Tutorial 2

Overview

- Matrix and Vector Product
- Matrix Dimensions
- Perceptron and XOR
- Activation Functions
- Loss Functions
- Partial Derivatives

Matrix and Vector Product

Can you multiply these?

$$W = \begin{bmatrix} 3 & 4 \\ 1 & 2 \\ 5 & 6 \end{bmatrix} \qquad x = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

What are their dimensions?

 $W \sim 3 \times 2$

 $x \sim 2 \times 1$

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 $x \sim 2 \times 1$

Wx ~?

What are their dimensions?

 $W \sim 3 \times 2$

 $x \sim 2 \times 1$

 $Wx \sim 3 \times 1$

Can you multiply them?

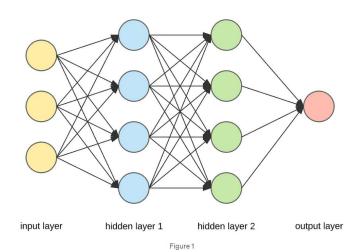
$$W = \begin{bmatrix} 3 & 4 & 5 & 1 \\ 1 & 7 & 7 & 3 \end{bmatrix} \qquad x = \begin{bmatrix} 3 \\ 5 \\ 4 \end{bmatrix}$$

In other words: addition of columns of W with weights of x

$$Wx = 3\begin{bmatrix} 3\\1 \end{bmatrix} + 5\begin{bmatrix} 4\\7 \end{bmatrix} + 4\begin{bmatrix} 5\\7 \end{bmatrix} + ??\begin{bmatrix} 1\\3 \end{bmatrix}$$

Matrix Dimensions

Describe the NN



Given:

- Input layer and hidden layers are row vectors (a horizontal vectors)
- Every node is calculated like this:

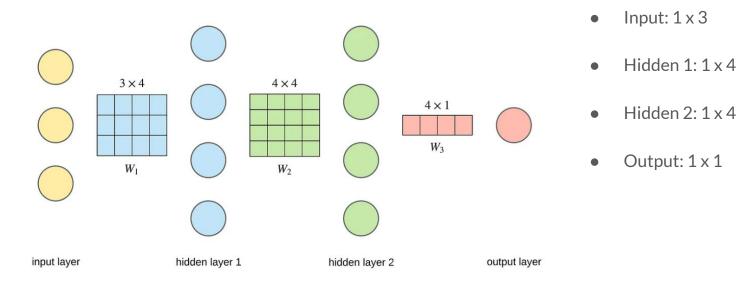
$$z = f(x \cdot w) = f\left(\sum_{i=1}^{n} x_i w_i\right)$$

Solve:

- Define the dimensions of all layers
- Define the dimensions of the weight matrices

https://towardsdatascience.com/applied-deep-learning-part-1-artificial-neural-networks-d7834f67a4f6#106c

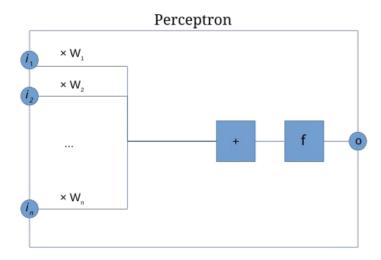
Describe the NN: Solution



https://towardsdatascience.com/applied-deep-learning-part-1-artificial-neural-networks-d7834f67a4f6#106c

Perceptron and XOR

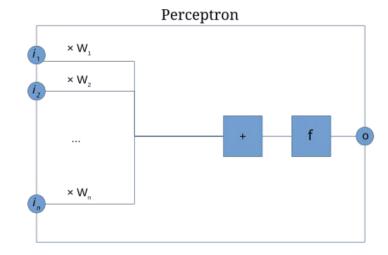
What is the perceptron?



