# CSC484 Assignment #1

# Andrii Osipa

# January 2017

# Problem 1.1

Solution. Let  $X_i$  be random variable with the following definition:

$$X_i = \begin{cases} 1, \text{there is no balls in } i^{th} \text{ bin} \\ 0, \text{otherwise} \end{cases}$$

. Then  $\sum_{i=1}^n X_i$  is the number of empty bins. If we have no balls in  $i^{th}$  bin then on each ball it was put into any other bin and this event has probability  $\frac{n-1}{n}$ . We have m balls therefore  $P(X_i=1)=\left(\frac{n-1}{n}\right)^m$ . So  $E[X_i]=\left(\frac{n-1}{n}\right)^m$ . Therefore  $E[X]=\sum_{i=1}^n E[X_i]=\sum_{i=1}^n \left(\frac{n-1}{n}\right)^m=n\left(\frac{n-1}{n}\right)^m=\frac{(n-1)^m}{n^{m-1}}$ .

# Problem 1.2

Solution. Probability that  $i^{th}$  and  $j^{th}$  elements are compared:

$$X_{ij} = \begin{cases} 1, i^{th} \text{ and } j^{th} \text{ elements were compared} \\ 0, \text{ otherwise} \end{cases}$$

Lets take a look at three cases:

- 1. i < k' < j: in this case two elements will be compared if one of them is chosen as pivot. Therefore  $P(X_{ij}=1)=\frac{2}{j-i+1}$ .
- 2. i < j < k': in this case we have the following: if pivot is between  $i^{th}$  and  $j^{th}$  elements it is obvious that they never will be compared as they will be in the different partitions.

If pivot is between  $j^{th}$  and  $k^{th}$  elements then they also will not be compared as after partition algorithm will not run Select(...) for subarray where both  $i^{th}$  and  $j^{th}$  elements are.

If pivot is  $k^{th}$  element then those two also will not be compared as algorithm will just return pivot.

If pivot is  $i^{th}$  or  $j^{th}$  element then they will be compared. Therefore

$$P(X_{ij} = 1) = \frac{2}{k - i + 1}.$$

3. k < i < j: this case is similar to previous one. Here we have

$$P(X_{ij} = 1) = \frac{2}{j - k + 1}.$$

Then for total number of comparations we have

$$X = \sum_{i < j} X_{ij} = \sum_{k \le i < j} X_{ij} + \sum_{i < k < j} X_{ij} + \sum_{i < j \le k} X_{ij}.$$

Lets calculate each of the sums:

$$\sum_{k \le i < j} E[X_{ij}] = \sum_{i=k}^{n} \sum_{j=i+1}^{n} \frac{2}{j-k+1} = \sum_{j=k+1}^{n} \sum_{i=k}^{j-1} \frac{2}{j-k+1} = \sum_{j=k+1}^{n} \frac{2(j-k)}{j-k+1} = \sum_{j=k+1}^{n} \frac{2(j-k)$$

$$=\sum_{j=1}^{n-k}\frac{2j}{j+1}=\sum_{j=1}^{n-k}\left(2-2\frac{1}{j+1}\right)=2(n-k+1)-2\sum_{j=1}^{n-k+1}\frac{1}{j}=O(n)-O(\ln n)=O(n)$$

 $\sum_{i < j \le k} E[X_{ij}] = O(n) \text{ and proof is very similar to the previous one.}$ 

$$\sum_{i < k < j} E[X_{ij}] = \sum_{i=1}^{k-1} \sum_{j=k+1}^{n} \frac{2}{j-i+1} = \sum_{i=1}^{k-1} \sum_{j=k+1-i}^{n-i} \frac{2}{j+1} \le \sum_{i=1}^{k-1} \sum_{j=2}^{n-1} \frac{2}{j+1} \le \sum_{i=1}^{k-1} \sum_{j=2}^{n-i} \frac{2}{j+1} \le \sum_{i=1}^{k-1$$

$$\leq (k-1)\ln(n-1) = O(\ln n).$$

Therefore for X we have  $E[X] = O(n) + O(n) + O(\ln n) = O(n)$ .

