

## 1 Ex 1.

**Claim 1.** Let  $A$  be a random matrix in  $M(\mathbb{F}_2^{k \times n})$  then for any non zero  $x \in \mathbb{F}$  we have that  $Ax$  distributed uniformly.

*Proof.* By the fact that  $x \neq 0$  there exists at least one coordinate  $i \in [k]$  such that  $x_i \neq 0$ . Thus we have

$$\begin{aligned}(Ax)_j &= \sum_k A_{jk}x_k = \sum_{i \neq k} A_{jk}x_k + A_{ji}x_i \\ &= \sum_{i \neq k} A_{jk}x_k + A_{ji}\end{aligned}$$

Notice that due to the fact that  $\mathbb{F}_2$  is a field, there is exactly one assignment that satisfies the equation conditioned on all the values  $A_{jk}$  where  $j \neq k$ .

$$\begin{aligned}\Pr[(Ax)_j = 1] &= \sum_{A_{jk}; k \neq i} \Pr[(Ax)_j = 1 | A_{jk}; k \neq i] \Pr[A_{jk}; k \neq i] \\ &= \frac{1}{2}\end{aligned}$$

therefore any coordinate of  $Ax$  distributed uniformly  $\Rightarrow Ax$  distributed uniformly.  $\square$

By the uniformity of  $Ax$  we obtain that the expected Hamming wight of  $Ax$  is :

$$\mathbf{E}[|Ax|] = \mathbf{E}\left[\sum_i^n (Ax)_i\right] = \frac{1}{2}n$$

As the coordinates of  $A_x$  are independent (each row of  $A$  is sampled separately) we can use the Hoff' bound to conclude that:

$$\Pr\left[||Ax| - \mathbf{E}[|Ax|]| \geq \left(\frac{1}{2} - \delta\right)n\right] \leq e^{-n(\frac{1}{2}-\delta)^2}$$

Now we will use the union bound to show that any  $x \in \mathbb{F}_2^k$ ,  $Ax$  is at weight at least  $\delta$ .

$$\Pr[|Ax| \geq \delta : \forall x \in \mathbb{F}_2^k] \geq 1 - |\mathbb{F}_2^k| \cdot e^{-n(\frac{1}{2}-\delta)^2}$$

Denote  $k = \rho n$  and notice that the above holds when  $\rho \geq \left(\frac{1}{2} - \delta\right)^2$

## 2 Ex 2.

**Claim 2.** Let  $v_1, v_2, \dots, v_m$  unit vectors in an inner-product space such that  $\langle v_i, v_j \rangle \leq -2\varepsilon$  for all  $i \neq j$ , then  $m \leq \frac{1}{2\varepsilon} + 1$ .

*Proof.* Let's us bound from both sides the norm of the summation  $|\sum_i v_i|$ . As the norm is by definition (construction) non-negative we are going to bound from the left by 0, on the other hand we have that:

$$0 \leq \left|\sum_{v_i} v_i\right| = m + 2 \sum_{i,j} \langle v_i, v_j \rangle \leq m - 2 \cdot \frac{m(m-1)}{2} \cdot 2\varepsilon$$

Thus we obtain  $m(2(m-1)\varepsilon - 1) \leq 0$  namely,  $m \leq \frac{1}{2\varepsilon} + 1$   $\square$

Now, define the following product for  $u, v \in \mathbb{F}_2^n$ ,  $\langle v, u \rangle = \sum_i (-1)^{v_i} (-1)^{u_i}$  observes that:

1.  $\langle v, v \rangle = \sum_i 1 = n \geq 0$ .
2.  $\langle v, u \rangle = \langle u, v \rangle$ .
3.  $\langle ax + by, z \rangle = (-1)^a \langle x, z \rangle + (-1)^b \langle y, z \rangle$ .

Now the  $v$ 's corresponds to code with distance at least  $d$  then, i.e for any codewords  $v$  and  $u$  disagree on at least  $d$  coordinates, and therefore  $\langle v, u \rangle \leq \text{agree} - \text{disagree} = n - 2 \text{ disagree} = n - 2d$ . Now consider the normal codewords  $\tilde{v}_1.. \tilde{v}_n$  and assume that

$$\langle \tilde{v}_i, \tilde{v}_j \rangle = (1 - 2\delta) = \frac{1}{n} (n - 2d(v_i, v_j)) \leq \varepsilon$$

So if  $d \geq \frac{1}{2} + \varepsilon$  we obtain the condition of the above claim.

### 3 Ex 3.

Consider the following process for decoding  $a$ ,

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1 for  $t \in [\tau]$  do
2   for  $i \in [n]$  do
3      $x \sim_u \mathbb{F}_2^n$ 
4      $a_i^{(t)} \leftarrow w(x) + w(\sigma_i(x))$ 
5   end
6 end
7 for  $i \in [n]$  do
8    $\hat{a}_i \leftarrow [\frac{1}{\tau} \sum_t a_i^{(t)}]$ 
9 end
10 return  $\hat{a}_0, \hat{a}_1, \hat{a}_2.. \hat{a}_n$ 

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**Claim 3.** For  $\tau = \Omega(\frac{1}{\varepsilon^4} \log(n))$  The above decoding success to decode  $w(x)$  with probability  $\geq 1 - \frac{1}{n}$ .

*Proof.* In this question we will say that  $w$  agree on  $x, \sigma_i(x)$  if both  $x, \sigma_i(x)$  were either flipped or unflipped. Clearly if  $w(x)$  agree with  $w(\sigma_i(x))$  then

$$\begin{aligned}
w(x) + w(\sigma_i(x)) &= H_a(x) + H_a(\sigma_i(x)) \\
&= \sum_{i \neq j} a_j(x_j + x_j) + a_i(x_j + 1 + x_j) = a_j \quad (\text{neither of them were flipped.}) \\
&= 1 + H_a(x) + 1 + H_a(\sigma_i(x)) = a_i \quad (\text{both flipped.})
\end{aligned}$$

Thus we can bound the probability that  $w(x) + w(\sigma_i(x)) \neq a_i$  by the probability that  $w$  disagree on  $x$  and that append at probability:

$$\xi := (1 - f(x)) f(\sigma_i(x)) + (1 - f(\sigma_i(x))) f(x)$$

Now as we want bound  $\xi$  we could think about the maximization problem under the restrictions that  $f(x), f(\sigma_i(x)) \leq \frac{1}{2} - \varepsilon$ . We know that the maximum lay on the boundary so we can assign  $\frac{1}{2} - \varepsilon$  for each of the probabilities to obtain an upper bound. That will yield  $\xi \leq 2 \cdot (\frac{1}{2} - \varepsilon) (\frac{1}{2} + \varepsilon)$ , namely  $\xi \leq \frac{1}{2} - 2\varepsilon^2$ . Now the probability that a coordinate  $i$  will rounded to the opposite side, that it  $\hat{a}_i \neq a_i$  mean that arithmetic mean over  $\tau$  experiments were  $2\varepsilon^2$  far from the expectation. Which by Hoff' bound is bounded by:  $e^{\tau 4\varepsilon^4}$ . So using the union bound we obtain:

$$\Pr[\text{decoding success}] \geq 1 - n \cdot e^{\tau 4\varepsilon^4}$$

Therefore it's enough to take  $\tau = O(\frac{1}{\varepsilon^4} \log(n))$  to obtain a decoder which run at time  $O(\frac{1}{\varepsilon^4} n \log(n))$  and success with heigh probability.  $\square$

4 Ex 4.