

Approximation Theory of Neural Networks

(1) Introduction

(2) An example

(3) Review of two classical approximation Thms
(Stone-Weierstrass, Kolmogorov Superposition Thm)

(4) Universal Approximation Thm.

(5) Error bounds of NN approximations

(6) Other topics

An example Given $g \in C([0,1])$. Construct a simple NN.

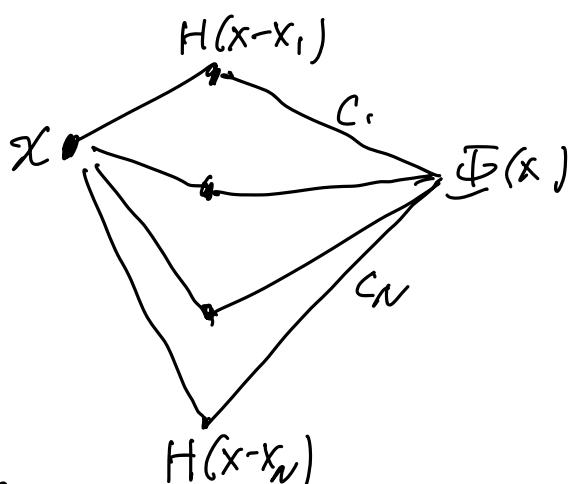
EGNN (H , $L=2$, $N_0=1$, $N_L=1$), to approximate g . Here,
 $H(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 & \text{if } x \geq 0 \end{cases}$ — the Heaviside function. ($H(0)=1$)

Thm $\forall \varepsilon > 0. \exists N \in \mathbb{N} (N \gg 1), \exists 0 \leq x_1 < \dots < x_N \leq 1,$

$\exists c_1, \dots, c_N \in \mathbb{R}$, s.t. $|\Phi_N(x) - g(x)| < \varepsilon \quad \forall x \in [0,1]$,

where

$$\Phi(x) = \frac{\Phi_N(x)}{N} = \sum_{j=1}^N c_j H(x-x_j) \in \mathcal{NN}(H, 2, 1, 1), \quad x \in [0,1].$$



$-x_j = \phi_j$: threshold/bias.

c_j — weights in the output

$$A_1 = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}_1, \quad b_1 = \begin{bmatrix} -x_1 \\ \vdots \\ -x_N \end{bmatrix}_1.$$

$$A_2 = [c_1 \dots c_N]_N, \quad b_2 = 0 \in \mathbb{R}.$$

Proof Since $g \in C([0,1])$ is unif. cont. on $[0,1]$, $\exists d > 0$
s.t. $|x-y| < d \implies |g(x) - g(y)| < \varepsilon$. Choose $N \in \mathbb{N}$ s.t. $N > \frac{1}{d}$.

[2]

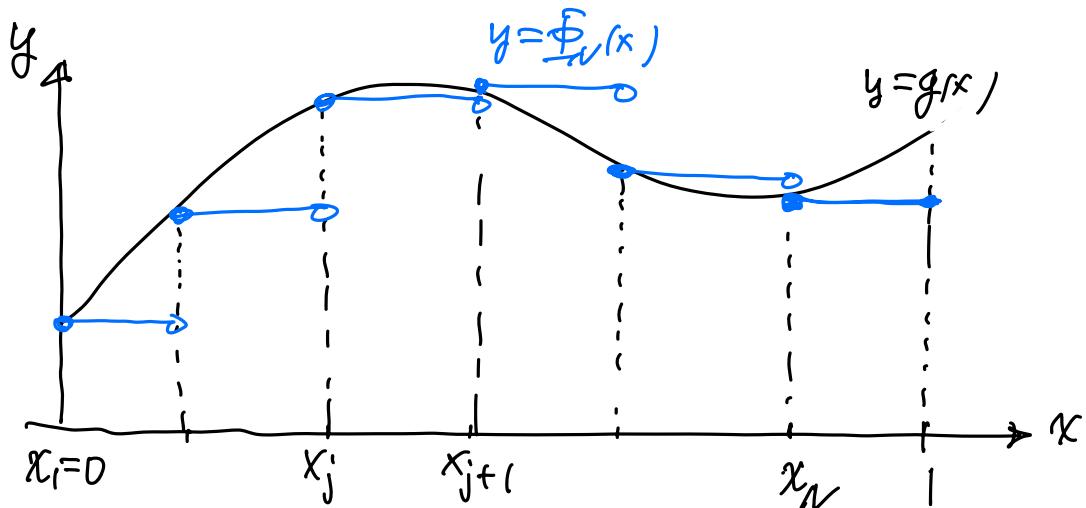
Define $x_j = (j-1)/N$, $j=1, \dots, N$. On each $[x_j, x_{j+1}]$.

