

# Prerequisites Test

Cryptography and Security 2024

# Exercise 1 Probability

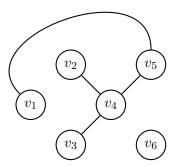


Figure 1: Example graph with 6 vertices  $(V = \{v_1, v_2, v_3, v_4, v_5, v_6\})$  and 4 edges  $(E = \{\{v_1, v_5\}, \{v_2, v_4\}, \{v_3, v_4\}, \{v_4, v_5\}\})$  where  $v_6$  is an isolated vertex.

A simple, undirected graph G = (V, E) with vertex set V and edge set E is a graph without any self loops or multiedges, and whose edges have no direction. In other words, for any two vertices a and b, there is at most one edge  $\{a,b\}$  (edge (a,b)=(b,a)) between them, and there is no edge if a=b. We say a vertex is isolated if there is no edge connected to it. See Figure 1 as an example.

Suppose for every two different vertices a and b, the edge  $\{a,b\}$  appears with probability p. Moreover, the appearances of any two different edges are independent.

#### Question 1.1 Expectation

Given |V| = n, use the linearity of expectation to compute the expected number of isolated vertices.

**Hint:** Define a boolean random variable  $X_i$  that indicates whether vertex i is isolated.

Answer. Let  $X_i$  be a boolean random variable that indicates whether vertex i is isolated. Then  $\Pr[X_i = 1] = (1 - p)^{n-1}$  since there are at most n - 1 edges that can connect to i, and each of them does not appear with probability 1 - p. Since  $X_i$  takes value in  $\{0, 1\}$ ,  $E[X_i] = \Pr[X_i] = (1 - p)^{n-1}$ .

Let  $X = \sum_i X_i$  be the random variable for the number of isolated vertices. By linearity of expectation,  $\mathrm{E}[X] = \sum_i \mathrm{E}[X_i] = n \cdot (1-p)^{n-1}$ 

#### Question 1.2 Variance

Compute the variance of number of isolated vertices.

Answer. We know that  $Var[X] = E[X^2] - E[X]^2$  and  $E[X] = n \cdot (1-p)^{n-1}$  from the previous question. It suffices to compute  $E[X^2]$ .

Again let  $X = \sum_{i} X_{i}$ . Then we compute

$$E[(\sum_{i} X_{i})^{2}] = \sum_{i,j} E[X_{i}X_{j}]$$
$$= \sum_{i \neq j} E[X_{i}X_{j}] + \sum_{i} E[X_{i}^{2}]$$

For the second term, notice that  $X_i^2 = X_i$  for Bernoulli random variable. Then for n addends,  $\sum_i \mathrm{E}[X_i^2] = n \mathrm{E}[X_i] = n \cdot (1-p)^{n-1}$ .

For the first term, we need some extra care. Since  $X_iX_j$  takes value in  $\{0,1\}$ , it suffices to calculate  $Pr[X_iX_j]$ . Also notice that  $X_iX_j=1$  iff both vertice i and j are isolated. We first ignore the edge  $\{i, j\}$ , the probability that vertex i does not have the remaining n-2 edges is  $(1-p)^{n-2}$ , same for vertex j. The probability that both vertex i and j are isolated except for edge  $\{i,j\}$  is then  $(1-p)^{2n-4}$ . Finally, the probability that  $\{i,j\}$  does not appear is (1-p). Putting everything together, the probability that  $X_i X_j = 1$  is  $(1-p)^{2n-3}$ . To conclude, the variance is  $n(1-p)^{n-1} + n(n-1)(1-p)^{2n-3} - n^2(1-p)^{2n-2}$ .

# Exercise 2 Algorithms

## Question 2.1 Randomized Algorithms

In this exercise, we are going to analyze the expected number of iterations for a simple randomized search algorithm. Let  $a_1, \ldots, a_N$  be N integers. Our goal is to search the index of a given target number T. Below is an algorithm  $\mathsf{RandSearch}(a_1, \ldots, a_N, N, T)$  that takes a list of integers as  $a_1, \ldots, a_N$ , the number of integers as N and the target integer as T, outputs an index in  $\{1, \ldots, N\}$  if T is found, 0 if not.

