrace{\sum_{u \in V} f(u,v)}$ flow into V $\underbrace{\sum_{u \in V} f(v,u)}$ flow out of V

■ DEF : Value : The value of a flow f $|f| = \sum_{v \in V} f(s,v)$ i.e. the sum of flow out of source

■ DEF : Saturated : an edge is saturated if $f(u,v) = c(u,v)$

■ DEF : Residual capacity: Given a flow f on a network $G = (V, E)$ for each pair of vertices u, v , in V . The residual capacity $C_f(u,v)$ is defined:

$$C_f(u,v) = \begin{cases} C(u,v) - f(u,v) & \text{if } (u,v) \in E \\ f(u,v) & \text{if } (v,u) \in E \\ 0 & \text{o/w: } (u,v), (v,u) \notin E \end{cases}$$

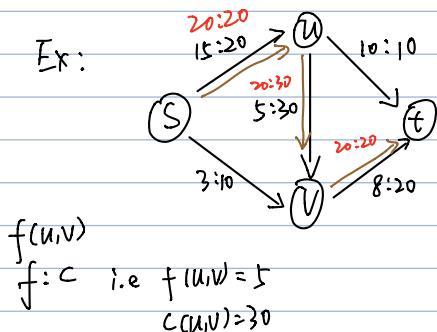
■ DEF : Residual Graph: let f be a flow on a network $G = (V, E)$
The residual graph $G_f = (V, E_f)$

$$E_f = \{(u,v) \in V \times V ; C_f(u,v) > 0\} \quad \text{NOTE: } G_f \text{ has edges whenever } C_f(u,v) > 0$$

■ DEF : Augmenting Path: an augmenting path P is a directed path $s \rightarrow t$ in residual graph G_f

Maximum Flow Problem : Given a network $G = (V, E, C, S, t)$ Find a flow of max value.

Ex:



$f(u,v)$

$$f: c \text{ i.e. } f(u,v) = 5 \\ c(u,v) = 30$$

IDEA:

① try Greedy

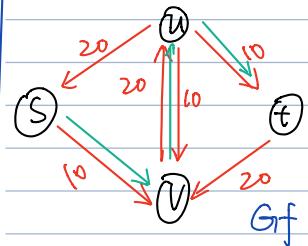
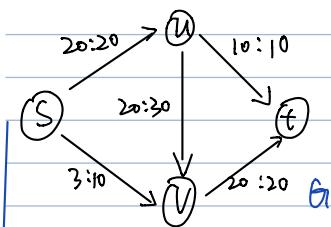
- 1: find an s to t directed path in G : always choose max capacity edge
- 2: find the min(capacity) of edges on the path

$$\text{capacity of path} = \min \{ c(uv) : (u,v) \text{ on path } P \}$$

3. Assign $f(e) = c(p)$ for all edges e on path

4. repeat until no s - t unsaturated path in G

IDEA 2: Ford - Fulkerson ALGORITHM



Forward edges:

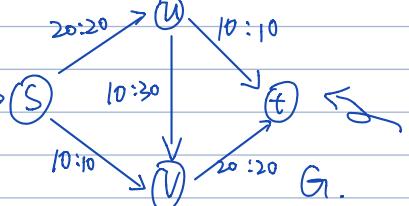
for each $(u,v) \in G$ for which $f(u,v) < c(u,v)$, create an (u,v) in G_f . Assign its capacity $C_f(u,v) = c(u,v) - f(u,v)$

backward edges:

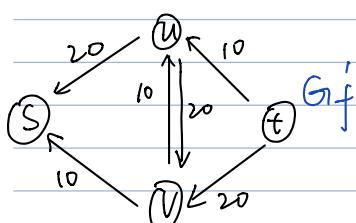
for each $(u,v) \in G$ for which $f(u,v) > 0$ create edge (v,u) in G_f with $C_f(v,u) = f(u,v)$.

