

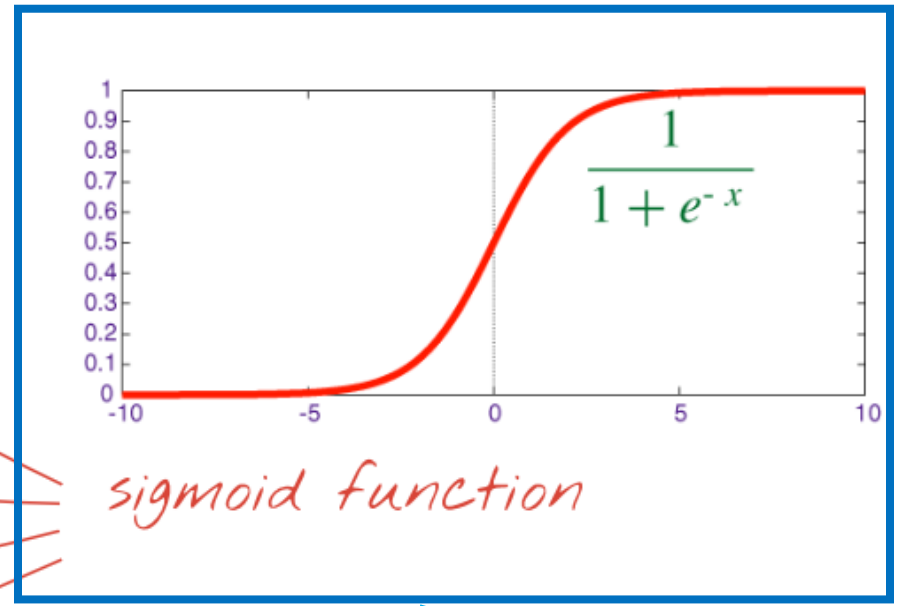
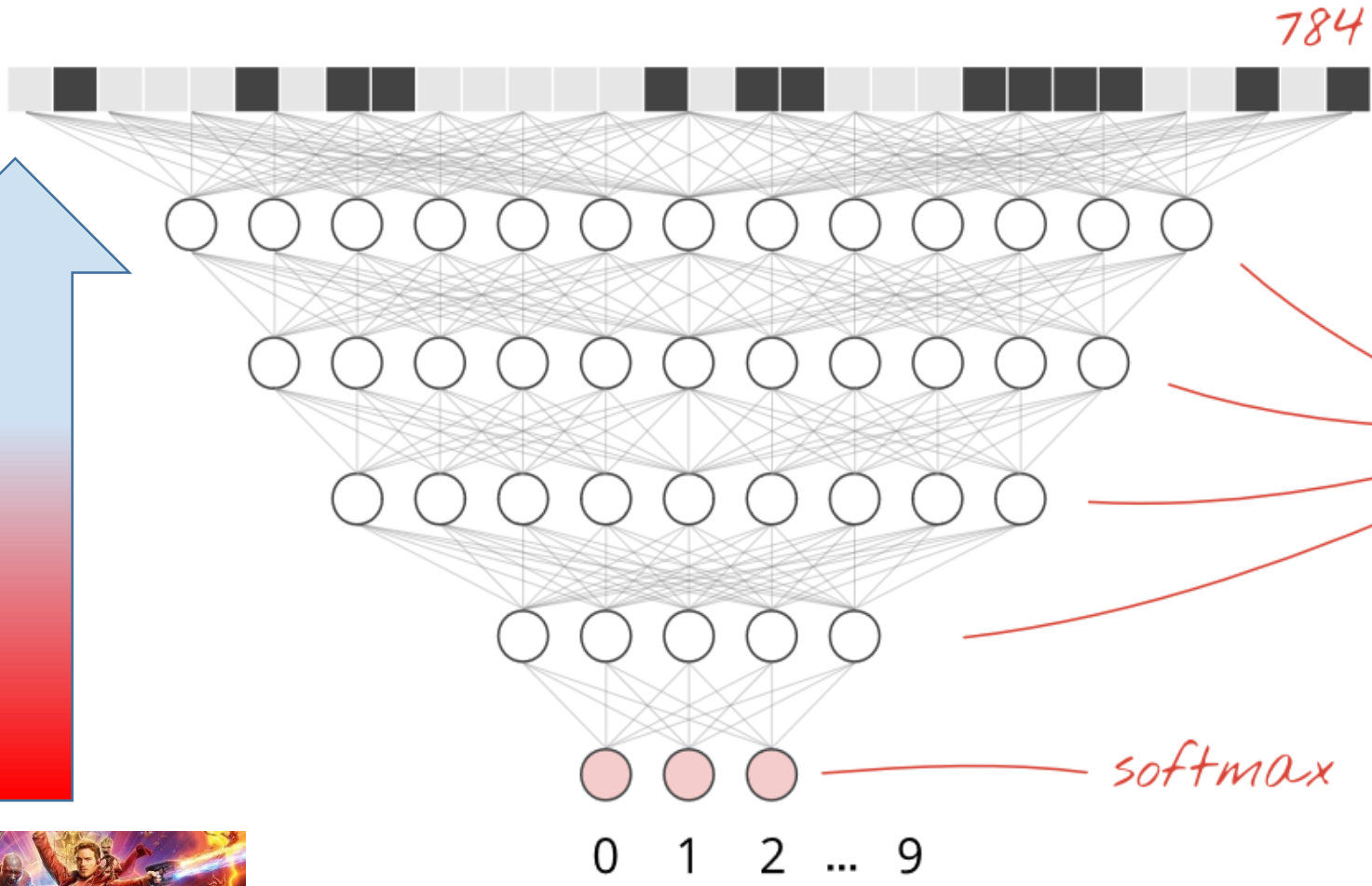
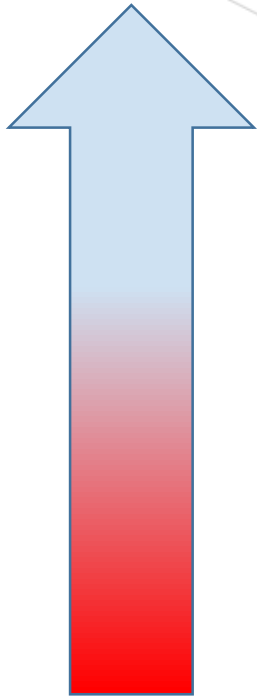
By Comdet Phaudphut, fb.com/comdet, comdet.p@gmail.com

