- 教材讨论
  - -JH第3章第2节第1、2小节

### 问题1: 伪多项式时间算法

• U和Value(h)-U之间有什么联系和区别? A是它们中哪个问题的算法? 复杂度是多少?

**Definition 3.2.1.1.** Let U be an integer-valued problem, and let A be an algorithm that solves U. We say that A is a pseudo-polynomial-time algorithm for U if there exists a polynomial p of two variables such that

$$Time_A(x) = O(p(|x|, Max-Int(x)))$$

for every instance x of U.

**Definition 3.2.1.2.** Let U be an integer-valued problem, and let h be a non-decreasing function from  $\mathbb{N}$  to  $\mathbb{N}$ . The h-value-bounded subproblem of U, Value(h)-U, is the problem obtained from U by restricting the set of all input instances of U to the set of input instances x with Max- $Int(x) \leq h(|x|)$ .

• 你能给出素数判定问题的一个伪多项式时间算法吗?

# 问题1: 伪多项式时间算法(续)

· 什么是KP? 在KP的DP算法中,以下概念各是什么含义?

```
 \begin{array}{ll} - \mid & I = (w_1, \ldots, w_n, c_1, \ldots, c_n, b) \\ - \mid_{\mathsf{i}} & I_i = (w_1, w_2, \ldots, w_i, c_1, c_2, \ldots, c_i, b) \\ - \mathsf{triple} & (k, W_{i,k}, T_{i,k}) \in \left\{0, 1, 2, \ldots, \sum\limits_{j=1}^i c_j\right\} \times \{0, 1, 2, \ldots, b\} \times Pot(\{1, \ldots, i\}) \\ - \mathsf{TRIPLE}_i & TRIPLE_i \end{array}
```

# 问题1: 伪多项式时间算法(续)

- 你能基于上述概念解释这个算法的基本过程吗?
- 你能证明这个算法的正确性吗?你会计算它的时间复杂度吗? (如何体现伪多项式?)

### Algorithm 3.2.2.2 ((DPKP)).

```
I = (w_1, w_2, \dots, w_n, c_1, c_2, \dots, c_n, b) \in (\mathbb{N} - \{0\})^{2n+1}, n a positive
Input:
          integer.
Step 1: TRIPLE(1) := \{(0,0,\emptyset)\} \cup \{(c_1,w_1,\{1\}) \mid \text{if } w_1 \leq b\}.
Step 2: for i = 1 to n - 1 do
             begin SET(i+1) := TRIPLE(i);
               for every (k, w, T) \in TRIPLE(i) do
                  if w + w_{i+1} \le b then
                 SET(i+1) := SET(i+1) \cup \{(k+c_{i+1}, w+w_{i+1}, T \cup \{i+1\})\};
              Set TRIPLE(i+1) as a subset of SET(i+1) containing exactly
              one triple (m, w', T') for every achievable profit m in SET(i+1)
              by choosing a triple with the minimal weight for the given m
            end
Step 3: Compute c := \max\{k \in \{1, \dots, \sum_{i=1}^n c_i\} \mid (k, w, T) \in TRIPLE(n)
          for some w and T.
Output: The index set T such that (c, w, T) \in TRIPLE(n).
```

# 问题1: 伪多项式时间算法(续)

- 什么是最大流问题?
- 你能解释Ford-Fulkerson算法的基本过程吗?
- 你会计算它的时间复杂度吗? (如何体现伪多项式?)

#### Algorithm 3.2.3.10 (The Ford-Fulkerson Algorithm).

```
Input: (V, E), c, s, t of a network H = ((V, E), c, \mathbb{Q}^+, s, t).
                                                            Step 1: Determine an initial flow function f of H (for instance, f(e) = 0 for
                                                                       all e \in E); HALT := 0
                                                            Step 2: S := \{s\}; \overline{S} := V - S;
                                                            Step 3: while t \notin S and HALT=0 do
                                                                         begin find an edge e = (u, v) \in E(S, \overline{S}) \cup E(\overline{S}, S) such that
                                                                                  res(e) > 0
                                                                                  -c(e) - f(e) > 0 if e \in E(S, \overline{S}) and f(e) > 0 if
                                                                                  e \in E(\overline{S}, S)";
                                                                                  if such an edge does not exist then HALT:= 1
                                     1/1
                                                                                  else if e \in E(S, \overline{S}) then S := S \cup \{v\}
                                                                                                          else S := S \cup \{u\}:
           3/3
                        4/15
10/10
                                                                                  \overline{S} := V - S
                                        7/10
                                                                         end
                    7/7
                                                            Step 4: if HALT=1 then return (f,S)
                                                                       else begin find an augmenting path P from s to t, which
                                                                                     consists of vertices of S only; -this is possible
                                                                                     because both s and t are in S'':
                                                                                     compute res(P);
                                                                                     determine f' from f as described in Lemma 3.2.3.9
                                                                       end:
```

