• Let A be an algorithm to a problem  $\Pi$ . For any instance I of  $\Pi$ , let be(I) be the binary encoding of I. Let |I| be the length of be(I), that is |I| = |be(I)| (ex. the length of a 0-1 knapsack instance is  $O(n((\lg \max_i v_i) + (\lg \max_i w_i)) + \lg W))$ . Also, let  $\beta$  be the length of unary representation of a maximum number in I (ex. W in a 0-1 knapsack instance), or 0 if no numbers occur in I. Significantly, the length of unary encoding of  $\beta$  is not necessarily bounded by  $poly(|be(\beta)|)$ .

In the time complexity of  $A^1$ ,

- if the number of objects in I is in the exponent, then A is called an *exponential time* algorithm (ex. naive  $O(2^n)$  time algo for vertex cover),
- if |I| is not in the exponent and  $\beta$  occurs in the base, noting that  $\beta = 2^{\lg \beta}$ , A is called a *pseudo-polynomial time* algorithm (ex. naive O(n) time algorithm for determining whether the given positive integer n is a prime; or the dynamic programming based algorithm for 0-1 knapsack, which takes  $O(Wn) = O(2^{\lg W}n)$  time),
- if both |I| and  $\lg \beta$  are not in the exponent, but  $\lg \beta$  does occur in the base, then A is a *weakly-polynomial time* algorithm (ex. computing gcd(a, b) for  $a > b \ge 1$  in  $O(\lg b)$  time)
- if  $\beta$  does not appear at all and the number *n* of objects in *I* appears in the base, then *A* is called a *strongly-polynomial time* algorithm, and (ex.  $O(n \lg n)$  time algorithm to merge sort *n* numbers)
- if the time complexity of A is  $2^{(\lg n)^c}$  for c > 1, then A is called a *quasi-polynomial time* algorithm (such algorithms are considered better to exponential time algorithms but not as efficient as polynomial time algorithms).
- It is desirable to have a strongly-polynomial time algorithm. Both the strongly- and weakly-polynomial time algorithms are polynomial time algorithms. However, if the algorithm's time complexity is a weakly-polynomial, then that algorithm is said to be a weakly-polynomial time algorithm; otherwise, we simply say that it is a polynomial time algorithm. Further, both the pseudo-polynomial time and exponential time algorithms are exponential in time complexity since the input complexity is in the exponent of time complexity; significantly, a pseudo-polynomial time algorithm takes exponential time when the input instance has large numbers.
- Assuming P ≠ NP, no NP-hard problem is solvable in polynomial time. However, the NP-hardness of a problem Π does not necessarily rule out the possibility of solving it with a pseudo-polynomial time algorithm. If Π is NP-hard and Π does not have numerical parameters, then Π cannot be solved by a pseudo-polynomial time algorithm unless P = NP. Thus, assuming P ≠ NP, only NP-complete problems that are potential candidates for being solved by pseudo-polynomial time algorithms are those problems that have numerical parameters. Some of the examples for numerical problems include partition, 0-1 knapsack, integer partition, bin packing, and TSP. While other problems being numerical problems is obvious, the TSP problem is a numerical since the edge weights are numerical parameters. On the other hand, the CLIQUE problem may not be considered as a numerical problem since the clique size k in any reasonable instance is upper bounded by n.

From the above definitions, an algorithm that solves a problem  $\Pi$  is called a pseudo-polynomial time algorithm for  $\Pi$  if its time complexity is  $O(poly(|I|, \beta))$ . For any decision problem  $\Pi$ , let  $\Pi_p$  denote

<sup>&</sup>lt;sup>1</sup>Though we use  $\lg n$  bits to label/index any object among *n* objects *I* has, in the word-RAM model of computation, usually, any of these labels or references to them fit in a word and hence occupy O(1) space.

the subproblem of  $\Pi$  obtained by restricting  $\Pi$  to only those instances I that satisfy  $\beta \leq poly(|I|)$ . (To remind, |I| is |be(I)|.) The problem  $\Pi$  is called *strongly NP-hard* if  $\Pi_p$  is NP-hard.

**Lemma.** Unless P = NP, there can be no pseudo-polynomial time algorithm for any strongly NP-hard problem.

*Proof.* For the sake of contradiction, assume a pseudo-polynomial time algorithm  $\mathcal{A}$  exists for a strongly NP-hard problem  $\Pi$ . That is,  $\Pi_p$  is NP-hard. Given any input string w, first check whether w encodes an instance I of  $\Pi$  satisfying  $\beta \leq poly(|I|)$ . (That is, we check whether w is an instance of  $\Pi_p$ .) If so, apply  $\mathcal{A}$  to I. For any  $I \in \Pi_p$ , the  $\mathcal{A}$  takes  $poly(|I|, \beta = O(poly(|I|))) = poly(|I|)$  time, contradicting the supposition  $P \neq NP$ .

In other words, a decision problem  $\Pi$  is strongly NP-hard if every problem in NP can be reduced to  $\Pi$  in polynomial time such that the length of every number in unary representation in the reduced instance f(w) is at most a polynomial in the length of the binary encoding of w. That is, to prove  $\Pi$  is strongly NP-hard, one needs to give a polynomial time reduction from any strongly NP-hard problem to  $\Pi_p$ . One such strongly NP-hard numerical problem is binpacking, thanks to polynomial time reduction from 3-dimensional matching. The TSP is a strongly NP-hard problem since it remains NP-hard even if each number involved in it is upper bounded by the number of nodes in the graph. (And, even if one considers the CLIQUE problem as a numerical problem due to number k in any instance, this problem is strongly NP-hard since it remains NP-hard even if k in the instance is upper bounded by the number of nodes in the graph.) A decision problem  $\Pi$  is strongly NP-complete if  $\Pi$  belongs to NP and  $\Pi$  is strongly NP-hard.

• An NP-hard problem that is not strongly NP-hard is said to be *weakly NP-hard*. Any weakly NP-hard problem can have a pseudo-polynomial time algorithm without disobeying  $P \neq NP$ . The example weakly NP-hard numerical problems include 0-1 knapsack, integer knapsack, subsetsum, and partition. A decision problem  $\Pi$  is *weakly NP-complete* if  $\Pi$  belongs to NP and  $\Pi$  is weakly NP-hard.

References:

"Strong" NP-Completeness Results: Motivation, Examples, and Implications by M. R. Garey and D. S. Johnson, JACM '78, Vol 25, No 3, pp 499-508.