Math 608D, Assignment 2

1. Let $\mathcal{X} \subset \mathbf{R}^n$ be a finite point cloud, i.e., $|\mathcal{X}| = N$. We wish we wish to find a nearly isometric embedding of \mathcal{X} into ℓ_1^m . To be precise, fix $0 < \epsilon < 1$. We wish to show there is a mapping $A : \mathbf{R}^n \to \mathbf{R}^m$ satisfying

$$(1-\epsilon) \|x-y\|_2 \le \|A(x-y)\|_1 \le (1+\epsilon) \|x-y\|_2$$
 for all $x, y \in \mathcal{X}$.

What mapping makes this embedding (with high probability)? How small can you choose m as a function of n, N, ϵ ?

2. Consider the *p*-norm, $p \ge 1$,

$$\|x\|_p := \left(\sum_i |x_i|^p\right)^{1/p}$$

Let $B_p^n := \{x \in \mathbb{R}^n : \|x\|_p \leq 1\}$ be the unit ball for this norm. Let $N(K, \|\cdot\|_p, \epsilon)$ be the covering number of the a set $K \subset \mathbb{R}^n$ using the *p*-norm, that is, the minimal cardinality of a subset $\mathcal{X} \subset K$ satisfying: for any $x \in K$ there is $y \in \mathcal{X}$ such that $\|x - y\|_p \leq \epsilon$. Bound the following:

- (a) $N(B_1^n, \|\cdot\|_1, \epsilon)$.
- (b) $N(B_p^n, \|\cdot\|_p, \epsilon).$
- (c) $N(K, \|\cdot\|_{\infty}, \epsilon)$, where $K \subset B_{\infty}^n$ is a polytope with N vertices. (Can you beat the bound in the previous problem when N is not too large?)
- (d) **Tensors.** Consider $\mathbf{R}^{n \times n \times n}$, i.e., arrays with 3 modes. (Picture a rubix cube.) We call elements of this set tensors. A rank-1 tensor takes the form

$$T = u \times v \times w$$
, where $u, v, w \in \mathbf{R}^n$.

In other words,

$$T_{i,j,k} = u_i \cdot v_j \cdot w_k$$

We say that a tensor has rank at most r if it may be written as the sum of r rank-1 tensors. The Frobenius norm of a tensor is $||T||_F = \sqrt{\sum_{i,j,k} T_{i,j,k}^2}$. Let K be the set of rank-r tensors with Frobenius norm at most 1. Bound $N(K, ||\cdot||_F, \epsilon)$. (I believe such a bound was made in the literature in the past few years, but I haven't read the details.)

- (e) The packing number P(K, ||·||, ϵ) of a set K is the maximal cadinality of a subset X ∈ K satisfying ||x - y|| > ϵ for all x, y ∈ X with x ≠ y. Lower bound P(K, ||·||_F, ϵ) with K as in the last problem. (I'm not sure whether this has been done.)
- 3. Single index model/1-layer neural net. A 1-layer neural net is a function $f : \mathbb{R}^n \to \mathbb{R}^m$ of the following form. Let $W \in \mathbb{R}^{m \times n}$ be a weight matrix and let $\sigma : \mathbb{R} \to \mathbb{R}$ be some activation function. Then

 $f(x) = \sigma(Wx)$ where σ acts elementwise.

In other words, the *i*th entry of f(x) is

$$(f(x))_i = \sigma(\langle W_i, x \rangle)$$
 where W_i is the *i*th row of W .

Note: The above model is also called the *single-index model* of statistics.

In this problem we explore the behaviour of f under the assumption that the weights are Gaussian, i.e., assume $W_{i,j} \sim N(0,1)$ and all entries are independent. We will check whether f preserves Euclidean norms.

(a) Suppose that σ is a rectified linear unit i.e.,

$$\sigma(x) = \max(x, 0).$$

- i. Fix $x \in \mathbb{R}^n$. Let $T_x = \|f(x)\|_2$. What is $\mathbb{E} T_x^2$?
- ii. Bound

$$\|T_x-\boldsymbol{E}\,T_x\|_{\Psi_2}\,.$$

iii. Let $K\subset S^{n-1}$ be a finite set, i.e., $|K|\leq\infty.$ Bound

$$\boldsymbol{E}\sup_{x\in K}\left|T_{x}-\boldsymbol{E}\,T_{x}\right|.$$

iv. We now pass to the infinite case. Bound

$$\boldsymbol{E}\sup_{x\in\boldsymbol{S}^{n-1}}\left|T_{x}-\boldsymbol{E}\,T_{x}\right|.$$

v. Now the more challenging infinite case. Let $K \subset S^{n-1}$ be an arbitrary infinite set. Bound

$$\boldsymbol{E}\sup_{x\in K}\left|T_{x}-\boldsymbol{E}\,T_{x}\right|.$$

(Your answer should depend on how "large" K is.)

(b) Let $\sigma : \mathbf{R} \to \mathbf{R}$ be a general function. Complete Items i-v above in this setting. Your answers should depend on properties of σ . (You may assume what you need.)