

ML/DL Study W02

- XOR – Neural Networks
- Forward/Back Propagation
- Sigmoid VS Relu
- Xavier VS He
- Drop out
- Batch Normalization

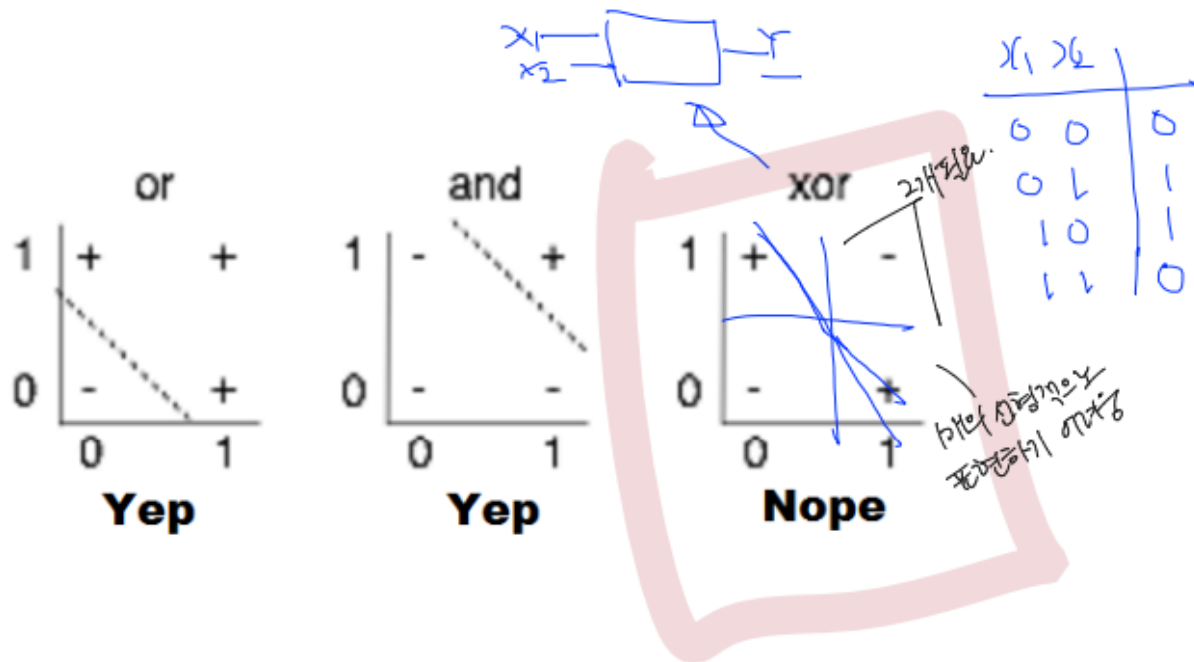
```
lookup.KeyValue  
f.constant(['em  
=tf.constant([0  
ce = tf.lookup.StaticV  
init,  
num_oov_buckets=5)
```

```
lookup.StaticVocabular  
initializer,  
num_oov_buckets,  
lookup_key_dtype=None  
name=None,  
experimental_is_open
```

