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### Exercise 3

Show that a multi-layer neural network with linear activation function  $s(x) = x$  is equivalent to a single layer linear network. Assume that in each layer the inputs follow a Normal distribution with mean zero and small variance, i.e.  $\sigma \ll 1$ . For which of the activation functions  $s(x) = 1/(1 + \exp(-x))$ ,  $s(x) = \tanh(x)$ ,  $s(x) = \text{relu}(x)$  and  $s(x) = \text{selu}(x)$  is a deep network equivalent to a linear network for the given distribution? The selu function is given by:

$$\text{selu}(x) = \begin{cases} \lambda x & \text{if } x > 0 \\ \alpha(\exp(x) - 1) & \text{otherwise} \end{cases} \quad (4)$$

where  $\lambda \approx 1.0507$  and  $\alpha \approx 1.75814$ . (Hint: consider the case  $\sigma \rightarrow 0$  using a Taylor expansion around 0.)

• Deep Linear = single linear

M2P (with  $L$  layers) has about:  $y = W^L \cdot W^{L-1} \cdot \dots \cdot W^2 \cdot W^1 \cdot x$

We can just bundle  $W^L \cdot W^{L-1} \cdot \dots \cdot W^2 \cdot W^1 = W_{\text{eff}} \rightarrow y = W_{\text{eff}} x$   
(single neuron!)

• for which activation functions is a deep network approximately linear when  $x \sim N(0, \sigma^2)$  with  $\sigma^2 \ll 1$ ?

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

Taylor expansion in  $x$ :  $s(x) \approx \frac{1}{2} + \frac{x}{4} - \frac{x^3}{48} + \dots$

for small  $\sigma$   $s(x) \approx \frac{1}{2} + \frac{x}{4}$

Not linear: constant bias  
breaks linearity over multiple layers  $x$

2)  $\tanh(x)$

Taylor expansion  $s(x) = x - \frac{x^3}{3} + \frac{2x^5}{15} - \dots$

for small  $\sigma$   $s(x) = x$

Linear ✓

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

$x > 0 \rightarrow \text{output } x$

$x \leq 0 \rightarrow \text{output } 0$

↳ normal distribution around 0 means that whole input is mapped into 0  
Non linearity!

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

Taylor expansion

$$\begin{cases} \lambda x & x \geq 0 \\ \lambda \alpha (x + \frac{x^2}{2} + \dots) \approx \lambda \alpha x & x < 0 \end{cases}$$

↳ this is piecewise linear (but not globally) slope differs between  $x \geq 0$  and  $x < 0$