

# Advanced Proof-Systems ; Problem Set 1

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## 1 Chernoff Bound and Sequential Repetition

### 1.1 Markov's Inequality

$$\begin{aligned}\mu &= \sum_r \Pr[x = r] \cdot r = \sum_{r \leq \alpha\mu} \Pr[x = r] \cdot r + \sum_{r > \alpha\mu} \Pr[x = r] \cdot r \\ &\geq \sum_{r > \alpha\mu} \Pr[x = r] \cdot r > \sum_{r > \alpha\mu} \Pr[x = r] \cdot \alpha\mu = \alpha\mu \sum_{r > \alpha\mu} \Pr[x = r] = \alpha\mu \cdot \Pr[x > \alpha\mu] \\ &\Rightarrow 1 > \alpha \Pr[x > \alpha\mu] \Rightarrow \Pr[x > \alpha\mu] < \frac{1}{\alpha}\end{aligned}$$

### 1.2 Chernoff Bound

2

3

## 4 Matrix Multiplication

### 4.1 Naive Algorithm

We can deduce an algorithm from the definition of matrix multiplication. Since  $\forall i, j : (A \cdot B)_{i,j} = \langle A_{i,:}, B_{:,j} \rangle = \sum_{k=1}^n A_{i,k} \cdot B_{k,j}$ , we can write the following algorithm:

```
1   C = zeros(n,n)
2   for i=1 to n:
3     for j=1 to n:
4       for k=1 to n:
5         C[i,j] += A[i,k] * B[k,j]
```

## 4.2 Random Subset Sum Principle

Lemma:

For any  $x \in \mathbb{F}^n \neq 0$  and uniformly sampled  $y \in \mathbb{F}^n$ ,  $\langle x, y \rangle$  is uniformly distributed over  $\mathbb{F}^n$ .

Proof of lemma: Since  $x \neq 0$ , it must have at least one non-zero entry - denote it with  $x_i$ . We can see that:

$$\langle x, y \rangle = \sum_{j=1}^n x_j y_j = x_i y_i + \sum_{j \neq i} x_j y_j := x_i y_i + \alpha \quad (*)$$

Note how  $\alpha$  is just a constant (with respect to  $i$ ) value in  $\mathbb{F}^n$ .

We have also seen in the lecture that for any non-zero  $a \in \mathbb{F}^n$ , and uniformly sampled  $b \in \mathbb{F}^n$  -  $a \cdot b$  is distributed uniformly - meaning so is  $x_i \cdot y_i$  in (\*).

Moreover, for any non-zero  $a \in \mathbb{F}^n$ , and uniformly sampled  $b \in \mathbb{F}^n$  -  $a + b$  is also uniformly distributed.

Meaning that in (\*),  $x_i \cdot y_i + \alpha$  is uniformly distributed, hence - so is  $\langle x, y \rangle$ .

Proof of claim:

Let  $x \in \mathbb{F}^n \setminus \{0\}$ . From the lemma, we know that  $\langle x, y \rangle$  is uniformly distributed over  $\mathbb{F}^n$ , thus the probability that  $\langle x, y \rangle = 0$  is  $\frac{|\{0\}|}{|\mathbb{F}^n|} = \frac{1}{|\mathbb{F}^n|}$ .

## 4.3 Randomized Verification of Matrix Multiplication

Algorithm  $P$ :

```

1  On input A,B,C
2  sample x uniformly from F^n
3  d := B*x
4  e := A*d
5  c := C*x
6  return e=c

```

Now it is easy to see how the complexity of this algorithm is  $O(n^2)$ , since in lines 3,4,5 we have operations of  $O(n^2)$  complexity, and the rest of the operations seen can be done in linear complexity.

As for the correctness:

- Completeness - If  $A \cdot B = C$ , then  $\Pr[P(A, B, C) = 1] = 1$ :  
When indeed  $A \cdot B = C$ , then multiplying both sides by  $x$  will yield the same value, meaning  $e = c$  and  $P(A, B, C) = 1$ .

- Soundness - If  $A \cdot B \neq C$ , then  $\Pr[P(A, B, C) = 1] \leq \frac{1}{2}$ :  
The conditions for the algorithm to accept are:

$$(AB)x = Cx \Leftrightarrow \forall i \in [n] : (AB)_{i,:} x_i = C_{i,:} x_i \Leftrightarrow \forall i \in [n] : ((AB)_{i,:} - C_{i,:}) x_i = 0$$

Now since  $A \cdot B \neq C$ , there must be at least one  $k$  such that:

$$(AB)_{k,:} - C_{k,:} \neq 0 \Rightarrow \Pr[(AB)_{k,:} - C_{k,:} \cdot x_k] \leq \frac{1}{|\mathbb{F}^n|}$$

So we can see:

$$\Pr[\forall i \in [n] : ((AB)_{i,:} - C_{i,:})x_i = 0] \leq \Pr[((AB)_{i,:} - C_{i,:})x_i] \leq \frac{1}{|\mathbb{F}^n|}$$

It seems fair to assume  $|\mathbb{F}^n| \geq 2$ , but just in case, we can handle  $\mathbb{F} = \{e\}$  by checking for this case specifically.