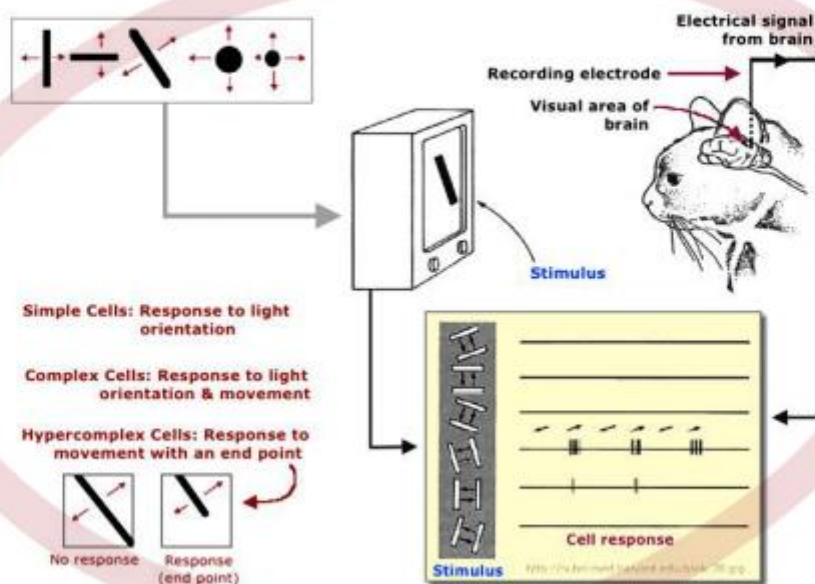
XOR – NN

XOR

(Simple) XOR problem: linearly separable?



Convolutional Neural Networks



22222 X

22222
X
X
X
X

42222 22222




$$[1 \ 1] \begin{bmatrix} 5 \\ 5 \end{bmatrix} - 8 = 5 + 5 - 8 = 2, \text{Sigmoid}(2) = 1$$

$$[1 \ 1] \begin{bmatrix} -7 \\ -7 \end{bmatrix} + 3 = -7 + -7 + 3 = -11, \text{Sigmoid}(-11) = 0$$

$$[1 \ 0] \begin{bmatrix} -11 \\ -11 \end{bmatrix} + 6 = -11 + 0 + 6 = -5, \text{Sigmoid}(-5) = 0$$

x_1	x_2	y_1	y_2	\bar{y}	XOR
0	0	0	1	0	0 ✓
0	1	0	0	1	1 ✓
1	0	0	0	1	1 ✓
✓ 1	1	1	0	0	0 ✓

3rd layer learning



```
K = tf.sigmoid(tf.matmul(X, W1) + b1)
Hypothesis = tf.sigmoid(tf.matmul(K, W2) + b2)
```

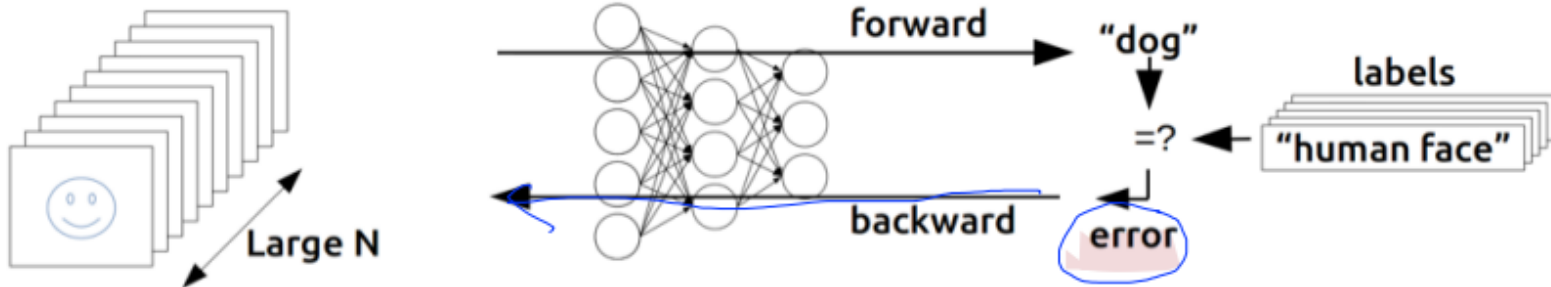
Forward/Back Propagation

Forward/Back Propagation

Backpropagation

(1974, 1982 by Paul Werbos, 1986 by Hinton)

