



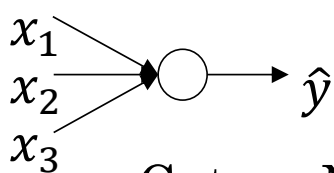
deeplearning.ai

Batch Normalization

Why does
on each mini-batch
Batch Norm work?

first reason: BN makes weights, later or deeper than your network. (speed up)

Learning on shifting input distribution

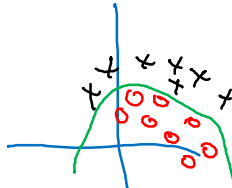
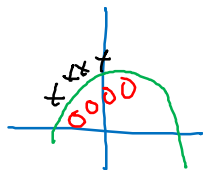
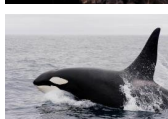
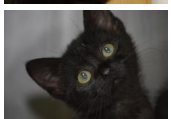
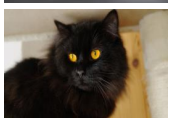
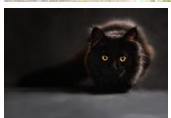


Cat

Non-Cat

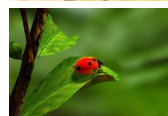
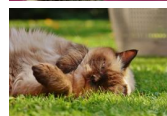
$y = 1$ ✓

$y = 0$



$y = 1$ ✓

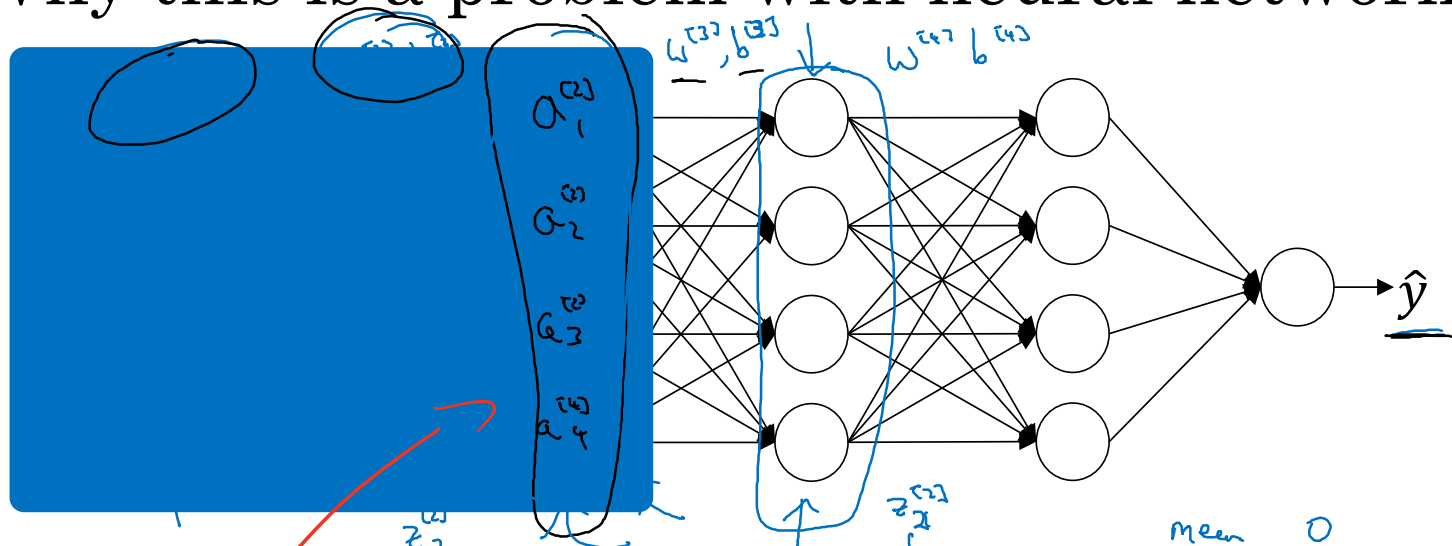
$y = 0$



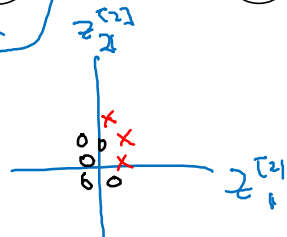
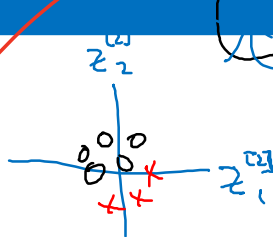
协变量转移
"Covariate shift"

$X \rightarrow Y$

Why this is a problem with neural networks?



from the perspective of the third hidden layer, these hidden unit values are changing all the time \rightarrow covariate shift.



mean 0
variance 1
 $\leadsto \beta^{[2,2]}, \beta^{[2,3]}$

forces the later layers to adapt to as early as layer changes is reduced

Batch Norm as regularization second reason X

- Each mini-batch is scaled by the mean/variance computed on just that mini-batch. $\tilde{z}^{[l]}$ $\{4, 128\}$ $z^{[l]}$ $X^{[t]}$
- This adds some noise to the values $z^{[l]}$ within that minibatch. So similar to dropout, it adds some noise to each hidden layer's activations. μ, σ^2
- This has a slight regularization effect.

mini-batch : 64 \longrightarrow 512
reduce effect