# 2020 FALL MAS583 HW4

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## 1 Problem 1

### 1.1 Question

Let X be a random variable taking integral nonnegative values, let  $\mathbb{E}[X^2]$  denote the expectation of its square, and let Var[X] denote its variance. Prove that

$$\Pr[X = 0] \le \frac{\operatorname{Var}[X]}{\mathbb{E}[X^2]}.$$

#### 1.2 Answer

The above inequality is equivalent to the following one,

$$\Pr[X \ge 1] \ge \frac{\mathbb{E}^2[X]}{\mathbb{E}[X^2]}.$$

Let  $P_i$  denote Pr[X = i] for any nonnegative integer i, we have

$$\Pr[X \ge 1] \mathbb{E}[X^2] - \mathbb{E}^2[X]$$

$$= (\sum_{i=1}^{\infty} P_i) (\sum_{i=1}^{\infty} i^2 P_i) - (\sum_{i=1}^{\infty} i P_i)^2$$

$$= \sum_{i,j} (j^2 P_i P_j) - \sum_{i,j} (ij P_i P_j)$$

$$= \frac{1}{2} (\sum_{i,j} (i^2 P_i P_j) + \sum_{i,j} (j^2 P_i P_j) - \sum_{i,j} (2ij P_i P_j))$$

$$= \frac{1}{2} \sum_{i,j} ((i-j)^2 P_i P_j)$$
>0.

completing the proof.

# 2 Problem 2

### 2.1 Question

(\*) Show that there is a positive constant c such that the following holds. For any n reals  $a_1, a_2, ..., a_n$  satisfying  $\sum_{i=1}^n a_i^2 = 1$ , if  $(\epsilon_1, ..., \epsilon_n)$  is a  $\{-1, 1\}$ -random vector obtained by

choosing each  $\epsilon_i$  randomly and independently with uniform distribution to be either -1 or 1, then

$$\Pr\left[\left|\sum_{i=1}^{n} \epsilon_i a_i\right| \le 1\right] \ge c.$$

#### 2.2 Answer

First let's consider the cases where for all i,  $|a_i| < 1/2$ . Let k be the minimum number such that  $\sum_{i=1}^k a_i^2 \ge 1/2$ , as k is minimum,  $\sum_{i=1}^{k-1} a_i^2 < 1/2$ , besides,  $|a_k| < 1/2$ , i.e.,  $a_k^2 < 1/4$ , thus,  $1/2 \le \sum_{i=1}^k a_i^2 < 3/4$  and  $1/4 < \sum_{i=k+1}^n a_i^2 \le 1/2$ .

Let  $X = \sum_{i=1}^k \epsilon_i a_i$  and  $Y = \sum_{i=k+1}^n \epsilon_i a_i$  who are independent with each other. Clearly,  $\mathbb{E}[X] = \mathbb{E}[Y] = 0$ , and  $\operatorname{Var}[X] = \sum_{i=1}^k a_i^2 \operatorname{Var}[\epsilon_i] = \sum_{i=1}^k a_i^2$ , similarly  $\operatorname{Var}[Y] = \sum_{i=k+1}^n a_i^2$ . Thus, as X and Y are symmetric on the both sides of zero,

$$\Pr[0 \le X \le 1] \ge \frac{1}{2} \Pr[|X| \le 1] \ge \frac{1}{2} (1 - \text{Var}[X]) > 1/8,$$

$$\Pr[-1 \le Y \le 0] \ge \frac{1}{2} \Pr[|Y| \le 1] \ge \frac{1}{2} (1 - \text{Var}[Y]) \ge 1/4,$$

finally, we have

$$\Pr\left[\left|\sum_{i=1}^{n} \epsilon_{i} a_{i}\right| \le 1\right] = \Pr[|X + Y| \le 1] \ge \Pr[0 \le X \le 1] \Pr[-1 \le Y \le 0] > 1/32.$$

