

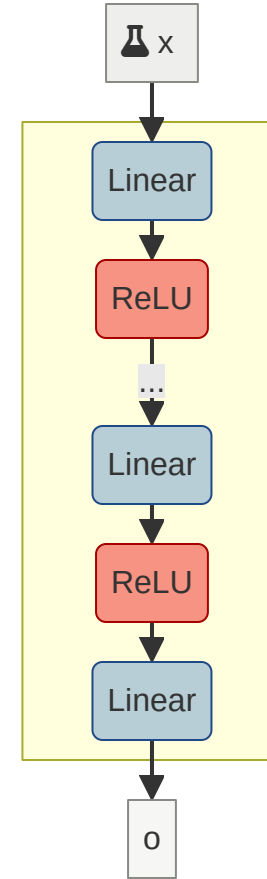
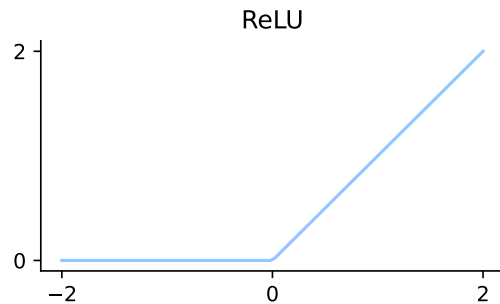
# Activation Functions

# Recap: Non-Linearities

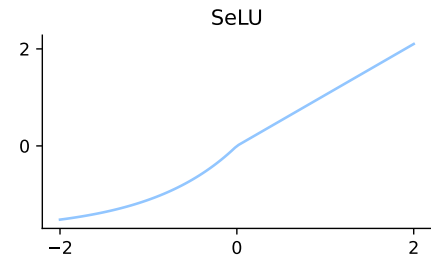
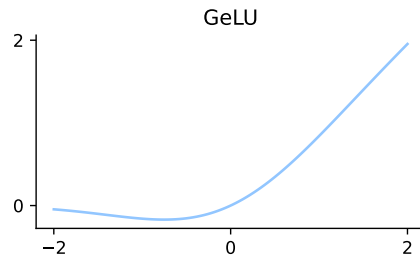
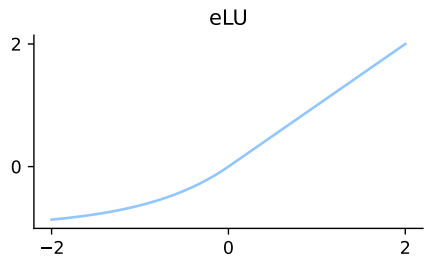
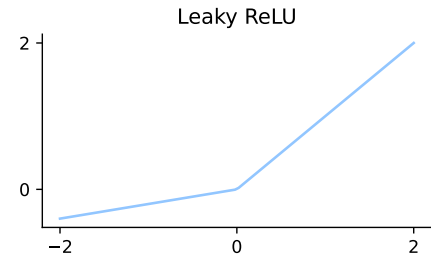
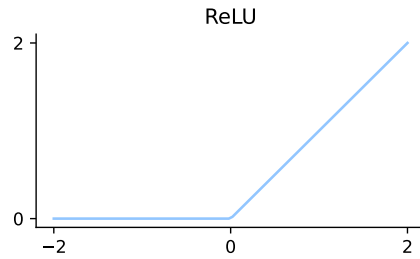
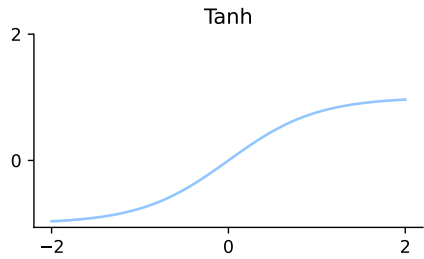
Rectified Linear Unit (ReLU)

