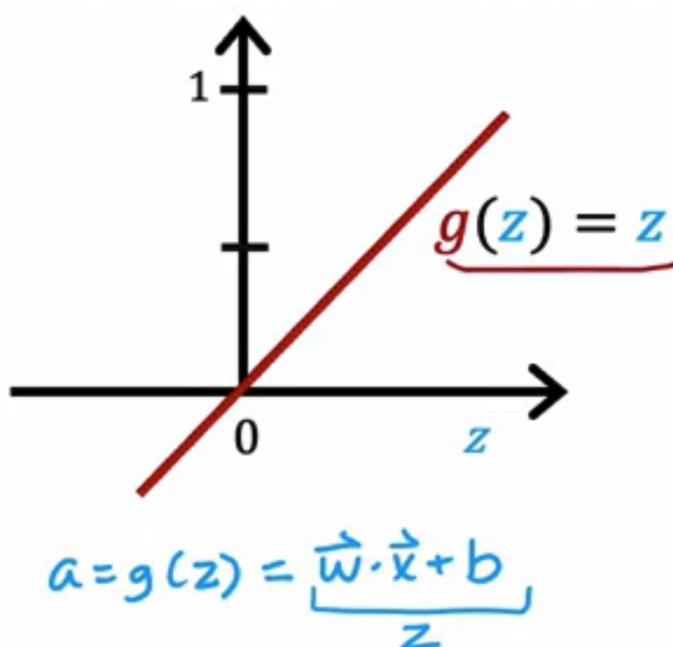
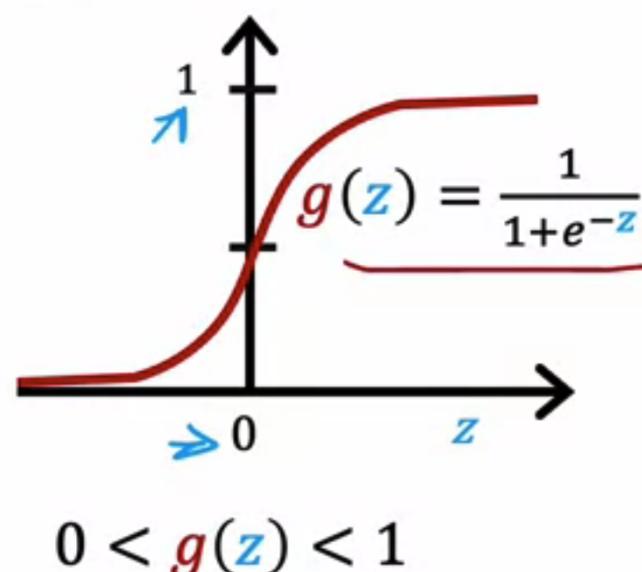