As for the cases where  $|a_i| \ge 1/2$  for some i, let l be any number such that  $|a_l| \ge 1/2$ , and let  $X = \sum_{i \ne l} \epsilon_i a_i$ ,  $Y = \epsilon_l a_l$ . Still clearly,  $\mathbb{E}[X] = \mathbb{E}[Y] = 0$ ,  $\operatorname{Var}[X] = \sum_{i \ne l} a_i^2 = 1 - a_l^2 \le 3/4$ ,  $\operatorname{Var}[Y] = a_l^2$ ,  $\operatorname{Pr}[|Y| \le 1] = 1$ . Thus, with the same symmetry,

$$\Pr[0 \le X \le 1] \ge \frac{1}{2} \Pr[|X| \le 1] \ge \frac{1}{2} (1 - \text{Var}[X]) \ge 1/8,$$
  
$$\Pr[-1 \le Y \le 0] \ge \frac{1}{2} \Pr[|Y| \le 1] = 1/2,$$

finally we have

$$\Pr\left[\left|\sum_{i=1}^{n} \epsilon_{i} a_{i}\right| \leq 1\right] = \Pr[|X + Y| \leq 1] \geq \Pr[0 \leq X \leq 1] \Pr[-1 \leq Y \leq 0] \geq 1/16,$$

completing the proof.

#### 2.3 Reference

https://people.math.gatech.edu/~tetali/MATH7018 SPR14/HW2.pdf

### 3 Problem 3

#### 3.1 Question

(\*) Show that there is a positive constant c such that the following holds. For any n vectors  $a_1, a_2, ..., a_n \in \mathbb{R}^2$  satisfying  $\sum_{i=1}^n \|a_i\|^2 = 1$  and  $\|a_i\| \leq 1/10$ , where  $\|\cdot\|$  denotes the

usual Euclidean norm, if  $(\epsilon_1, ..., \epsilon_n)$  is a  $\{-1, 1\}$ -random vector obtained by choosing each  $\epsilon_i$  randomly and independently with uniform distribution to be either -1 or 1, then

$$\Pr\left[\left\|\sum_{i=1}^n \epsilon_i a_i\right\| \le 1/3\right] \ge c.$$

#### 3.2 Answer

First, we design an algorithm which gives the following lemma.

**Lemma 3.1** Given any t vectors  $v_1, ..., v_t \in \mathbb{R}^2$  with each  $||v_i|| \leq d$ , for some given constant d > 0. We can find  $\epsilon_1, ..., \epsilon_t \in \{-1, 1\}$  such that

$$\left\| \sum_{i=1}^{t} \epsilon_i v_i \right\| \le 3d.$$

Equivalently, if each  $\epsilon_i$  is independently chosen in  $\{-1,1\}$ , uniformly at random, then we have

$$\Pr\left[\left\|\sum_{i=1}^{t} \epsilon_i v_i\right\| \le 3d\right] \ge 1/2^t$$

**Proof.** First, we assume that each  $v_i \neq 0$  and divide  $\mathbb{R}^2 \setminus 0$  into 6 sections  $S_1, ..., S_6$ , where for  $i \in [6]$ ,  $S_i$  contains all the vectors who have argument in  $[(i-1)\pi/3, i\pi/3)$ . For each vector v in  $S_2$ ,  $S_4$ , or  $S_6$ , we can consider -v who's in  $S_1$ ,  $S_3$ , or  $S_5$ , then we have t vectors where each of them is in  $S_1$ ,  $S_3$ , or  $S_5$ . Now, at each step, we arbitrarily choose two vectors who are in the same section until all vectors are in different sections, when the total number of vectors is no more than 3 as we will guarantee that in the whole process, each vector is in  $S_1$ ,  $S_3$ , or  $S_5$ . As the two vectors we choose, say x and y, are in the same section, their difference in argument is no more than  $\pi/3$ , we can easily check that among the four possible results x + y, x - y, -x + y, -x - y, there exists one that is 0 or is in  $S_1$ ,  $S_3$ , or  $S_5$  with norm no more than the maximum of ||x|| and ||y||. Therefore, when the algorithm ends, we have at most 3 vectors with norm no more than d, thus, any linear combination of them with coefficient -1 or 1 has norm no more than 3d, completing the proof.

**Theorem 3.2 (McDiarmid's Inequality)** Let  $x_1, ..., x_n$  be independent random variables taking on values in a set A and let  $c_1, ..., c_n$  be positive real constants. If  $\phi : A^n \to \mathbb{R}$  satisfies

$$\sup_{x_1,...,x_n,x_i' \in A} |\phi(x_1,...,x_i,...,x_n) - \phi(x_1,...,x_i',...,x_n)| \le c_i,$$

for  $i \in [n]$ , then for any  $\epsilon > 0$ ,

$$\Pr[\phi(x_1, ..., x_n) - \mathbb{E}[\phi(x_1, ..., x_n)] \ge \epsilon] \le e^{-2\epsilon^2 / \sum_{i=1}^n c_i^2}.$$