NOTE: G_f has edges whenever $C_f(u,v) > 0$

find augmenting path : $C_f(P) = 10$



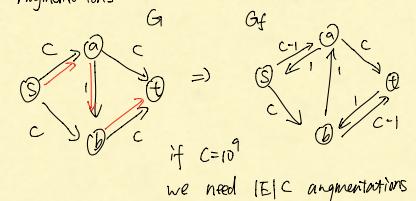
Augment flow f in G using Path P and $C_f(P)$. $f'(e) = f(e) + C_f(P)$
 if e is a forward edge . $f'(e) = f(e) - C_f(P)$ if e is a backward edge



Ford - Fulkerson Algorithm

- 01 $f(u,v) = 0 \quad \forall u,v \in V$
- 02 Build residual graph G_f
- 03 while (there exist an augmenting path)
- 04 find augmenting path P
- 05 compute capacity $C_f(P)$ of P
- 06 augment flow by $C_f(P)$ along P

Analysis: $O(|E|)$ time per augmentation
 - capacities integers $\in [0, c]$ then at most $|E|c$ augmentations



Prove: Does ALG terminate having found max flow?

analyze running time

Claim: if capacities are integers, then ALG terminates

Proof: $|f'| > |f|$. in fact, $|f'| = |f| + C_f(P)$ $C_f(P) > 0$

\therefore flow is strictly increasing

There is only finite capacity coming out of S

By capacity constraint $f(S) \leq c(S)$: S all edges (S, x) , $x \in V$, $C_f(P) \geq 1$

\therefore smallest the flow can increase on each iteration is 1

\therefore with finite bound on $C(S)$, will come to halt and cannot increase flow away more. \square

Irrational = 2wicky

DEF: Cut: partition of vertices V into 2 subsets: $S, V-S$

DEF: S-t Cut: a cut with $S \in S, t \in V-S$

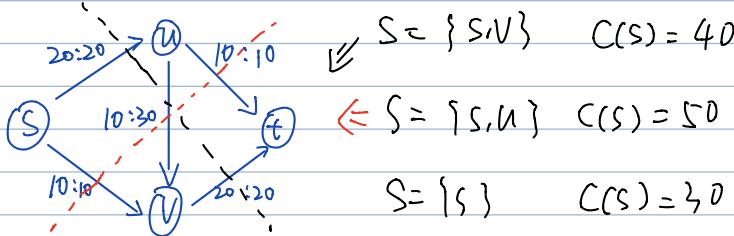
DEF: Capacity of a cut: $C(S,T) = \sum_{\substack{u \in S \\ v \in T}} c(u,v)$
e out of S

DEF: Flow across cut: $f(S,T) = \sum_{u \in S} \sum_{v \in T} f(u,v) - \sum_{v \in T} \sum_{u \in S} f(v,u)$ Claim: $f(S,T) \leq C(S,T)$

Corollary: let (S,T) be any s-t cut and f any s-t flow Then $|f| \leq C(S,T)$
 \rightarrow value of flow \uparrow capacity of cut

Theorem: for any flow network G , $|f| = C(S,T)$ for some f and some cut (S,T)
Maxflow = Mincut

- (1) f is a max flow
- (2) G_f has no augmenting path
- (3) $|f| = C(S,T)$ for some cut (S,T)



Edmonds-Karp

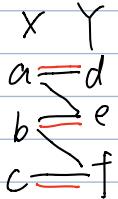
1. use BFS to find augmenting path
2. use dijkstra "fastest path"

$O(|VE|)$ augmentations

$O(|VE|^2)$ running time

Maximum Bipartite Matching

DEF: a bipartite graph (undirected) $V = X \cup Y$, $X \cap Y = \emptyset$, and every edge e in E has one end-point in X and one end point in Y



A matching M in $G = (V, E)$ $M \subseteq E$ such that for all vertices $v \in V$, at most one edge of M is incident on v

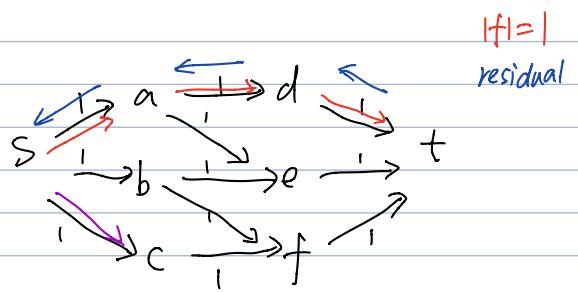
A maximum matching is a matching M such that for any matching M' , $|M| \geq |M'|$

max matching
 $|M|=3$

Given: undirected bipartite graph $G = (V, E)$ $V = X \cup Y$

- construct a flow network $G' = (V', E')$ for G

- Direct all edges in G from X to Y
- Add $s, t \in V$ $V' = V \cup \{s, t\}$
- Create new directed edges from s to each $x \in X$
- Create new directed edges from each $y \in Y$ to t
- Assign unit capacity to each edge in E'



