HSE UNIVERSITY, MASTER'S PROGRAM 'DATA SCIENCE'

## Counting problems (#P)

We are going to discuss *counting* versions of NP problems. It will be more convenient for us to use the definition of NP in terms of deterministic polynomial computations with hints. Namely, for an NP-problem A,

$$x \in A \iff (\exists y) (|y| < p(|x|) \text{ and } R(x, y)),$$

where R is a polynomial-time computable predicate.

The corresponding counting problem, #A, is the question how many hints, or witnesses, y, for a given x, satisfy R(x, y). Each NP-problem has its counting counterpart. For example, SAT is the question "does the given Boolean formula have a satisfying assignment?", while #SAT is "how many satisfying assignments does this Boolean formula have?"; HAMPATH is "is there a Hamiltonian path in the given (directed) graph?" and #HAMPATH is "how many Hamiltonian paths does this graph have?" etc. These counting counterparts of NP decision problems form the #P class.

Clearly, the counting problem #A is at least as hard as the decision problem A: if we can count the number of witnesses, we can then compare it with zero and answer whether there exists at least one witness. In particular, if  $P \neq NP$  and A is NP-complete, then #A cannot be solved in polynomial time.

There are, however, situations when  $A \in \mathsf{P}$ , but #A is hard (i.e., not polynomial-time solvable, unless  $\mathsf{P} = \mathsf{NP}$ ). In order to establish results of this sort, one develops the theory of  $\#\mathsf{P}$ -completeness, which runs in parallel with the theory of NP-completeness.

**Definition.** A polynomial-time *counting reduction* of #A to #B is a pair of functions f and g, where f maps inputs of #A to inputs of #B, and  $g: \mathbb{N} \to \mathbb{N}$ , such that

$$#A(x) = g(#B(f(x))).$$

Notation:  $#A \leq_c^P #B$ .

Counting reduction is indeed a reduction in the following sense: if #B is polynomial-time solvable and  $\#A \leq_c^P \#B$ , then #A is also polynomial-time solvable.

A specific kind of counting reduction is *parsimonious reduction*, in which g is the identity function (i.e., #A(x) = #B(f(x)), just as in *m*-reductions of decision problem).

**Definition.** A counting problem #B is #P-complete, if for any  $\#A \in \#P$  we have  $\#A \leq_c^P \#B$ .

The reduction in the proof of Cook–Levin theorem and Tseitin transformations can be performed parsimoniously; thus, we establish the fact that #SAT and #3-SAT are #P-complete.

Using only parsimonious reductions, however, does not give interesting results. The reason is that each parsimonious reduction for counting problems induces an *m*-reduction on decision problems. Thus, establishing #P-completeness by parsimonious reductions implies NPcompleteness for corresponding decision problems (which itself already yields hardness of the counting problem without any reference to the theory of #P-completeness).

More general counting reductions, however, could yield new results. Our first example is #DNF-SAT, the counting problem for satisfying assignments of Boolean formulae in DNF. The corresponding decision problem, DNF-SAT, is polynomial-time decidable. For the counting

problem the situation is different, namely, #3-SAT  $\leq_c^P$  #DNF-SAT, whence #DNF-SAT is #P-complete.

Indeed, if  $\varphi$  is a 3-CNF, then its negation,  $\neg \varphi$ , can be easily (polynomial-time) transformed into an equivalent DNF. Next, an assignment is satisfying for  $\varphi$  iff it is not satisfying for  $\neg \varphi$ , therefore

$$\#$$
SAT $(\varphi) = 2^n - \#$ SAT $(\neg \varphi)$ 

where n is the number of variables  $(2^n \text{ is the total number of assignments})$ . Thus, #3-SAT is reduced to #DNF-SAT by the following counting reduction:

$$f: \varphi \mapsto \neg \varphi;$$
  
$$g: k \mapsto 2^n - k.$$

Now, if  $P \neq NP$ , then #DNF-SAT is not polynomial-time solvable. Indeed, polynomial-time solvability of #DNF-SAT, by the aforementioned reduction, would yield that of #3-SAT, and thus of 3-SAT, which is impossible.

The more famous example of #P-complete problem is the *permanent* problem. We consider  $n \times n$  Boolean matrices (i.e., constructed 0's and 1's, with operations modulo 2). Recall that the *determinant* of a matrix is

$$\det(a_{ij}) = \sum_{\sigma \in \mathbf{S}_n} (-1)^{\operatorname{sign}(\sigma)} \prod_{i=1}^n a_{i\sigma(i)}.$$

(Here  $\mathbf{S}_n$  is the set (group) of all permutations of  $\{1, \ldots, n\}$ .) The determinant can be polynomially computed using, say, Gaussian elimination procedure.

The permanent is defined similarly, but without alternation of signs:

$$\operatorname{perm}(a_{ij}) = \sum_{\sigma \in \mathbf{S}_n} \prod_{i=1}^n a_{i\sigma(i)}.$$

Computing the permanent can be seen as a counting problem: for a given matrix  $(a_{ij})$ , count the number of permutations  $\sigma \in \mathbf{S}_n$  such that  $a_{i\sigma(i)} = 1$  for all *i*. Thus, PERM  $\in \#P$ . Valiant (1979) showed #P-hardness of PERM, but using a more general notion of Turing reduction; Ben-Dor and Halevi (1993) did it by counting reduction. For proof details, see their paper: A. Ben-Dor, S. Halevi (1993), "Zero-one permanent is #P-complete, a simple proof."

Interestingly enough, #P-completeness of PERM yield that of #2-SAT (Valiant 1979), even in its fragment with only monotone 2-CNF's (without negations). Recall that the 2-SAT decision problem is polynomial. For proof, see L. G. Valiant (1979), "The complexity of enumeration and reliability problems."