

# Approximation by Superpositions of a Sigmoidal Function

M. Ruby, C. Salinas

Department of Mathematics  
Purdue University

Purdue University Machine Learning Seminar, 2019

# Introduction

In many Machine Learning applications the goal is to approximate a continuous function  $f: \mathbf{R}^n \rightarrow \mathbf{R}$  by an expression of the form

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

where  $y_j \in \mathbf{R}^n$  and  $\alpha_j, \theta_j \in \mathbf{R}$  are fixed.

In Machine Learning terminology, the  $y$  represent weights and the  $\theta$  biases. Equation  $(\star)$  models a fully connected network with one hidden layer.

# The Main Result

The author demonstrates that, theoretically, such an architecture is sufficient to approximate all continuous functions on the unit cube. That is,

## Theorem 1 (Theorem 2 in [C2])

*For  $\sigma$  a continuous sigmoidal function, finite sums of the form*

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

*are dense in  $C(I_n)$ .*

# Sigmoidal Functions are Universal Approximators

The proof will rely on the following more general result

## Theorem 2 (Theorem 1 in [C2])

*Let  $\sigma$  be continuous and discriminatory. Then the set of finite sums of the form*

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

*is dense in  $C(I_n)$ .*

With this theorem at our disposal we need only show that sigmoidal functions are discriminatory; that is, we will show that

## Lemma 3 (Lemma 1 in [C2])

*Any continuous sigmoidal function is discriminatory.*

# A Corollary of the Hahn–Banach Theorem

To prove Theorem 2 we will need a little bit of Functional Analysis. In particular, we will need the following corollary of the Hahn–Banach Theorem

## Corollary 4

*If  $V$  is a normed vector space,  $W$  a subspace of  $V$ , and  $v_0 \in V \setminus W$  with  $\text{dist}(v_0, W)$  nonzero, then there exists a bounded linear functional  $L$  such that  $L(v_0) = 1$ , but  $L(v) = 0$  for all  $v \in W$ .*

## Proof.

The proof follows from the Hahn–Banach Theorem applied to the quotient  $V/W$ . For details, see Corollary 6.8 in [C1], p. 79.  $\square$

# The Riesz Representation Theorem

We will also be needing use of the Riesz Representation Theorem

## Theorem 5 (Riesz Representation Theorem)

*If  $X$  is locally compact and  $\mu$  a measure on  $X$ , the map*

$$\mu \longmapsto \int_X f \, d\mu$$

*is an isometric isomorphism from the space of measures on  $X$ ,  $M(X)$ , to the dual of  $C_0(X)$ .*

## Proof.

See [C1], Theorem 5.7. □

What the Riesz Representation Theorem is saying is that every bounded linear functional  $L$  on  $C_0(X)$  is of the form  $L(f) = \int_X f \, d\mu$  for some measure  $\mu$  on  $X$ .

# Discriminatory and Sigmoidal Functions

Let's define a couple of these terms.

## Definition 6 (Discriminatory functions)

We say  $\sigma$  is *discriminatory* if given a measure  $\mu$  on  $I_n$

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

for all  $y \in \mathbf{R}^n$  implies  $\mu = 0$ .

And of course.

## Definition 7 (Sigmoidal functions)

We say that  $\sigma$  is sigmoidal if

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

# Proof of Theorem 1, Part 1

- ▶ Let  $S = \left\{ \sum_{j=1}^N \alpha_j \sigma(y_j^T x + \theta_j) \right\} \subset C(I_n)$ . Clearly  $S$  is a subspace of  $C(I_n)$ .
- ▶ The author proceeds by showing that the closure,  $R$ , of  $S$  is  $C(I_n)$ :
- ▶ By Corollary 4, there exists a functional  $L$  which is zero on  $S$  but nonzero on  $C(I_n) \setminus S$ .
- ▶ By the Riesz Representation Theorem,  $L$  is of the form

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

for some  $\mu$  measure on  $I_n$ .



# Proof of Theorem 1, Part 2

- ▶ Since  $\sigma(y^T x + \theta)$  is in  $R$  for all  $y, \theta$ ,

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

for all  $y, \theta$ .

- ▶ However, since  $\sigma$  was assumed to be discriminatory,  $\mu = 0$ .
- ▶ This contradicts assumption that  $L$  is nonzero.
- ▶ Therefore,  $R = C(I_n)$ .

