## PCP - Huji Course, Ex 1.

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## 1 Ex 1.

Let A be a random matrix in  $M(\mathbb{F}_2^{k \times n})$ . For any non-zero  $x \in \mathbb{F}$ , we have that Ax is distributed uniformly.

#### Claim 1.

*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

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

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  $j \neq k$ .

$$\mathbf{Pr}\left[(Ax)_{j} = 1\right] = \sum_{A_{jk}; k \neq i} \mathbf{Pr}\left[(Ax)_{j} = 1 \mid A_{jk}; k \neq i\right] \mathbf{Pr}\left[A_{jk}; k \neq i\right]$$
$$= \frac{1}{2}$$

Therefore, any coordinate of Ax is distributed uniformly  $\Rightarrow Ax$  is distributed uniformly.

By the uniformity of Ax, we obtain that the expected Hamming weight of Ax is:

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

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

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

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

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

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

## 2 Ex 2.

Claim 2. Let  $v_1, v_2, \ldots, v_m$  be 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 us bound the norm of the summation  $|\sum_i v_i|$  from both sides. As the norm is nonnegative by definition, we will bound it from the left by 0. On the other hand, we have that:

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

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

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

- 1.  $\langle v, v \rangle = \sum_{i} 1 = n \ge 0$ .
- 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, if the v's correspond to a code with distance at least d, then, for any codewords v and u that disagree on at least d coordinates, we have that  $\langle v,u\rangle \leq \text{agree-disagree} = n-2$  disagree = n-2d. Now consider the normal codewords  $\tilde{v_1}..\tilde{v_n}$  and assume that

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

Therefore, 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,

**Claim 3.** For  $\tau = \Omega\left(\frac{1}{\varepsilon^4}\log\left(n\right)\right)$  The above decoding success to decode  $w\left(x\right)$  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 filliped or unflipped. Clearly if w(x) agree with  $w(\sigma_i(x))$  than

$$\begin{split} w\left(x\right) + w\left(\sigma_{i}(x)\right) &= H_{a}(x) + H_{a}(\sigma_{i}(x)) \\ &= \sum_{i \neq j} a_{j}(x_{j} + x_{j}) + a_{i}(x_{j} + 1 + x_{j}) = a_{j} \quad \text{( neither of them were flipped.)} \\ &= 1 + H_{a}(x) + 1 + H_{a}(\sigma_{i}(x)) = a_{i} \quad \text{( both flipped.)} \end{split}$$

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  $\xi \leq 2 \cdot \left(\frac{1}{2} - \varepsilon\right) \left(\frac{1}{2} + \varepsilon\right)$ , 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:

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

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.

#### 4 Ex 3.

Consider the following process for decoding a:

Claim 4. For  $\tau = \Omega\left(\frac{1}{\varepsilon^4}\log\left(n\right)\right)$  the above decoding succeeds in decoding  $w\left(x\right)$  with probability  $\geq 1 - \frac{1}{n}$ .

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

$$w(x) + w(\sigma_i(x)) = H_a(x) + H_a(\sigma_i(x))$$

$$= \sum_{i \neq j} a_j(x_j + x_j) + a_i(x_j + 1 + x_j) = a_j \quad (\text{ neither of them were flipped.})$$

$$= 1 + H_a(x) + 1 + H_a(\sigma_i(x)) = a_i \quad (\text{ both flipped.})$$

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

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

Now, to bound  $\xi$ , we can think about the maximization problem under the restrictions that  $f(x), f(\sigma_i(x) \le \frac{1}{2} - \varepsilon)$ . We know that the maximum lies on the boundary, so we can assign  $\frac{1}{2} - \varepsilon$  for each of the probabilities to obtain an upper bound. That will yield  $\xi \le 2 \cdot \left(\frac{1}{2} - \varepsilon\right) \left(\frac{1}{2} + \varepsilon\right)$ , namely  $\xi \le \frac{1}{2} - 2\varepsilon^2$ . Now, the probability that a coordinate i will be rounded to the opposite side, i.e.  $\hat{a}_i \ne a_i$ , means that the arithmetic mean over  $\tau$  experiments is  $2\varepsilon^2$  far from the expectation. According to Hoff's bound, this is bounded by  $e^{\tau^4\varepsilon^4}$ . Thus, using the union bound, we obtain:

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

Therefore, it is enough to take  $\tau = O(\frac{1}{\varepsilon^4} \log(n))$  to obtain a decoder that runs in time  $O(\frac{1}{\varepsilon^4} n \log(n))$  and succeeds with high probability.

#### 5 Ex 4.

#### 5.1 (a)

We will prove that if for any x, f interpolate well on x, x+1, ..., x+d+1 than any it interpolate well on every coordinates set at size d+1. Denote by  $J \subset \mathbb{F}_q$  at size d+1. Let's continue by induction on  $\max J$ . The base case  $\max J = d+1 \Rightarrow J = \{1, 2..., d+1\}$  follow straightforwardly from the assumption. Assume the correctness for any J such that  $\max J \leq x_0$  and consider J' such that  $\max J' = x_0 + .$  Now it given that  $S = \{x_0 - d, x_0 - d+1, ...x_0 + 1\}$  is well interpolating set, so there exists coefficients  $a_1, a_{d+1}$  such that  $a_{d+1}f(x_0+1) = \sum_{x_i \in S/(x_0+1)} a_i f(x_i)$  On the overhand, for ant any  $x_i \in S/(x_0+1)$  the union  $K = x_i \cup J/(x_0+1)$  is subset of  $\mathbb{F}_q$  at size d+1 such that  $\max K \leq x_0$ . Hence by induction assumption K is well interpolating set and we can exchange any  $f(x_i)$  for  $x_i \in S$  by a linear combination of  $f(x_i)$  for  $x_i \in J/(x_0+1)$ . So in overall we obtain that J is depended set, namely f is well interpolate on J.

