

## Homework 6 Solution

**Problem 1.** *Prove the following:*

- (a) *Let  $X = (X_1, \dots, X_n)$  and  $Y = (Y_1, \dots, Y_n)$  be independent symmetric Bernoulli random vectors in  $\mathbb{R}^n$ , i.e. all coordinates  $X_i$  and  $Y_i$  are independent random variables that take values  $\pm 1$  with probability  $1/2$ . Show that:*

$$\mathbb{P}\{|\langle X, Y \rangle| \geq 0.001n\} \leq 2\exp(-cn)$$

- (b) *Deduce that the angle  $\angle(X, Y)$  between the vectors  $X$  and  $Y$  satisfies*

$$\mathbb{P}\left\{\left|\angle(X, Y) - \frac{\pi}{2}\right| \geq 0.01\pi\right\} \leq 2\exp(-c_1n)$$

- (c) *Prove that for every dimension  $n > C$ , there exist  $N \geq \frac{1}{2}\exp(cn)$  vectors  $v_1, \dots, v_N$  in  $\mathbb{R}^n$  so that all pairwise angles between these vectors are between  $89^\circ$  and  $91^\circ$ .*

- (a) for every  $i$ ,  $X_i Y_i$  is a symmetric Bernoulli random variable, and the inner product can be written as a sum of independent symmetric Bernoulli random variables. Applying Hoeffding's inequality:

$$\mathbb{P}\left\{\left|\sum_{i=1}^n X_i Y_i\right| \geq 0.001n\right\} \leq 2\exp(-5 \cdot 10^{-7}n)$$

- (b) Denoting the angle with  $\theta$ ,  $\cos \theta = \frac{\langle X, Y \rangle}{n}$ . Further

$$\begin{aligned} \mathbb{P}\left\{\left|\theta - \frac{\pi}{2}\right| \geq 0.01\pi\right\} &= \mathbb{P}\{|\cos \theta| \geq \sin(0.01\pi)\} \\ &= \mathbb{P}\{|\langle X, Y \rangle| \geq n \sin(0.01\pi)\} \\ &\leq 2\exp(-c_1n) \quad (\text{Hoeffding's}) \end{aligned}$$

- (c) Denoting the angle between every pair of vectors as  $\theta_{ij}$ .

$$\mathbb{P}\left\{\left|\theta_{ij} - \frac{\pi}{2}\right| \geq \frac{\pi}{180}\right\} \leq 2\exp(-c'n)$$

Thus the probability of existing a pair with an angle not lying in the given interval is bounded above by  $2\exp(-c'n)N^2$  using union bound. We want at least  $\frac{\exp(cn)}{2}$  such vectors. Setting an upper bound of 0.01, we want

$$\exp(n(2c - c')) \leq 2 \cdot 10^{-2}$$

picking  $c < \frac{c'}{2}$ , the inequality holds for  $n > C$  for some  $C$  since the exponential function is not bounded.

□

**Problem 2.** *Prove the following:*

- (a) *Let  $z$  be a fixed unit vector in  $\mathbb{R}^d$ . Show that the  $X = Gz$  is a random vector in  $\mathbb{R}^n$  whose all coordinates  $X_j$  are independent random variables, which satisfy*

$$\mathbb{E}X_j = 0, \quad \text{Var}(X_j) = 1, \quad \|X_j\|_{\psi_2} \leq C_1$$

- (b) *Prove a thin-shell inequality for  $X = Gz$ :*

$$\mathbb{P}\{0.99\sqrt{n} \leq \|X\|_2 \leq 1.01\sqrt{n}\} \geq 1 - 2\exp(-cn)$$

- (c) *Let  $x_1, \dots, x_N$  be any set of fixed vectors in  $\mathbb{R}^d$ . Let  $G$  be an  $n \times d$  Bernoulli random matrix, and set  $T = \frac{G}{\sqrt{n}}$ . Prove that if  $n = C \log N$ , then the map  $T$  approximately preserves the pair-wise geometry of the data, namely that following event holds with positive probability:*

$$0.99\|x_i - x_j\|_2 \leq \|Tx_i - Tx_j\|_2 \leq 1.01\|x_i - x_j\|_2 \quad \text{for all } i, j \in [N]$$

- (a) Denoting the  $i$ th row of  $G$  with  $G_i$

$$\mathbb{E}X_j = \mathbb{E}G_j z = \sum_{i=1}^n \mathbb{E}G_{ji} z_i = 0$$

$$Var(X_j) = \mathbb{E}X_j^2 = \sum_{k,i=1}^n \mathbb{E}[G_{ji}G_{jk}]z_i z_k = \sum_{i=1}^n z_i^2 = 1 \quad (unit)$$

For the third one, we know that  $X_j$  is a sum of independent symmetric Bernoulli random variables with coefficients  $z_i$ , thus we can apply Hoeffding's inequality, which yields

$$\mathbb{P}\{|G_j z| \geq t\} \leq 2\exp\left(\frac{-t^2}{2\|z\|_2^2}\right) = 2\exp\left(\frac{-t^2}{2}\right)$$

which implies that  $\|X_j\|_{\psi_2} \leq \sqrt{2}$

(b)

$$\mathbb{E}\|X\|_2^2 = \sum_{i=1}^n \mathbb{E}X_i^2 = n$$

$$\mathbb{P}\{|\|X\|_2^2 - n| \geq 0.01n\} = \mathbb{P}\left\{\left|\sum_{i=1}^n (X_i^2 - 1)\right| \geq 0.01n\right\}$$

As shown before, each  $X_i$  is subgaussian, which makes  $X_i^2 - 1$  sub-exponential for each  $i$ .

