## Lecture 0: Mathematical Preliminaries

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In this note we will cover some basics of probability theory and algebra which will be used during the course.

### 0.1 Some useful inequalities

We will use capital letters to denote random variables. The notations $\mathbb{E}(X)$ and $\operatorname{Var}(X)$ stand for expectation and variance of the random variable $X$, respectively.

Lemma 0.1.1 (Markov's inequality). Let $X$ be any non-negative random variable. Then,

$$
\operatorname{Pr}[X \geq \alpha] \leq \frac{\mathbb{E}(X)}{\alpha}
$$

Lemma 0.1.2 (Chebyshev's inequality). Let $X$ be any random variable and $\alpha>0$. Then,

$$
\operatorname{Pr}[|X-\mathbb{E}(X)| \geq \alpha] \leq \frac{\mathbb{V a r}(X)}{\alpha^{2}}
$$

Markov's inequality can be used to prove Chebyshev's inequality. From Markov's inequality, one can also obtain the following strong tail bound for independent random variables.

Lemma 0.1.3 (Chernoff bound). Let $X_{1}, X_{2}, \ldots, X_{n}$ be i.i.d random variables and $\forall i X_{i} \in$ $\{0,1\}$. Let $X=\sum_{i=1}^{n} X_{i}$. Then,

$$
\operatorname{Pr}[|X-\mathbb{E}(X)| \geq \alpha \mathbb{E}(X)] \leq 2 e^{-\alpha^{2} \mathbb{E}(X)}
$$

Exercise 1. You are given a fair unbiased coin. The coin is tossed $n$ times independently. Use all the above inequalities and compute the probability of the following events.

1. More than $3 n / 4$ heads are observed.
2. More than $n / 2+2 \sqrt{n}$ heads are observed.

Comment on which inequalities are applicable and comment on which inequality gives the best bounds.

Exercise 2. You are given $n$ independent random variables $X_{1}, X_{2}, \ldots, X_{k}$. For every $i \in[n], \operatorname{Pr}\left[X_{i}=1\right] \geq 3 / 4$ and $\operatorname{Pr}\left[X_{i}=0\right] \leq 1 / 4$. Let $X=\sum_{i} X_{i}$. In terms of $k$ compute the probability of the event $X \leq k / 4$.

Exercise 3. Let $b\{0,1\}$ be a fixed bit. We generate bits $X_{1}, X_{2}, \ldots, X_{k}$ from by tossing independent coins. Each coin comes up HEAD with probability $3 / 4$ and TAIL with probability $1 / 4$. If the $i$ th coin toss comes out to be HEAD then $X_{i}=b$ else $X_{i}=1-b$. Let $X=$ majority $_{i}\left(X_{i}\right)$, that is $X=1$ iff $\sum_{i} X_{i} \geq k / 2$. What is the probability that $X \neq b$ ?

### 0.2 Abstract algebra

A field $\mathbb{F}=(S,+, *)$ is a set $S$ with two binary operators, + and $*$ with the following properties:

- Closure: For all $a, b \in S, a+b \in S$ and $a * b \in S$.
- Associativity: For all $a, b, c \in S a+(b+c)=(a+b)+c$ and $a *(b * c)=(a * b) * c$.
- Identity: There exist two special elements $i_{0}, i_{1} \in S$ such that for all $a \in S a+i_{0}=$ $i_{0}+a=a$ and $a * i_{1}=i_{1} * a=a$. Here, $i_{0}$ is called the additive identity and $i_{1}$ is called the multiplicative identity.
- Inverses: For each element $a \in S$ there exist $a^{\prime}, a^{\prime \prime} \in S$ such that $a+a^{\prime}=a^{\prime}+a=i_{0}$ and $a * a^{\prime \prime}=a^{\prime \prime} * a=i_{1}$.
- Distributivity: For all $a, b, c \in S, a *(b+c)=a * b+a * c$.

A field is called a finite field if $|S|$ is finite.
Exercise 4. Let $p$ be a prime and let $\mathbb{F}_{p}$ denote $(\{0,1, \ldots, p-1\},+(\bmod p), \times(\bmod p))$. Prove that $\mathbb{F}_{p}$ is a finite field. Here,$+(\bmod p)$ and $\times(\bmod p)$ represent addition and multiplication modulo $p$.

Is $\mathbb{F}_{6}$ a finite field? Justify your answer.
Exercise 5. Let $p$ be a prime and let $\mathbb{F}_{p}[x]=\left(S(x), \oplus_{p}, \otimes_{p}\right)$ be a structure defined so that $S=\{$ polynomials over the indeterminate $x$ with coefficients from $\{0,1, \ldots, p-1\}\}$, for two polynomials $r(x), q(x) \in S, r(x) \oplus_{p} q(x)$ defined as addition of two polynomials with coefficients modulo $p$ and $r(x) \otimes_{p} q(x)$ defined as multiplication of two polynomials with coefficients modulo $p$. Prove that $\mathbb{F}_{p}[x]$ is not a field.

