

A STUDY OF SIMULTANEOUS APPROXIMATION BY NEURAL NETWORKS

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ABSTRACT. This paper shows the degree of simultaneous neural network approximation for a target function in $C^r[-1, 1]$ and its first derivative. We use the Jackson's theorem for differentiable functions to get a degree of approximation to a target function by algebraic polynomials and trigonometric polynomials. We also make use of the de La Vallée Poussin sum to get an approximation order by algebraic polynomials to the derivative of a target function. By showing that the divided difference with a generalized translation network can be arbitrarily closed to algebraic polynomials on $[-1, 1]$, we obtain the degree of simultaneous approximation.

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1. Introduction

This paper is related to the problem of simultaneous approximation by a three layered feedforward neural network. A three layered feed forward network has an input layer, a hidden layer and an output layer. A feedforward network with one hidden layer is of the form

$$\sum_{i=1}^n c_i \psi(a_i x + b_i) \quad (1.1)$$

where the weight a_i , the threshold b_i and c_i are real numbers for $1 \leq i \leq n$ and ψ is an activation function. For a natural number n , $\Psi_{\psi, n}$ denotes the set of all such functions.

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Most papers related to the neural network approximation investigate the density problems and complexity problem. The density problem was proved in the many papers [5, 6]. In [5], Leshno, Lin, Pinkus and Schocken proved a complete characterization of the activation function. Under some weak conditions for an activation function, any continuous functions on a compact set can be approximated arbitrarily closed by neural networks.

Recently, the complexity problem was discussed in many papers [3, 7, 11]. Those papers investigated the relation between the number of neurons and its approximation capacity.

More recently, the simultaneous approximation of a function and its derivatives was also investigated by many researchers [1, 6, 12]. In [6], Li proved that a function and its derivatives on a compact set can be simultaneously approximated by neural networks. On the other hand, Cao, Xu and Li [1] proved the degree of simultaneous approximation by neural networks in the pointwise sense, but the pointwise convergence is not so useful in the applications. In this paper, we show the degree simultaneous approximation by neural networks in the uniform sense.

2. Preliminaries

Throughout the paper, n , m , r and k are natural numbers. For each n , we denote the class of all algebraic polynomials of degree not exceeding n by Π_n and the class of all trigonometric polynomials of degree not exceeding n by Π_n^* . In each case, we define

$$E_n(f) := \inf_{P \in \Pi_n} \|f - P\|_{(\infty, [-1, 1])} \quad (2.1)$$

and

$$E_n^*(g) := \inf_{P^* \in \Pi_n^*} \|g - P^*\|_{(\infty, [-\pi, \pi])} \quad (2.2)$$

for $f \in C[-1, 1]$ and a 2π periodic function $g \in C[-\pi, \pi]$, respectively. Jackson's theorems for differentiable functions are the followings [2].

Theorem 2.1. (1) For $f \in C^r[-1, 1]$ with $\|f^{(r)}\|_{(\infty, [-1, 1])} < \infty$,

$$E_n(f) \leq \frac{C_r}{n^r}$$

where the constant C_r is independent of n .

(2) For a 2π periodic function $g \in C^r[-\pi, \pi]$ with $\|g^{(r)}\|_{(\infty, [-\pi, \pi])} < \infty$,

$$E_n^*(g) \leq \frac{C_r^*}{n^r}$$

where the constant C_r^* is independent of n .

Note that $g' \in C^{r-1}[-\pi, \pi]$ for $g \in C^r[-\pi, \pi]$ and so we get

$$E_n^*(g') \leq \frac{C_r^{**}}{n^{r-1}}. \tag{2.3}$$

For a 2π periodic function $g \in C[-\pi, \pi]$, $S_n(g, t) := S_n(t)$ denotes the n th partial sum of its Fourier series. For each n , the de La Vallée Poussin sum τ_n is defined by

$$\tau_n(g, t) := \tau_n(g) = \frac{1}{n} \sum_{i=n}^{2n-1} S_i(t).$$

Note that the de La Vallée Poussin sum τ_n of a 2π periodic function g has the following properties [8, 9].

