

On the Approximation of Functional Classes Equipped with a Uniform Measure Using Ridge Functions*

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We introduce a construction of a uniform measure over a functional class \mathcal{B}^r which is similar to a Besov class with smoothness index r . We then consider the problem of approximating \mathcal{B}^r using a manifold M_n which consists of all linear manifolds spanned by n ridge functions, i.e., $M_n = \{\sum_{i=1}^n g_i(a_i \cdot x) : a_i \in S^{d-1}, g_i \in L_2([-1, 1])\}$, $x \in B^d$. It is proved that for some subset $A \subset \mathcal{B}^r$ of probabilistic measure $1 - \delta$, for all $f \in A$ the degree of approximation of M_n behaves asymptotically as $1/n^{r/(d-1)}$. As a direct consequence the probabilistic (n, δ) -width for nonlinear approximation denoted as $d_{n, \delta}(\mathcal{B}^r, \mu, M_n)$, where μ is a uniform measure over \mathcal{B}^r , is similarly bounded. The lower bound holds also for the specific case of approximation using a manifold of one hidden layer neural networks with n hidden units. © 1999 Academic Press

1. INTRODUCTION

We consider the problem of approximating a functional class \mathcal{B}^r similar to a Besov class using a manifold of ridge functions $M_n = \{\sum_{i=1}^n g_i(a_i \cdot x) : a_i \in S^{d-1}, g_i \in L_2([-1, 1])\}$, defined on the unit ball $B^d = \{x \in \mathbb{R}^d : \|x\|_2 := (\sum_{i=1}^d x_i^2)^{1/2} \leq 1\}$ in the space \mathbb{R}^d . Here $S^{d-1} = \{x \in B^d : \|x\|_2 = 1\}$ is the

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unit sphere in \mathbb{R}^d . The degree of approximation of $f \in \mathcal{B}^r$ by M_n in the space L_2 is defined by the expression

$$\text{dist}(f, M_n, L_2) = \inf_{h \in M_n} \|f - h\|_{L_2},$$

where $\|f\|_{L_2}$ denotes the L_2 -norm of f on B^d .

Vostrecov and Kreines [27] and Lin and Pinkus [8, 9] studied issues of fundamentality of ridge functions in functional spaces. The specific case in which the ridge functions are sigmoidal, e.g., $g_i(y) = \sigma(y) = 1/(1 + e^{-y})$, $1 \leq i \leq n$, and translations are permitted corresponds to a manifold $\mathcal{H}_n = \{\sum_{i=1}^n c_i \sigma(a_i \cdot x + b_i) : a_i \in \mathbb{R}^d, b_i, c_i \in \mathbb{R}\}$, of one hidden layer neural networks with n hidden units. There have been many investigations concerning the approximation properties of \mathcal{H}_n , e.g., Barron [1], Mhaskar [15], Girosi *et al.* [4–6], DeVore *et al.* [2], Petrushev [19], and Maiorov and Meir [13].

Recently a series of results was obtained for estimates of approximation of functions by the ridge-manifold M_n in the two-dimensional case, $d=2$ (see Oskolkov [18], and Temlyakov [23]). In particular, Oskolkov showed that for $d=2$ the orders of approximation of radial functions by the ridge-manifold M_n and by the space of algebraic polynomials of degree n coincide. In Maiorov [12], the asymptotic behavior of the distance

$$\text{dist}(W_2^{r,d}, M_n, L_2) \asymp n^{-r/(d-1)}$$

for the Sobolev class $W_2^{r,d}$, $d \geq 2$, was obtained.

In this work we are interested in assessing how massive is the subset of functions in \mathcal{B}^r such that for *all* functions in this subset a certain degree of approximation holds. In order to formalize the statement that a high percentage of the functions in \mathcal{B}^r are approximated by M_n to a certain degree we construct a uniform measure over \mathcal{B}^r . The volume of a subset A of the unit ball in \mathbb{R}^n which is equipped with a uniform probability measure is proportional to the probability of A . Similarly if \mathcal{B}^r is equipped with the uniform measure then a subset in \mathcal{B}^r of high probability is interpreted as being massive in the sense of occupying almost all of \mathcal{B}^r .

To proceed we first construct a uniform measure μ over \mathcal{B}^r . We then calculate lower and upper bounds on the degree of approximation by M_n which holds for all functions in some subset $A \subset \mathcal{B}^r$ of probability $1 - \delta$. Specifically we obtain a degree of approximation such that for some $A \subset \mathcal{B}^r$, with $\mu(A) \geq 1 - e^{-\alpha(n)}$ and $\alpha(n) = c_0 n^{d/(d-1)}$, then for all $f \in A$, $c_1/n^{r/(d-1)} \leq \text{dist}(f, M_n, L_2) \leq c_2/n^{r/(d-1)}$, for some constants $c_0, c_1, c_2 > 0$ depending on r and d , but not on n .

In [12], upper and lower bounds on the distance

$$\text{dist}(W_2^{r,d}, M_n, L_2) = \sup_{f \in W_2^{r,d}} \text{dist}(f, M_n, L_2)$$

for a Sobolev class $W_2^{r,d}$ were obtained. However, this type of result only guarantees the existence of a function in $W_2^{r,d}$ for which the lower bound holds. That is, it is a “worst case” result. In this work we extend that result by obtaining tight lower and upper bounds that hold for all functions in a subset of large measure in \mathcal{B}^r .