Let  $k_0 = 0$ , and for i = 1, 2, ..., let  $k_i$  be the minimum number such that

$$\sum_{j=k_{i-1}+1}^{k_i} \|a_j\|^2 \ge 1/10000,$$

until  $k_i = n$  or

$$\sum_{j=k_{i-1}+1}^{n} \|a_j\|^2 < 1/10000$$

where we let  $k_i = n$ . For each i, as  $||a_i|| \le 1/10$ , i.e.,  $||a_i||^2 \le 1/100$ ,

$$\sum_{j=k_{i-1}+1}^{k_i} \|a_j\|^2 < \frac{1}{100} + \frac{1}{10000}.$$

Therefore, the last i where  $k_i = n$  satisfies that  $99 \le i \le 10000$ . Let

$$X_i = \phi_i(\{\epsilon_j\}_{j=k_{i-1}+1}^{k_i}) = \left\| \sum_{j=k_{i-1}+1}^{k_i} \epsilon_j a_j \right\|,$$

we have

$$\mathbb{E}[X_i^2] = \sum_{j=k_{i-1}+1}^n \|a_j\|^2,$$

thus,

$$\mathbb{E}[X_i] \le \sqrt{\mathbb{E}[X_i^2]} < \sqrt{\frac{1}{100} + \frac{1}{10000}} < 0.1005.$$

As we have

$$|\phi_i(a_{k_{i-1}+1},...,a_m,...,a_{k_i}) - \phi_i(a_{k_{i-1}+1},...,a_m',...,a_{k_i})| \le 2||a_m||,$$

by McDiarmid's Inequality,

$$\begin{aligned} \Pr[X_i \leq 2/19] &\geq 1 - \Pr[X_i \geq 2/19] \\ &= 1 - \Pr[X_i - \mathbb{E}[X_i] \geq 2/19 - \mathbb{E}[X_i]] \\ &\geq 1 - \Pr[X_i - \mathbb{E}[X_i] \geq 2/19 - \sqrt{1/50}] \\ &\geq 1 - e^{-2(2/19 - 0.1005)^2 / \sum_{j=k_{i-1}+1}^{k_i} (2||a_m||)^2} \\ &\geq 1 - e^{-(2/19 - 0.1005)^2 / 0.201} \\ &\geq 1/10000. \end{aligned}$$

Therefore, with probability at least  $1/20000^{10000}$ , we have

$$\left\| \sum_{j=k_{i-1}+1}^{k_i} \epsilon_j a_j \right\| \le 2/19,$$

for each i, where each probability is halved as the two choices where every two counterparts are inverse number are taken into consideration. And the algorithm (Lemma 1.1) gives us a way to finally get a vector whose norm is no more than  $\frac{6}{19} < 1/3$ , which has a probability at least  $1/2^{10000}$  given all these sub-sum of vectors with desired norm bound. Thus, we have

$$\Pr\left[\left\|\sum_{i=1}^{n} \epsilon_i a_i\right\| \le 1/3\right] \ge 1/40000^{10000},$$

completing the proof.

### 3.3 A Failed Attempt

**Theorem 3.3 (McDiarmid's Inequality)** Let  $x_1, ..., x_n$  be independent random variables taking on values in a set A and let  $c_1, ..., c_n$  be positive real constants. If  $\phi : A^n \to \mathbb{R}$  satisfies

$$\sup_{x_1,...,x_n,x_i' \in A} |\phi(x_1,...,x_i,...,x_n) - \phi(x_1,...,x_i',...,x_n)| \le c_i,$$

for  $i \in [n]$ , then

$$\Pr[\phi(x_1, ..., x_n) - \mathbb{E}[\phi(x_1, ..., x_n)] \ge \epsilon] \le e^{-2\epsilon^2 / \sum_{i=1}^n c_i^2}$$

In our case, we have  $\epsilon_1, ..., \epsilon_n$  that are independent random variables taking on values in  $A = \{-1, 1\}$  and let  $\phi(\epsilon_1, ..., \epsilon_n) = \|\sum_{i=1}^n \epsilon_i a_i\|$ , then

$$\sup_{\epsilon_1,...,\epsilon_n,\epsilon_i'\in A} |\phi(\epsilon_1,...,\epsilon_i,...,\epsilon_n) - \phi(\epsilon_1,...,\epsilon_i',...,\epsilon_n)| \le 2||a_i|| \le \frac{1}{5}.$$