- (1) $|\tau_n(g, t)| \leq \max |g(t)|$ for $t \in [-\pi, \pi]$.
- (2) $\|\tau_n(g) - g\|_{(\infty, [-\pi, \pi])} \leq 4E_n^*(g)$.
- (3) $\tau_n'(g, t) = \tau_n(g', t)$.

In addition, $\tau_n(T_k, t) = T_k(t)$ for any trigonometric polynomial $T_k(t)$ for $k \leq n$.

3. Main results

First of all, we get the degree of simultaneous approximation by algebraic polynomials. The basic idea comes from Theorem 2.1.

Theorem 3.1. *Assume that $f \in C^r[-1, 1]$ with $\|f^{(r)}\|_{(\infty, [-1, 1])} < \infty$. For each n , there is an algebraic polynomial $P_{2n-1} \in \Pi_{2n-1}$ such that*

$$\|f - P_{2n-1}\|_{(\infty, [-1, 1])} \leq \frac{C^*}{n^r} \quad \text{and} \quad \|f' - P'_{2n-1}\|_{(\infty, [-1, 1])} \leq \frac{C^{**}}{n^{r-1}} \tag{3.1}$$

where C^* and C^{**} are the positive constants which are independent of n .

Proof. In [9], there exists an extension $\tilde{f} \in C^r$ of f on $[-2, 2]$ such that $\tilde{f}(x) = f(x)$ for $x \in [-1, 1]$. For $x \in [-2, 2]$, we define

$$g(t) := \tilde{f}(2 \cos t) = \tilde{f}(x).$$

Then $g \in C^r[-\pi, \pi]$. From the elementary differentiation

$$\tilde{f}'(x) = \frac{d\tilde{f}(x)}{dx} = \frac{d\tilde{f}(2 \cos t)}{dt} \cdot \frac{dt}{dx} = g'(t) \cdot \frac{1}{-2 \sin t},$$

we obtain, for some C ,

$$\|\tilde{f}'\|_{(\infty,[-1,1])} = \|f'\|_{(\infty,[-1,1])} = \left\| g' \cdot \frac{1}{-2 \sin t} \right\|_{(\infty, [\frac{\pi}{3}, \frac{2\pi}{3}])} \leq C \|g'\|_{(\infty, [-\pi, \pi])}. \quad (3.2)$$

By (2.3) and the properties (2) and (3) of the de La Vallée Poussin sum, we have

$$\begin{aligned} \|g' - \tau'_n(g)\|_{(\infty, [-\pi, \pi])} &= \|g' - \tau_n(g')\|_{(\infty, [-\pi, \pi])} \\ &\leq C_1 E_n^*(g') \\ &\leq \frac{C^{**}}{n^{r-1}}. \end{aligned} \quad (3.3)$$

We define the algebraic polynomial P_{2n-1} by

$$P_{2m-1}(\tilde{f}, x) := P(\tilde{f}, 2 \cos t) = \tau_n(g, t).$$

Then $P_{2n-1} \in \Pi_{2n-1}$. From Theorem 2.1 and the property (2) of the de La Vallée Poussin sum,

$$\begin{aligned} \|f - P_{2n-1}\|_{(\infty, [-1, 1])} &\leq \|\tilde{f} - P_{2n-1}\|_{(\infty, [-2, 2])} \\ &= \|g - \tau_n(g)\|_{(\infty, [-\pi, \pi])} \\ &\leq C_2 E_n^*(g) \\ &\leq \frac{C^*}{n^r}. \end{aligned} \quad (3.4)$$

In addition, we get the following result from (3.3).

$$\begin{aligned} \|f' - P'_{2n-1}\|_{(\infty, [-1, 1])} &\leq \|g' - \tau'_n(g)\|_{(\infty, [-\pi, \pi])} \\ &\leq C_3 E_n^*(g') \\ &\leq \frac{C^{**}}{n^{r-1}}. \end{aligned} \quad (3.5)$$

This completes the proof. \square

Now, we show that any polynomials can be simultaneously approximated by neural networks.