As a consequence, we obtain asymptotically tight lower and upper bounds on the distance between \mathcal{B}^r and M_n measured by a probabilistic (n, δ) -width which is defined as

$$d_{n,\delta}(\mathcal{B}^r, \mu, M_n) = \inf_{\substack{A \subset \mathcal{B}^r \\ \mu(A) = 1 - \delta}} \sup_{f \in A} \text{dist}(f, M_n, L_2), \quad (1)$$

where $0 \leq \delta \leq 1$ and the infimum runs over all subsets A of \mathcal{B}^r with probability $\mu(A) = 1 - \delta$. From the construction of the class \mathcal{B}^r one can see that for any $0 \leq \delta \leq 1$ there exists a subset $A \in \mathcal{B}^r$ such that $\mu(A) = 1 - \delta$. Quantities similar to (1) were considered in [25, 11, 14] where μ was taken to be a Gaussian or Wiener measure and the approximation was linear.

From (1) the next inverse formulation follows

$$\mu\{f \in \mathcal{B}^r : \text{dist}(f, M_n, L_2) \geq d_{n,\delta}\} = \delta,$$

where $d_{n,\delta} = d_{n,\delta}(\mathcal{B}^r, \mu, M_n)$. Indeed, from (1) it follows that there exists the subset A in \mathcal{B}^r such that $\mu(A) = 1 - \delta$ and

$$\begin{aligned} & \mu\{f \in \mathcal{B}^r : \text{dist}(f, M_n, L_2) \geq d_{n,\delta}\} \\ &= \mu\{f \in \mathcal{B}^r : \text{dist}(f, M_n, L_2) \geq \sup_{h \in A} \text{dist}(h, M_n, L_2)\} \end{aligned}$$

and hence

$$\mu\{f \in \mathcal{B}^r : \text{dist}(f, M_n, L_2) \geq d_{n,\delta}\} = \mu\{\mathcal{B}^r \setminus A\} = 1 - \mu(A) = \delta.$$

The main contributions of this paper are threefold: (i) the construction of a uniform measure over a functional class \mathcal{B}^r which is similar to a Besov class. (ii) Proving a lower bound on the degree of approximation by ridge functions which holds for all functions in some subset of \mathcal{B}^r of probability measure $1 - \delta$ with respect to the uniform measure. (iii) Introducing a probabilistic width $d_{n,\delta}$ for *nonlinear* approximation and estimating $d_{n,\delta}(\mathcal{B}^r, \mu, M_n)$ for a uniform measure μ .

2. PRELIMINARIES

We begin by introducing some notation. For an integer $m \geq 1$ let $\mathbb{Z}_m = \{1, 2, \dots, m\}$. Consider the ball of radius r in \mathbb{R}^m denoted by $B^m(r) = \{x \in \mathbb{R}^m: \|x\|_2 \leq r\}$ and set $B^m = B^m(1)$. For a vector $z \in \mathbb{R}^m$, $\text{sgn}(z) = (\text{sgn}(z_1), \dots, \text{sgn}(z_m))$, $\text{sgn}(z_i) = 1$ for $z_i \geq 0$, $\text{sgn}(z_i) = -1$ for $z_i < 0$. We denote by $\|v\|_{l_p^m}$ or simply $\|v\|_p$, $p \geq 1$, the l_p^m Euclidean norm of $v \in \mathbb{R}^m$. For any Euclidean sets A and B in \mathbb{R}^m we use a distance function $\text{dist}(a, B, l_p^m) = \inf_{b \in B} \|a - b\|_{l_p^m}$ for any $a \in A$, and $\text{dist}(A, B, l_p^m) = \sup_{a \in A} \text{dist}(a, B, l_p^m)$.

Define the space of functions

$$L_2 = L_2(B^d) = \left\{ f: \|f\|_{L_2} := \left(\int_{B^d} |f(x)|^2 dx \right)^{1/2} < \infty \right\}.$$

We write $\int_{B^d} f(x) dx$ where $x = (x_1, \dots, x_d)$, and $dx = dx_1 \cdots dx_d$.

The notation $a_n \asymp b_n$ in this paper means that there exist constants $c_1, c_2 > 0$ which depend only on the smoothness parameter r of the class \mathcal{B}^r and the dimensionality d of the domain B^d such that for every $n \geq 1$, $c_1 \leq a_n/b_n \leq c_2$.

We define the class of functions \mathcal{B}^r using the classical means of approximation, namely, algebraic polynomials. Consider the space $\mathcal{P}_s = \text{span}\{x_1^{k_1} \cdots x_d^{k_d}: |k| = k_1 + \cdots + k_d \leq s\}$, $s = 0, 1, \dots$, consisting of all algebraic polynomials on \mathbb{R}^d of total degree at most s . Let $\mathcal{P}_s^h = \text{span}\{x_1^{k_1} \cdots x_d^{k_d}: |k| = s\}$ be the subspace of \mathcal{P}_s consisting of homogeneous polynomials of degree s . Set $m_s = \dim \mathcal{P}_s^h$. It is known (cf. [22]) that $m_s = \binom{d+s-1}{d-1} \asymp s^{d-1}$.

Let the set of polynomials $Q_s = \{q_l\}_{l=1}^{m_s}$ be a basis in \mathcal{P}_s^h . The set of polynomials $\bigcup_{s=0}^{\infty} Q_s$ is a complete system of functions in the space L_2 . Using the method of orthogonalization in L_2 we can construct a complete orthogonal system of polynomials in L_2

$$P = \bigcup_{s=0}^{\infty} \{p_{s,1}, \dots, p_{s,m_s}\},$$

such that the set $P_s^h = \{p_{s,1}, \dots, p_{s,m_s}\}$ is a complete orthonormal system of functions in the subspace \mathcal{P}_s^h . Note in particular that in [12] we constructed one specific orthonormal system of algebraic polynomials in L_2 .