## RandSearch $(a_1, \ldots, a_N, N, T)$

1: pick a uniformly random permuation of  $\sigma$  of  $a_1, \ldots, a_N$ 

2: **for** i = 1 to N:

3: **if**  $a_{\sigma(i)} = T$ :

4: return  $\sigma(i)$ 

5: return 0

1. Compute the expected number of iterations of the for loop in line 2 of the RandSearch algorithm.

Answer. Assuming  $a_i$  are distinct and there exists  $i, 1 \leq i \leq N$  s.t.  $T = a_i$ . The grading was a bit relaxed due to this assumption being announced mid exam and each question was graded based on its own assumption.

**Note:** Albeit being simple, this analysis appears quite often in the course while analyzing the complexity of brute force attacks.

Let X be the random variable for the number of iterations of the RandSearch algorithm. We have

$$E[X] = \sum_{i=1}^{N} E[Pr[C = a_{\sigma(i)}]] \cdot i$$

Since  $\sigma$  is a uniformly random permutation, we have

$$E[Pr[C = a_{\sigma(i)}]] = \frac{1}{N}$$

for all i Hence, we have

$$E[X] = \sum_{i=1}^{N} \frac{1}{N} \cdot i = \frac{N+1}{2}$$

**Common mistake:** X is not a geometric distribution, it would be the case if we were sampling an element from  $\{a_1, \ldots, a_N\}$  at each iteration. However, in our case we are iterating over these elements meaning that we only visit them once in a random order.

#### Question 2.2 Asymptotic Notation

Assume that f(n) and g(n) are two positive functions.

**Definition 1** (O(.)). We write f(n) = O(g(n)) if and only if there exist constants N > 0 and C > 0 such that for all  $n \ge N$ ,  $f(n) \le C \cdot g(n)$ .

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**Definition 2**  $(\Omega(.))$ . We write  $f(n) = \Omega(g(n))$  if and only if there exist constants N > 0 and C > 0 such that for all  $n \geq N$ ,  $f(n) \geq C \cdot g(n)$ .

**Definition 3**  $(\Theta(.))$ . We write  $f(n) = \Theta(g(n))$  if and only if f(n) = O(g(n)) and  $f(n) = \Omega(g(n))$ .

**Definition 4** (o(.)). We write f(n) = o(g(n)) if and only if for all constants C > 0 there exists N > 0 such that for all  $n \ge N$ ,  $f(n) < C \cdot g(n)$ .

1. Show that if  $f = \Omega(g)$  then f is not in o(g).

Answer. Assume the contrary that  $f = \Omega(g)$  and f = o(g), we have the following:

(a) by definition of o, for all constants c > 0 there exists  $N_1 > 0$  such that for all  $n \ge N_1$  we have

$$f(n) < c \cdot g(n)$$

(b) by definition of  $\Omega$ , there exists constants  $c_2>0$  and  $N_2>0$  such that for all  $n\geq N_2$  we have

$$f(n) > c_2 \cdot q(n)$$

Note that since the first condition holds for all positive constants, it holds for  $c_2$  as well. Moreover, both conditions when n is greater than  $N_1$  and  $N_2$ . Hence, we obtain a contradiction as required.

2. Show that if  $f(n) + g(n) = \Theta(\max\{f(n), g(n)\})$ .

Answer. By definition, we need to show that  $f(n) + g(n) = O(\max\{f(n), g(n)\})$  and  $f(n) + g(n) = \Omega(\max\{f(n), g(n)\})$ . Note that both f and g are positive functions. Hence, for a fixed for any n > 0 we have f(n) > 0 and g(n) > 0.

- Due to the positiveness of f and g we have  $f(n) + g(n) > 1 \cdot max\{f(n), g(n)\}$  so  $f(n) + g(n) = \Omega(max\{f(n), g(n)\})$  with C = 1.
- Again, due to positiveness, we have  $f(n) + g(n) \le 2 \cdot max\{f(n), g(n)\}$  so  $f(n) + g(n) = O(max\{f(n), g(n)\})$  with C = 2.

# Exercise 3 Number Theory & Algebra

The goal of this exercise is to demonstrate the connection between two mathematical concepts: Mersenne primes and perfect numbers.

