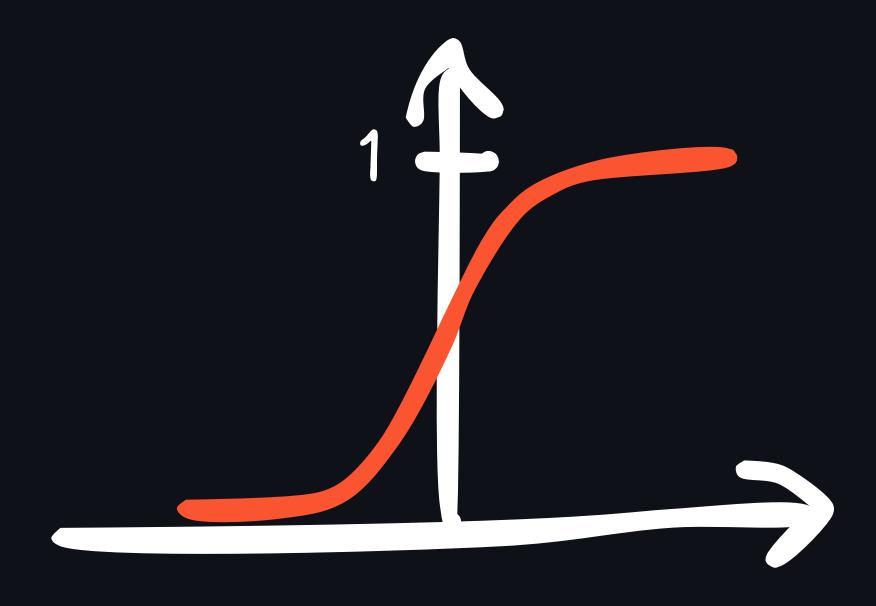
The Sigmoid activation function.



$$a\left(z
ight)=rac{1}{1+e^{-z}}$$

- Perfect for classical binary classification where the output is either 0 or 1.
- Output the probability of belonging to a certain class.

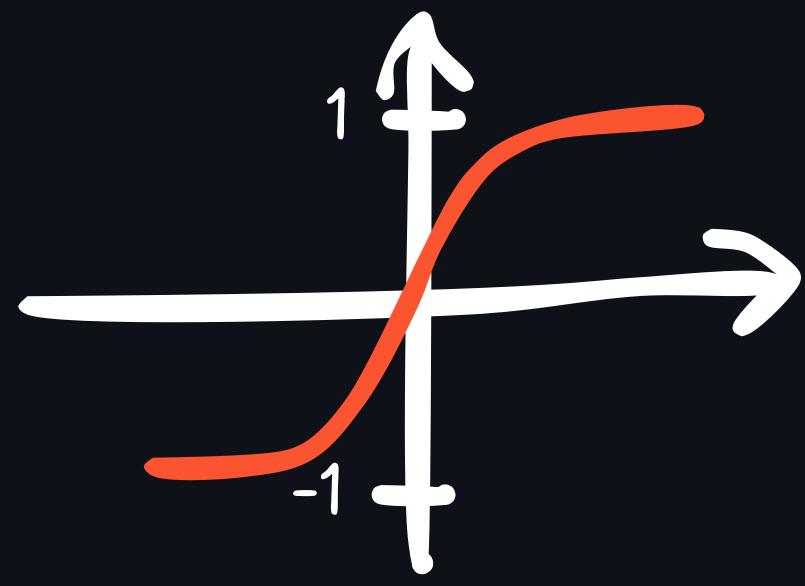
The Softmax activation function.

Output Layer Probabilities  $\begin{pmatrix} 2.1 \\ 3 \\ 4.8 \\ 3.2 \\ 0.1 \end{pmatrix} \longrightarrow a(z)_j = \frac{e^{z_j}}{\sum_{k=1}^K e^{z_k}} \longrightarrow \begin{pmatrix} 0.8 \\ 0.11 \\ 0.03 \\ 0.04 \\ 0.02 \end{pmatrix}$ 

$$a(z)_j = rac{e^{z_j}}{\sum_{k=1}^K e^{z_k}} ext{ for } j=1,\ldots,K ext{ and } \mathbf{z}=(z_1,\ldots,z_K)$$

- Generalization of sigmoid for multi-class classification tasks like image classification.
- Output probabilities for each class from 1 to K.