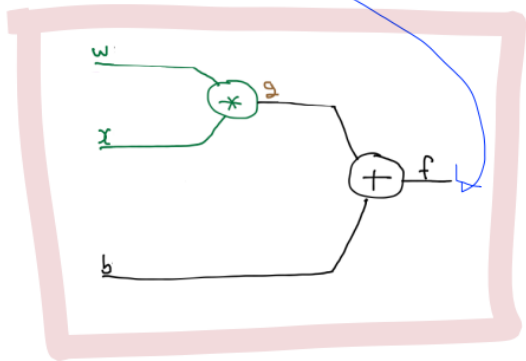
Training



Forward/Back Propagation

Back propagation (chain rule)

$$f = wx + b, g = wx, f = g + b$$



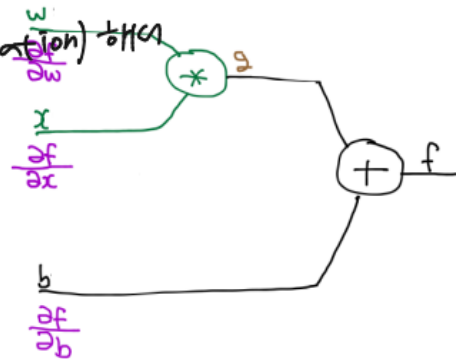
Back propagation (chain rule)



역전파.

$$f = wx + b, g = wx, f = g + b$$

내 target과 실제 model이 계산한 output 사이
cost를 구하고 이걸
뒤로(back) 전파(propagation) 해서
각 node가 갖는
변수를 갱신해내는 AI.

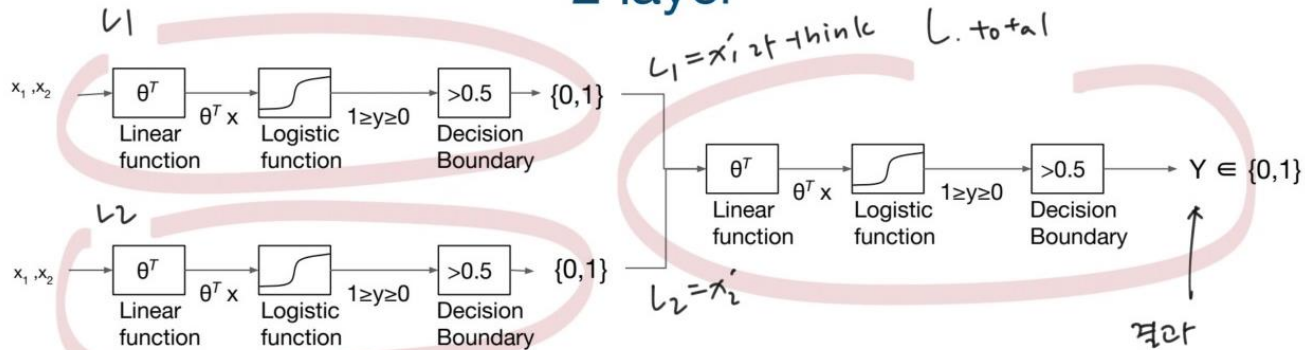


Sigmoid VS Relu

Sigmoid

Neural Net

2 layer

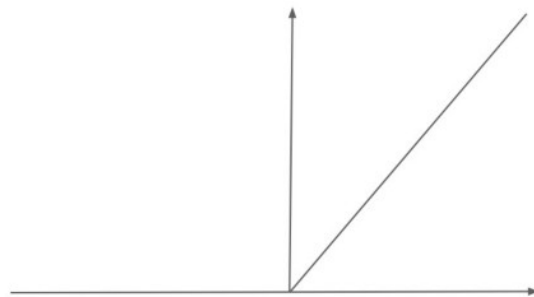
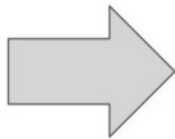
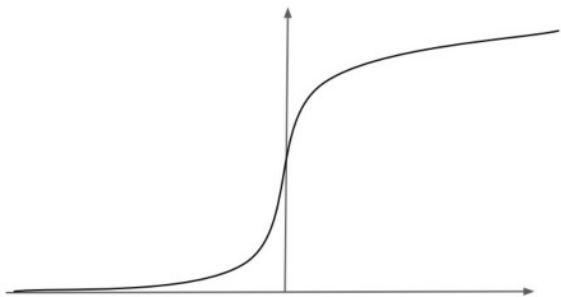