$$|g(x) - g(x_j)| < \varepsilon \quad \forall x \in [x_j, x_{j+1}].$$

Define $C_1 = g(x_1)$, $C_j = g(x_j) - g(x_{j-1})$, $j=2, \dots, N$. Then.

$$\begin{aligned}\Phi_N(x) &= \sum_{j=1}^N C_j H(x-x_j) \\ &= \begin{cases} g(x_1) & \text{if } x \in [x_1, x_2], \\ g(x_L) & \text{if } x \in [x_L, x_S], \\ \vdots & \vdots \\ g(x_{j+1}) & \text{if } x \in [x_j, x_{j+1}], \\ g(x_N) & \text{if } x \in [x_N, 1]. \end{cases}\end{aligned}$$

Then $|\Phi_N(x) - g(x)| < \varepsilon$. $\forall x \in [0, 1]$. QED



The First Weierstrass Approximation Theorem. Let $a, b \in \mathbb{R}$ with $a < b$. Let $f \in C([a, b])$ and $\varepsilon > 0$. Then there exists a polynomial p such that

$$\max_{a \leq x \leq b} |f(x) - p(x)| < \varepsilon. \quad (*)$$

Remarks

① Define for $g \in C([a, b])$,

$$\|g\| = \|g\|_{ab} = \|g\|_{C([a, b])} = \|g\|_C = \max_{a \leq x \leq b} |g(x)|.$$

Then $(C([a, b]), \|\cdot\|)$ is a Banach space. The inequality (*) is the same as $\|f - p\| < \varepsilon$.

- ① Equivalently. $\lim_{n \rightarrow \infty} E_n(f) = 0$, where
 $E_n(f) = \min_{g \in P_n} \|f - g\|$, (\min is attained). [3]
 $P_n = \{ \text{all polynomials of deg } \leq n \}$.

- ② The theorem provides no info about P .

The Second Weierstrass Approximation Theorem Let $f \in C_{\mathbb{R}}$ and $\varepsilon > 0$ then there exists a trigonometric polynomial T such that $\max_{-\pi \leq x \leq \pi} |f(x) - T(x)| < \varepsilon$.

The Stone-Weierstrass Theorem Let X be a compact Hausdorff space and $C(X, \mathbb{R})$ or $C(X)$ the space of all real continuous functions equipped with the uniform norm. Let $\mathcal{A} \subseteq C(X, \mathbb{R})$. Assume

- ① \mathcal{A} is a subalgebra;
- ② \mathcal{A} separates points of X ; and
- ③ \mathcal{A} contains the constant functions.

Then $\overline{\mathcal{A}} = C(X, \mathbb{R})$ (i.e., \mathcal{A} is dense in $C(X, \mathbb{R})$)

Remarks

- ① \mathcal{A} is a subalgebra means: \mathcal{A} is a vector subspace of $C(X, \mathbb{R})$ and \mathcal{A} is closed in multiplication: $f, g \in \mathcal{A} \Rightarrow f+g \in \mathcal{A}; f \in \mathcal{A}, \alpha \in \mathbb{R} \Rightarrow \alpha f \in \mathcal{A}; f, g \in \mathcal{A} \Rightarrow f \cdot g \in \mathcal{A}$.