$s \xrightarrow{\quad} T \Rightarrow$ {4 non-graph
4 graph}

Week 9 NP-complete Problems

MST in Graph

Problems with running times $O(n^k)$ for some $k > 0$
 n : input size

efficiently solvable or tractable

Shortest Paths in Graph

Problems whose running time cannot be bounded
By $O(n^k)$ for any $k > 0$ are intractable

Max Flows in Networks

Matchings in Bipartite Graphs (N!)

1. Satisfiability (SAT)

Given a Boolean formula in CNF (i.e. conjunctions of clauses), each clause is the disjunction of several literals, where a literal is a boolean variable or its negation

$$\text{E.X. } (x_1 \vee x_2 \vee x_3) \wedge (\bar{x}_1 \vee \bar{x}_2) \wedge (x_2 \vee \bar{x}_3) \wedge (\bar{x}_1 \vee x_2) \wedge (\bar{x}_1 \vee \bar{x}_2 \vee \bar{x}_3)$$

① ② ③ ④ ⑤

Verify one solution

$O(mn)$: n : # variables
 m : # clauses

A satisfying truth assignment (TVA) is an assignment of true or false to each variable so that every clause contains a literal whose value is true

SAT Problem: Given a Boolean Formula $f(x_1, x_2, \dots, x_n)$ on n variables, is there a TVA to x_1, x_2, \dots, x_n that makes $f(x_1, x_2, \dots, x_n) = \text{TRUE}$?

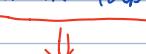
(Take e.g. ② ③ ④ force $\Rightarrow x_1, x_2, x_3 = T \Rightarrow$ ① or ⑤ false) \therefore E.X. false
or $x_1, x_2, x_3 = F$

Finding Solution:

2^n Rows = 2^n Assignment $O(2^n)$

Exponential

How decide in general? \Rightarrow Truth Table:



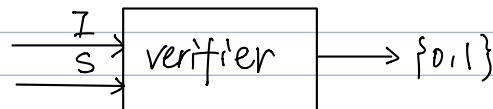
Better Method? \Rightarrow No currently

Special case: 2 SAT $(x_1 \vee \bar{x}_2) \wedge (\bar{x}_1 \vee x_3) \wedge (x_1 \vee x_2)$ $T \bar{T} T \Rightarrow$ polynomial time
(Advanced Algorithm)

2. P and NP Problem

DEF: Class of all problems that can be solved in polynomial is P

DEF: A problem is in class NP if there exists a poly-time verifying ALG that takes an instance I of X and a proposed solution S and outputs true
I.F.F. S is a solution to solution I. i.e. $V(S, I) = 1$



Note: Any problem in P is also in NP $\therefore P \subseteq NP$

IS $P = NP$? (If " $=$ " \Rightarrow every problem in the world can be solved in poly-time)

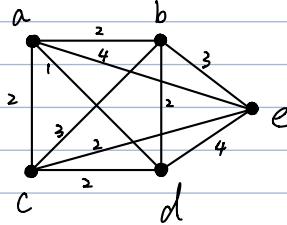
Consensus: No \Rightarrow Not proved

■ Traveling Salesman Problem (TSP)

Given n vertices $1, \dots, n$ and all $\binom{n(n-1)}{2}$ distance between them, and a budget b .
 Is there a tour, i.e. a cycle that passes through each vertex exactly once of total cost $\leq b$?
 i.e. permutation $\tau(1), \dots, \tau(n)$ of vertices such that total distance is at most b :

$$d[\tau(1), \tau(2)] + \dots + d[\tau(n-1), \tau(n)] + d[\tau(n), \tau(1)] \leq b$$

Ex.



$$b = 10. \text{ Is there a tour of cost } \leq 10?$$

Yes: (a.d.b.e.c.a)

How decide in general? exhaustive search: try all $(n-1)!$ tours

DP ALG: $O(n^2 2^n)$: still exponential

TSP \in NP

MST: Given a distance (weight) matrix and a bound b is there a spanning tree with total weight $\sum_{(i,j) \in E} d(i,j) \leq b$? MST \in P
 MST \subseteq NP

What make the huge difference?