Examples of Activation Functions

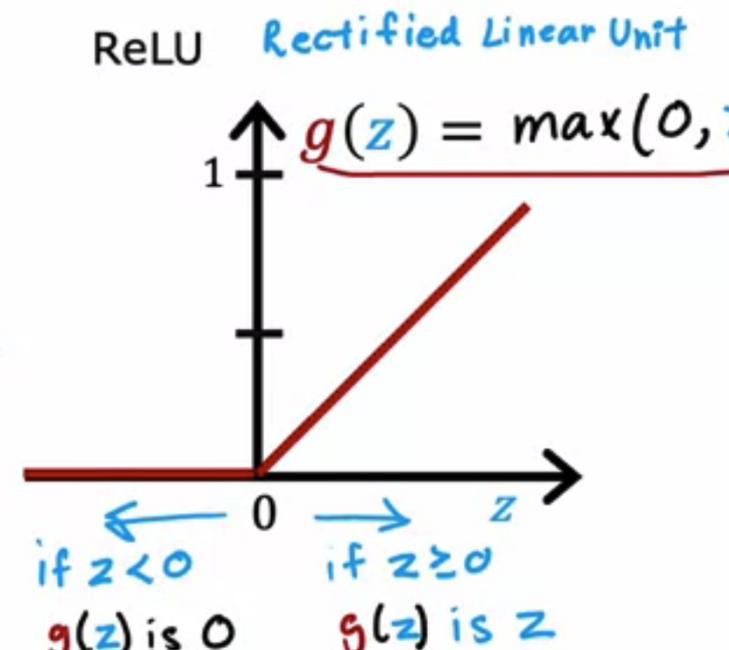
Linear activation function



Sigmoid

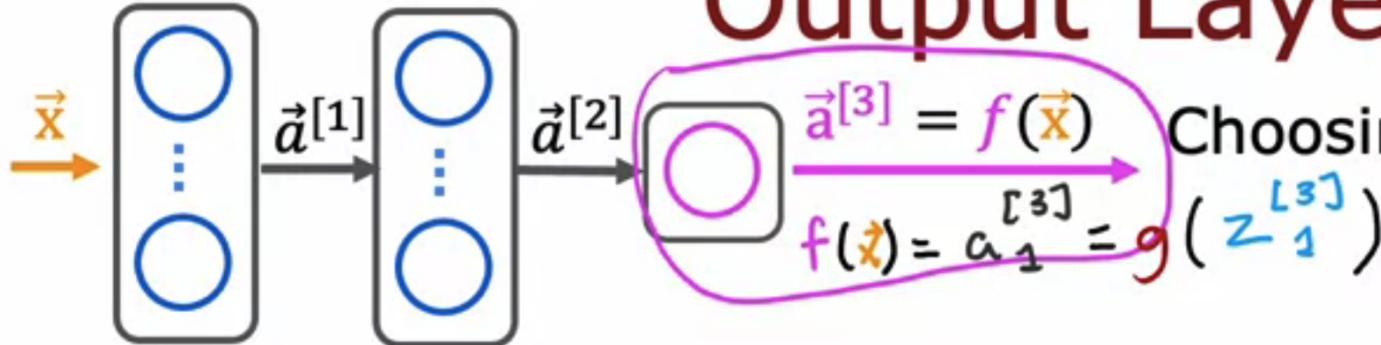


ReLU

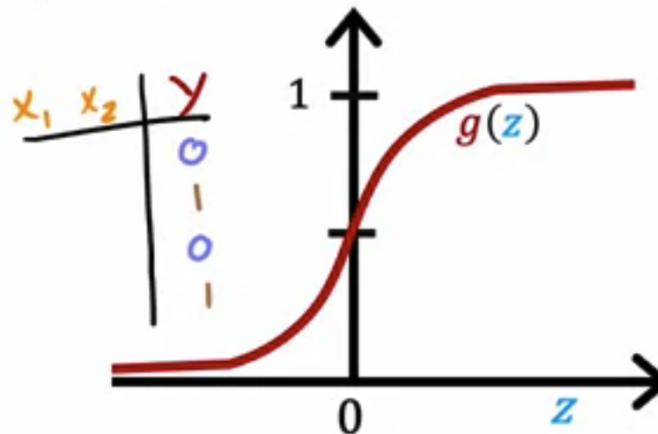


Rectified Linear Unit

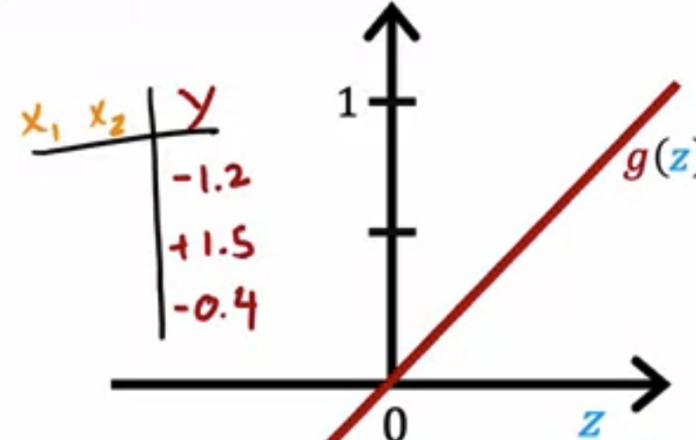
Output Layer



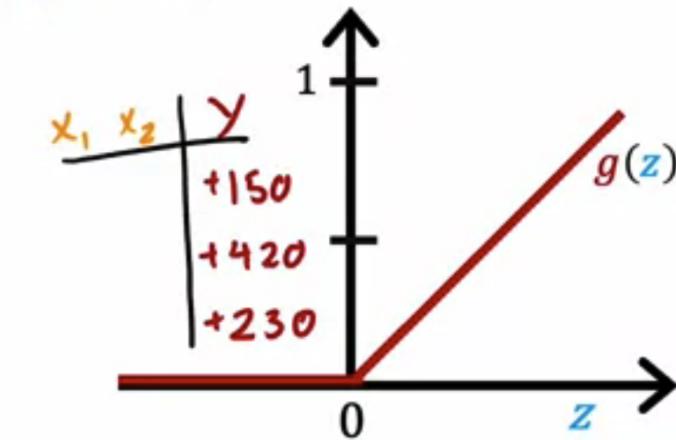