- ② \mathcal{A} separates points of X : $\forall x, y \in X, x \neq y \exists f \in \mathcal{A}$ s.t. $f(x) \neq f(y)$.

- ③ $\overline{\mathcal{A}}$ is the closure of \mathcal{A} w.r.t. the uniform

4

nam. $\overline{\mathcal{A}} = C(X, \mathbb{R})$ means that $\forall f \in C(X, \mathbb{R})$ there exist $g_k \in \mathcal{A}$ ($k=1, 2, \dots$) such that $g_k \rightarrow f$ as $k \rightarrow \infty$. i.e.,

$$\|f - g_k\| = \max_{x \in X} |f(x) - g_k(x)| \rightarrow 0 \text{ as } k \rightarrow \infty.$$

(*) You may consider X to be a compact subset of \mathbb{R}^n . Compact here = closed + bounded.

Example $X = \prod_{j=1}^n [a_j, b_j]$. ($a_j, b_j \in \mathbb{R}, a_j < b_j, j=1, \dots, n$) or $X = \overline{\text{co}(X_1, \dots, X_m)}$ the convex hull of m

points in \mathbb{R}^n ($X_1, \dots, X_m \in \mathbb{R}^n$). $\mathcal{A} = \{ \text{all real multivariable polynomials } p = p(x_1, \dots, x_n) \}$.

Then $\overline{\mathcal{A}} = C(X, \mathbb{R})$, i.e., $\forall f \in C(X)$ $\forall \varepsilon > 0$, there exists a polynomial $p: \mathbb{R}^n \rightarrow \mathbb{R}$ s.t.

$$|f(x) - p(x)| < \varepsilon \quad \forall x \in X.$$

The classical Weierstrass Thm.

Corollary. Let $X = \prod_{j=1}^n [a_j, b_j]$ with all $a_j, b_j \in \mathbb{R}$ and $a_j < b_j$ ($j=1, \dots, n$). Let $f \in C(X)$ and $\varepsilon > 0$. Then, $\exists N \in \mathbb{N}$, and $f_{ij} \in C([a_j, b_j])$ ($i=1, \dots, N; j=1, \dots, n$) such that

$$\max_{x=(x_1, \dots, x_n) \in X} |f(x) - \sum_{i=1}^N \prod_{j=1}^n f_{ij}(x_j)| < \varepsilon.$$

Proof Let $\mathcal{A} = \left\{ \sum_{i=1}^N \prod_{j=1}^n g_{ij}(x_j) : g_{ij} \in C([a_j, b_j]), j=1, \dots, n, i=1, \dots, N, N=1, 2, \dots \right\}$.

Apply the Stone-Weierstrass Thm. QED
(Fill in details!)

Remark. We can choose $f_{i,j} \in \mathcal{P}$ (the set of all one-variable real polynomials). 5

The Kolmogorov Thm (or Kolmogorov-Arnold Thm)

Kolmogorov, 1956, Arnold 1957, Kolmogorov 1957.

Thm (Kolmogorov 1957) For any integer $n \geq 2$ and and $f \in C([0,1]^n)$ there exist $\psi^{p,q} \in C([0,1])$ ($p=1, \dots, n$, $q=1, \dots, 2^n+1$) and $x_q \in \mathbb{C}(\mathbb{R})$ ($q=1, \dots, 2^n+1$) such that

$$f(x_1, \dots, x_n) = \sum_{q=1}^{2^n+1} x_q \left(\sum_{p=1}^n \psi^{p,q}(x_p) \right) \quad \forall x = (x_1, \dots, x_n) \in [0,1]^n.$$

Remarks

- Originated from Hilbert's 13th prob. (1900).
- Constructive proof is given by Braess and Griebel (Constr. Approx. 2009), after Sprecher (1996, 1997) and Köppen (2002).