[Tensorflow Code]

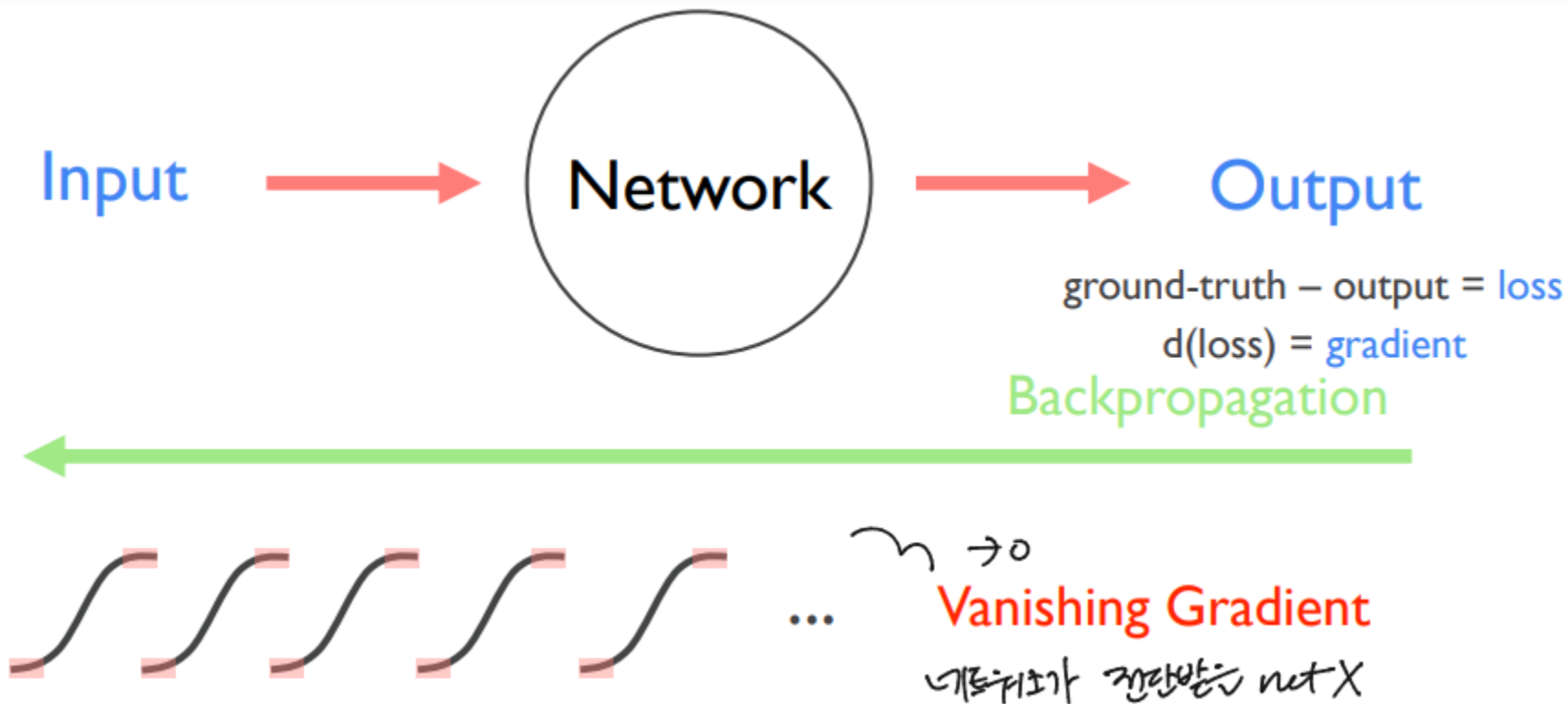
```
def neural_net(features):
    layer1 = tf.sigmoid(tf.matmul(features, W1) + b1) # W1=[2,1], b1=[1]
    layer2 = tf.sigmoid(tf.matmul(features, W2) + b2) # W2=[2,1], b2=[1]
    hypothesis = tf.sigmoid(tf.matmul(tf.concat([layer1, layer2], -1), W3) + b3) # W3=[2,1], b3=[1]
    return hypothesis
```