Binary classification
Sigmoid
 $y=0/1$



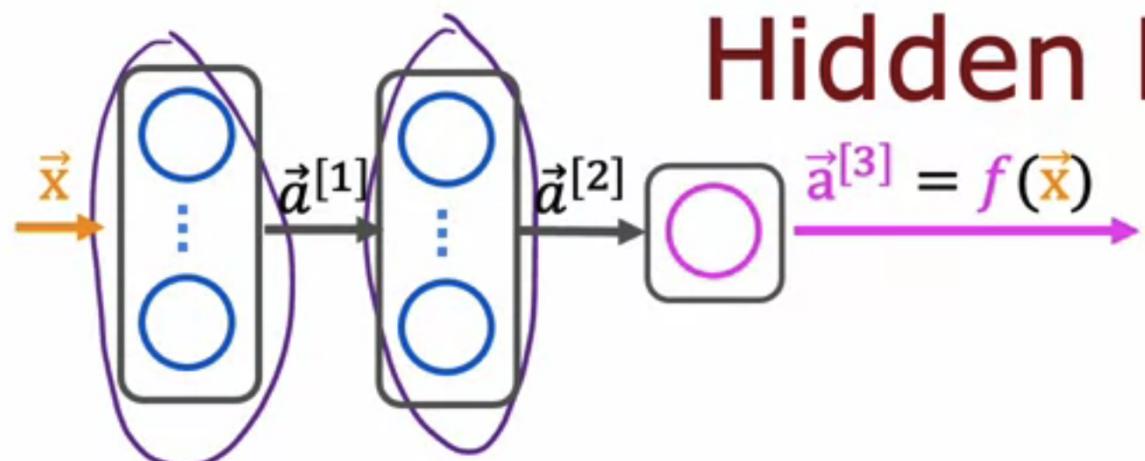
Regression
Linear activation function
 $y = +/-$



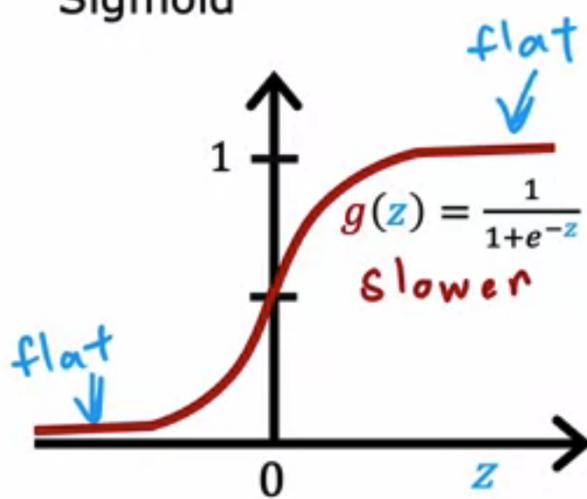
Regression
ReLU
 $y = 0 \text{ or } +$



Hidden Layer



Sigmoid



$J(W, B)$

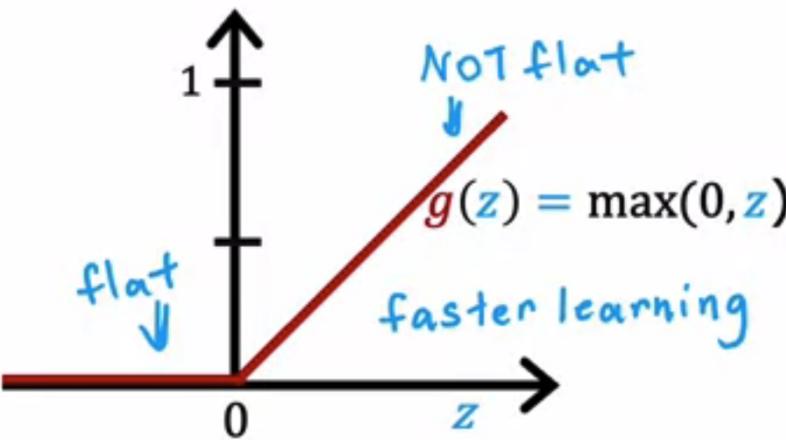
$$\frac{\partial}{\partial W} J(W, B) \approx 0$$