For any natural N we denote the set of multi-indexes

$$A_N = \{(s, l): s = 2^N + 1, \dots, 2^{N+1}, l = 1, \dots, m_s\}.$$

Introduce the subspace $\Phi_N = \text{span}\{p_{s,l}: (s,l) \in \mathcal{A}_N\}$. Let G_N^r , $r > 0$, be the ball with radius 2^{-rN} in the space Φ_N , that is,

$$G_N^r = \left\{ \sum_{(s,l) \in \mathcal{A}_N} c_{s,l} p_{s,l} \in \Phi_N : \left(\sum_{(s,l) \in \mathcal{A}_N} |c_{s,l}|^2 \right)^{1/2} \leq 2^{-rN} \right\}.$$

Denote by \mathcal{B}^r , the set of all functions $f \in L_2(B^d)$ which can be represented as infinite sums of functions from G_N^r , namely

$$\mathcal{B}^r = \left\{ f: f = \sum_{N=0}^{\infty} f_N, f_N \in G_N^r, N = 0, 1, \dots \right\}.$$

It is not hard to see that the class \mathcal{B}^r is essentially equivalent to the class H^r , consisting of all functions f for which the best approximation by algebraic polynomials of degree 2^N satisfies the inequality

$$\text{dist}(f, \mathcal{P}_{2^N}, L_2) \leq 2^{-rN} \quad (N = 0, 1, \dots).$$

From Jackson's Theorem (see [24]), it follows that the Sobolev class $W_2^{r,d}$ belongs to the class H^r and hence also to the class $c\mathcal{B}^r$, for some constant c . Observe also that the latter class (discussed also in [23]) is analogous to the Besov class [26] which is defined using trigonometric polynomials.

As an approximating function class we will use the following nonlinear manifold

$$M_n = \left\{ h(x) = \sum_{l=1}^n h_l(a_l \cdot x) : a_l \in S^{d-1}, h_l \in L_2([-1, 1]) \right\} \quad (x \in B^d) \quad (2)$$

which represents the union of all linear manifolds that are spanned by n ridge functions from the space $L_2([-1, 1])$ of square-integrable functions on the segment $[-1, 1]$.

3. UNIFORM MEASURE CONSTRUCTION

The construction of a uniform measure over a functional class is non-trivial. For example, it is not possible to construct such a measure over a Sobolev or Besov class. For this reason we consider the class \mathcal{B}^r which permits such a construction.

Let $P = \{p_{s,l}\}$ be a complete system of orthonormal polynomials in L_2 , as constructed in Section 2. Then we can express the class \mathcal{B}^r as

$$\mathcal{B}^r = \left\{ f \in L_2 : f(x) = \sum_{N=0}^{\infty} \sum_{(s,l) \in \mathcal{A}_N} c_{s,l} p_{s,l}(x), \right. \\ \left. \left(\sum_{(s,l) \in \mathcal{A}_N} |c_{s,l}|^2 \right)^{1/2} \leq 2^{-rN}, \text{ for all } N \geq 0 \right\}. \quad (3)$$

Consider the subspace $\Phi_N = \text{span}\{p_{s,l} : (s,l) \in \mathcal{A}_N\}$. We have that Φ_N is orthogonal to $\Phi_{N'}$, for all $N \neq N'$, and \mathcal{B}^r is isomorphic to the set D^r of infinite sequences of finite dimensional vectors, i.e.,

$$\mathcal{B}^r \simeq D^r := \prod_{N=0}^{\infty} B^{|\mathcal{A}_N|}(2^{-rN}) \\ := \{c = (c^0, \dots, c^N, \dots) : c^N \in B^{|\mathcal{A}_N|}(2^{-rN})\}, \quad (4)$$

where $c^N := (c_{s,l})_{(s,l) \in \mathcal{A}_N}$, and $|\mathcal{A}_N|$ is the cardinality of \mathcal{A}_N , $N \geq 0$.

Note that the cardinality of \mathcal{A}_N satisfies the asymptotic

$$|\mathcal{A}_N| = \sum_{s=2^N+1}^{2^{N+1}} \dim \mathcal{P}_s^h = \sum_{s=2^N+1}^{2^{N+1}} m_s \asymp \sum_{s=2^N+1}^{2^{N+1}} s^{d-1} \asymp 2^{dN}.$$

Let $b_n \equiv B^{|\mathcal{A}_n|}(2^{-rn})$ be the ball of radius 2^{-rn} in $\mathbb{R}^{|\mathcal{A}_n|}$, and denote the volume of b_n by $\text{vol}(b_n)$. Let $v_n(dc^n) = dc^n / \text{vol}(b_n)$ be the normed Lebesgue measure on b_n , $v_n(b_n) = 1$, and

$$D_N^r = \prod_{n=0}^N b_n.$$

For $c = (c^0, \dots, c^N) \in D_N^r$ define the measure on D_N^r as

$$\lambda_N(dc) = \prod_{n=0}^N v_n(dc^n).$$

Now, let $B \subset D_N^r$. We have

$$\lambda_{N+1}(B \times b_{N+1}) = \int_{B \times b_{N+1}} \lambda_N(dc) v_{N+1}(dc^{N+1}) \\ = \frac{1}{\prod_{n=0}^N \text{vol}(b_n)} \frac{1}{\text{vol}(b_{N+1})} \int_{B \times b_{N+1}} dx dy \\ = \frac{1}{\prod_{n=0}^{N+1} \text{vol}(b_n)} \int_B \int_{b_{N+1}} dy dx$$

which equals $\text{vol}_{(B)}/\prod_{n=0}^N \text{vol}_{(b_n)} = \lambda_N(B)$. It follows from the Kolmogorov Extension of Measure Theorem (see, for example, Shirayev [21, Theorem 3, and observation, p. 163]) that there exists a unique probability measure λ on D^r such that for every $B \subset D_N^r$

$$\lambda((c^0, \dots, c^N, \dots) \in D^r : (c^0, \dots, c^N) \in B) = \lambda_N(B).$$

This uniform measure λ on D^r induces a uniform measure μ on \mathcal{B}^r , which will now be used to establish our main result.