Applying Bernstein's inequality

$$\mathbb{P}\left\{\left|\sum_{i=1}^n (X_i^2 - 1)\right| \geq 0.01n\right\} \leq 2\exp\left(-c \min\left(\frac{10^{-4}n^2}{\sigma^2}, \frac{0.01n}{k}\right)\right) \quad (1)$$

Where

$$\sigma^2 = \sum_{i=1}^n \|X_i - 1\|_{\psi_1}^2 \leq nC$$

$$k = \max_{i \in [n]} \|X_i - 1\|_{\psi_1}^2 \leq C$$

since all variables are sub-exponential. Substituting in (1), we have

$$\mathbb{P}\{|\|X\|_2^2 - n| \leq 0.01n\} \geq 1 - 2\exp(-c'n)$$

which implies that

$$0.99\sqrt{n} \leq \|X\|_2 \leq 1.01\sqrt{n}$$

with probability at least  $1 - 2\exp(-c'n)$

- (c) Fix a pair  $x_i$  and  $x_j$ . Normalizing  $\frac{x_i - x_j}{\|x_i - x_j\|}$ , we can apply the thin-shell we found in previous section, yielding

$$\begin{aligned} \mathbb{P} \left\{ 0.99\sqrt{n} \leq \frac{\|Gx_i - Gx_j\|_2}{\|x_i - x_j\|_2} \leq 1.01\sqrt{n} \right\} = \\ \mathbb{P} \{ 0.99\|x_i - x_j\|_2 \leq \|Tx_i - Tx_j\|_2 \leq 1.01\|x_i - x_j\|_2 \} \geq \\ 1 - 2\exp(-cn) \end{aligned}$$

Applying union bound on all pairs, the probability of existing a pair not lying in the interval is bounded above by  $2\exp(2 \log N - cn)$ . Setting  $n \geq \frac{4 \log N}{c}$  would give us a high confidence.

□

### Problem 3.

- (a) Check that the Gram matrix  $G$  of any system of vectors is positive semidefinite.
- (b) Conversely, prove that any  $n \times n$  positive semidefinite matrix  $G$  is a Gram matrix of some system of vectors  $v_1, \dots, v_n$  in  $\mathbb{R}^n$ .

- (a) Assume we have a set of vectors  $v_1, \dots, v_n$  whose gram matrix is  $G$ .

$$V = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$$

$$G = \begin{bmatrix} \langle v_1, v_1 \rangle & \dots & \langle v_1, v_n \rangle \\ \vdots & \ddots & \vdots \\ \langle v_n, v_1 \rangle & \dots & \langle v_n, v_n \rangle \end{bmatrix} = VV^\top$$

let  $x$  be a non-zero vector.

$$x^\top Gx = x^\top VV^\top x = (V^\top x)^\top V^\top x = \langle V^\top x, V^\top x \rangle = \|V^\top x\|^2 \geq 0$$

where  $\|\cdot\|$  denotes the norm induced by the inner product.

- (b) According to the real spectral theorem, we can pick an orthonormal basis  $v_1, \dots, v_n$  of  $\mathbb{R}^n$  and write

$$G = V\Sigma V^\top$$

where  $V$  is the column vector  $V = [v_1, \dots, v_n]$  and  $\Sigma$  is the diagonal matrix of eigenvalues. expanding the relation, we get  $G = \sum_{i=1}^n \lambda_i v_i v_i^\top$ . Setting  $A = \sum_{i=1}^n \sqrt{\lambda_i} v_i v_i^\top$ ,

$$A^2 = \sum_{i,j=1}^n \sqrt{\lambda_i \lambda_j} v_i \langle v_i, v_j \rangle v_j^\top = \sum_{i=1}^n \lambda_i v_i v_i^\top = G$$

Note that the square root is real-valued since all eigenvalues are non-negative. Further, it's easy to check that  $A^2 = AA^\top$ , which is the gram matrix of the rows of  $A$  as shown in part (a). So  $G$  is the gram matrix of the rows of  $A$ .

□

**Problem 4.** Let  $\theta$  be a random vector uniformly distributed on the unit circle  $S^1 = \{x \in \mathbb{R}^2 : x_1^2 + x_2^2 = 1\}$ . Prove that for any pair of vectors  $u, v \in S^1$ , we have

$$\mathbb{E} \text{sign}(\langle u, \theta \rangle) \text{sign}(\langle v, \theta \rangle) = \frac{2}{\pi} \arcsin(\langle u, v \rangle)$$

Setting  $\angle(u, v) = \alpha$ , we have  $\alpha = \cos^{-1}(\langle u, v \rangle)$

$$\begin{aligned} \mathbb{E} \text{sign}(\langle u, \theta \rangle) \text{sign}(\langle v, \theta \rangle) &= 1 - \frac{2\alpha}{\pi} \\ &= \frac{2}{\pi} \sin^{-1}(\langle u, v \rangle) \end{aligned}$$

□