

2-3 Activation Functions

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WISE and SOE, XMU, 2025

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Why activation?

1. Neural networks mainly involve two steps for each neuron
 - Linear transformation
 - **Activation** (nonlinear transformation)
2. If there is no activation, neural networks are simple linear models!
3. **Activation** empowers neural networks to learn complex structures

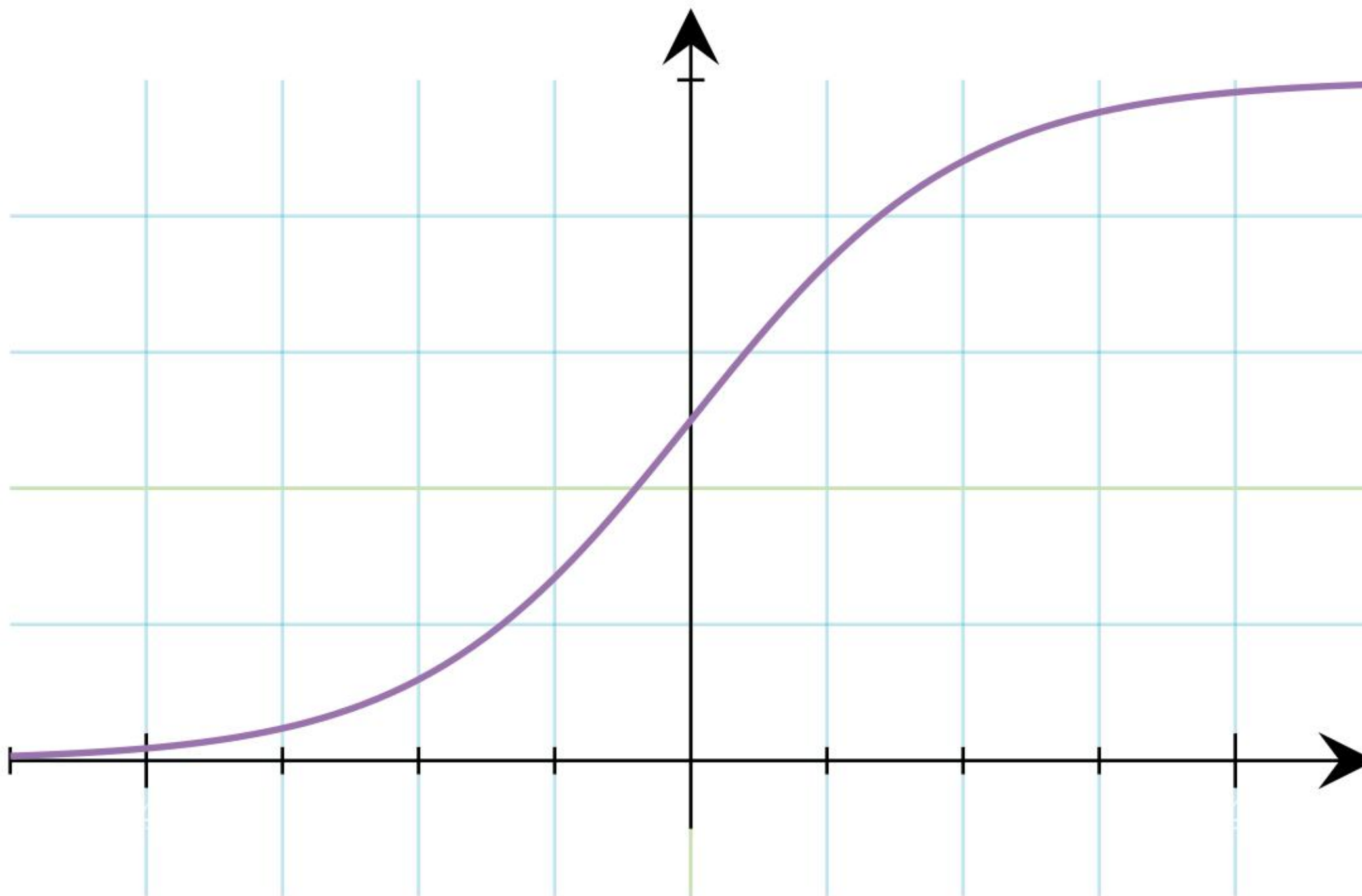
Sigmoid activation function

1. Form

$$\sigma(z) = \frac{1}{1 + \exp(-z)}$$

- Range: $(0, 1)$
- Derivative: $\sigma'(z) = \sigma(z)\{1 - \sigma(z)\}$

Sigmoid activation function



Sigmoid activation function

1. Advantages

- Range is $(0, 1)$, and it is commonly used for binary classification (last layer).
- Gradient vanishes (actually it is not good) if z is large, so it is robust to outliers
- Easy to obtain its derivative for backpropagation.

