

Activation Funⁿ

Drawback - if Activation funⁿ are not there, then, entire neural Network graph will be just a linear relationship.

Key role of Activation funⁿ is to introduce Non-linearity in your relationships

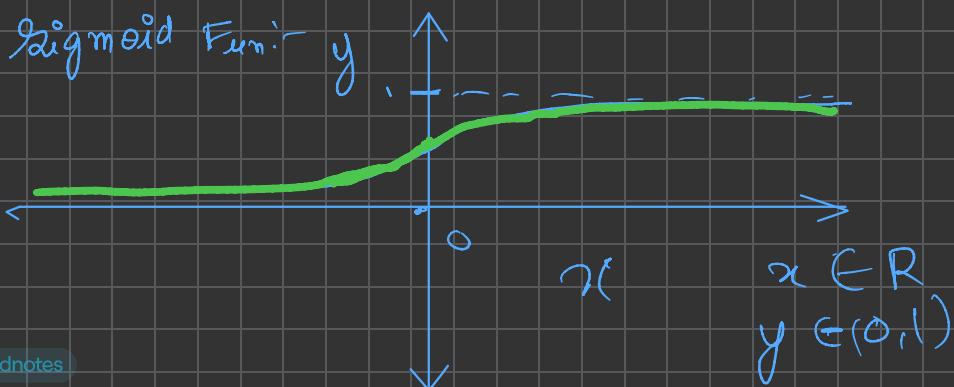
3 Most used Activation funⁿ:

$$1) \text{Sigmoid} = \frac{1}{1+e^{-z}}$$

$$2) \tanh = \frac{e^z - e^{-z}}{e^z + e^{-z}}$$

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

1) Sigmoid Funⁿ:



Derivation

we need to prove,

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

??

~~if~~

Probability = $\frac{\# \text{ of outcomes for an event}}{\text{total } \# \text{ of outcomes}}$

(How likely is the event)

$$H=20, T=30, P(H) = 20/50$$

$$P(T) = 30/50$$

$$P(H) = 1 - P(T)$$

$$H=20, T=30$$

Q How much more likely Head will come compare to Tail?

$$\Delta \Rightarrow \frac{H}{T} = \frac{20}{30} = \frac{2}{3}$$

If Tail comes once, then Head also will occur once.

Equally likely H & T to occur.

$$H = 80, T = 20$$

Q How much more likely head will come from Tail?

Ans $\frac{H}{T} = \frac{80}{20} = 4$

If Tail comes once, then Head will come 4x more

Q How much more likely Tail will come from Head?

$$\rho = \frac{T}{H} = \frac{20}{80} = \frac{1}{4} = 0.25$$

If H comes, 4x times, then Tail comes 1x time, i.e., 0.25,

If H comes 1x time, then Tail comes 0.25x times

So, in conclusion, there is a event which is happening & which is not happening

$$\text{odds} = H/T \text{ & } T/H$$

$\text{odds} = \frac{\# \text{ of time of event}}{\# \text{ of time of event not happening}}$

$\frac{\# \text{ of time of event happening}}{\# \text{ of time of event not happening}}$

(Dividing by total section)

$$\text{odds} = \frac{P(\text{event happening})}{P(\text{event not happening})}$$

$$\text{odds} = \frac{P}{1-P}$$



$$P(0,1)$$

$$P=0, \text{ odds} = 0$$

$$P=1, \text{ odds} = \infty$$

$$(0,1) \rightarrow (0,\infty)$$

map

understanding $\log(x)$ significance?

$$x \in (0, \infty) \quad y \in (-\infty, \infty) \\ y = \log(x)$$



even for $10^{32} \rightarrow \text{No.} \rightarrow \log \rightarrow 32$

$10^{-32} \rightarrow \text{No.} \rightarrow \log \rightarrow -32$

∴ \log is introduced for squashing in neural network

$$\text{odds} = \frac{p}{1-p}$$

(taking \log on both sides)

$$\log \left(\frac{p}{1-p} \right) = z$$

$$\frac{p}{1-p} = e^z$$

$$\frac{1}{1+e^{-z}} = f(z)$$

$$\cancel{p} = e^z \Rightarrow p = (1-p)e^z$$

$$p = e^z - pe^z \Rightarrow p = \frac{e^z}{1+e^z}$$

$$p = \frac{e^z}{1+e^z} \approx \frac{1}{1+e^{-z}}$$

$$\begin{aligned} & \left(\cancel{\frac{e^z}{1+e^z}} * \cancel{\frac{e^z}{e^z}} \right) \\ & \xrightarrow{\quad \uparrow \quad} \\ & \frac{1}{\cancel{e^z}} \\ & \downarrow \\ & \left(\frac{1}{e^z} + 1 \right) \\ & \downarrow \\ & \frac{1}{1+e^{-z}} \end{aligned}$$

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

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