4. MAIN RESULTS

Let $r > 0$ and an integer $d \geq 1$ be given. Fix an integer $n \geq 1$, and set $\alpha(n) = c_1 n^{d/(d-1)}$, for some constant $c_1 > 0$ depending only on r and d . Let μ be the uniform measure over \mathcal{B}^r constructed in Section 3.

THEOREM 1.

$$\mu \left\{ f \in \mathcal{B}^r : \text{dist}(f, M_n, L_2) \geq \frac{c_2}{n^{r/(d-1)}} \right\} \geq 1 - e^{-\alpha(n)}$$

for some constant $c_2 > 0$ depending only on r and d .

THEOREM 2. For all $f \in \mathcal{B}^r$

$$\text{dist}(f, M_n, L_2) \leq \frac{c_3}{n^{r/(d-1)}},$$

where $c_3 > 0$ is some constant depending only on r and d .

From Theorems 1 and 2 we have the following corollary which estimates the probabilistic width defined in (1).

COROLLARY 1. Let $0 \leq \delta < 1 - 2e^{-\alpha(n)}$. Then

$$\frac{c_2}{n^{r/(d-1)}} \leq d_{n, \delta}(\mathcal{B}^r, \mu, M_n) \leq \frac{c_3}{n^{r/(d-1)}}$$

for some constants $c_2, c_3 > 0$ depending only on r and d .

Indeed let $0 \leq \delta < 1 - 2e^{-\alpha(n)}$ be any number. Then for any set $A \subset \mathcal{B}^r$ with the measure $\mu(A) = 1 - \delta$ we have $\mu(A) \geq 2e^{-\alpha(n)}$. Therefore from

Theorem 1 it follows that there exists a function $f \in A$ such that $\text{dist}(f, M_n, L_2) \geq c_2 n^{-r/(d-1)}$. Hence

$$d_{n,\delta}(\mathcal{B}^r, \mu, M_n) \geq \text{dist}(f, M_n, L_2) \geq \frac{c_2}{n^{r/(d-1)}}.$$

The upper bound in Corollary 1 follows directly from Theorem 2.

We note that Traub *et al.* [25] consider also the so called average case setting which introduces the notion of an average distance with respect to a measure over a functional space in our case defined for $0 < p < \infty$ as

$$d_n^{\text{avg}}(\mathcal{B}^r, \mu, M_n)_p = \left(\int_{f \in \mathcal{B}^r} |\text{dist}(f, M_n, L_2)|^p \mu(df) \right)^{1/p}.$$

The following corollary follows easily from Theorems 1 and 2.

COROLLARY 2. *For any $0 < p < \infty$,*

$$\frac{c_2}{n^{r/(d-1)}} \leq d_n^{\text{avg}}(\mathcal{B}^r, \mu, M_n)_p \leq \frac{c_3}{n^{r/(d-1)}}$$

for some constants $c_2, c_3 > 0$ depending only on r, d , and p .

We proceed to prove Theorem 1, first stating several auxiliary lemmas. From the definition of the orthonormal system $P = \{p_{s,l}\}$ it follows that an $h \in M_n$ can be expressed as a sum $\sum_{N=0}^{\infty} \sum_{(s,l) \in \Delta_N} c_{s,l}(h) p_{s,l}(x)$ with the coefficients, $c_{s,l}(h) = \langle h, p_{s,l} \rangle = \int_{B^d} h(x) p_{s,l}(x) dx$. Let $N \in \mathbb{Z}_+$ be some number, and $I \subset \Delta_N$ be any subset. Consider the set of sign-valued vectors

$$\Gamma_n^I := \{(\text{sgn}(c_{s,l}(h)))_{(s,l) \in I} : h \in M_n\}. \quad (5)$$

We will use the next lemma which follows from Lemma 3 of [12].

LEMMA 1. *Assume that N and n are such that $|\Delta_N| = \lfloor c_5 n^{d/(d-1)} \rfloor$, for some absolute constant $c_5 > 0$. Then for any subset $I \subset \Delta_N$ with $|I| \geq |\Delta_N|/10$ we have*

$$|\Gamma_n^I| \leq 2^{c_4 |I|} \leq 2^{c_6 n^{d/(d-1)}},$$

where $c_4 = 0.23$, and $c_6 = c_4 c_5$.

The next lemma then follows.

LEMMA 2. Let $|\mathcal{A}_N| = \lceil c_5 n^{d/(d-1)} \rceil$, and let $I \subset \mathcal{A}_N$, $|I| \geq |\mathcal{A}_N|/10$. Introduce the sets of sign-valued vectors $E^{|I|} = \{-1, +1\}^{|I|}$, and $\hat{E}^{|I|} = \{\varepsilon \in E^{|I|} : \text{dist}(\varepsilon, \Gamma_n^I, l_2^{|I|}) \geq 2\sqrt{|I|/3}\}$. Then

$$|\hat{E}^{|I|}| \geq 2^{|I|} - 2^{c_7 |I|}$$

for some absolute constant $0 < c_7 < 1$.

