Social and Technological Networks	Edinburgh, 2019
Notes 1.	
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**Optimizations.** In class we discussed optimization problems and approximations. For a given function  $f(\cdot)$ , a *maximization* problem is to find the input x to f that achieves the maximum possible f(x). Similarly, a *minimization* problems about finding the input to minimize f.

For the influence maximization problem, the input x too the form of subsets of V. This creates a computational challenge, since there are many possible subsets of V. We were looking for subsets of size k, but even then, there are  $\binom{n}{k}$  possibilities, which is can be very large for a large k.

**Q 1.** Suppose k = n/2. Show that the number of possible subsets of size k is at least  $2^{\Omega(n)}$ .

**Approximations.** For a maximization problem, if *f* achieves its maximum value for input  $x^*$  and  $f(x^*) = OPT$ , then a *c*-approximation algorithm finds an *x* such that  $f(x) \ge c \cdot OPT$ . The value for *c* in this case will be a positive fraction less than 1.

For a minimization problem, if f achieves its minimum value for input  $x^*$  and  $f(x^*) = OPT$ , then a c-approximation algorithm finds an x such that  $f(x) \le c \cdot OPT$ . The value of c will be greater than 1.

In both cases, *c* is called the approximation factor.

**Q 2.** What is the range of possible values for the approximation factor of an approximation algorithm for the influence maximization problem?

**Q 3.** Suppose we want to find shortest paths. Is this is a maximization or minimization problem? What is the range of possible values for the approximation factor of an approximation algorithm for this problem?

Two very useful inequalities:

$$\left(1+\frac{1}{x}\right)^x \le e$$
$$\left(1-\frac{1}{x}\right)^x \le \frac{1}{e}$$

We will make use of these many times in the course.

**Q 4.** Suppose a tortoise is at distance n meters from its destination. It is getting tired with time, and in each hour, it covers 1/2 of the remaining distance to the destination. How long does it take the tortoise to get to the destination?

**Q** 5. How long does it take it to get to within 1 meter of the destination?

**Q 6.** Now suppose instead the tortoise covers 1/k of the remaining distance to the destination in each hour. How long does it take it to get to within 1 meter?

**Q 7.** If the tortoise covers 1/k of the remaining distance to the destination in each hour, what fraction of the distance remains to be covered after 1 hr? What fraction remains to be covered after *k* hours?

**Q 8.** In the independent activation model, suppose node u has neighbors  $v_1, v_2, ...$  and the corresponding edges have associated probabilities  $p_1, p_2, ...$  Suppose the probability of u being activated is p(u). If all neighbors  $v_1, v_2, ...$  are active, can we say that  $p(u) \le p_1 + p_2 + ...$ ? What is the exact probability of u becoming active?

**Q 9.** Suppose for a node x in a network, an edge to each other n - 1 nodes exists with probability  $\frac{\ln n}{n-1}$ . Show that the probability that x has no edge is  $\leq \frac{1}{n}$  [hint: use  $(1 - \frac{1}{x})^x$  with  $x = \frac{\ln n}{n-1}$ .]

**Problem instances** An *instance* of a problem is the problem asked for a particular dataset. For example, finding shortest path is an algorithmic problem, while finding the shortest path between a specific pari of nodes on a particular network is an instance of the shortest path problem.

**NP hardness.** (optional. not in exam). There are certain problems that are considered NP-hard, and belong to the class of problems called NP-hard. Let us refer to this set as *NPH*.

It can be shown that for any pair of problems  $A, B \in NPH$ , any given instance of A can be "reduced" to an instance of B in polynomial time. Meaning that given an instance of A, we can convert it to an instance of B in polynomial time, and then any solution of the instance of B can be converted back to a solution of the instance of A in polynomial time.

This also implies that if there is a polynomial time solution to *B*, then that can be used to get a polynomial time solution of *A* via the reduction. Thus, if any problem in *NPH* has a polynomial time solution, then every problem in NPH will have a polynomial time solution via the reduction. It is however generally believed that NP-hard problems cannot have polynomial time solutions. Though a proof of this fact is not known.

When we encounter a new problem C, the usual way to show that it is NP-hard is to take a known problem  $B \in NPH$  and show that a polynomial time reduction exists from B to C. This implies that C is NP hard, and a polynomial time algorithm is unlikely. If a poly time algorithm is found for C, that would imply that all problems in NPH has poly time algorithms. Thus, once we have shown a problem to be NP-hard, finding ploy time algorithms for it is really unlikely.