Universal Approximations

Question: Given a function/map: $f: X \rightarrow \mathbb{R}^m$, are there NNMs Φ_k ($k=1, 2, \dots$) such that $\Phi_k \rightarrow f$? Here $X \subseteq \mathbb{R}^d$ and the convergence is w. r. t. some norm or metric or some topology.

More questions

- What is the class of functions that can be approximated well by NNMs?
- The norm/metric? Uniform norm?

L^1, L^2, L^p ($1 \leq p < \infty$), L^∞ ? $W^{k,p}$?

① Sequence of NNs w.r.t. depth ($\gg 1$, with $W \gg 1$?)

② NN approximations vs. classical approximations (by polynomials, piecewise polynomials, trigonometric polynomials, splines, wavelets, finite elements, etc. and data fitting, e.g., interpolation, least-squares, etc.)?

Some observations / considerations

① Only need to consider the output dimension = 1? Yes, for fixed L . But generally?

Suppose $\Phi^{(j)} \in \text{NN}(\sigma, L, N_0, N_L=1)$ and $\Phi^{(j)} \approx f^{(j)}: X (\subseteq \mathbb{R}^{N_0}) \rightarrow \mathbb{R}^1$, $j=1, \dots, m$. Then by parallelization with shared inputs, we can construct $\Phi \in \text{NN}(\sigma, L, N_0, N_L=m)$ such that $(\Phi(x))_j = \Phi^{(j)}(x) \forall x \in X$ for $j=1, \dots, m$. Thus, $\Phi \approx f$ with $f = \begin{bmatrix} f^{(1)} \\ \vdots \\ f^{(m)} \end{bmatrix}$.

In deep NN approximations, we may have $\Phi^{(j)} \in \text{NN}(\sigma, L_j, N_0, N_L=1)$ ($j=1, \dots, m$). How to construct Φ from these $\Phi^{(j)}$ so that Φ can approximate $f: X (\subseteq \mathbb{R}^{N_0}) \rightarrow \mathbb{R}^{\max(N_1, \dots, N_m)}$? This leads to the question: how to add one layer to an existing NN to get a new NN which is the same function? For some σ , it may be

fine. But in general, unknown or may not [7] be possible.

The way is: for each component of $f: X \rightarrow \mathbb{R}^m$, we approximate it by a sequence of ANNs.

① Any polynomial will not work (as an activation function). Since, if σ is a polynomial of degree n , then any $\text{ANN}(\sigma, N_0=1, N_L=1)$ is also a polynomial of degree $\leq n$. Therefore $\text{ANN}(\sigma, N_0=1, N_L=1)$ is a finite-dimensional space, and cannot approximate $C([0,1])$ which is infinitely dimensional.

② Proof of universal approximations — abstract vs. constructive.

Early works

- ① G. Cybenko, Math. Control Signals Systems, 1989.
- ① K.-I. Funahashi, Neural Networks, 1989.
- ① K. Hornik, M. Stinchcombe, and H. White, Neural Networks, 1989.
- ① K. Hornik, Neural Networks, 1991.
- ① Leshno et al. Neural Networks. 1993.

Some notations / definitions

① The space $C(K)$.

② $d \in \mathbb{N}$: input dimension.

③ $K \subseteq \mathbb{R}^d$: compact (i.e., bounded + closed, e.g., $K = [0,1]^d$, K = a closed ball in \mathbb{R}^d ; etc.)

① Jones 1990
constructive proof.
① Carroll-Dickson (1990)
Also constructive

③ $C(K) = \{ \text{all real-valued continuous functions } K \rightarrow \mathbb{R} \}$

④ $\|f\| = \|f\|_C = \|f\|_{C(K)} = \|f\|_\infty = \max_{x \in K} |f(x)|$.
Unif. norm. or max. norm.