Let  $t = \mathbb{E}[\phi(\epsilon_1, ..., \epsilon_n)]$ , by McDiarmid's Inequality, we have

$$\Pr\left[\left\|\sum_{i=1}^{n} \epsilon_{i} a_{i}\right\| \leq 1/3\right]$$

$$= \Pr\left[\phi(\epsilon_{1}, ..., \epsilon_{n}) - t \leq 1/3 - t\right]$$

$$\geq 1 - e^{-2(1/3 - t)^{2}/4 \sum_{i=1}^{n} \|a_{i}\|^{2}}$$

$$= 1 - e^{-(1/3 - t)^{2}/2}.$$

As each  $||a_i||$  is bounded, t is also bounded (Actually, we need to prove t < 1/3, on which I failed), therefore, there must be some positive c such that

$$\Pr\left[\left\|\sum_{i=1}^{n} \epsilon_i a_i\right\| \le 1/3\right] \ge c,$$

completing the proof.

# 3.4 Another Failed Attempt

To show

$$\left\| \sum_{i=1}^{n} \epsilon_i a_i \right\| \le 1/3,$$

suffices to show

$$\langle \sum_{i=1}^{n} \epsilon_i a_i, v \rangle \le 1/3, \forall v \in \mathbb{R}^2 \text{ with } ||v|| = 1.$$

Given any  $v \in \mathbb{R}^2$  with ||v|| = 1, let  $f_v : \{-1, 1\}^n \to \mathbb{R}$ , where

$$f_v(\epsilon = (\epsilon_1, ..., \epsilon_n)) = \langle \sum_{i=1}^n \epsilon_i a_i, v \rangle = \sum_{i=1}^n \epsilon_i \langle a_i, v \rangle$$

Theorem 3.4 (Azuma's inequality) Let  $X_0, ..., X_n$  be a Martingale with  $|X_t - X_{t-1}| \le a_t$  for any  $t \in [n]$ , then for any  $\lambda \ge 0$  one has  $\Pr[|X_n - X_0| > \lambda ||a||] \le 2e^{-\lambda^2/4}$ , where  $a = (a_1, ..., a_n)$ .

Let  $X_0 = 0$  and  $X_i = \sum_{k=1}^i \epsilon_k \langle a_k, v \rangle$  for any  $i \in [n]$ , easy to check  $X_0, ..., X_n$  is a Martingale, where

$$|X_t - X_{t-1}| = \langle a_t, v \rangle, \forall t \in [n].$$

Note that  $X_n = f_v(\epsilon)$ . By Azuma's inequality and the symmetric of  $f_v$ , we have

$$\Pr[X_n > 1/3] = \frac{1}{2} \Pr[|X_n| > 1/3] \le e^{-\lambda^2/4},$$

where  $\lambda = (3\sqrt{\sum_{i=1}^{n} \langle a_i, v \rangle^2})^{-1} \ge (3\sqrt{\sum_{i=1}^{n} \|a_i\|^2})^{-1} = 1/3$ . Therefore, we have

$$\Pr\left[\left\langle \sum_{i=1}^{n} \epsilon_i a_i, v \right\rangle \le 1/3\right] \ge 1 - e^{-1/36},$$

completing the proof (I want to claim that if we let v has the same direction with  $\sum_{i=1}^{n} \epsilon_i a_i$  then we can complete the proof but I feel that it's not so clear).

#### 3.5 Some Other Ideas

First, show that when  $n \leq c$ , where c is a large constant, say  $10^{100}$ , we can find at least 1 possible choice of  $(\epsilon_1, ..., \epsilon_n)$  such that  $\|\sum_{i=1}^n \epsilon_i a_i\| \leq 1/3$ . Then we have

$$\Pr\left[\left\|\sum_{i=1}^{n} \epsilon_i a_i\right\| \le 1/3\right] \ge \frac{1}{2^{10^{100}}},$$

for all  $n \leq 10^{100}$ . When  $n > 10^{100}$ , we need to show there are at least  $2^{n-10^{100}}$  choices of  $(\epsilon_1, ..., \epsilon_n)$  satisfy the desired condition. Intuitively, we may expect that we can find  $n-10^{100}$  a<sub>i</sub> such that for all the  $2^{n-10^{100}}$  choices of corresponding  $\epsilon_i$ , we can find at least 1 possible solution for the remaining  $\epsilon_i$ . However, it's obvious that when  $n \to \infty$ , we can hardly prove this.