### 5.2 (b)

Define the function  $g(x) = f(t^{-1}(x-s))$ . Note that q is prime, thus  $(\mathbb{F}_q/0,\cdot)$  and  $(\mathbb{F}_q,+)$  are groups and the inverse elements  $-s,t^{-1}$  are exist and uniqs. Suppose that y is a zero of  $g \Rightarrow f(t(y+s)) = g(y) = 0$ , Hence the number of zeros of f equals to the number of zeros of g, which means that their degree are equal  $\Rightarrow g$  is also a polynomial at degree at most  $d, \Rightarrow a_1, a_2..a_d$  are also the interpolation coefficients respecting to the interpolation set  $\{tx_1 + s, tx_2 + s, ..., tx_d + s\}$ .

## 6 Ex (5).

#### 6.1 (a)

As shown in the previous section, by the fact that q is prime, we have that  $g_{u,v}$  acts on  $\mathbb{F}_q^m$  by  $g_{u,v}(x) = u + vx$ , (for any  $v \neq 0$ ). Thus,  $f(g_{u,v}(x))$  is just a permutation over the values of f. As the number of zeros remains the same, we have that  $f(g_{u,v}(x))$  is also a degree d polynomial. Therefore, the restricted polynomial  $f|_L$  corresponds to the restriction L' of another polynomial obtained by taking u' = 0 and v to be supported only on a single coordinate. Hence, the restricted polynomial can have at most d zeros.

#### 6.2 (b)

As we have that f is a polynomial of degree exactly d, there must be a monomial  $x_1^{d_1}x_2^{d_2}...x_k^{d_k}$  such that  $\sum_i d_i = d$ . Denote by g the sum of all those monomials, and by  $v \in \mathbb{F}_q^n$  a coordinate on which  $g(v) \neq 0$  (if there is no such v, then we could write f as a sum of monomials, each of degree at most d-1).

Now, as g(v),  $t \neq 0$ , we obtain that g(vt) is equal exactly to  $t^d \cdot c$  where c is the sum of coefficients of each monomial of g (also c = g(1)). As f(x) - g(x) is a polynomial of degree d - 1, it holds from the previous section that (f - g)(vt) is also a polynomial of degree at most d - 1. Thus, it cannot zero out  $g(vt) \Rightarrow f(vt) = (f - g)(vt) + g(vt)$  is also a polynomial of degree d.

# 7 Ex (6).

#### 7.1 (a) and (b).

Let F be a function from  $\{0, 1..d\}^2 \to \mathbb{F}_q$ , we are going to define a d-degree polynomial  $f : \mathbb{F}_q^2 \to \mathbb{F}_q$  that agree with F and show that is uniq. Notice that any polynomial could be written as  $\sum_{i,j} a_i a_j x^i y^j$ . Thus, the assignments of  $(d+1)^2$  points define  $(d+1)^2$  equations over  $(d+1)^2$  variables. In addition, as the determinant of the matrix equals

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0^{2} \\ 1 & 1 & 1 \cdot 0 & 0 & 1 \cdot 1 \\ 1 & 1 & 1 & 1 \cdot 1 & 1^{2}1 & d & d & d^{2} & d^{2} \end{bmatrix} \cdot \begin{bmatrix} a_{00} \\ a_{01} \\ a_{10} \\ a_{11} \\ a_{20} \\ a_{dd} \end{bmatrix} = \begin{bmatrix} F(0,0) \\ F(0,1) \\ F(1,0) \\ F(1,1) \\ F(2,0) \\ F(d,d) \end{bmatrix}$$

Figure 1: Illustration of the equations system. The left system is a vadermonde matrix in which the ((x,y),(i,j)) entry corresponds to  $x^iy^j$  where (x,y) are one of the points in  $\mathbb{F}_q^2$ .

$$\begin{split} \sum_{\sigma \in S_n} (-1)^{\sigma(\pi)} \prod_{i,j} (x^{\sigma(i)_1} y^{\sigma(j)_2}) &= \prod_{i < j} (x_i - x_j) \prod_{i < j} (y_i - y_j) \\ &= \prod_{i < j} (x_i - x_j) \prod_{i < j} (y_i - y_j) &= \prod i^{d-i} \prod j^{d-j} \not | q \end{split}$$

So the detriment is not zero, thus we can solve that system by gauss elimination and obtain unique solution. The solution is uniq and define the coefficients of f.

## 7.2 (c).

Now consider a function  $f: \mathbb{F}_q^2 \to \mathbb{F}_q$  which any restriction of f to a line is a polynomial at degree at most d