The above is a very useful structure and we may encounter it many times during the course.

Exercise 6. Let us consider the following structure: $\left(\{1,0, x, 1+x\},+(\bmod 2), \times\left(\bmod x^{2}+\right.\right.$ $x+1)$ ). Prove that this is a finite field. This finite field is often denoted as $\mathbb{F}_{2^{2}}$, as this is a finite field with 4 elements.

In the Exercise 6 we have constructed a field of size $2^{2}$ by performing additions modulo 2 and multiplications modulo a certain fixed polynomial of degree 2 . In the same way, if we were to create fields of size $2^{k}$, we can do this by performing additions modulo 2 and multiplications modulo a certain fixed polynomial of degree $k$.

### 0.3 Linear algebra

Exercise 7. Let $Q$ be a $2 \times n 0-1$ matrix. Suppose the rank of $Q$ is 2, then

$$
\operatorname{Pr}_{\alpha \in\{0,1\}^{n}}\left[Q \alpha=\left[\begin{array}{l}
0 \\
0
\end{array}\right]\right]=\operatorname{Pr}_{\alpha \in\{0,1\}^{n}}\left[Q \alpha=\left[\begin{array}{l}
0 \\
1
\end{array}\right]\right]=\operatorname{Prr}_{\alpha \in\{0,1\}^{n}}\left[Q \alpha=\left[\begin{array}{l}
1 \\
0
\end{array}\right]\right]=\operatorname{Pr}_{\alpha \in\{0,1\}^{n}}\left[Q \alpha=\left[\begin{array}{l}
1 \\
1
\end{array}\right]\right]
$$

### 0.4 Pairwise independence

Let us define a set of functions $\mathcal{H}=\left\{h:\{0,1\}^{m} \rightarrow\{0,1\}^{n}\right\}$. The number of functions in this set is $\left(2^{n}\right)^{2^{m}}$. The number of bits required to pick a random function from this family is $\log \left(\left(2^{n}\right)^{2^{m}}\right)$, i.e. $O\left(2^{m} n\right)$.

Definition 0.4.1. We call a set of functions $\mathcal{F} \subseteq \mathcal{H}$ a pairwise independent family of functions if $\forall x \neq y \in\{0,1\}^{m}$ and for any fixed $u, v \in\{0,1\}^{n}$ we have

$$
\operatorname{Pr}_{f \in \mathcal{F}}[f(x)=u \wedge f(y)=v]=\frac{1}{2^{2 n}}
$$

In class we will often design an algorithm which picks a purely random function from a set of all functions, i.e. from $\mathcal{H}$. We will then analyze the algorithm and observe that we only need to pick a purely random function from a set of pairwise independent family of functions. We will first give a construction of such a family and then observe some useful properties of such families.

Exercise 8 (Pairwise independent hash functions). Let $A \in\{0,1\}^{m \times n}$ and $b \in\{0,1\}^{n}$. Let us define a function $f_{A, b}:\{0,1\}^{m} \rightarrow\{0,1\}^{n}$ as $f_{A, b}(x)=A x+b$, where all additions and multiplications are defined modulo 2. Let $\mathcal{F}=\left\{f_{A, b} \mid A \in\{0,1\}^{n \times m}, b \in\{0,1\}^{n}\right\}$. Prove that $\mathcal{F}$ is a family of pairwise independent hash functions. Formally, prove that $\forall x \neq y \in\{0,1\}^{m}$ and for any fixed $u, v \in\{0,1\}^{n}$ we have

$$
\operatorname{Pr}_{A \in\{0,1\}^{m \times n}, b \in\{0,1\}^{n}}\left[f_{A, b}(x)=u \wedge f_{A, b}(y)=v\right]=\frac{1}{2^{2 n}}
$$

Proof. Let $x \neq y \in\{0,1\}^{m}$ be any two fixed vectors and let $u, v \in\{0,1\}^{n}$ also be two fixed vectors. For an $n \times m$ matrix $A$, let the $i$ th row of the matrix be denoted as $\tilde{a}_{i}$. For a vector $b$ let $b_{i}$ denote its $i$ th bit. Then the condition $A x+b=u$ can be written as $\wedge_{i=1}^{n}\left(\left\langle\tilde{a}_{i}, x\right\rangle+b_{i}=u_{i}\right)$. Therefore,

$$
\operatorname{Pr}\left[f_{A, b}(x)=u \wedge f_{A, b}(y)=v\right]=\operatorname{Pr}\left[\wedge_{i=1}^{n}\left(\left\langle\tilde{a}_{i}, x\right\rangle+b_{i}=u_{i} \wedge\left\langle\tilde{a}_{i}, y\right\rangle+b_{i}=v_{i}\right)\right]
$$