Sigmoid

What's Next?
Sigmoid -> Relu



Problem of Sigmoid

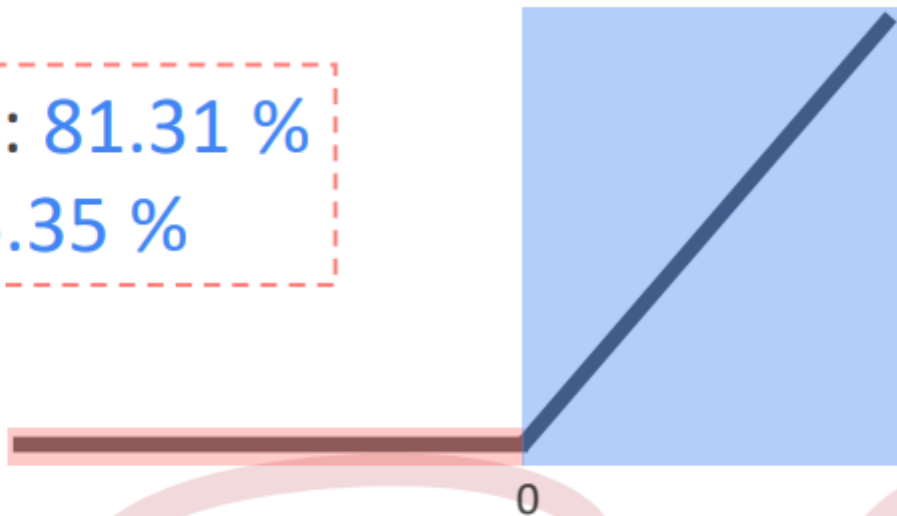


Why Relu?