Proposition $(C(K), \|\cdot\|)$ is a Banach space.

Let $\mathcal{A} \subseteq C(K)$. \mathcal{A} is dense in $C(K)$ if: $\forall f \in C(K)$ there exist $\bar{\Phi}_k \in \mathcal{A}$ ($k=1, 2, \dots$) such that $\|\bar{\Phi}_k - f\| \rightarrow 0$ as $k \rightarrow \infty$. Equivalently, $\forall f \in C(K) \ \forall \varepsilon > 0 \ \exists \bar{\Phi} \in \mathcal{A}$ such that $\|\bar{\Phi} - f\| < \varepsilon$. Equivalently, $f \in \overline{\mathcal{A}}$, the closure of \mathcal{A} in $C(K)$.

Remark A NN function is defined on the entire space of the input space \mathbb{R}^d . If we want to approximate $f \in C(K)$, $K \subseteq \mathbb{R}^d$: compact, we can approximate f by some $\tilde{f} \in C(Q)$, $Q = [a, b]^d \supseteq K$, -as $a < b < \infty$, by using mollifiers and approximations to get $\tilde{f} \in C_c(\mathbb{R}^d)$, even that $\tilde{f} \in C_c^\infty(\mathbb{R}^d)$, $\text{supp } \tilde{f} \subseteq Q$. So, we can consider Q instead of K .

Question: If NNs approx. $C(K)$, can they approx. $C(K')$ for any different K' ? (K, K' : compact).

① $NN(\sigma, L, d, 1) = \text{all NNs with the activation function } \sigma, \text{ depth } L (\# \text{hidden layers} = L-1), \text{ input dimension } d, \text{ and output dimension } 1$.

A Universal Approximation Theorem

Theorem (Cybenko 1989, Funahashi 1989, and Hornik, Stinchcombe, and White 1989) Let $\sigma \in C(\mathbb{R})$ be such that $\sigma(-\infty) = 0$ and $\sigma(+\infty) = 1$. Then $NN(\sigma, 2, d, 1)$ is dense in $C(K)$ for any compact $K \subseteq \mathbb{R}^d$ and any $d \in \mathbb{N}$.

Remarks

○ Three papers published in the same year! Proved the same result using different methods — so there are three different methods. Note: Funahashi (1989) and Hornik, Stinchcombe, & White (1989) require σ to be increasing. Also, Carroll and Dickinson (1990) gave a constructive proof of the result using Radon transforms.

Here, we detail Cybenko's proof. We also sketch Funahashi's proof which is based on integral formula of Irie-Miyake (1988). We also briefly describe the ideas of proof by Hornik, Stinchcombe, and White.

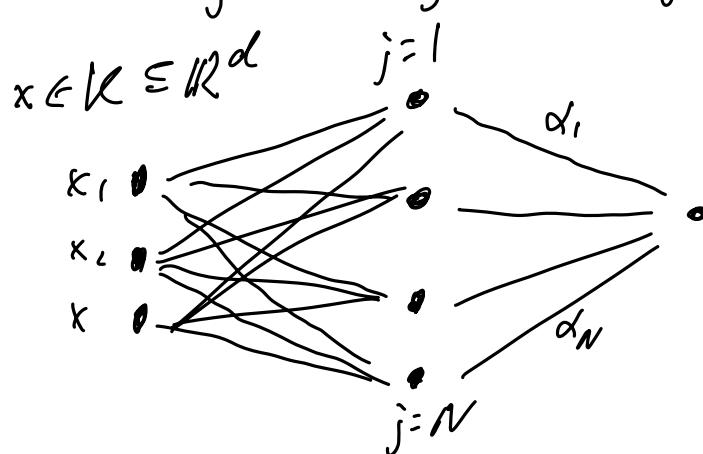
○ Note that the ReLU is not covered in this theorem.