DEEP LEARNING CONT.



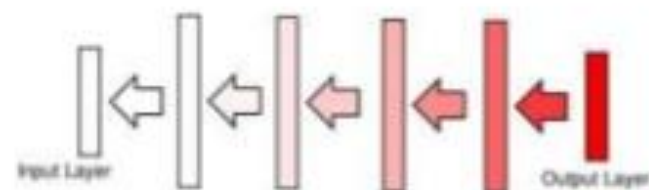
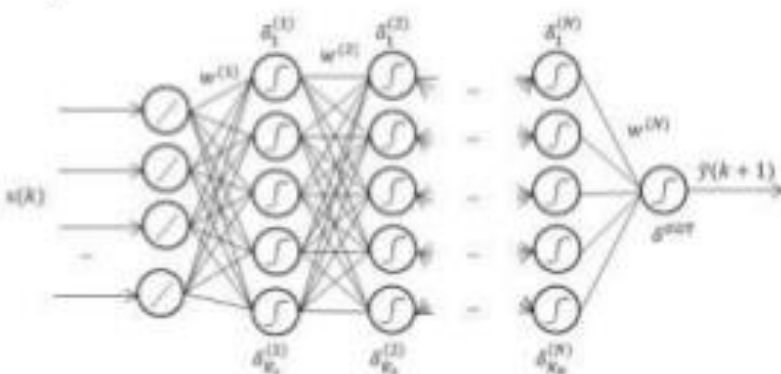
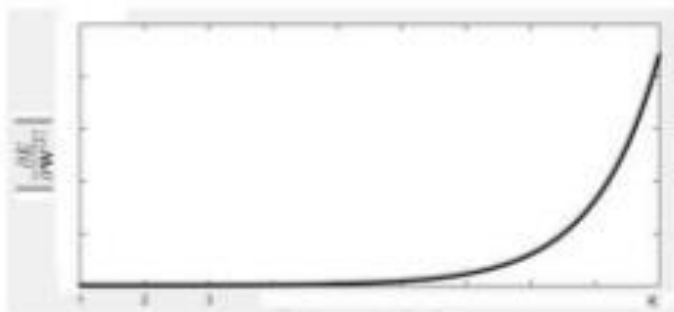
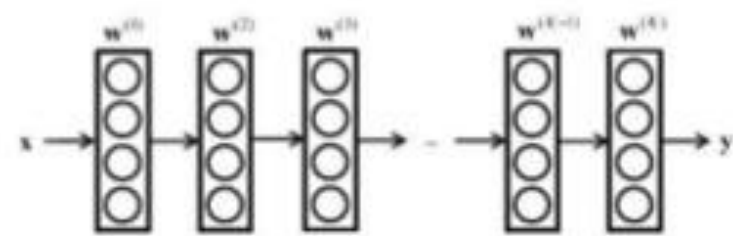
Review Prev Lab