# Sigmoidal Functions are Discriminatory

Next the author shows that the sigmoidal functions that we care about are in fact discriminatory. More generally, he shows that

## Lemma 8

*Any bounded, measurable sigmoidal function is discriminatory. In particular, any continuous sigmoidal function is discriminatory.*

# Proof of Lemma 3, Part 1

- First note that

$$\sigma(\lambda(y^T x + \theta) + \phi) \begin{cases} \rightarrow 1 & \text{for } y^T x + \theta > 0 \text{ as } \lambda \rightarrow +\infty, \\ \rightarrow 0 & \text{for } y^T x + \theta < 0 \text{ as } \lambda \rightarrow +\infty, \\ = \sigma(\phi) & \text{for } y^T x + \theta = 0 \text{ for all } \lambda. \end{cases}$$

- So  $\sigma_\lambda(x) := \sigma(\lambda(y^T x + \theta) + \phi)$  converges pointwise and boundedly to the function

$$\gamma(x) = \begin{cases} 1 & \text{for } y^T x + \theta > 0, \\ 0 & \text{for } y^T x + \theta < 0, \\ \sigma(\phi) & \text{for } y^T x + \theta = 0, \end{cases}$$

as  $\lambda \rightarrow \infty$ .

## Proof of Lemma 3, Part 2

► Let

$$\Pi_{y,\theta} = \{x: y^T x + \theta = 0\},$$

$$H_{y,\theta} = \{x: y^T x + \theta > 0\}.$$

► By the Lebesgue Bounded Convergence Theorem,

$$\begin{aligned} 0 &= \int_{I_n} \sigma_\lambda(x) d\mu(x) \\ &= \int_{I_n} \gamma(x) d\mu(x) \\ &= \sigma(\phi) \mu(\Pi_{y,\theta}) + \mu(H_{y,\theta}) \end{aligned}$$

for all  $\phi, \theta, y$ .

## Proof of Lemma 3, Part 3

- ▶ The author next shows that if  $\mu(H_{y,\theta}) = 0$  for all  $y, \theta$  then  $\mu = 0$ .
- ▶ Fix  $y$  and define  $F$  by

$$F(h) = \int_{I_n} h(y^T x) d\mu(x).$$

Note that  $F$  is a bounded functional on  $L^\infty(\mathbf{R})$  since  $\mu$  is a finite (signed) measure.

- ▶ Let  $h = \chi_{[\theta, \infty)}$ . Then

$$F(h) = \int_{I_n} \chi_{[\theta, \infty)}(y^T x) = \mu(\Pi_{y, -\theta}) + \mu(H_{y, -\theta}) = 0.$$

- ▶ By linearity  $L(h) = 0$  for  $h = \chi_I$  for  $I$  any interval in  $\mathbf{R}$ .

## Proof of Lemma 3, Part 4

- ▶ Since the simple functions are dense in  $L^\infty(\mathbf{R})$ ,  $F = 0$ .
- ▶ Let  $s(u) = \sin(mu)$  and  $c(u) = \cos(mu)$ .
- ▶ Then

$$\begin{aligned} L(c(m^T x) + is(m^T x)) &= \int_{I_n} \cos(m^T x) + i \sin(m^T x) d\mu(x) \\ &= \int_{I_n} \exp(im^T x) d\mu(x) \\ &= 0 \end{aligned}$$

for all  $m$ .

- ▶ Since the Fourier transform of  $\mu$  is 0,  $\mu$  itself must be 0. For further details, consult [R], Ch. 7.
- ▶ Therefore,  $\sigma$  is discriminatory.

# Conclusion

By Lemma 3 and Theorem 2, the main result follows

## Theorem 1 (Theorem 2 in [C2])

*For  $\sigma$  a continuous sigmoidal function, finite sums of the form*

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

*are dense in  $C(I_n)$ .*

# References



J.B. Conway. *A Course in Functional Analysis*. Graduate Texts in Mathematics. Springer New York, 2013. ISBN: 9781475738285. URL: <https://books.google.com/books?id=ccEGCAAAQBAJ>.



G. Cybenko. *Approximations by Superpositions of a Sigmoidal Function*. Report. University of Illinois at Urbana-Champaign, Center for Supercomputing Research and Development, 1989. URL: <https://books.google.com/books?id=Xm3HtgAACAAJ>.



W. Rudin. *Functional Analysis*. International series in pure and applied mathematics. McGraw-Hill, 1991. ISBN: 9780070542365. URL: [https://books.google.com/books?id=Sh%5C\\_vAAAAAAAJ](https://books.google.com/books?id=Sh%5C_vAAAAAAAJ).