Hornik (1991) extended the result to any $\sigma \in C(\mathbb{R})$ that is bounded and non constant. The most general result is given by Leshno, Lin, Pinkus, and Schocken (1993): $\sigma \in C(\mathbb{R})$ leads to the universal approximation property $\Leftrightarrow \sigma$ is not a polynomial. We will sketch the proof of this general result.

○ Representation of $G \in NN(\sigma, 2, d, 1)$:

$$G(x) = \sum_{j=1}^N \alpha_j \sigma(w_j^T x + \theta_j) + \beta, \quad \forall x \in \mathbb{R}^d, \quad (*)$$

where $\alpha_j \in \mathbb{R}$, $w_j \in \mathbb{R}^d$, $\theta_j \in \mathbb{R}$ ($j=1, \dots, N$), $\beta \in \mathbb{R}$.



$$l=2, N_o=d, N_i=1, W=N.$$

$$A_1 = \begin{bmatrix} w_1^T \\ \vdots \\ w_N^T \end{bmatrix}_{N \times d}, \quad b_1 = \begin{bmatrix} \theta_1 \\ \vdots \\ \theta_N \end{bmatrix}.$$

$$A_2 = [\alpha_1 \dots \alpha_N]_{1 \times N}$$

$$b_2 = \beta \in \mathbb{R}^1$$

Note: Since $\sigma \neq \text{const.}$, by choosing all $w_j=0$, we can absorb β . i.e., $NN(\sigma, 2, d, 1)$ consists of all G with $\beta=0$.

Proof (Cybenko's proof)

Step 1 $N_\sigma := \mathcal{N}(\sigma, L=2, d, 1)$ is a vector subspace of $C(K)$ by elementary constructions. (Or by direct verification.)

Step 2 If $f \in C(K) \setminus \overline{N_\sigma}$ ($\overline{N_\sigma}$ = the closure of N_σ in $C(K)$), then by the Hahn-Banach Thm, $\exists \varphi \in C(K)^*$, such that $\varphi(f) \neq 0$ and $\varphi = 0$ on $\overline{N_\sigma}$.

By Riesz's Thm, $\exists \mu \in \mathcal{M}(K)$ — the space of Radon measures on K , s.t. $\varphi(g) = \int_K g d\mu \quad \forall g \in C(K)$.

Step 3. For any $g(x) = \sigma(w^T x + \varrho)$ ($x \in \mathbb{R}^d$), where $w \in \mathbb{R}^d$, $\varrho \in \mathbb{R}$, we have $g \in \overline{N_\sigma}$, hence $\varphi(g) = 0$. i.e.,

$$\varphi(g) = \int_K g d\mu = \int_K \sigma(w^T x + \varrho) d\mu(x) = 0.$$

Extend μ to $\mathcal{M}(\mathbb{R}^d)$ (finite, signed Radon measures on \mathbb{R}^d) trivially, i.e., $\forall A \in \mathcal{B}_{\mathbb{R}^d}$ (all Borel sets of \mathbb{R}^d), $\mu(A) = \mu(A \cap K)$. Then $\mu \in \mathcal{M}_c(\mathbb{R}^d)$ (i.e., $\mu \in \mathcal{M}(\mathbb{R}^d)$ with compact support). Moreover

$$\int_{\mathbb{R}^d} \sigma(w \cdot x + \varrho) d\mu(x) = 0 \quad \forall w \in \mathbb{R}^d, \forall \varrho \in \mathbb{R}.$$

By a lemma (next lecture), the assumptions of σ lead to $\mu = 0$. Hence $\varphi(f) = \int_K f d\mu = 0$, a contradiction. (The continuity of σ is needed for $g(x) = \sigma(w^T x + \varrho)$ to be in $C(K)$ so the $\varphi(g) = \int_K g d\mu$.) QED

Remark The continuity of σ is needed in the theorem so that $\mathcal{N}(\sigma, 2, d, 1) \subseteq C(K)$.