$$f(x) = \max(0, x)$$

Sigmoid : 81.31 %

Relu : 85.35 %



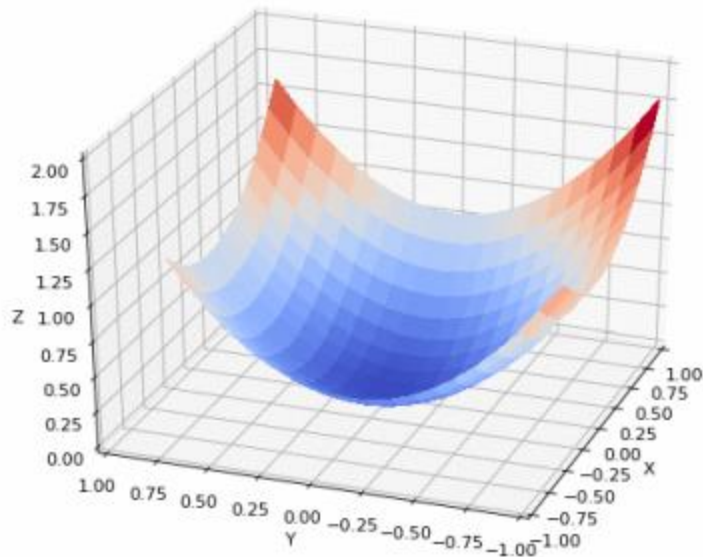
`tf.keras.activations`

sigmoid, tanh
relu, elu, selu

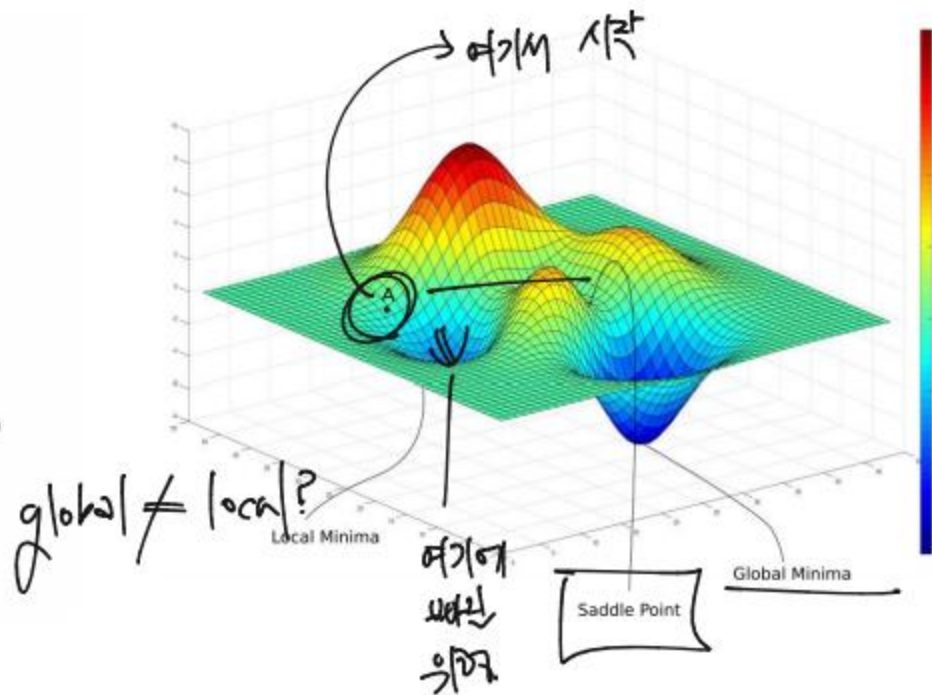
functional or you can use

Weight Initialization

Xavier VS He



global min.



Xavier VS He

Create network

```
weight_init = tf.keras.initializers.RandomNormal()
```

```
weight_init = tf.keras.initializers.glorot_uniform()
```

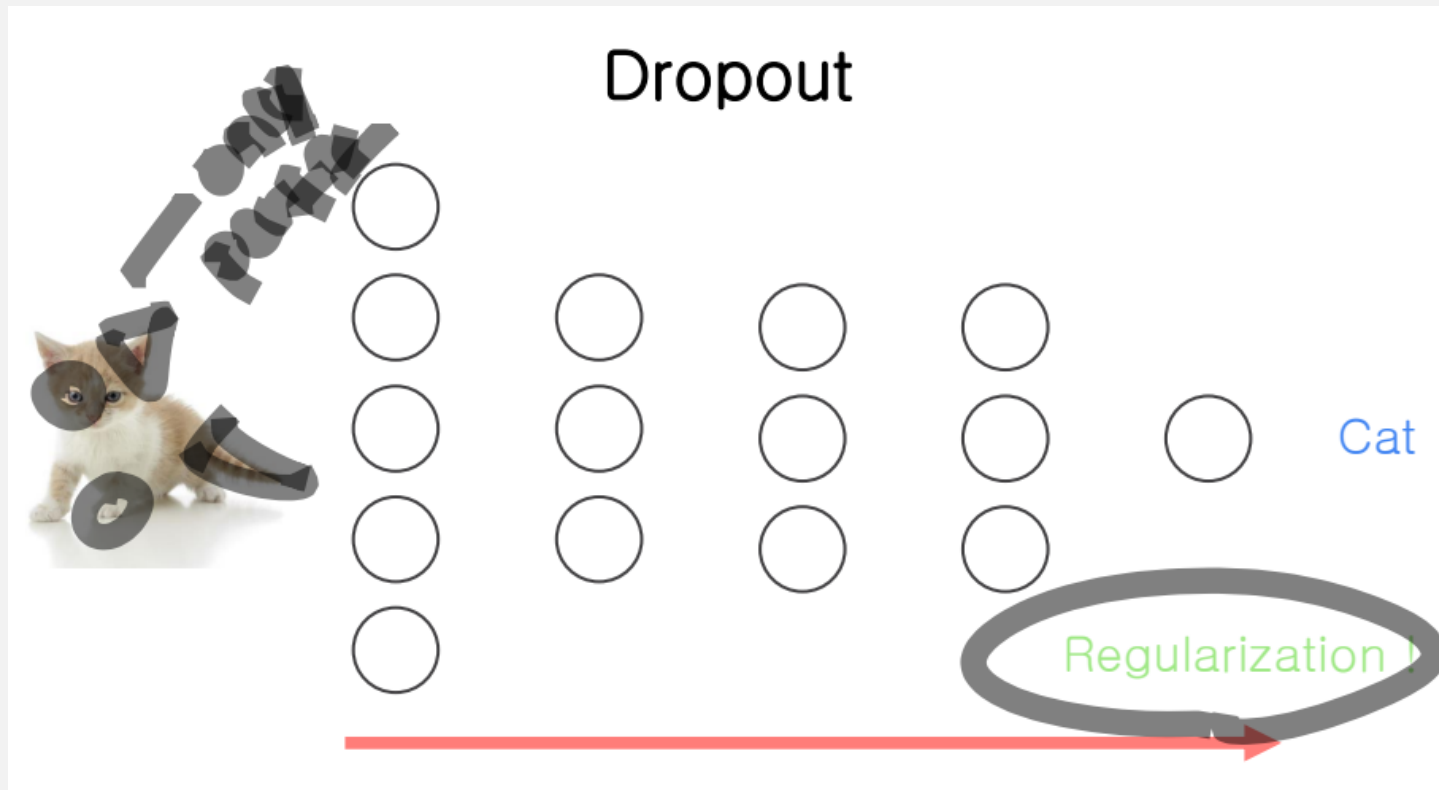
```
weight_init = tf.keras.initializers.he_uniform()
```