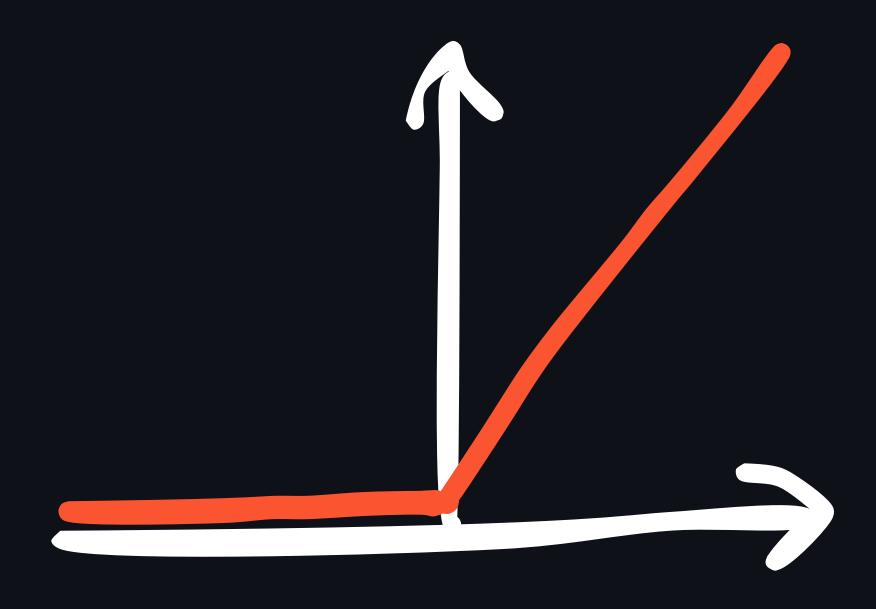
The TanH (Hyperbolic Tangent) activation function.



$$a(z)=rac{e^z-e^{-z}}{e^z+e^{-z}}$$

Often used in hidden layers of a neural network. It centers the data, improving the learning for subsequent layers.

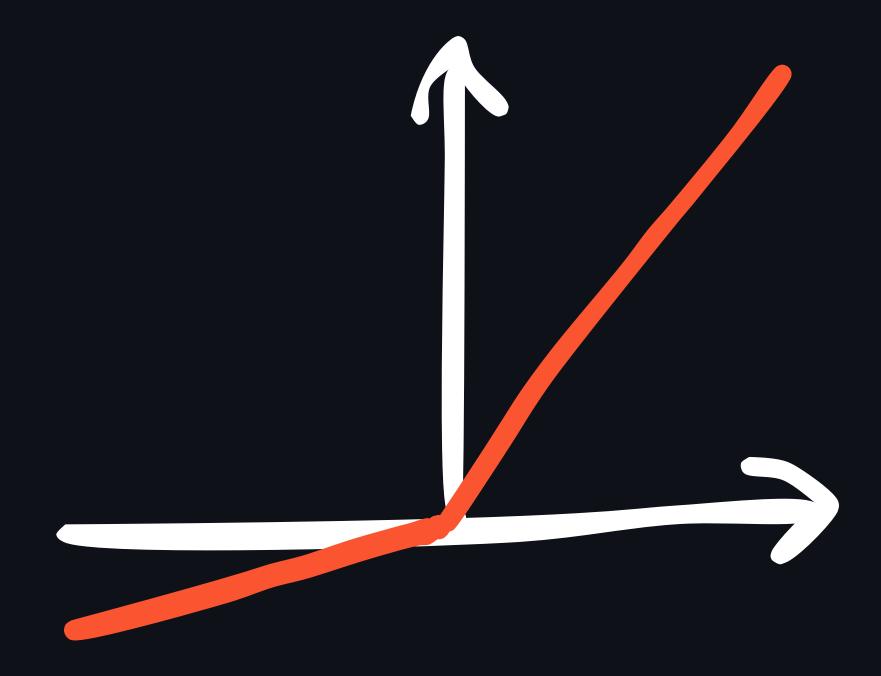
The ReLU (Rectified Linear Unit) activation function.



$$a\left(z\right)=max\left(0,x\right)$$

Widely used in CNNs for tasks like image recognition due to its computational efficiency.