2. Disadvantages

- **Gradient vanishing.**
 - ▷ For a neural network with 5 layers, we may have $0.25^5 \approx 0.001!$
- Output is **not symmetric about 0.**
- **Heavy computation** for the derivative due to the exponential.

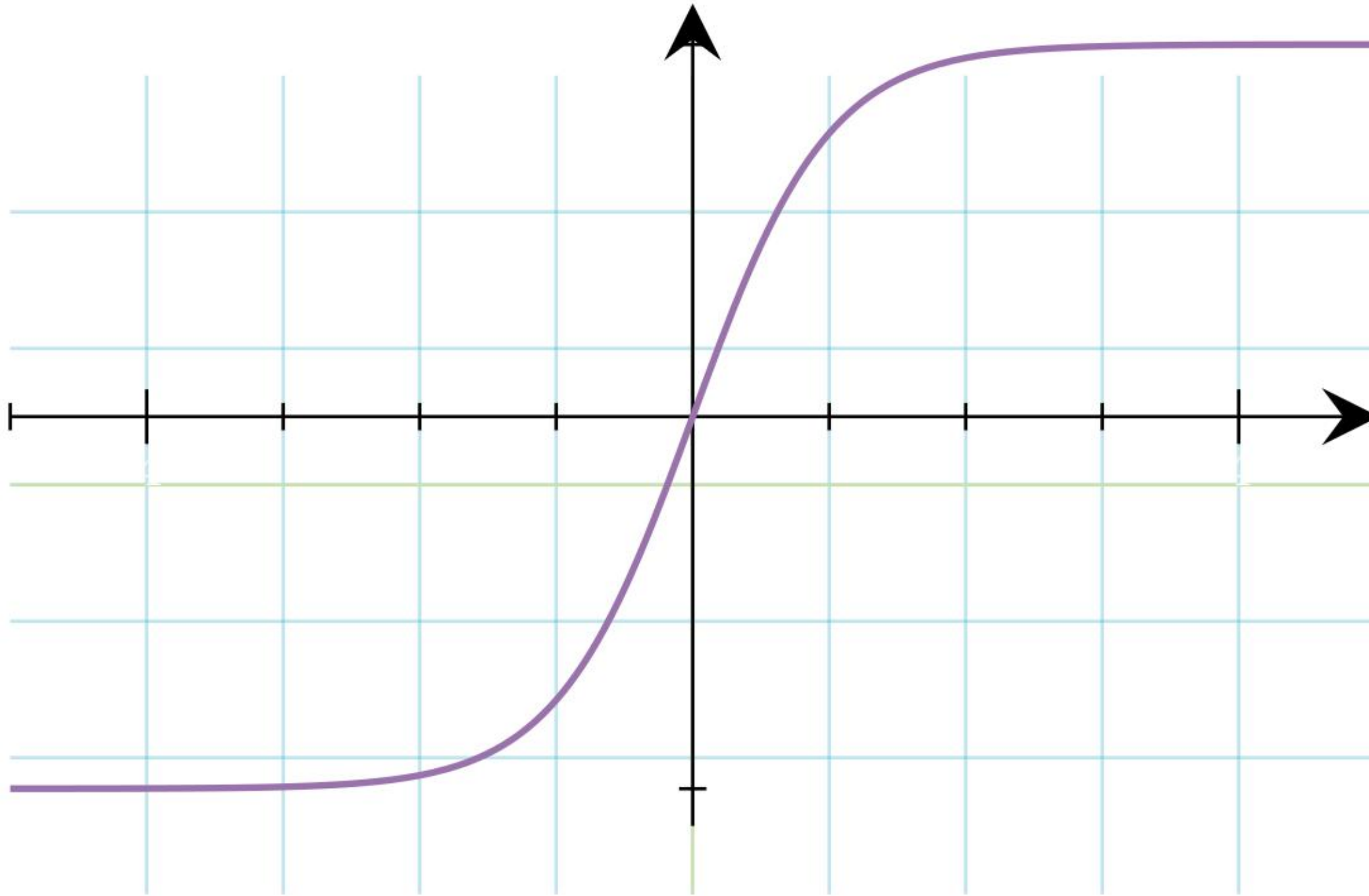
Tanh activation function

1. Form

$$\sigma(z) = \frac{2}{1 + \exp(-2z)} - 1$$

- Range: $(-1, 1)$
- Derivative: $\sigma'(z) = 1 - \sigma^2(z)$

Tanh activation function



Tanh activation function

1. Advantages

- Mean zero for this activation function
- Compared with sigmoid, the magnitude of gradient is larger near 0.
- Function is symmetric about 0.

