

# Neural network is universal approximator

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Plan:

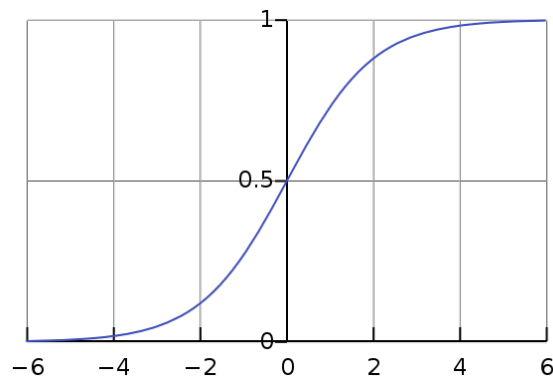
- wstęp o sigmoidzie wraz z grafik
- definicje i twierdzenia (hahn-banach, iesz)
- twierdzenie i dowód o gęstości kombinacji liniowej sigmoid
- dowód graficzny
- cytowania

## 0.0.1

Neural networks with sigmoidal activation functions can approximate to arbitrary accuracy any functional continuous mapping from one finite-dimensional space to another, provided the number  $N$  of hidden units is sufficiently large.

wiki: "A sigmoid function is a mathematical function having a characteristic "S"-shaped curve or sigmoid curve. Often, sigmoid function refers to the special case of the logistic function defined by the formula"

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$



Let  $I_n$  denote the  $n$ -dimensional unit cube,  $[0, 1]^n$ . The space of continuous functions on  $I_n$  is denoted by  $C(I_n)$  and we use  $\|f\|$  to denote the supremum norm of an  $f \in C(I_n)$ . The space of finite, signed regular Borel measures on  $I_n$  is denoted by  $M(I_n)$ .

**Definition 0.1.** We say that  $\sigma$  is sigmoidal if

$$\sigma(x) \rightarrow \begin{cases} 1 & \text{as } x \rightarrow +\infty \\ 0 & \text{as } x \rightarrow -\infty \end{cases}$$

**Definition 0.2.** We say that  $\sigma$  is discriminatory if for a measure  $\mu \in M(I_n)$

$$\int_{I_n} \sigma(y^T x + \theta) d\mu(x) = 0$$

for all  $y \in \mathbf{R}$  and  $\theta \in \mathbf{R}$  implies that  $\mu = 0$ .

Hahn-Banach theorem shows how to extend linear functionals from subspaces to whole spaces. Moreover, we can do it in a way that respects the boundedness properties of the given functional. The most general formulation of the theorem requires a preparation

**Definition 0.3.** A sublinear functional is a function  $f : V \rightarrow \mathbf{R}$  on a vector space  $V$  which satisfies subadditivity (1) and positive homogeneity conditions (2)

$$f(x + y) \leq f(x) + f(y) \quad \forall x, y \in V \quad (1)$$

$$f(\alpha x) = \alpha f(x) \quad \forall \alpha \geq 0, x \in V \quad (2)$$

**Theorem 0.1** (Hahn-Banach theorem for real vector spaces). *If  $p : V \rightarrow \mathbf{R}$  is a sublinear function, and  $\psi : U \rightarrow \mathbf{R}$  is a linear functional on a linear subspace  $U \subset V$ , and satisfying  $\psi(x) \leq p(x) \forall x \in U$ . Then there exists a linear extension  $\Psi : V \rightarrow \mathbf{R}$  of  $\psi$  to the whole space  $V$ , such that*

- $\Psi(x) = \psi(x) \forall x \in U$
- $\Psi(x) \leq p(x) \forall x \in V$

*Rudin 1991, Th 3.2*

**Theorem 0.2** (Riesz representation theorem). *Let  $H$  be a Hilbert space over  $\mathbf{R}$ , and  $T$  a bounded linear functional on  $H$ . If  $T$  is a bounded linear functional on a Hilbert space  $H$  then there exist some  $g \in H$  such that for every  $f \in H$  we have (<http://www.math.jhu.edu/~lindblad/632/riesz.pdf>)*

$$T(f) = \langle f, g \rangle \quad \forall f \in H$$

*Any bounded linear functional  $T$  on the space of compactly supported continuous functions on  $X$  is the same as integration against a measure  $\mu$ . (<http://mathworld.wolfram.com/RieszRepresentationTheorem.html>)*

$$Tf = \int f d\mu$$

**Theorem 0.3.** *Let  $\sigma$  be any continuous discriminatory function. Then finite sums of the form*

$$G(x) = \sum_{j=1}^N \alpha_j \sigma(y_j^T x + \theta_j)$$

*are dense in  $C(I_n)$ . In other words, given any  $f \in C(I_n)$  and  $\epsilon > 0$ , there is a sum,  $G(x)$ , of the above form, for which*

$$|G(x) - f(x)| < \epsilon \quad \forall x \in I_n$$

*Proof.* Let  $S \subset C(I_n)$  be the set of functions of the form  $G(x)$ . Clearly  $S$  is a linear subspace of  $C(I_n)$ . We claim that the closure of  $S$  is all of  $C(I_n)$ .

Assume that closure of  $S$  is not all of  $C(I_n)$ . Then the closure of  $S$ , say  $R$ , is a closed proper subspace of  $C(I_n)$ . By the Hahn-Banach theorem, there is a bounded linear functional on  $C(I_n)$ , call it  $L$ , with the property that  $L \neq 0$  but  $L(R) = L(S) = 0$ .

By the Riesz Representation Theorem, this bounded linear functional,  $L$ , is of the form

$$L(h) = \int_{I_n} h(x) d\mu(x)$$

for some  $\mu \in M(I_n)$ , for all  $h \in C(I_n)$ . In particular, since  $\sigma(y^T x + \theta)$  is in  $R$  for all  $y$  and  $\theta$ , we must have that

$$\int_{I_n} \sigma(y^T x + \theta) d\mu(x) = 0$$

for all  $y$  and  $\theta$ .

However, we assumed that  $\sigma$  was discriminatory so that this condition implies that  $\mu = 0$  contradicting our assumption. Hence, the subspace  $S$  must be dense in  $C(I_n)$ .

This demonstrates that sums of the form

$$G(x) = \sum_{j=1}^N \alpha_j \sigma(y_j^T x + \theta_j)$$

are dense in  $C(I_n)$  providing that  $\sigma$  is continuous and discriminatory.  $\square$

## 0.1 visual proof