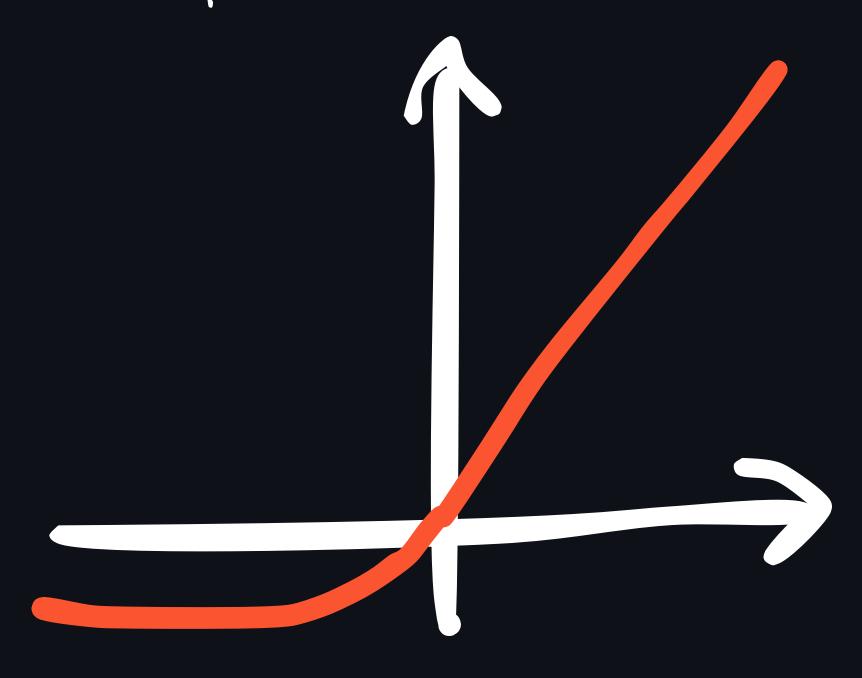
The leaky ReLU activation function.



$$a(z) = egin{cases} z & ext{if } z > 0, \ lpha z & ext{otherwise}. \end{cases}$$

Helps prevent dead neurons in ReLU. Useful in deep learning models where avoiding the dying ReLU problem is critical.

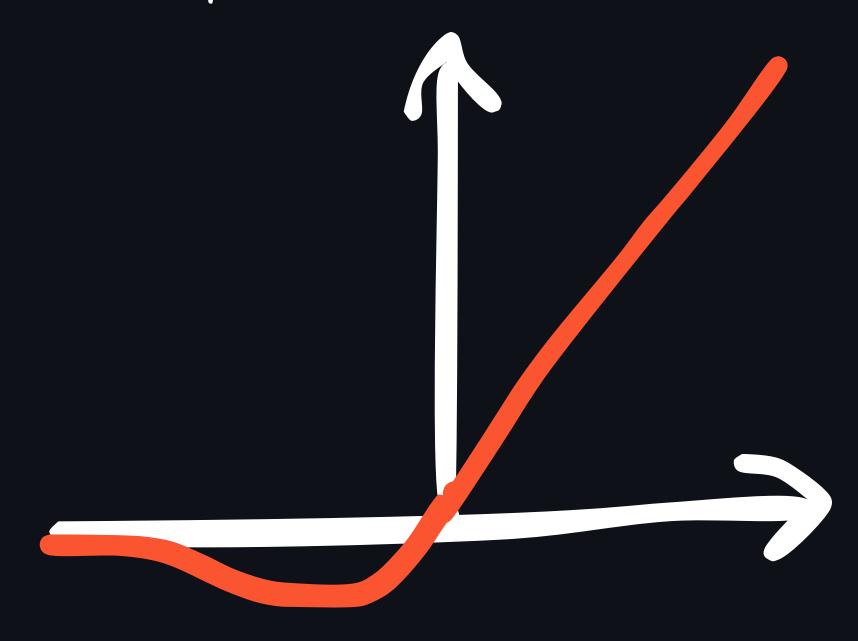
The ELU (Exponential Linear Unit) activation function.



$$a(z) = egin{cases} z & ext{if z} > 0, \ lpha(e^z - 1) & ext{if z} \leq 0. \end{cases}$$

ELU improves accuracy for deep neural networks by allowing negative values for activations.

The Gelu (Gaussian Error Linear Unit) activation function.



$$a(z) = zP(Z \le z)$$

Non-linear activation function that approximates the expectation of a Gaussian distribution.

# Thanks for Reading

I hope this helps!