As $A, b$ are chosen independently and uniformly at random, we get that

$$
\left.\operatorname{Pr}\left[\wedge_{i=1}^{n}\left(\left\langle\tilde{a}_{i}, x\right\rangle+b_{i}=u_{i} \wedge\left\langle\tilde{a}_{i}, y\right\rangle+b_{i}=v_{i}\right)\right)\right]=\prod_{i=1}^{n} \operatorname{Pr}_{\tilde{a}_{i}, b_{i}}\left[\left\langle\tilde{a}_{i}, x\right\rangle+b_{i}=u_{i} \wedge\left\langle\tilde{a}_{i}, y\right\rangle+b_{i}=v_{i}\right]
$$

Suppose we are able to prove that for every $i, \operatorname{Pr}_{\tilde{a}_{i} \in\{0,1\}^{n}, b_{i}\{0,1\}}\left[\left\langle\tilde{a}_{i}, x\right\rangle+b_{i}=u_{i}\right]=\frac{1}{4}$ then we will be done. To prove that consider the following:
$\left[\begin{array}{c}x_{1}, x_{2}, \ldots, x_{n}, \\ y_{1}, y_{2}, \ldots, y_{n}, \\ y_{2}\end{array}\right]\left[\tilde{a}_{i 1}, \tilde{a}_{i 2}, \ldots, \tilde{a}_{i n}, b_{i}\right]^{T}=\left[\begin{array}{c}u_{i} \\ v_{i}\end{array}\right]$
As $x \neq y$, there exists $j \in[n]$ such that $x_{j} \neq y_{j}$. Therefore the matrix $\left[\begin{array}{ll}x_{1}, x_{2}, \ldots, x_{n}, & 1 \\ y_{1}, y_{2}, \ldots, y_{n}, & 1\end{array}\right]$ has full row rank.

Therefore, using Exercise 7, we get $\operatorname{Pr}_{\tilde{a}_{i} \in\{0,1\}^{n}, b_{i} \in\{0,1\}}\left[\left\langle\tilde{a}_{i}, x\right\rangle+b_{i}=u_{i}\right]=\frac{1}{4}$.

Note that $|\mathcal{F}|$ is $2^{m n}+2^{n}$. Therefore, the number of bits required for pick a random function from $\mathcal{F}$ is $O(m n)$.

Here is one useful property of pairwise independent 0-1 random variables.
Exercise 9. Let $X_{1}, X_{2}, \ldots, X_{n}$ be pairwise independent 0-1 random variables. Let $X=$ $\sum_{i=1}^{n} X_{i}$. Then $\operatorname{Var}(X) \leq E(X)$.
Proof.

$$
\begin{array}{rlr}
\mathbb{E}\left(X^{2}\right) & =\mathbb{E}\left(\left(\sum_{i=1}^{n} X_{i}\right)^{2}\right) \\
& =\mathbb{E}\left(\sum_{i=1}^{n} X_{i}^{2}+\sum_{i \neq j} X_{i} X_{j}\right) & \\
& =\mathbb{E}\left(\sum_{i=1}^{n} X_{i}+\sum_{i \neq j} X_{i} X_{j}\right) & \text { (As } X_{i} \text { s are 0-1 valued) } \\
& =\mathbb{E}(X)+\sum_{i \neq j} \mathbb{E}\left(X_{i} X_{j}\right) & \text { (By linearity of expectation) } \\
& =\mathbb{E}(X)+\sum_{i \neq j} 1 \cdot \operatorname{Pr}\left[X_{i}=1 \wedge X_{j}=1\right] & \text { (By the definition of expectation) } \\
& =\mathbb{E}(X)+\sum_{i \neq j} \operatorname{Pr}\left[X_{i}=1\right] \operatorname{Pr}\left[X_{j}=1\right] & \text { (By pairwise independence of } \left.X_{i} \mathrm{~s}\right) \\
& =\mathbb{E}(X)+\sum_{i \neq j} \mathbb{E}\left(X_{i}\right) \mathbb{E}\left(X_{j}\right) & \text { (By the definition of expectation) }
\end{array}
$$

Similarly, we can evaluate $(\mathbb{E}(X))^{2}$ as follows:

$$
\begin{array}{rlr}
(\mathbb{E}(X))^{2} & =\left(\mathbb{E}\left(\sum_{i=1}^{n} X_{i}\right)\right)^{2} & \\
& =\left(\sum_{i=1}^{n} \mathbb{E}\left(X_{i}\right)\right)^{2} & \text { (By linearity of expectation) } \\
& =\sum_{i=1}^{n} \mathbb{E}\left(X_{i}\right)^{2}+\sum_{i \neq j} \mathbb{E}\left(X_{i}\right) \mathbb{E}\left(X_{j}\right) &
\end{array}
$$

Therefore, $\operatorname{Var}(X)=\mathbb{E}\left(X^{2}\right)-(\mathbb{E}(X))^{2}=\mathbb{E}(X)-\sum_{i=1}^{n} \mathbb{E}\left(X_{i}\right)^{2} \leq \mathbb{E}(X)$.