MST: spanning tree

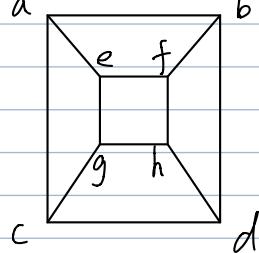
Tsp: spanning tour \Leftarrow tree that not allow to branch

Approximation ALG for TSP: MST + Matching $\frac{3}{2}$ of optimal

■ 3. Hamilton Cycle

A hamilton cycle in undirected graph $G = (V, E)$ is a simple cycle that goes through every vertex exactly once.

Ex. a



$$\Rightarrow a, e, g, c, d, h, f, b, a$$

cycle problem: No poly time ALG know for this problem.

Euler Cycle? goes every edge exactly once.
 (every vertices has even edgeree)
 Linear time

■ Table of problems

Hard
 SAT, 3SAT
 TSP

HAM cycle

Easy (P)
 2SAT

MST

Euler cycle

Reductions

Reduce A to B

$$A \leq_p B$$

"reduce to"

$$A \leq_p B \leftarrow \text{Also hard}$$

"reduce to"

i.e. is P \neq NP?

Reduction Compose: if $A \leq_p B$ and $B \leq_p C$, Then $A \leq_p C$

DEF: A problem A is NP-complete if

1. A is in NP

2. for all problems Ω in NP, $\Omega \leq_p A$

NP-Hard

$$\Omega_1 \leq_p \Omega_2 \leq_p \dots \leq_p \Omega_k \leq_p A$$

↑

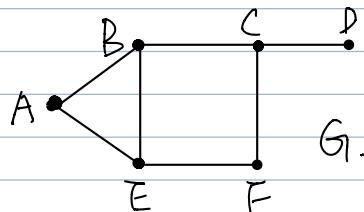
1st problem: SAT (Cook, 1971)

3 SAT \leq_p IND SET \Rightarrow IND SET is NP-complete

Reduction Example

3 set \leq Independent Set

Def: an independent set in an undirected graph $G = (V, E)$ is a subset $S \subseteq V$ such that for all $x, y \in S$, $(x, y) \notin E$. No two vertices in S are joined by an edge.



{A, D, F} is an IS in G_1 .

Independent set problem: Given $G_1 = (V, E)$ positive integer $k \leq |V|$, is there an ISet $S \subseteq V$ of size k ?

3 SAT: 3 CNF Formula to find a satisfying TVA.

must pick one literal from each clause and give it truth value
TRUE choices must be consistent

$$E \cdot X: (x_1 \vee x_2 \vee x_3) \wedge (x_2 \vee \bar{x}_3 \vee \bar{x}_4) \wedge (\bar{x}_1 \vee x_3 \vee x_4) \wedge (x_1 \vee \bar{x}_2 \vee \bar{x}_4)$$

4 choices: TVA: $x_1=T, x_2=T, x_3=T, x_4=T$

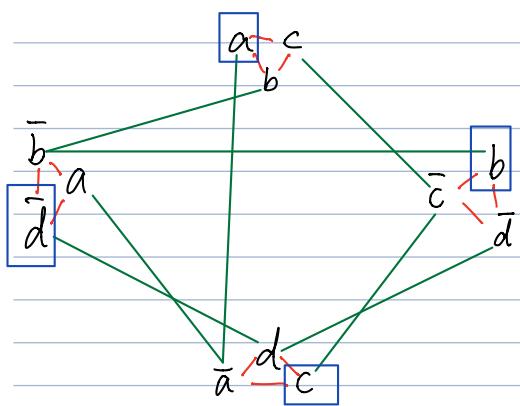
↓

n: variables

m: clauses

Reduction: Graph: each clause becomes a triangle

$$E \cdot P: (a \vee b \vee c) \wedge (b \vee c \vee d) \wedge (\bar{a} \vee c \vee d) \wedge (a \vee \bar{b} \vee \bar{d}) \quad k=4$$



$$S = \{a, b, c, \bar{d}\}$$

Claim: G contains an IS of size exactly k .
 I.F.F. the original formula Φ is satisfiable