- We add 3 (someone 4) hidden layers.
- But we didn't gain more accuracy as expected.
- Why ?
 - code errors or bugs?
 - machine mistake?
 - Take a look closer, what happen.



This function make farther layer from output small amplitude
We called "Gradient Vanishing Problem"

Bad effect of vanishing (exploding) gradients: a problem

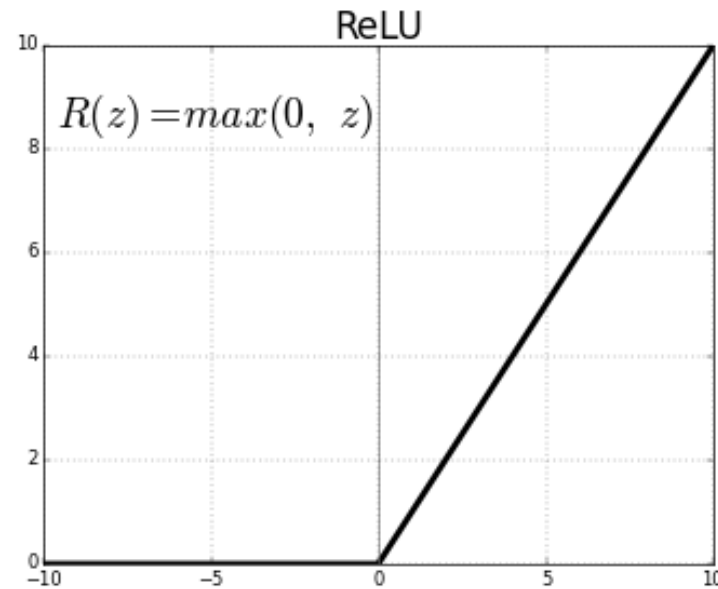


$$\delta_j^{(m)} = f_j^{(m-1)} \sum_i w_{ij}^{(m)} \delta_i^{(m+1)}, \quad \Rightarrow$$

$$\frac{\partial E(k)}{\partial w_{ji}^{(m)}} \rightarrow 0 \text{ for } m \rightarrow 1$$

$$\frac{\partial E(k)}{\partial w_{ji}^{(m)}} = \delta_j^{(m)} z_i^{(m-1)},$$

We use new invented activation function “ReLU” (Rectified Linear Units)



If $x > 0$ $y = x$
else $y = 0$
what it so easy?

- Invented from biological motivation
- But It can describe by proof math ... ummm no!
- Let change our code