Source: https://en.wikipedia.org/wiki/Perceptron

What is the perceptron?

$$f(x) = h(Wx + b)$$



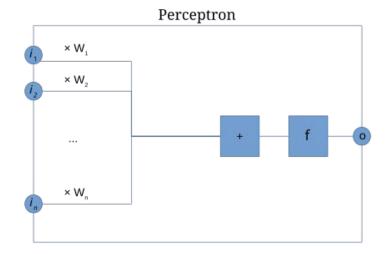
Source: https://en.wikipedia.org/wiki/Perceptron

What is the perceptron?

$$f(x) = h(Wx + b)$$

Where h = Heaviside step function

$$H(x):= \left\{egin{array}{ll} 1, & x\geq 0 \ 0, & x<0 \end{array}
ight.$$



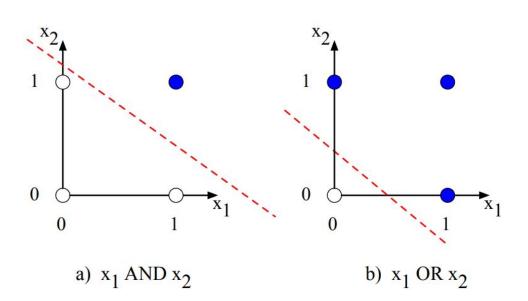
Source: https://en.wikipedia.org/wiki/Perceptron

Logical operators

AND			OR			XOR		
x 1	x2	у	x1	x2	у	x1	x2	у
0	0	0	0	0	0	0	0	0
0	1	0	0	1	1	0	1	1
1	0	0	1	0	1	1	0	1
1	1	1	1	1	1	1	1	0

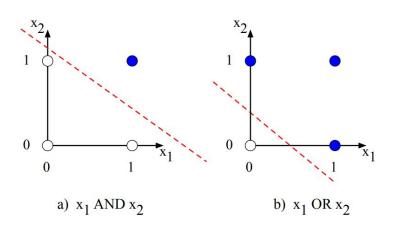
Jurafsky & Martin, chapter 7: https://web.stanford.edu/~jurafsky/slp3/7.pdf

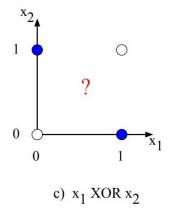
AND, OR



Jurafsky & Martin, chapter 7: https://web.stanford.edu/~jurafsky/slp3/7.pdf

AND, OR, XOR

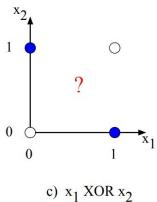




Jurafsky & Martin, chapter 7: https://web.stanford.edu/~jurafsky/slp3/7.pdf

AND, OR, XOR



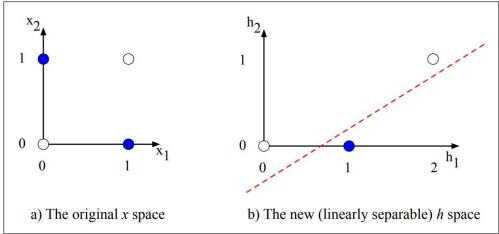


Jurafsky & Martin, chapter 7: https://web.stanford.edu/~jurafsky/slp 3/7.pdf

How do you solve the XOR problem?

How do you solve the XOR problem?

Nonlinearity + Hidden layers!



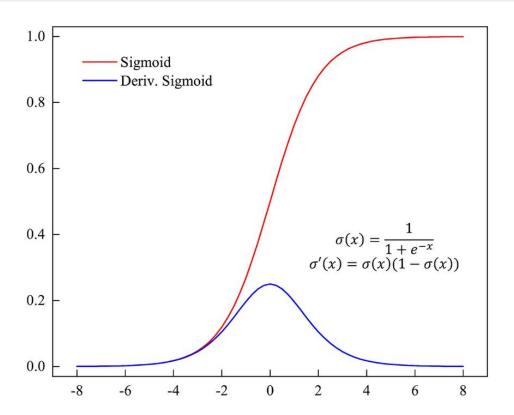
Jurafsky & Martin, chapter 7: https://web.stanford.edu/~jurafsky/slp3/7.pdf

Figure 7.7 The hidden layer forming a new representation of the input. (b) shows the representation of the hidden layer, \mathbf{h} , compared to the original input representation \mathbf{x} in (a). Notice that the input point [0, 1] has been collapsed with the input point [1, 0], making it possible to linearly separate the positive and negative cases of XOR. After Goodfellow et al. (2016).