goto Step 2

# 问题2: strongly NP-hard

• 一个strongly NP-hard问题可能存在伪多项式时间算法吗?为什么?

**Definition 3.2.4.1.** An integer-valued problem U is called **strongly NP-hard** if there exists a polynomial p such that the problem Value(p)-U is NP-hard.

- 如何证明一个问题是strongly NP-hard? 你能以TSP为例来说明吗?
- 为什么这句话成立?
  - Every weighted version of an optimization graph problem (e.g., WEIGHT-VCP) is strongly NP-hard if the original "unweighted" version (e.g., MIN-VCP) is NPhard.

### 问题3:参数化

从不同的角度,谈谈你对参数化的理解、 参数化的意义以及与伪多项式算法的关系

**Definition 3.3.1.1.** Let U be a computing problem, and let L be the language of all instances of U. A parameterization of U is any function Par:  $L \to \mathbb{N}$  such that

- (i) Par is polynomial-time computable, and
- (ii) for infinitely many  $k \in \mathbb{N}$ , the k-fixed-parameter set

$$Set_U(k) = \{x \in L \mid Par(x) = k\}$$

is an infinite set.

We say that A is a Par-parameterized polynomial-time algorithm for U if

- (i) A solves U, and
- (ii) there exists a polynomial p and a function f : N → N such that, for every x ∈ L,

$$Time_A(x) \leq f(Par(x)) \cdot p(|x|).$$

# 问题3:参数化(续)

- 对于一个问题,可能有多种参数化方法,如何评价其好坏?
  - Capture the inherent difficulty of particular input instances.
  - One can design a practical parameterized polynomial-time algorithm.
  - Most of the problem instances occurring in the considered application have this parameter reasonably small.

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- (i) A solves U, and
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$$Time_A(x) \le f(Par(x)) \cdot p(|x|).$$

### 问题3:参数化(续)

- · 什么是VC问题?
- 你理解它的两个参数化算法了吗?为什么说第二种更好?

**Algorithm 3.3.2.4.** Input: (G, k), where G = (V, E) is a graph and k is a positive integer.

Step 1: Let H contain all vertices of G with degree greater than k. if |H|>k, then  $\operatorname{output}(\text{"reject"})$  {Observation 3.3.2.2}; if  $|H|\leq k$ , then m:=k-|H| and G' is the subgraph of G obtained

by removing all vertices of H with their incident edges.

Step 2: if G' has more than m(k+1) vertices [|V-H| > m(k+1)] then output ("reject") {Observation 3.3.2.3}.

Step 3: Apply an exhaustive search (by backtracking) for a vertex cover of size at most m in  $G^\prime$ .

if G' has a vertex cover of size at most m, then output("accept"), else output("reject"). We consider the following divide-and-conquer strategy. Let (G, k) be an input instance of the vertex cover problem. Take an arbitrary edge  $\{v_1, v_2\}$  of G. Let  $G_i$  be the subgraph of G obtained by removing  $v_i$  with all incident edges from G for i = 1, 2. Observe that

$$(G, k) \in VC \text{ iff } [(G_1, k-1) \in VC \text{ or } (G_2, k-1) \in VC].$$

Obviously,  $(G_i, k-1)$  can be constructed from G in time O(|V|). Since, for every graph H, (H, 1) is a trivial problem that can be decided in O(|V|) time and the recursive reduction of (G, k) to subinstances of (G, k) can result in solving at most  $2^k$  subinstances of (G, k), the complexity of this divide-and-conquer algorithm is in  $O(2^k \cdot n)$ .