$$\text{ReLU}(x) = \max(x, 0)$$

Allows deep networks to model arbitrary differentiable functions



# Zoo of Activation Functions

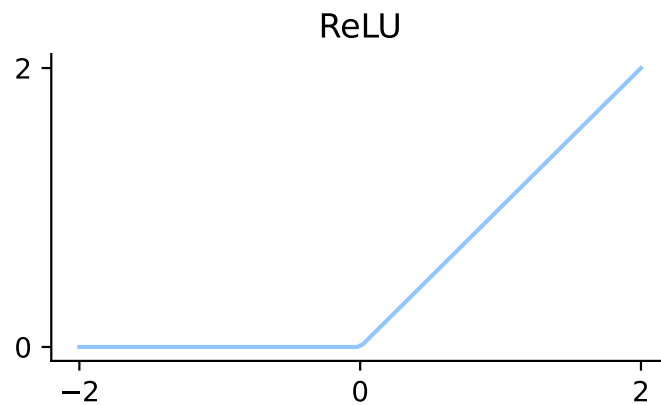


# ReLU

$$\text{ReLU}(x) = \max(x, 0)$$

✓ Simple

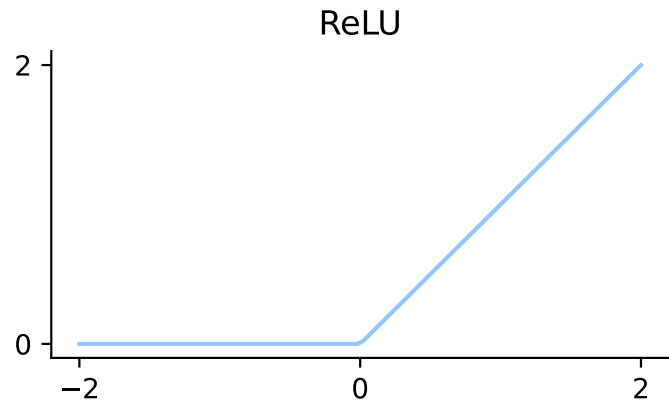
✗ ReLU units can be fragile during training and "die"



# Dead ReLUs

How can we prevent dead ReLUs?

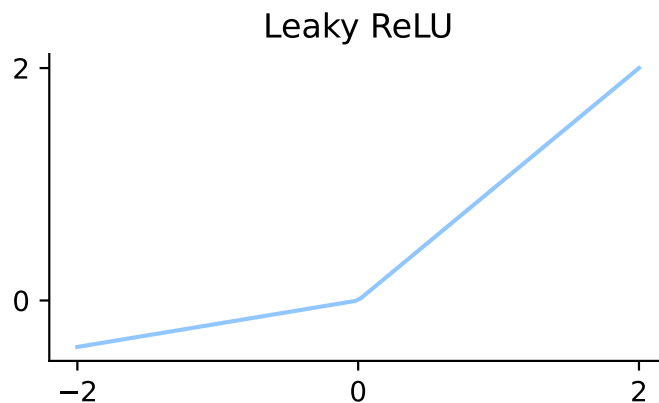
- Initialize network carefully
- Decrease the learning rate



# Leaky ReLU

$$\text{LeakyReLU}(x) = \max(x, \alpha x)$$

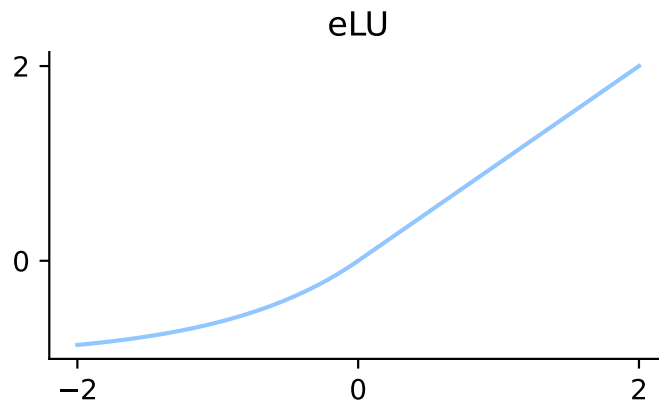
- Where  $0 < \alpha < 1$
  - Called **PReLU** if  $\alpha$  is learned
- ✓ Non-negative gradient for negative inputs
- ✗ The slope  $\alpha$  needs to be tuned
- ✗ Cannot wipe the negative signal out



# Elu

$$\text{ELU}(x) = \begin{cases} x & \text{if } x \geq 0 \\ \alpha(e^x - 1) & \text{if } x < 0 \end{cases}$$

- ✓ Non-negative gradient for negative inputs
- ✗  $\alpha$  needs to be tuned
- ✗ Exponential is computationally expensive