Activation functions

Sigmoid

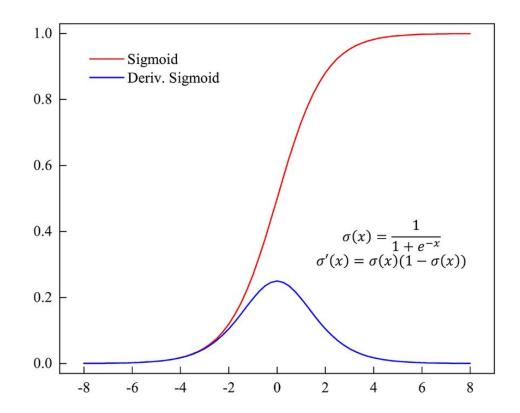
Properties of Sigmoid



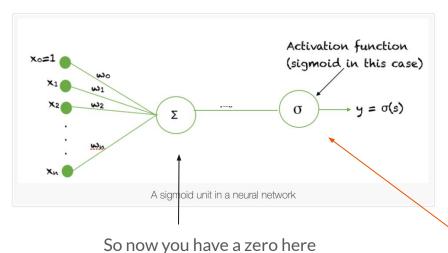
Sigmoid

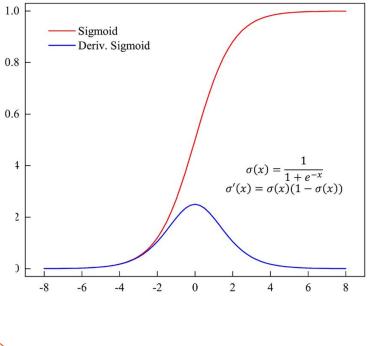
Properties of Sigmoid

- Range (0, 1)
- Linear around 0
- Outliers squashed to 0 and 1
- ...



Sigmoid: calculation





Your goal: calculate this!

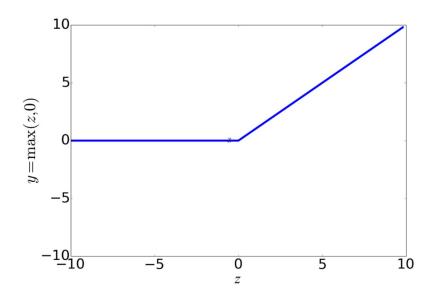
image source(left): https://machinelearningmastery.com/a-gentle-introduction-to-sigmoid-function/