**Definition 5** (Mersenne primes). A prime number q is called a Mersenne prime if  $q = 2^n - 1$  for some  $n \in \mathbb{N}$ .

For example,  $7 = 2^3 - 1$  and  $31 = 2^5 - 1$  are Mersenne primes. Note that q being prime implies n is prime but not all numbers of the form  $2^p - 1$  are Mersenne primes. For instance,  $2^{11} - 1 = 2047 = 23 \times 89$ .

**Definition 6** (Perfect numbers). A number  $n \in \mathbb{N}$  is called *perfect* if the sum of all its positive divisors equals 2n.

For example, 6 is the smallest perfect number, as its divisors are 1, 2, 3 and 6 and their sum is equal to 12.

To link both concepts, we use Euler's  $\sigma$  function, which is similar to Euler's totient function. It is defined as follows:

$$\sigma: \mathbb{N} \to \mathbb{N}$$

$$\sigma(n) = \sum_{d|n,d>0} d$$

### Question 3.1 Perfect Numbers

Here are some calculations to reinforce your understanding in those notions:

- Are 28 and 42 perfect numbers? Show why.
- What is the value of  $\sigma(p^n)$ , where p is prime and  $n \in \mathbb{N}$ ?
- Show that  $\sigma$  is multiplicative, i.e., for  $a, b \in \mathbb{N}$  such that  $\gcd(a, b) = 1$ , we have

$$\sigma(a \cdot b) = \sigma(a)\sigma(b)$$

Answer. • 28 is a perfect number. Since  $28 = 7 \times 4$ , we have

$$\sigma(28) = 1 + 2 + 4 + 7 + 14 + 28 = 56 = 2 \times 28.$$

However, 42 is not perfect. Since  $42 = 2 \times 3 \times 7$ , we have

$$\sigma(42) = 1 + 2 + 3 + 6 + 7 + 14 + 21 + 42 = 96 \neq 84 = 2 \times 42.$$

• Since divisors of  $p^n$  are of the form  $p^m$  with  $0 \le m \le n$ , we have

$$\sigma(p^n) = \sum_{i=0}^{n} p^i = \frac{p^{n+1} - 1}{p - 1}.$$

• Since a and b are coprime, all divisor d of ab can be uniquely writen as  $d = d_a d_b$  with  $d_a | a$  and  $d_b | b$ . Therefore,

$$\sigma(ab) = \sum_{d|ab} d = \sum_{d_a|a,d_b|b} d_a d_b = \sum_{d_a|a} \sum_{d_b|b} d_a d_b = \left(\sum_{d_a|a} d_a\right) \left(\sum_{d_b|b} d_b\right) = \sigma(a)\sigma(b).$$

## Question 3.2 Euclid-Euler Theorem

Prove the following equivalence, known as the Euclid-Euler Theorem:

 $n=2^{p-1}(2^p-1)$  with  $2^p-1$  a Mersenne prime  $\iff n$  is an even perfect number.

**Hint:** Let  $n = 2^k x$  with x odd and n perfect, show that x and  $\frac{x}{2^{k+1}-1}$  muss be the only divisors of x.

Answer.

Let  $n = 2^{p-1}(2^p - 1)$  with  $2^p - 1$  prime. Then,

$$\sigma(n) = \sigma(2^{p-1})\sigma(2^p - 1) = \frac{2^p - 1}{2 - 1}(2^p - 1 + 1) = (2^p - 1)(2^p) = 2n.$$

Thus, n is a perfect number.

Now, suppose  $n = 2^k x$  with x odd, and n is perfect. Then,

$$2^{k+1}x = 2n = \sigma(n) = (2^{k+1} - 1)\sigma(x).$$

Since  $2^{k+1} - 1$  is odd, it must divide x. Hence,  $\frac{x}{2^{k+1} - 1}$  must be a divisor of x. Therefore,

$$\frac{2^{k+1}x}{2^{k+1}-1} = \sigma(x) = x + \frac{x}{2^{k+1}-1} + \dots$$
 other divisors.

The equality holds only if x has no other divisors, meaning x is prime. Additionally,  $1 = \frac{x}{2^{k+1}-1}$  implies  $x = 2^{k+1} - 1$ , which shows x is a Mersenne prime.