Random : 85.35 %

Xavier : 96.50 %

Drop out

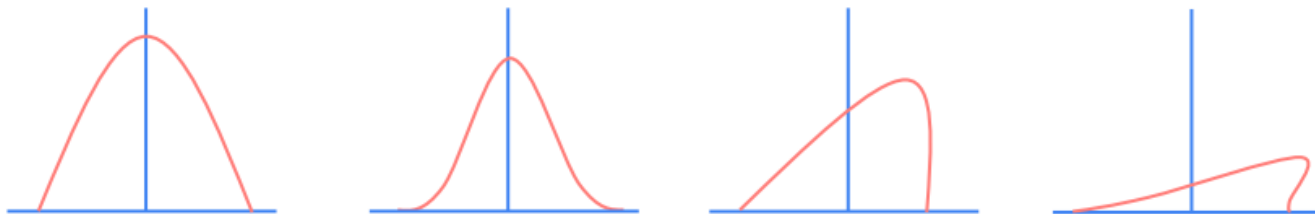
Drop out



Batch Normalization

Batch Normalization

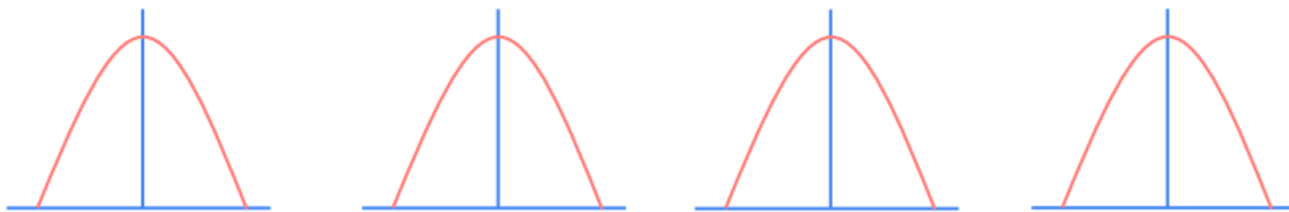
Batch Normalization



Internal Covariate Shift

Batch Normalization

Batch Normalization



$$\bar{x} = \frac{x - \mu_B}{\sqrt{\sigma_B^2 + \epsilon}}$$

$$\hat{x} = \gamma \bar{x} + \beta$$