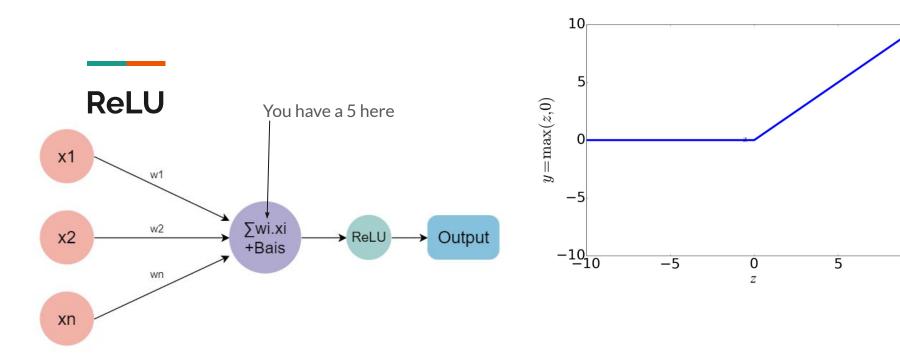
ReLU

Properties of ReLU

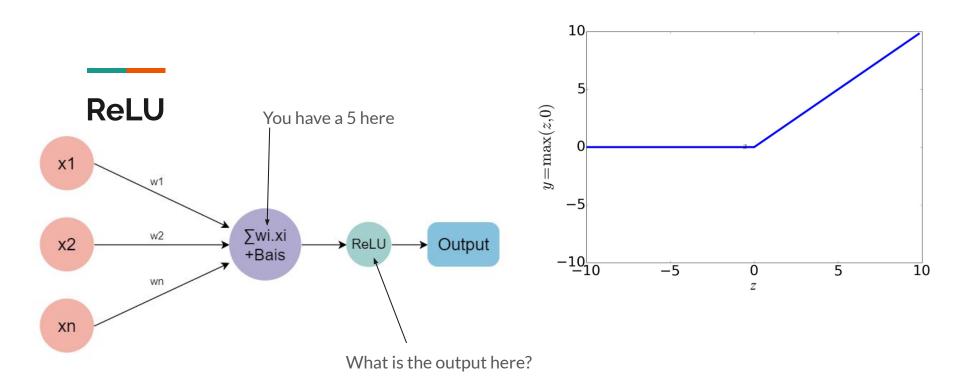
Range [0, +inf)

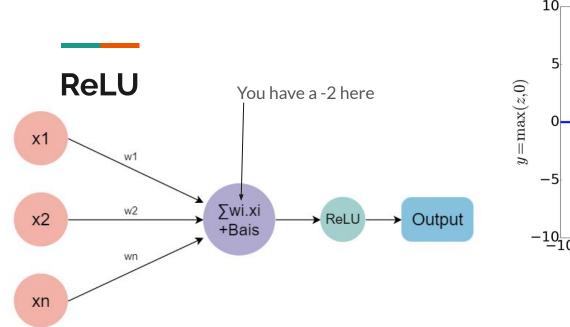
ReLU: Rectified Linear Unit

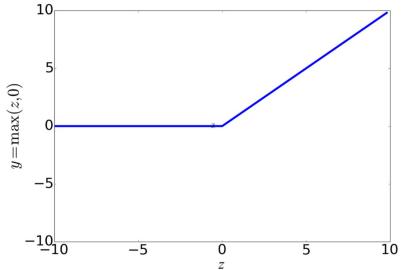


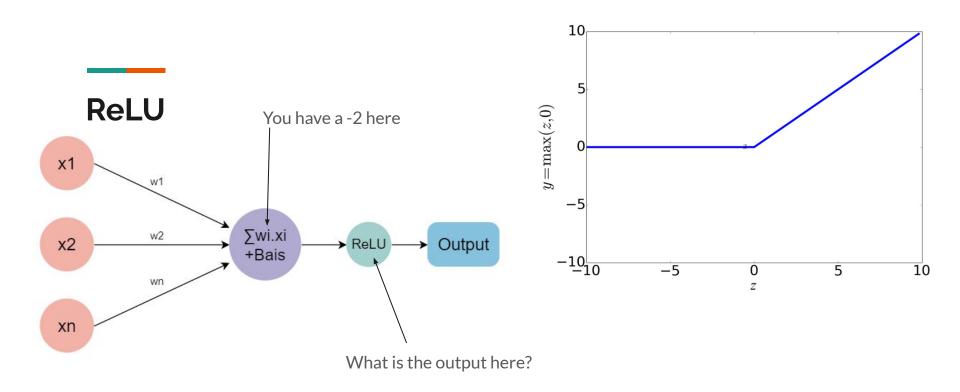


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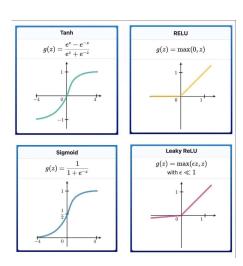








More



And a lot more...