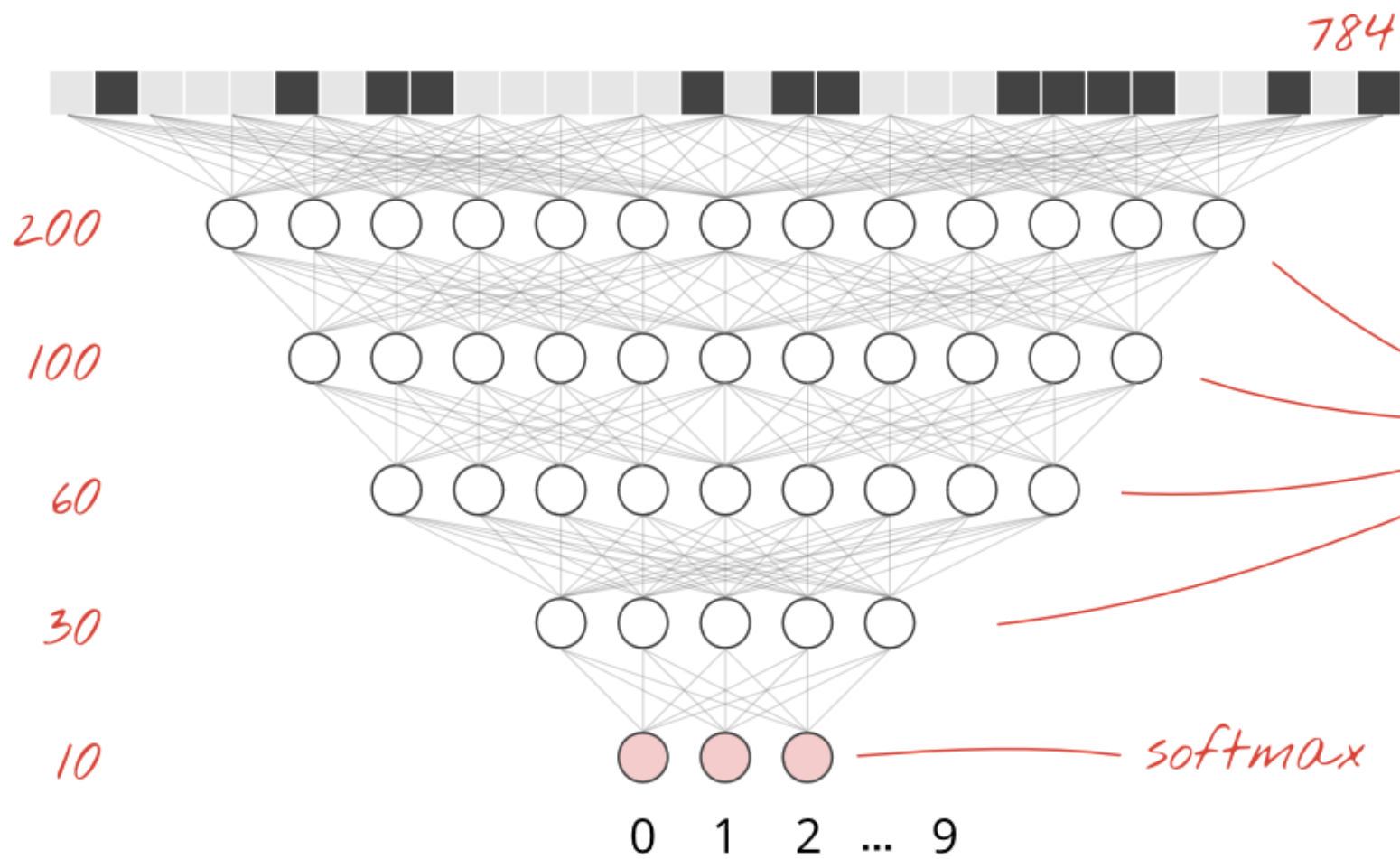
From sigmoid

```
1 #model
2 fc1 = tf.nn.sigmoid(tf.matmul(X, W1) + b1)
3 fc2 = tf.nn.sigmoid(tf.matmul(fc1, W2) + b2)
4 fc3 = tf.nn.sigmoid(tf.matmul(fc2, W3) + b3)
5 fc4 = tf.nn.sigmoid(tf.matmul(fc3, W4) + b4)
6 Ylogits = tf.matmul(fc4, W5) + b5
```

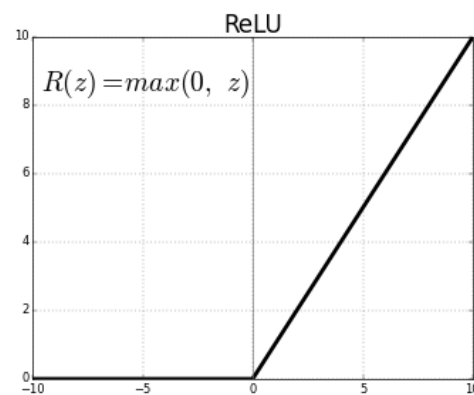


to relu

```
1 #model
2 fc1 = tf.nn.relu(tf.matmul(X, W1) + b1)
3 fc2 = tf.nn.relu(tf.matmul(fc1, W2) + b2)
4 fc3 = tf.nn.relu(tf.matmul(fc2, W3) + b3)
5 fc4 = tf.nn.relu(tf.matmul(fc3, W4) + b4)
6 Ylogits = tf.matmul(fc4, W5) + b5
```



ReLU



Run & open tensorboard
tensorboard --logdir="logs"
and see what happen!