Proof. Set $k = |I|$. From Lemma 1 it follows that the cardinality $|\Gamma_n^I| \leq 2^{c_4 k}$. Fix any $\varepsilon^* \in E^k$. Denote by

$$D_{\varepsilon^*} = \left\{ \varepsilon \in E^k : \|\varepsilon - \varepsilon^*\|_{l_2^k}^2 \geq \frac{4k}{9} \right\}.$$

Now $|D_{\varepsilon^*}|$ is independent of the specific choice of $\varepsilon^* \in E^k$. As such $|D_{\varepsilon^*}| = |\{\varepsilon \in E^k : \|\varepsilon - \underline{1}\|_{l_2^k}^2 \geq 4k/9\}|$ where $\underline{1} = [1, \dots, 1] \in E^k$. The latter equals $\sum_{i \geq k/9} \binom{k}{i}$ and is bounded from below by $2^k - 2^{c_8 k}$, $c_8 = 1 - 2(7/18)^2 \log_2 e = 0.55\dots$, where we used an upper bound on the tails of the binomial distribution (cf. [3]).

Set $\bar{D}_{\varepsilon} = E^k \setminus D_{\varepsilon}$. Then $|\bar{D}_{\varepsilon^*}| = |E^k \setminus D_{\varepsilon^*}| \leq 2^{c_8 k}$. We also have $\hat{E}^k := \bigcap_{\varepsilon^* \in \Gamma_n^I} D_{\varepsilon^*} = E^k \setminus (\bigcup_{\varepsilon^* \in \Gamma_n^I} \bar{D}_{\varepsilon^*})$. It follows that

$$|\hat{E}^k| \geq |E^k| - \left| \bigcup_{\varepsilon^* \in \Gamma_n^I} \bar{D}_{\varepsilon^*} \right| \geq |E^k| - |\Gamma_n^I| 2^{c_8 k} \geq 2^k - 2^{c_4 k} 2^{c_8 k}.$$

Set $c_7 = c_4 + c_8 = 0.78\dots$. Thus $|\hat{E}^k| \geq 2^k - 2^{c_7 k}$, which proves the lemma.

DEFINITION 1. Let B^m denote the unit ball in \mathbb{R}^m . For any set $A \subset B^m$ denote the volume of A as $\text{vol}(A)$. The uniform measure over the ball denoted by ν is defined such that for every $A \subset B^m$, $\nu(A) = \text{vol}(A)/\text{vol}(B^m)$.

Denote by

$$A := \left\{ x \in B^m : |x_k| > \frac{3}{8\sqrt{m}}, \text{ for at least } \frac{m}{10} \text{ coordinates } k \right\}.$$

We will use the following lemma.

LEMMA 3. For any $m \geq 1$

$$\nu(A) \geq 1 - 3e^{-c_9 m}$$

for some absolute constant $c_9 > 0$.

Proof. We aim at finding a lower bound on $v(A)$ by first expressing the measure of the set A under the uniform measure over B^m as the measure of another set under the Gaussian measure over \mathbb{R}^m . Introduce the auxiliary set in \mathbb{R}^m

$$\hat{A} = \left\{ x \in \mathbb{R}^m : |x_k| > \frac{3}{4\sqrt{m}} \|x\|_2, \text{ for at least } \frac{m}{10} \text{ coordinates } k \right\}.$$

Denote by $B^m(\alpha, \beta) := B^m(\beta) \setminus B^m(\alpha)$. We have

$$v(A) \geq v(B^m(\tfrac{1}{2}, 1) \cap \hat{A}) \geq v(\hat{A} \cap B^m) - (\tfrac{1}{2})^m. \quad (6)$$

Let $\chi_A(x)$ denote the indicator function of the set A . Switching to polar coordinates we have, since $x \in \hat{A}$ implies $ax \in \hat{A}$ for all $a \neq 0$

$$\begin{aligned} v(\hat{A} \cap B^m) &= \frac{1}{\text{vol}(B^m)} \int_{B^m} \chi_{\hat{A}}(x) dx \\ &= \frac{1}{\text{vol}(B^m)} \int_0^1 r^{m-1} dr \int_{S^{m-1}} \chi_{\hat{A}}(s) ds \quad (s \in S^{m-1}), \end{aligned} \quad (7)$$

where ds is the Lebesgue measure on S^{m-1} . Assume that m is even (for m odd the proof is analogous). The volume of the unit ball $\text{vol}(B^m) = \pi^{m/2}/(m/2)!$. It is known (cf. [17]) that $\int_0^\infty x^{m-1} e^{-x^2} dx = \frac{1}{2} \Gamma(m/2)$. Hence it follows that

$$\frac{1}{\text{vol}(B^m)} \int_0^1 r^{m-1} dr = \pi^{-m/2} \int_0^\infty r^{m-1} e^{-r^2} dr.$$

Therefore using once more polar coordinates we obtain from (7)

$$\begin{aligned} v(\hat{A} \cap B^m) &= \pi^{-m/2} \int_0^\infty r^{m-1} e^{-r^2} dr \int_{S^{m-1}} \chi_{\hat{A}}(s) ds \\ &= \pi^{-m/2} \int_{\mathbb{R}^m} \chi_{\hat{A}}(x) e^{-|x|^2} dx. \end{aligned} \quad (8)$$

Define a Gaussian measure over \mathbb{R}^m as $\gamma(G) = \pi^{-m/2} \int_G e^{-|x|^2} dx$, $G \subset \mathbb{R}^m$. From (8) it is seen that $v(\hat{A} \cap B^m) = \gamma(\hat{A})$. Let

$$D = \left\{ x \in \mathbb{R}^m : |x_k| \geq \frac{3}{2}, \text{ for at least } \frac{m}{10} \text{ coordinates } k \right\}.$$