Loss function

Why is it needed?

The training procedure:

- we use our model to make predictions on some input data,
- compare those predictions to the actual observed outcomes (or labels, targets, ground truths)
- use a loss (or cost) function to measure how "wrong" our model's predictions are
- parameter update: Based on the loss, we adjust the model's parameters to try to reduce this loss in future predictions
- repeat the process for certain repetition / some result has been achieved / model no longer improves

Cross-Entropy Loss

For binary classification
$$L_{CE}(\hat{y}, y) = -\log p(y|x) = -[y\log \hat{y} + (1-y)\log(1-\hat{y})]$$

More: https://ml-cheatsheet.readthedocs.io/en/latest/loss_functions.html

Calculation exercise

For binary classification
$$L_{CE}(\hat{y},y) = -\log p(y|x) = \left[- \left[y \log \hat{y} + (1-y) \log (1-\hat{y}) \right] \right]$$

Now your label is 1, and your prediction is 0.9, what should the above formula be?

(just plug in the numbers)

Cross-Entropy Loss

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Cross-Entropy Loss

For binary classification
$$L_{CE}(\hat{y},y) = -\log p(y|x) = \left| -[y\log \hat{y} + (1-y)\log(1-\hat{y})] \right|$$

Now your label is 1, and your prediction is 0.9, what should the above formula be?

$$-1*log0.9 + (1-1)*log(1-0.9) = -1*log0.9 + 0 = -log0.9$$

Partial Derivatives

Calculate the following derivatives...

$$f(x) = 3x^2$$

 $f(x) = ax^2$, where a is a constant

f(y) = ay, where a is a constant

 $f(y) = a^2y$, where a is a constant

Calculate the following derivatives...

$$f(x) = 3x^2$$

$$f(x) = ax^2$$
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$$f(y) = ay$$
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$$f(y) = a^2y$$
, where a is a constant

$$f'(x) = 3.2x = 6x$$

$$f'(x) = a \cdot 2x = 2ax$$

$$f'(y) = a$$

$$f'(y) = a^2$$

Calculate the following derivatives...

$$f(x) = 3x^2$$

$$f(x) = ax^2$$
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$$f(y) = ay$$
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$$f(y) = a^2y$$
, where a is a constant

$$f(x, y) = 3x^2y, w.r.t. x and y$$

$$f'(x) = 3.2x = 6x$$

$$f'(x) = a \cdot 2x = 2ax$$

$$f'(y) = a$$

$$f'(y) = a^2$$

3

Partial derivatives

How to write the partial derivatives for: $f(x, y) = 3x^2y$

The partial derivative with respect to x

The partial derivative with respect to y

Partial derivatives

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is written $\frac{\partial}{\partial x} 3x^2y$ The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$

$$\frac{\partial}{\partial x} 3x^2 y$$

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is written $\frac{\partial}{\partial x} 3x^2y$ What are the constants? The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$

$$\frac{O}{2}3x^2y$$
 What are the const

$$\frac{\partial}{\partial u} 3x^2 y$$

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is written $\frac{\partial}{\partial x} 3x^2y$.

The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$.

$$\frac{\partial}{\partial x} 3x^2 y \blacktriangleleft$$

 $\frac{\partial}{\partial x} 3x^2 y = 3y \frac{\partial}{\partial x} x^2$

What are the constants?

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is written $\frac{\partial}{\partial x} 3x^2y$

The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$

$$\frac{\partial}{\partial x} 3x^{2}y \longrightarrow \frac{\partial}{\partial x} 3x^{2}y = 3y \frac{\partial}{\partial x} x^{2}$$

$$\frac{\partial}{\partial y} 3x^{2}y = 3y \frac{\partial}{\partial x} x^{2} = 3y \times 2x = 6xy$$

$$f(x, y) = 3x^2y$$

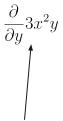
The partial derivative with respect to x is 6xy

The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$



What are the constants here?