Theorem 3.2. *Suppose that $\psi \in C^\infty(\mathbb{R})$ and there exists $b \in \mathbb{R}$ in some open interval $(b - \delta, b + \delta)$ in \mathbb{R} such that*

$$\psi^{(k)}(b) \neq 0$$

for any k . For $f \in C^r[-1, 1]$ with $\|f^{(r)}\|_{(\infty, [-1, 1])} < \infty$ and a natural number k , there exists a neural network $N_{k,h} \in \Psi_{\phi,k}$ such that

$$\|x^k - N_{k,h}\|_{(\infty, [-1, 1])} \leq M^* \cdot h \quad \text{and} \quad \|(x^k)' - N'_{k,h}\|_{(\infty, [-1, 1])} \leq M^{**} \cdot h \quad (3.6)$$

where M^* and M^{**} are positive constants.

Proof. We follow the idea in [7]. For each natural number k , the neural network

$$N_{k,h}(\phi, x) := \frac{1}{\phi^{(k)}(b)} \frac{1}{h^k} \sum_{j=0}^k (-1)^{(k-j)} \binom{k}{j} \phi(hjx + b) \quad (3.7)$$

represents a divided difference for x^k . Thus, as it is proved in [7],

$$\|G_{k,h}(\phi, x) - x^k\|_{(\infty, [-1, 1])} \leq M^* \cdot h \quad (3.8)$$

for some positive constant M^* . Note that

$$\begin{aligned} N'_{k,h}(\phi, x) &= \frac{1}{\phi^{(k)}(b)} \frac{1}{h^k} \sum_{j=0}^k (-1)^{(k-j)} \binom{k}{j} \phi'(hjx + b)hj \\ &= \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^k (-1)^{(k-j)} \binom{k-1}{j-1} \phi'(hjx + b) \\ &= \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \phi'(h(j+1)x + b). \end{aligned} \quad (3.9)$$

Using Taylor theorem for an integer m with $m > k - 1$, the equation (3.9) can be written as

$$\begin{aligned} \phi'(h(j+1)x + b) &= \phi'(hjx + b) + \sum_{i=1}^{m-1} \frac{\phi^{(i+1)}(hjx + b)}{i!} (hx)^i \\ &\quad + \frac{\phi^{(m+1)}(hjx + b + \xi)}{m!} (hx)^m \end{aligned} \quad (3.10)$$

where ξ is a value between $hjx + b$ and $h(j+1)x + b$. Thus, the equation (3.9) can be rewritten as

$$\begin{aligned} N'_{k,h}(\phi, x) &= \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \phi'(hjx + b) \\ &= \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \left(\sum_{i=1}^{m-1} \frac{\phi^{(i+1)}(hjx + b)}{i!} (hx)^i \right) \\ &= \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \left(\frac{\phi^{(m+1)}(hjx + b + \xi)}{m!} (hx)^m \right). \end{aligned} \quad (3.11)$$

The second part of (3.11) is

$$\frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \phi'(h j x + b) = k \cdot N_{k-1,h}(\phi', x). \quad (3.12)$$

The third part of (3.11) is

$$\begin{aligned} & \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \left(\sum_{i=1}^{m-1} \frac{\phi^{(i+1)}(h j x + b)}{i!} (h x)^i \right) \\ &= \sum_{i=1}^{m-1} h^i \left(\frac{1}{\phi^{(k)}(b)} \cdot \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} \binom{k-1}{j} (-1)^{(k-j-1)} \phi^{(i+1)}(h j x + b) \right) \\ &= \sum_{i=1}^{m-1} h^i (k x^{k-1} + M_1 \cdot h) \\ &\leq M_2 \cdot h \end{aligned} \quad (3.13)$$

where M_1 and M_2 are positive constants.

