why not linear activation function in all layers ?

$$g(z) = z$$

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(because everything is outputted to z)

⇒ if all g(z) one linear, it will be no different than linear negression.

basically ,

$$\alpha^{\Gamma 2} = \omega_{1}^{\Gamma 2} \times + b_{1}^{\Gamma 2}$$

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$$a^{22} = \omega_{1}(\omega_{1}^{21} \times + b_{1}^{21}) + b_{1}^{22}$$

$$\Rightarrow a^{[2]} = \frac{\omega_i + \omega_j}{1 + \omega_j} + \frac{\omega_j}{1 + \omega_j} + \frac{\omega_j}{1 + \omega_j}$$

$$\Rightarrow a \quad \text{In ease regression model } i$$

$$\Rightarrow a \quad \text{In ease regression model}$$

Even if we change the output layer activation function and keep linear function for all layers, we'll get whatever function we have for the output layer.

$$g(z) = z$$

$$\vec{a}^{[1]} \underbrace{\vec{a}^{[2]}}_{\vec{a}^{[3]}} \underbrace{\vec{a}^{[4]}}_{\vec{a}^{[4]}} \longrightarrow \underbrace{\text{output layor}}_{is sigmoid}$$

$$all linear regression$$

$$cquivalent to linear regression$$

⇒ hidden layers will have no influence and we basically get a logistics regression model.

In Short,

Don't use linear activation function in hidden layer.