2. Disadvantages

- Gradient vanishing.
- Heavy computation for the derivative due to the exponential.
- Do not induce sparsity

ReLU activation function

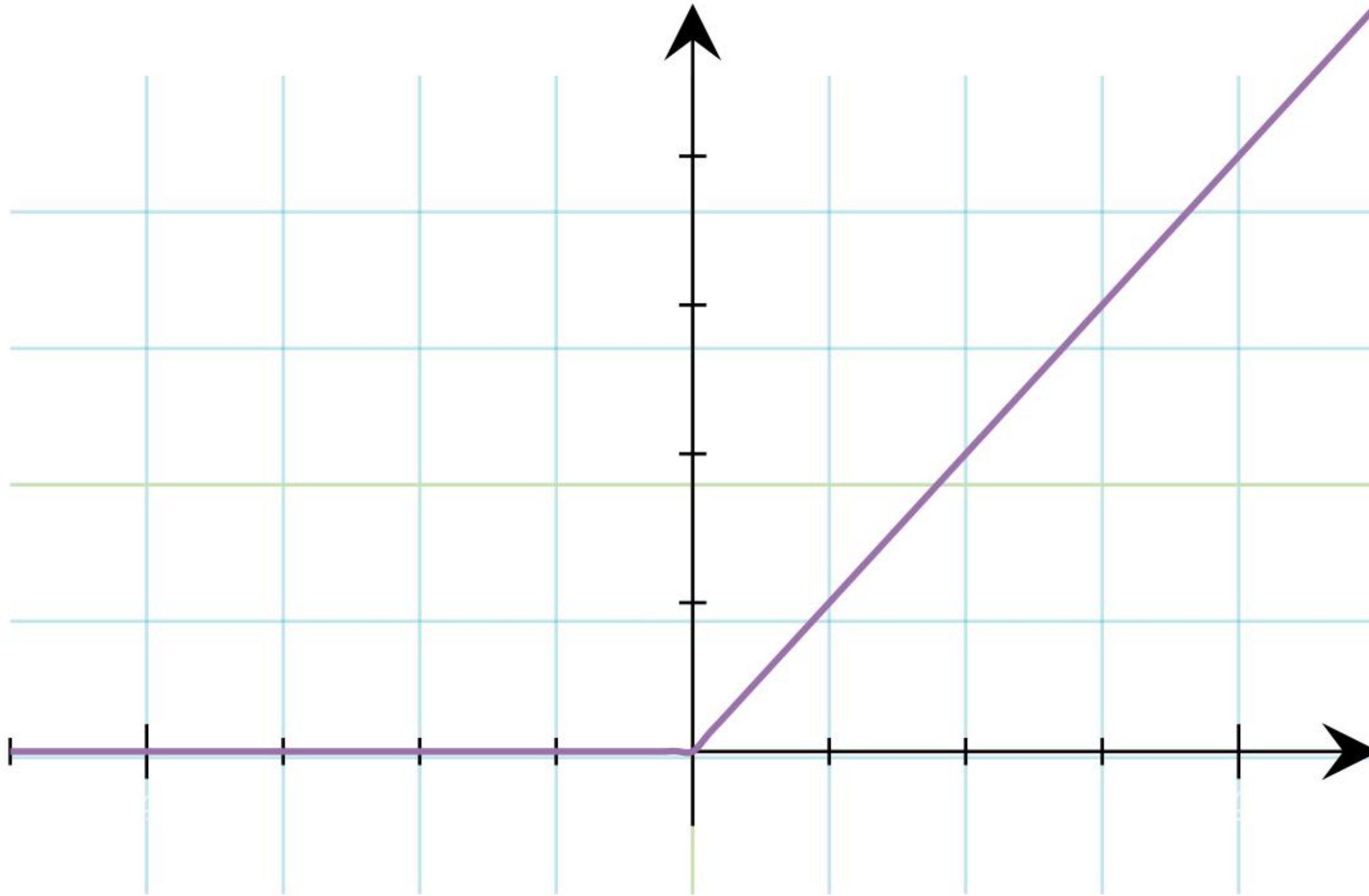
1. Form (Rectified Linear Unit)

$$\sigma(z) = \max\{0, z\}$$

- Range: $[0, \infty)$
- Derivative:

$$\sigma'(z) = \begin{cases} 0 & z \leq 0 \\ 1 & z > 0 \end{cases}$$

ReLU activation function



ReLU activation function

1. Advantages

- High computation efficiency
- Alleviate the gradient vanishing problem in some sense
- Sparse activation
- Fast convergence rate

2. Disadvantages

- Dying ReLU.
- Output is not symmetric about 0
- Not differentiable at $z = 0$

Leaky ReLU activation function

1. Form (Leaky Rectified Linear Unit)

$$\sigma(z) = \begin{cases} \alpha z & z \leq 0 \\ z & z > 0 \end{cases}$$

- α : a small number. For example, $\alpha = 0.01$
- Range: $(-\infty, \infty)$
- Derivative:

$$\sigma'(z) = \begin{cases} \alpha & z \leq 0 \\ 1 & z > 0 \end{cases}$$

Leaky ReLU activation function