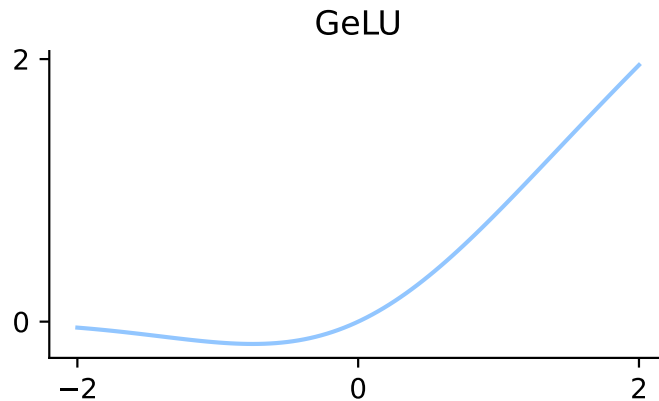
# GeLU

$$\text{GeLU}(x) = x \times \Phi(x)$$

- Where  $\Phi(x)$  is the CDF of the standard Gaussian
- $\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{t^2}{2}} dt$

✓ Non-zero gradient for negative inputs

✗ Requires more computation





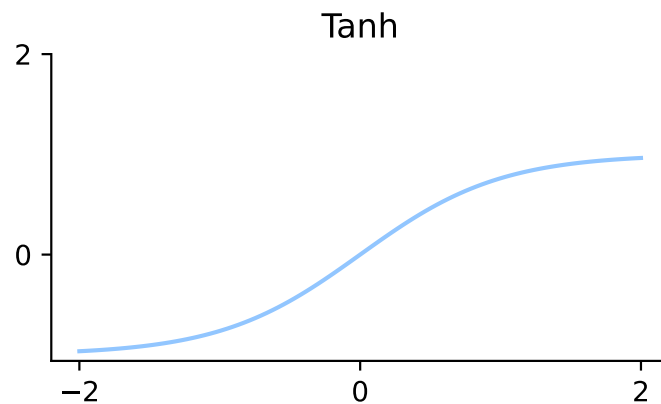
# Sigmoid

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

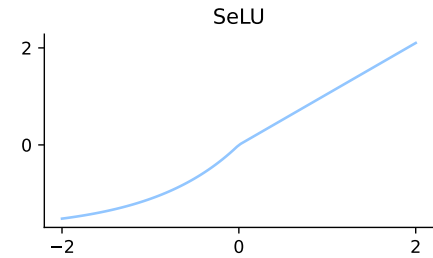
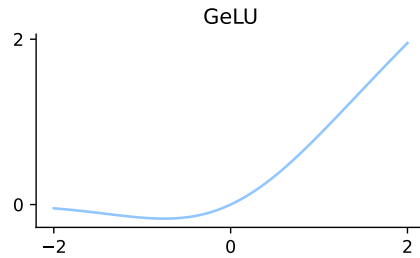
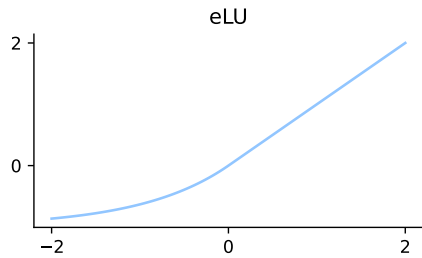
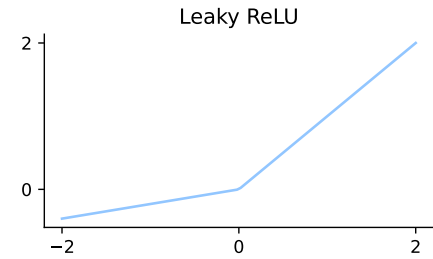
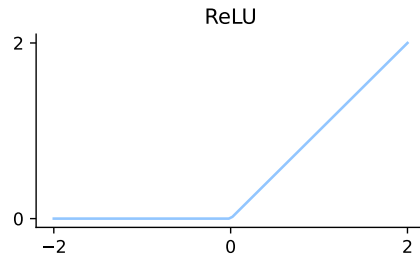
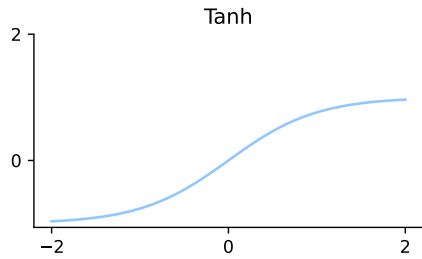
- Same as  $\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$

✗ Saturates on both ends

✗ Do **not** use sigmoid/tanh



# Which Activation to Choose?



# Activation Functions - TL;DR

Use ReLU with careful initialization and small learning rate

If ReLU fails, try Leaky ReLU or PReLU

Avoid Sigmoid and Tanh

Use GeLU for sophisticated models