Then it follows that

$$v(\hat{A} \cap B^m) = \gamma(\hat{A}) \geq \gamma(\hat{A} \cap B^m(2\sqrt{m})) \geq \gamma(D \cap B^m(2\sqrt{m})),$$

and therefore

$$\begin{aligned} v(\hat{A} \cap B^m) &\geq \gamma(D) + \gamma(B^m(2\sqrt{m})) - \gamma(D \cup B^m(2\sqrt{m})) \\ &\geq \gamma(D) + \gamma(B^m(2\sqrt{m})) - 1. \end{aligned} \quad (9)$$

Let $I \subset \mathbb{Z}_m = \{1, 2, \dots, m\}$. Consider the subset in D

$$D_I = \{x \in D: |x_i| \geq \frac{3}{2} \text{ for all } i \in I, |x_i| < \frac{3}{2} \text{ for all } i \in \mathbb{Z}_m \setminus I\}.$$

We have

$$\gamma(D) = \sum_{I \subset \mathbb{Z}_m} \gamma(D_I) = \sum_{l=1}^m \sum_{I \subset \mathbb{Z}_m, |I|=l} \gamma(D_I) \geq \sum_{l=m/10}^m \sum_{I \subset \mathbb{Z}_m, |I|=l} \gamma(D_I).$$

For $|I| = l$

$$\gamma(D_I) = p^l (1-p)^{m-l},$$

where

$$\frac{1}{\sqrt{\pi}} \int_{|t| \geq 3/2} e^{-t^2} dt = 0.134 \equiv p.$$

Hence from the definition of the Gaussian measure γ it follows that

$$\gamma(D) \geq \sum_{l=m/10}^m \binom{m}{l} p^l (1-p)^{m-l} > 1 - e^{-c_{10}m} \quad (10)$$

for some $0 < c_{10} < 1$ where we used a bound on the tail of the binomial distribution [3].

We now estimate $\gamma(B^m(2\sqrt{m}))$. We will show that

$$\gamma(B^m(2\sqrt{m})) \geq 1 - e^{-c_{11}m} \quad (11)$$

for some absolute constant $c_{11} > 0$.

Indeed using polar coordinates we have

$$\begin{aligned} \gamma(B^m(2\sqrt{m})) &= \pi^{-m/2} \int_{B^m(2\sqrt{m})} e^{-|x|^2} dx \\ &= 1 - \pi^{-m/2} \int_{\mathbb{R}^m \setminus B^m(2\sqrt{m})} e^{-|x|^2} dx \\ &= 1 - \pi^{-m/2} d(S^{d-1}) \int_{2\sqrt{m}}^{\infty} r^{m-1} e^{-r^2} dr, \end{aligned}$$

where $d(S^{d-1})$ is the Lebesgue measure of the sphere S^{d-1} . Using the substitution $r = \sqrt{mt/2}$, and the estimate $\int_8^\infty t^{k-1} e^{-kt} dr \leq (1/7k) e^{-8k} 8^{k+1/2}$, $k \geq 1$ (see [10, p. 471, form. (6.5)]), we obtain

$$\int_{2\sqrt{m}}^\infty r^{m-1} e^{-r^2} dr = \frac{1}{2} (m/2)^{m/2} \int_8^\infty t^{m/2-1} e^{-(m/2)t} dt \leq \frac{1}{2} (m/2)^{m/2} e^{-c'_{11}m},$$

where $c'_{11} = 4 - \frac{3}{2} \ln 2$. Since $d(S^{d-1}) = \text{vol}(B^m)/m = \pi^{m/2}/m\Gamma(m/2) \asymp \pi^{m/2} \times e^{m/2}/(m(m/2)^{m/2} \sqrt{2\pi m})$, then

$$\nu(B^m(2\sqrt{m})) \geq 1 - \pi^{-m/2} d(S^{d-1}) \frac{1}{2} (m/2)^{m/2} e^{-c'_{11}m} \geq 1 - e^{-c_{11}m},$$

where $c_{11} = \frac{3}{2}(1 - \ln 2)$.

Using (6), (9), (10), and (11) we obtain that

$$\nu(A) \geq 1 - e^{-c_{10}m} - e^{-c_{11}m} - 2^{-m} \geq 1 - 3e^{-c_9m},$$

for absolute constant $c_9 = \min\{c_{10}, c_{11}, \ln 2\}$. ■

We now proceed with finding a lower bound on the measure stated in Theorem 1.

4.1. Proof of Theorem 1

The proof of Theorem 1 is based on the following observation. Let $m = |A_N|$. In the space \mathbb{R}^m , consider the set $E^m = \{-1, +1\}^m$ endowed with a uniform discrete measure α , and let $\Gamma_n^{A_N}$ be the subset in E^m defined in (5). From Lemma 2 it follows that the measure of elements in E^m which are “badly” approximated by the manifold $\Gamma_n^{A_N}$, i.e., the α measure of set $G = \{\varepsilon \in E^m: \text{dist}(\varepsilon, \Gamma_n^{A_N}, l_2^m) \geq 2\sqrt{m/3}\}$ satisfies the inequality

$$\alpha(G) \geq 1 - 2^{-cm}$$

for $c > 0$. This implies that almost all elements from E^m , in the sense of the induced probabilistic measure over E^m , are “badly” approximated by $\Gamma_n^{A_N}$. The statement of the theorem follows upon making use of the isomorphism (4).