when $g(z)$ is flat

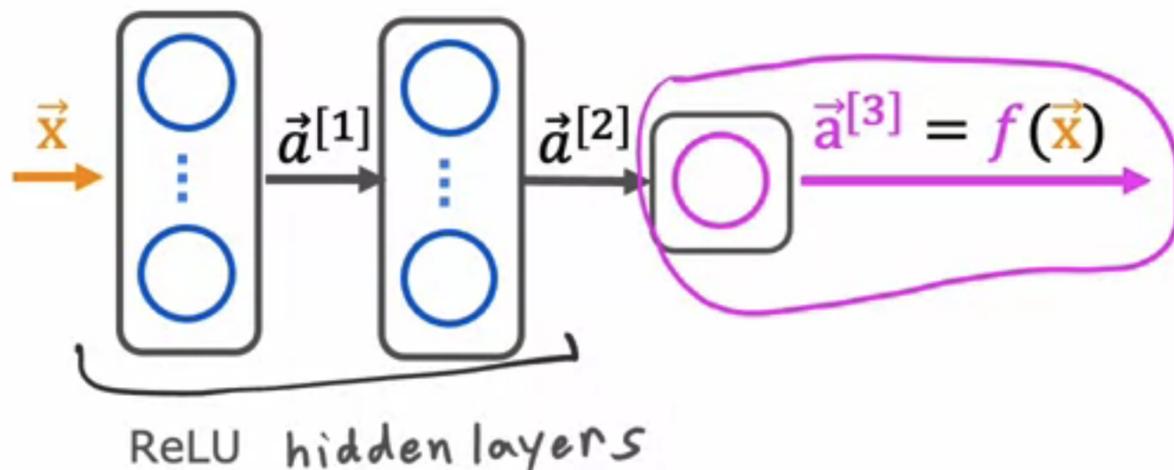
most common choice

ReLU

faster



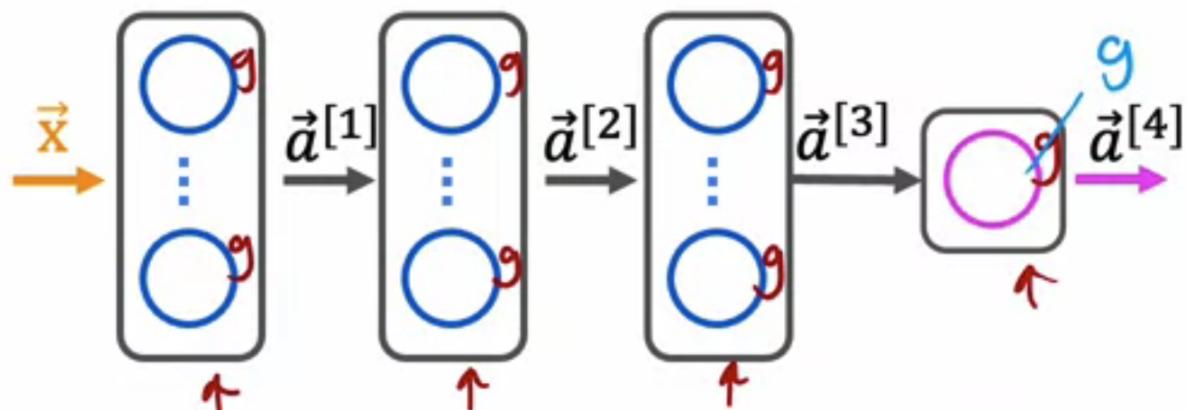
Choosing Activation Summary



```
from tf.keras.layers import Dense  
model = Sequential([  
    Dense(units=25, activation='relu'), layer1  
    Dense(units=15, activation='relu'), layer2  
    Dense(units=1, activation='sigmoid') layer3  
])  
or 'linear'  
or 'relu'
```

binary classification
activation='sigmoid'
regression y negative/
activation='linear' positive
regression $y \geq 0$
activation='relu'

Example



$$g(z) = z$$

$$\vec{a}^{[4]} = \vec{w}_1^{[4]} \cdot \vec{a}^{[3]} + b_1^{[4]}$$

all linear (including output)
↳ equivalent to linear regression

$$\vec{a}^{[4]} = \frac{1}{1+e^{-(\vec{w}_1^{[4]} \cdot \vec{a}^{[3]} + b_1^{[4]})}}$$

output activation is sigmoid
(hidden layers still linear)
↳ equivalent to logistic regression

Don't use linear activations in hidden layers (use ReLU)