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$



What are the constants here? $3, x^2$

$$\frac{\partial}{\partial y} 3x^2 y = 3x^2 \frac{\partial}{\partial y} y$$

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

The partial derivative with respect to y is written $\frac{\partial}{\partial y} 3x^2y$ $\frac{\partial}{\partial y} 3x^2y = 3x^2\frac{\partial}{\partial y}y$ \downarrow $\frac{\partial}{\partial y} 3x^2y = 3x^2\frac{\partial}{\partial y}y = 3x^2$

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

The partial derivative with respect to y is $3x^2$

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

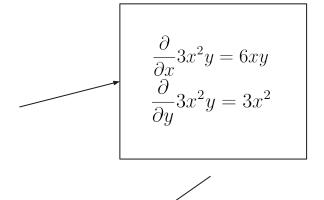
The partial derivative with respect to y is $3x^2$

Rewrite them!

$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

The partial derivative with respect to y is $3x^2$

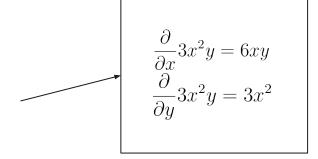


Next task: put them together

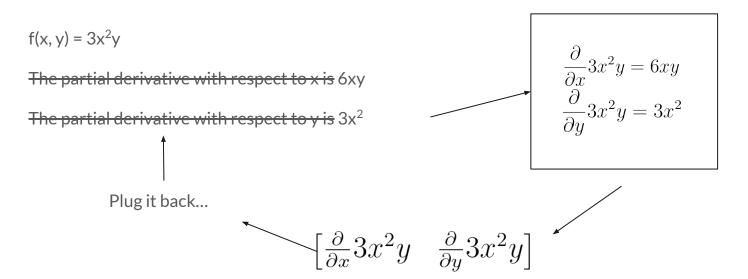
$$f(x, y) = 3x^2y$$

The partial derivative with respect to x is 6xy

The partial derivative with respect to y is $3x^2$



$$\begin{bmatrix} \frac{\partial}{\partial x} 3x^2 y & \frac{\partial}{\partial y} 3x^2 y \end{bmatrix}$$



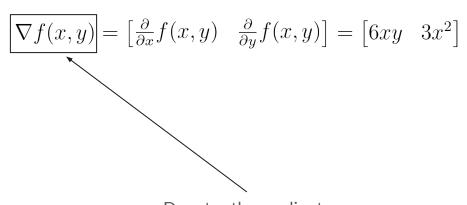
$$\begin{bmatrix} \frac{\partial}{\partial x} 3x^2y & \frac{\partial}{\partial y} 3x^2y \end{bmatrix} = \begin{bmatrix} 6xy & 3x^2 \end{bmatrix}$$
 Plug it back...
$$\begin{bmatrix} \frac{\partial}{\partial x} 3x^2y & \frac{\partial}{\partial y} 3x^2y \end{bmatrix}$$

$$\begin{bmatrix} \frac{\partial}{\partial x} 3x^2y & \frac{\partial}{\partial y} 3x^2y \end{bmatrix} = \begin{bmatrix} 6xy & 3x^2 \end{bmatrix}$$
Now let's write this $3x^2y$ as $f(x, y)$

$$\nabla f(x,y) = \begin{bmatrix} \frac{\partial}{\partial x} f(x,y) \end{bmatrix} \quad \frac{\partial}{\partial y} f(x,y) = \begin{bmatrix} 6xy & 3x^2 \end{bmatrix}$$

Done! We represented it by f(x,y)

$$\nabla f(x,y) = \begin{bmatrix} \frac{\partial}{\partial x} f(x,y) & \frac{\partial}{\partial y} f(x,y) \end{bmatrix} = \begin{bmatrix} 6xy & 3x^2 \end{bmatrix}$$
 nabla symbol



Denotes the gradient (the gradient in "gradient descent")