We proceed with the detailed proof. Let $N^* > 0$ be some integer which will be taken later to be sufficiently large. Since

$$\begin{aligned} \text{dist}(f, M_n, L_2)^2 &= \inf_{h \in M_n} \sum_{N=0}^{\infty} \sum_{(s,l) \in A_N} |c_{s,l}(f) - c_{s,l}(h)|^2 \\ &\geq \inf_{h \in M_n} \sum_{(s,l) \in A_{N^*}} |c_{s,l}(f) - c_{s,l}(h)|^2, \end{aligned}$$

then for an arbitrary $\varepsilon > 0$

$$\begin{aligned} & \mu\{f \in \mathcal{B}^r: \text{dist}(f, M_n, L_2) > \varepsilon\} \\ & \geq \mu\left\{f \in \mathcal{B}^r: \inf_{h \in M_n} \sum_{(s, l) \in A_{N^*}} |c_{s, l}(f) - c_{s, l}(h)|^2 > \varepsilon^2\right\}. \end{aligned} \quad (12)$$

Let m , N^* and n be such that $m = |A_{N^*}| = c_5 n^{d/(d-1)}$. To any $h \in M_n$ there corresponds a vector $\hat{h} \in \mathbb{R}^m$ defined as

$$\hat{h} = (c_{s, l}(h))_{(s, l) \in A_{N^*}}. \quad (13)$$

Denote by

$$\hat{M}_n = \{\hat{h} = (\hat{h}_1, \dots, \hat{h}_m) \in \mathbb{R}^m: h \in M_n\}.$$

Due to the isomorphism statement of (4) the approximation problem is now reduced to approximation in an m -dimensional Euclidean space.

Let $\varepsilon = 2^{rN^*}/4$ in (12). We have

$$\begin{aligned} \Sigma &:= \mu\left\{f \in \mathcal{B}^r: \inf_{h \in M_n} \sum_{k \in A_{N^*}} |c_k(f) - c_k(h)|^2 > \varepsilon^2\right\} \\ &= \nu\left\{y \in B^m: \inf_{\hat{h} \in \hat{M}_n} \sum_{i=1}^m |y_i - \hat{h}_i|^2 > \frac{1}{4}\right\}. \end{aligned}$$

Let $I \subseteq \mathbb{Z}_m$. Define the set

$$Q_I = \left\{x \in B^m: |x_i| \geq \frac{3}{8\sqrt{m}}, \text{ for all } i \in I, |x_i| \leq \frac{3}{8\sqrt{m}} \text{ for all } i \in \mathbb{Z}_m \setminus I\right\}.$$

From the definition of Q_I we have $\bigcup_{I \in \mathbb{Z}_m} Q_I = B^m$. Thus

$$\Sigma = \sum_{I \in \mathbb{Z}_m} \nu\left\{y \in Q_I: \inf_{\hat{h} \in \hat{M}_n} \sum_{i=1}^m |y_i - \hat{h}_i|^2 > \frac{1}{4}\right\}.$$

For all $I \subset \mathbb{Z}_m$, $|I| \geq m/10$, and $y \in Q_I$ we have

$$\sum_{i=1}^m |y_i - \hat{h}_i|^2 \geq \sum_{i \in I} |y_i - \hat{h}_i|^2 \geq \frac{9}{64m} \sum_{i \in I} \left| \frac{y_i}{|y_i|} - \frac{\hat{h}_i}{|\hat{h}_i|} \right|^2.$$

Denote by $\varepsilon_i(y) = y_i/|y_i| = \text{sgn}(y_i)$. Then using the fact that for any $a \in \mathbb{R}$ and $\delta \in \{-1, +1\}$ the inequality $|\delta - a| \geq \frac{1}{2} |\delta - \text{sgn}(a)|$ holds we have

$$\sum_{i=1}^m |y_i - \hat{h}_i|^2 \geq \frac{9}{256m} \sum_{i \in I} |\varepsilon_i(y) - \text{sgn}(\hat{h}_i)|^2.$$

We then have for $b = 64/9$

$$\begin{aligned} \Sigma &\geq \sum_{I \in \mathbb{Z}_m: |I| \geq m/10} \nu \left\{ y \in Q_I: \inf_{\hat{h} \in \hat{M}_n} \sum_{i \in I} |\varepsilon_i(y) - \text{sgn}(\hat{h}_i)|^2 > bm \right\} \\ &= \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I| = m/10 + j} \nu \left\{ y \in Q_I: \inf_{\hat{h} \in \hat{M}_n} \sum_{i \in I} |\varepsilon_i(y) - \text{sgn}(\hat{h}_i)|^2 > bm \right\}. \end{aligned}$$

For $I \in \mathbb{Z}_m$ let $E^{|I|} = \{-1, +1\}^{|I|}$. Define

$$\Gamma_n^{|I|} = \{(\text{sgn}(\hat{h}_i))_{i \in I}: h \in M_n\}.$$

Denote by $\|y\|_{l_2^{|I|}} = (\sum_{i \in I} |y_i|^2)^{1/2}$. Let

$$\hat{E}^{|I|} = \{\varepsilon \in E^{|I|}: \min_{\delta \in \Gamma_n^{|I|}} \|\varepsilon - \delta\|_{l_2^{|I|}}^2 \geq bm\}. \quad (14)$$

For any $\varepsilon = (\varepsilon_i)_{i \in I} \in E^{|I|}$ define the set

$$Q_{I, \varepsilon} = \{y \in Q_I: \text{sgn}(y_i) = \varepsilon_i, \text{ for all } i \in I\}.$$

