



1) $h = \text{id}$ - lin. regr.

2) $h = \sigma$ - lg. regr.

3) $h = \tanh$

4) $h = \text{ReLU}$

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

5) $h = \text{LReLU}, \text{ELU}$

6) NAS - neural arch. search

$$h = \text{swish}_\beta(x) = x \cdot \sigma(\beta x) = \frac{x}{1 + e^{-\beta x}}$$

$h = \text{Mish}(x)$

$$7) \mathcal{S}_\beta(x_1, \dots, x_n) = \frac{\sum x_i e^{\beta x_i}}{\sum e^{\beta x_i}} \quad (\text{September 2020})$$

$$\begin{aligned} \mathcal{S}_\beta(g(x), h(x)) &= g(x) \frac{e^{\beta g(x)}}{e^{\beta g(x)} + e^{\beta h(x)}} + h(x) \frac{e^{\beta h(x)}}{e^{\beta g(x)} + e^{\beta h(x)}} = \dots \\ &= (g(x) - h(x)) \sigma(\beta(g(x) - h(x))) + h(x) \end{aligned}$$

$$g(x) = x, \quad h(x) = 0:$$

hard max

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

\mathcal{S}_β

$$\text{swish}(x) = x \cdot \sigma(\beta x)$$

$$g(x) = x \quad h(x) = ax$$

LReLU

$$(1-a)x \cdot \sigma(\beta(1-a)x) + ax$$