#### 3.6 Reference

Rothvoss, Thomas. "Probabilistic Combinatorics." (2019).

Kutin, Samuel. "Extensions to McDiarmid's inequality when differences are bounded with high probability." Department Computer Science, University of Chicago, Chicago, IL. Technical report TR-2002-04 (2002).

Besides, I got some ideas from Prof. Liu's lectures when he mentioned Martingale.

## 4 Problem 4

### 4.1 Question

Let X be a random variable with expectation  $\mathbb{E}[X] = 0$  and variance  $\sigma^2$ . Prove that for all  $\lambda > 0$ ,

$$\Pr[X \ge \lambda] \le \frac{\sigma^2}{\sigma^2 + \lambda^2}.$$

#### 4.2 Answer

For any  $\lambda > 0$ , let  $Y_{\lambda} = \sigma^2 + \lambda X$ , clearly  $\mathbb{E}[Y_{\lambda}] = \sigma^2$  for all  $\lambda > 0$ , thus by Markov's inequality,

$$\Pr[X \ge \lambda] = \Pr[Y_{\lambda} \ge \sigma^2 + \lambda^2] \le \frac{\sigma^2}{\sigma^2 + \lambda^2},$$

completing the proof.

# 5 Problem 5

### 5.1 Question

Let  $v_1 = (x_1, y_1), ..., v_n = (x_n, y_n)$  be n two-dimensional vectors, where each  $x_i$  and each  $y_i$  is an integer whose absolute value does not exceed  $2^{n/2}/(100\sqrt{n})$ . Show that there are two disjoint sets  $I, J \subset \{1, 2, ..., n\}$  such that

$$\sum_{i \in I} v_i = \sum_{j \in J} v_j.$$

#### 5.2 Answer

First, note that "disjoint" actually makes no difference because if  $\sum_{i \in I} v_i = \sum_{j \in J} v_j$  but  $I \cap J \neq \emptyset$ , we can let  $\tilde{I} = I \setminus (I \cap J)$  and  $\tilde{J} = J \setminus (I \cap J)$ , which are disjoint and whose corresponding summations are still equal. When n = 1, clearly  $v_1 = (0,0)$  and  $I = \{1\}, J = \emptyset$  are the desired sets.

Suppose for all  $I, J \subset \{1, 2, ..., n\}$  such that  $I \neq J$ ,  $\sum_{i \in I} v_i \neq \sum_{j \in J} v_j$ . As  $\{1, 2, ..., n\}$  has  $2^n$  subsets,  $|\{\sum_{i \in I} v_i : I \subset \{1, 2, ..., n\}\}| = 2^n$ . On the other hand, let  $\{\epsilon_i\}_{i=1}^n$  be a sequence of random variables where each one is independently chosen in  $\{0, 1\}$  uniformly at random, and let  $X = \sum_{i=1}^n \epsilon_i x_i$  and  $Y = \sum_{i=1}^n \epsilon_i y_i$ , we have

$$Var[X] = \sum_{i=1}^{n} \frac{1}{4} x_i^2 \le \frac{2^n}{40000},$$

similarly,

$$Var[Y] \le \frac{2^n}{40000}.$$

Thus, by Chebyshev's inequality,

$$\Pr[|X - \mathbb{E}[X]| \ge 2\frac{2^{n/2}}{200}] \le \frac{1}{4},$$

similarly,

$$\Pr[|Y - \mathbb{E}[Y]| \ge 2\frac{2^{n/2}}{200}] \le \frac{1}{4},$$

by the independence between X and Y, which follows

$$\Pr[|X - \mathbb{E}[X]| \le \frac{2^{n/2}}{100}, |Y - \mathbb{E}[Y]| \le \frac{2^{n/2}}{100}] \ge \frac{1}{2}.$$

Note that (X,Y) is  $\sum_{i\in[n],\epsilon_i=1}v_i$ , which means among all the  $2^n$  choices of  $I\subset[n]$ , at least  $2^{n-1}$  should satisfy  $|\sum_{i\in I}x_i-\mathbb{E}[X]|\leq \frac{2^{n/2}}{100}$  and  $|\sum_{i\in I}y_i-\mathbb{E}[Y]|\leq \frac{2^{n/2}}{100}$ , where  $\sum_{i\in I}x_i$  and  $\sum_{i\in I}y_i$  should be distinct integers. However, the size of feasible set of this linear program is at most  $((2^{n/2}/50)+1)^2<2^{n-1}$  for all  $n\geq 2$ , which implies that there must be  $I\neq J$  such that  $\sum_{i\in I}v_i=\sum_{j\in J}v_j$ , completing the proof.