Then continuing from above we have

$$\begin{aligned} \Sigma &\geq \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I| = m/10 + j} \nu \{ y \in Q_I: \min_{\delta \in \Gamma_n^{|I|}} \|\varepsilon(y) - \delta\|_{l_2^{|I|}}^2 > bm \} \\ &= \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I| = m/10 + j} \sum_{\varepsilon \in E^{|I|}} \nu \{ y \in Q_{I, \varepsilon}: \min_{\delta \in \Gamma_n^{|I|}} \|\varepsilon(y) - \delta\|_{l_2^{|I|}}^2 > bm \} \end{aligned}$$

and since $\hat{E}^{|I|} \subset E^{|I|}$ then

$$\Sigma \geq \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I| = m/10 + j} \sum_{\varepsilon \in \hat{E}^{|I|}} \nu \{ y \in Q_{I, \varepsilon}: \min_{\delta \in \Gamma_n^{|I|}} \|\varepsilon(y) - \delta\|_{l_2^{|I|}}^2 > bm \}$$

Now from (14) for all $\varepsilon \in \hat{E}^{|I|}$ the condition $\min_{\delta \in \Gamma_n^{|I|}} \|\varepsilon - \delta\|_{l_2^{|I|}}^2 > bm$ is satisfied. We therefore have

$$\Sigma \geq \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I| = m/10 + j} \sum_{\varepsilon \in \hat{E}^{|I|}} \nu \{ y \in Q_{I, \varepsilon} \}.$$

Note that $v(y \in Q_{I,\varepsilon})$ does not depend on ε . Denote by $a_I := v\{y \in Q_{I,\varepsilon}\}$. Thus the latter becomes

$$\sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I|=m/10+j} \sum_{\varepsilon \in \hat{E}^{|I|}} a_I = \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I|=m/10+j} a_I |\hat{E}^{|I|}|.$$

From Lemma 2 it follows that for I such that $|I|=m/10+j$ the cardinality $|\hat{E}^{|I|}| \geq 2^{m/10+j} - 2^{c_7(m/10+j)}$ for some constant $0 < c_7 < 1$. We therefore have

$$\begin{aligned} \Sigma &\geq \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I|=m/10+j} a_I 2^{m/10+j} (1 - 2^{-(1-c_7)(m/10+j)}) \\ &\geq (1 - 2^{-(1-c_7)m/10}) \sum_{j=0}^{9m/10} \sum_{I \in \mathbb{Z}_m: |I|=m/10+j} a_I 2^{m/10+j}. \end{aligned}$$

Since $a_I 2^{m/10+j} = |E^{|I|}| a_I = v(Q_I)$ then

$$\Sigma \geq (1 - 2^{-(1-c_7)m/10}) \times v \left\{ y \in B^m: |y_k| > \frac{3}{8\sqrt{m}}, \text{ for at least } \frac{m}{10} \text{ coordinates } k \right\}.$$

Using Lemma 3 we have

$$\Sigma \geq (1 - 2^{-(1-c_7)m/10})(1 - 3e^{-c_9 m}) \geq 1 - e^{-c_{12} m}$$

for some absolute constants $c_9, c_{12} > 0$. Finally, from before, $m = 2^{dN^*}$ and $\rho = 2^{-rN^*}$ then $\rho \asymp m^{-r/d}$. Also, the condition of Lemma 1 has $m = c_5 n^{d/(d-1)}$ thus $\rho = c_{13}/n^{r/(d-1)}$ and therefore

$$\mu \left\{ f \in \mathcal{B}^r: \text{dist}(f, M_n, L_2) > \frac{c_{13}}{4n^{r/(d-1)}} \right\} \geq 1 - e^{-\alpha(n)},$$

where $\alpha(n) = c_5 c_{11} n^{d/(d-1)}$. This completes the proof of Theorem 1. \blacksquare

4.2. Proof of Theorem 2

Let \mathcal{P}_s and \mathcal{P}_s^h be as defined in Section 2. Choose n such that $n = \dim(\mathcal{P}_s^h)$. Then from Proposition 2 of [12] it follows that $\mathcal{P}_s \subset M_n$. Let N' be the integer such that $2^{N'-1} \leq s \leq 2^{N'}$. Considering the definition of \mathcal{B}^r we have for all $f \in \mathcal{B}^r$, $f(x) = \sum_{N=0}^{\infty} \sum_{(s,l) \in A_N} c_{s,l} P_{s,l}(x)$ and

$$\text{dist}(f, M_n, L_2)^2 \leq \text{dist}(f, \mathcal{P}_s, L_2)^2 \leq \left\| \sum_{N \geq N'} \sum_{(s,l) \in A_N} c_{s,l} P_{s,l} \right\|_{L_2}^2.$$

Therefore from the Parseval equality and the definition of class \mathcal{B}^r we obtain

$$\begin{aligned} \text{dist}(f, M_n, L_2)^2 &\leq \sum_{N \geq N'} \left\| \sum_{(s, l) \in A_N} c_{s, l} p_{s, l} \right\|_{L_2}^2 \\ &\leq \sum_{N \geq N'} 2^{-2rN} \leq c_{14} 2^{-2rN'} = c_{14} s^{-2r}, \end{aligned}$$

for some constant $c_{14} > 0$. It is known (cf. [22]) that $\dim(\mathcal{P}_s^h) = \binom{d+s-1}{d-1} \asymp s^{d-1}$. Since $n \asymp s^{d-1}$ then $s \asymp n^{1/(d-1)}$. Thus

$$\text{dist}(f, M_n, L_2) \leq c_{15} n^{-r/(d-1)},$$

which proves the theorem.

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