### Problem 1.3

Solution. Let  $A_i$  be event that i was chosen during step of updating X.

$$A_i = \begin{cases} 1, \text{i was selected on some step} \\ 0, \text{ otherwise} \end{cases}$$

It is easy to see that each value for X can be selected only once. If some  $k \in \{0, ..., n-1\}$  was selected then on every following step randomly selected number will be smaller than k. We have i from 0 to n-1. Number of steps in our algorithm is 1 plus number of updates of X.  $A = \sum_{i=0}^{n-1} A_i$  is total number of updates of X.

 $P(A_i = 1) = \frac{1}{i+1}$ : if was selected any number smaller that i then i will never be selected. Therefore  $E[A_i] = \frac{1}{i+1}$ . And  $E[A] = \sum_{i=0}^{n-1} E[A_i] = \sum_{i=0}^{n-1} \frac{1}{i+1} = O(\ln n)$ .

#### Problem 1.4. Bonus.

Solution. Let  $A_i$  be event that i was chosen during step of updating X.

$$A_i = \begin{cases} 1, \text{i was selected on some step} \\ 0, \text{ otherwise} \end{cases}$$

Each X can be chosen many times during algorithm, therefore  $\sum_{i=0}^{\infty} E[A_i]$  is much less then algorithm runtime. The only possibility that some specific value of X was not chosen at all means that 0 was chosen. Because in all other cases it is still possibility to select X in further steps. Therefore  $P(A_i = 1) = \frac{i}{i+1}$ .

Then 
$$E[A] = \sum_{i=1}^{\infty} E[A_i] = \sum_{i=1}^{\infty} \frac{i}{i+1} = \infty$$
 < runtime.

## Problem 1.5

Solution.  $a_0, ..., a_{k-1} \in \{0, ..., p-1\}$   $X_g = a_0 + a_1 g + a_2 g^2 + ... + a_{k-1} g^{k-1} \mod p, g \text{ from } 0 \text{ to } n-1.$  $P(X_0 = t_0, ..., X_{n-1} = t_{n-1}) = P(X_0 = t_0)...P(X_{n-1} = t_{n-1})?$ 

Obvious fact that  $P(X_0 = t_0) = P(a_0 = t_0) = \frac{1}{p}$ . For any i > 0 also

holds that  $P(X_i = t_i) = \frac{1}{p}$ . We have  $X_i = a_0 + a_1 i + a_2 i^2 + ... + a_{k-1} i^{k-1}$ 

mod p. Suppose  $a_0, ..., a_{r-1}, a_{r+1}, ..., a_{k-1}$  are fixed. Then only for  $a_r * r^i = t_i - a_1 i - a_2 i^2 - ... - a_{k-1} i^{k-1} \mod p$  we have that  $X_i = t_i$ . And here exists only one solution for  $a_r$ .

Proof: suppose there are no solution  $\Leftrightarrow \exists l \in \{0,...,p-1\}: xr^i \neq l \mod p \Rightarrow \exists y: xr^i = yr^i \mod p \Rightarrow (x-y)r^i = 0 \mod p \text{ and } p \text{ is not divisible by any } r: r \neq 1 \text{ and we have that } r \leq p \text{ therefore } x = y \text{ and we have contradiction.}$  Now we proved that solution exists. From same proof it easy to see that solution is the only one.

Therefore same fact holds for any index: we can have randomly selected k-2 indexes and the last one can be picked in the only way that  $X_i = t_i$  holds. So

we have 
$$P(X_i = t_i) = \frac{1}{p}$$
.

Now suppose  $X_0 = t_0, \dots, X_{n-1} = t_{n-1}$ .  $\begin{bmatrix} 1 & 0 & 0^2 & \dots & 0^{k-1} \\ 1 & 1 & 1^2 & \dots & 1^{k-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & n-1 & (n-1)^2 & \dots & (n-1)^{k-1} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{k-1} \end{bmatrix} = \begin{bmatrix} X_0 \\ X_1 \\ \vdots \\ X_{n-1} \end{bmatrix} = \begin{bmatrix} t_0 \\ t_1 \\ \vdots \\ t_{n-1} \end{bmatrix}$ Lets look at this as a system for  $a_0, \dots, a_{k-1}$ .