### 5.3 A Failed Attempt

First, if  $\exists i \neq j$  such that  $v_i = v_j$ , then we can just simply let  $I = \{i\}$  and  $J = \{j\}$  and the condition is satisfied. So we can only consider cases where each  $v_i$  is distinct. Let  $N = \lfloor 2^{n/2}/(100\sqrt{n}) \rfloor$  and  $\tilde{v}_i = (2N+1)(x_i+N)+(y_i+N)+1$ . Clearly, each  $\tilde{v}_i \in [(2N+1)^2] \subset [2^n/1000n]$  and is distinct. Furthermore,  $\sum_{i \in I} v_i = \sum_{j \in J} v_j$  if and only if (WRONG!!! This only holds when |I| = |J|)  $\sum_{i \in I} \tilde{v}_i = \sum_{j \in J} \tilde{v}_j$ . Suppose such I and J don't exist, i.e., for any two disjoint sets  $I, J \subset \{1, 2, ..., n\}$ ,  $\sum_{i \in I} v_i \neq \sum_{j \in J} v_j$ , we have  $\{\tilde{v}_i\}_{i=1}^n \subset [2^n/1000n]$  has distinct sums ("disjoint" actually makes no difference as if  $\sum_{i \in I} v_i = \sum_{j \in J} v_j$  but  $I \cap J \neq \emptyset$ , we can let  $\tilde{I} = I \setminus (I \cap J)$  and  $\tilde{J} = J \setminus (I \cap J)$ , which are disjoint and whose summations are still equal), thus we should have  $2^n < (2^n/1000n)n$ , which obviously doesn't hold. Thus, such I and J must exist.

# 6 Problem 6

### 6.1 Question

(\*) Prove that for every set X of at least  $4k^2$  distinct residue classes modulo a prime p, there is an integer a such that the set  $\{ax \pmod p : x \in X\}$  intersects every interval in  $\{0, 1, ..., p-1\}$  of length at least p/k.

### 6.2 Answer

Let  $P = \{0, 1, ..., p-1\}$  and  $Y = Y_a = \{ax \pmod{p} : x \in X\}$ , to prove the desired result, we need to find such a so that Y intersects every interval of length  $\lceil p/k \rceil$  in P. Let  $J_1, ..., J_{2k}$  be a fixed covering of P by 2k intervals of length  $\lceil p/2k \rceil$  each. Note that if Y intersects with each  $J_i$  then it certainly satisfies the desired property because every interval of length at least  $\lceil p/k \rceil$  in P must fully contain at least one  $J_i$ . Fix  $i \in [2k]$ , for any  $x \in X$ , let  $S_x^i$  be the indicating random variable of  $ax \pmod{p} \in J_i$ , then  $S_i = \sum_{x \in X} S_x^i$  is 0 if and only if Y doesn't intersect  $J_i$ . By linearity of expectation,

$$\mathbb{E}[S^i] = \sum_{x \in X} \mathbb{E}[S_x^i] = \frac{4k^2 \lceil p/2k \rceil}{p}.$$

Note that the random variables  $S_x^i$  for  $x \in X$  are pairwise independent, hence

$$\operatorname{Var}[S^i] = \sum_{x \in X} \operatorname{Var}[S_x^i] = \frac{4k^2 \lceil p/2k \rceil}{p(1 - \lceil p/2k \rceil/p)} = \frac{\mathbb{E}[S^i]}{1 - \lceil p/2k \rceil/p}.$$

Therefore, by Chebyshev's Inequality,

$$\Pr(S^i = 0) \le \Pr(|S^i - \mathbb{E}[S^i]| \ge \mathbb{E}[X^i]) \le \frac{\operatorname{Var}[S^i]}{\mathbb{E}^2[S^i]} < \frac{1}{2k},$$

since there are 2k possible choices of i, we have that with positive probability,  $S^i > 0$  for all  $i \in [2k]$ , completing the proof.

# 6.3 Reference

Alon, Noga, Igor Kriz, and J. Nesetril. "How to color shift hypergraphs." Studia Scientiarum Mathematicarum Hungarica 30.1 (1995): 1-12.

Alon, Noga, and Yuval Peres. "Uniform dilations." Geometric & Functional Analysis GAFA 2.1 (1992): 1-28.