The last part of (3.11) is

$$\begin{aligned} & \frac{1}{\phi^{(k)}(b)} \frac{k}{h^{k-1}} \sum_{j=0}^{k-1} (-1)^{(k-j-1)} \binom{k-1}{j} \left(\frac{\phi^{(m+1)}(h j x + b + \xi)}{m!} (h x)^m \right) \\ &\leq M_3 \cdot h \end{aligned} \quad (3.14)$$

since $m - k + 1 > 0$. From (3.12), (3.13) and (3.14), we have

$$\| (x^k)' - N'_{k,h} \|_{(\infty, [-1,1])} \leq M^{**} \cdot h. \quad (3.15)$$

By (3.8) and (3.15), we completes the proof. \square

The following is the main theorem of this paper.

Theorem 3.3. *Suppose that $\psi \in C^\infty(\mathbb{R})$ and there exists $b \in \mathbb{R}$ in some open interval $(b - \delta, b + \delta)$ in \mathbb{R} such that*

$$\psi^{(k)}(b) \neq 0$$

for any k . For $f \in C^r[-1, 1]$ with $\|f^{(r)}\|_{(\infty, [-1,1])} < \infty$ and a natural number n , there exists a neural network $N_{n,h} \in \Psi_{\phi,n}$ such that

$$\|f - N_{n,h}\|_{(\infty, [-1,1])} \leq \frac{D_1}{n^r} \quad \text{and} \quad \|f' - N'_{k,h}\|_{(\infty, [-1,1])} \leq \frac{D_2}{n^{r-1}} \quad (3.16)$$

where D_1 and D_2 are positive constants which are independent of n .

Proof. Let $n \geq 2$ be given. We choose the largest integer m such that $2m \leq n$. By Theorem 3.1, there exists an algebraic polynomial $P_{2m-1} \in \Pi_{2m-1}$ such that

$$\|f - P_{2m-1}\|_{(\infty,[-1,1])} \leq \frac{C^*}{m^r} \quad \text{and} \quad \|f' - P'_{2m-1}\|_{(\infty,[-1,1])} \leq \frac{C^{**}}{m^{r-1}}$$

where C^* and C^{**} are constants which are independent of n . We set $P_{2m-1}(x) = \sum_{j=0}^{2m-1} a_j x^j$. For each j with $0 \leq j \leq 2m - 1$, we choose h_j so small that

$$\|x^j - N_{j,h_j}\|_{(\infty,[-1,1])} \leq \frac{E_1}{\sum |a_j| m^r} \quad \text{and} \quad \|(x^j)' - N'_{j,h_j}\|_{(\infty,[-1,1])} \leq \frac{E_2}{\sum |a_j| m^{r-1}}$$

where E_1 and E_2 are constants which are independent of m . Now, we set

$$N_n = \sum_{j=0}^{2m-1} N_{j,h_j} \in \Psi_{\psi,2m} \subset \Psi_{\psi,n}.$$

Since $\frac{1}{n} \leq \frac{1}{2} \cdot \frac{1}{m} \leq \frac{1}{n-1} \leq \frac{2}{n}$ for $n \geq 2$, $\frac{1}{m^r} \leq \frac{4^r}{n^r}$. Therefore, from (3.4) and (3.8), we have

$$\begin{aligned} \|f - N_n\|_{(\infty,[-1,1])} &\leq \|f - P_{2m-1}\|_{(\infty,[-1,1])} + \|P_{2m-1} - N_n\|_{(\infty,[-1,1])} \\ &\leq \frac{C^*}{m^r} + \frac{E_1}{m^r} \leq \frac{D_1}{n^r} \end{aligned}$$

and also have

$$\begin{aligned} \|f' - N'_n\|_{(\infty,[-1,1])} &\leq \|f' - P'_{2m-1}\|_{(\infty,[-1,1])} + \|P'_{2m-1} - N'_n\|_{(\infty,[-1,1])} \\ &\leq \frac{C^{**}}{m^{r-1}} + \frac{E_2}{m^{r-1}} \leq \frac{D_2}{n^{r-1}} \end{aligned}$$

from (3.5) and (3.15). This completes the proof. □

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