First matrix is Vandermonde matrix. Determinant of this matrix nonzero as every row contains powers of different numbers from 0 to n-1 and if  $\exists k,l \in \{0,...,n-1\}: k=l \mod p \Rightarrow k=l$  as both k and l are less then p.

Det is nonzero  $\Rightarrow$  system is defined and has only one solution  $a'_0,...,a'_{k-1}$ . Therefore  $X_0=t_0,...,X_{n-1}-t_{n-1}\Leftrightarrow a_0=a'_0,...,a_{k-1}=a'_{k-1}$ . And  $P(a_0=t_0,...,a_{k-1})$  $a'_0, ..., a_{k-1} = a'_{k-1} = \left(\frac{1}{n}\right)^k$ .

As we showed before  $\prod_{i=0}^{n} n - 1P(X_i = t_i) = \left(\frac{1}{p}\right)^k$ . Therefore  $X_0,...,X_{n-1}$  are k-wise independent.

## Problem 2.1

Solution.  $A, B \subseteq U, |A| = \Theta(n), |B| = \Theta(n)$  and  $A \cap B = \emptyset$ . Let's denote by  $c_A$ and  $c_B$  constants such that  $|A| \leq c_A n$  and  $|B| \leq c_B n$ .  $P(x \in R) = \Theta(\frac{1}{n})$  and

and 
$$c_B$$
 constants such that  $|A| \leq c_A n$  and  $|B| \leq c_B n$ .  $P(x \in R) = \Theta(\frac{\pi}{n})$  and let  $c_R$  be constant s.t.  $P(x \in R) \geq \frac{c_R}{n}$ . 
$$P(A \cap R = \emptyset \land B \cap R = \emptyset) = \prod_{x \in A \cup B} P(x \notin R) = \prod_{x \in A \cup B} (1 - P(x \in R)) \geq \prod_{x \in A \cup B} \left(1 - \frac{c_R}{n}\right) = \left(1 - \frac{c_R}{n}\right)^{|A| + |B|} \geq \left(1 - \frac{c_R}{n}\right)^{c_A n + c_B n} = \left(1 - \frac{c_R}{n}\right)^{n(c_A + c_B)}$$
.

The latter is strictly decreasing with the following property, known from calcu-

$$\lim_{n \to \infty} \left( 1 - \frac{c_R}{n} \right)^{n(c_A + c_B)} = e^{-c_R(c_A + c_B)} =: c.$$

Therefore we showed that exists some constant c > 0 which is lower bound for  $P(A \cap R = \emptyset \land B \cap R = \emptyset).$ 

**Problem 2.2**  $X_1, X_2, \dots$  random variables.  $P(X_i = 1) = p$  and  $P(X_i = -1) = p$ 1-p. T is smallest t s.t.  $\sum_{i=1}^{t} X_i < 0$ . E[T] = ?

Solution. It is easy to see that T can not be even. Suppose T=2k for some k.  $\sum_{i=1}^{L} X_i < 0$  therefore there must be at least k+1 negative ones and k positive

ones and so 
$$\sum_{i=1}^{T} X_i = -2m$$
 for some  $m$ . If  $X_T$  is  $-1$  then  $\sum_{i=1}^{T-1} X_i = -2m+1 < 0$ 

therefore T is not the smallest one. If  $X_T$  is 1 then  $\sum_{i=1}^{T-1} X_i = -2m-1 < 0$ therefore T is not the smallest one. This gives us next fact: P(T=2k)=0.

For odd Ts we have: obviously, that  $X_T$  is -1, otherwise T is not the smallest.

Therefore  $\sum_{i=1}^{T-1} X_i = 0$ , otherwise T also not the smallest one.

 $E[T] = \sum_{k=0}^{\infty} c_k p^{k-1} (1-p)^k$ , where  $c_k$  is number of sequences of 1 and -1 of length 2k-1 and such that at each point < 2k-1 sum of sequence is  $\ge 0$  and total sum is -1.  $c_1 = 1$ ,  $c_2 = 1$ ,  $c_3 = 2$ ,  $c_4 = 5$ , etc.  $c_k = ?$ . ...