

AI and Deep Learning

Multi-Layer Neural Networks and Non-linear Decision Boundary

Jeju National University

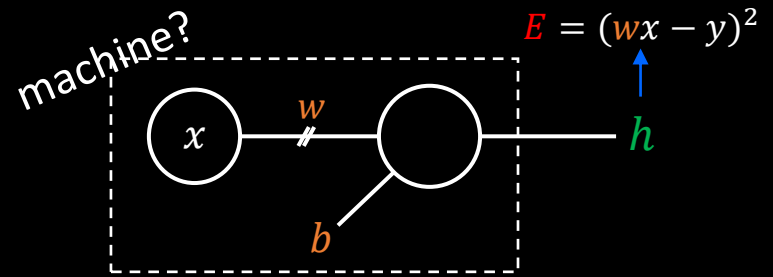
Yungcheol Byun

Agenda

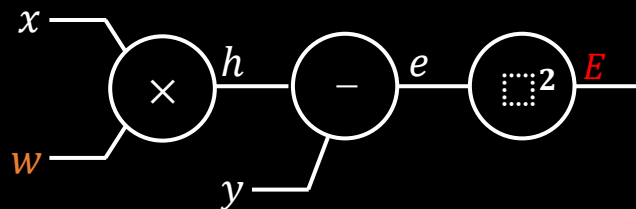
- Machine Learning revisited
- Linear decision boundary
- XOR problem
- Non-linearity and multi-layer neural networks
- Complex nonlinear boundary as you like

Machine Learning

A neuron



- ① Parameters(w, b) initialization randomly
- ② Building **computation graph** of E by TensorFlow framework
- ③ Foreword propagation by putting values into the graph and calculate E
- ④ Back-propagation to get the influence of w, b on the error (applying chain rules)
- ⑤ **Update w, b to adjust a decision boundary**
- ⑥ go to ③



How to change the decision boundary

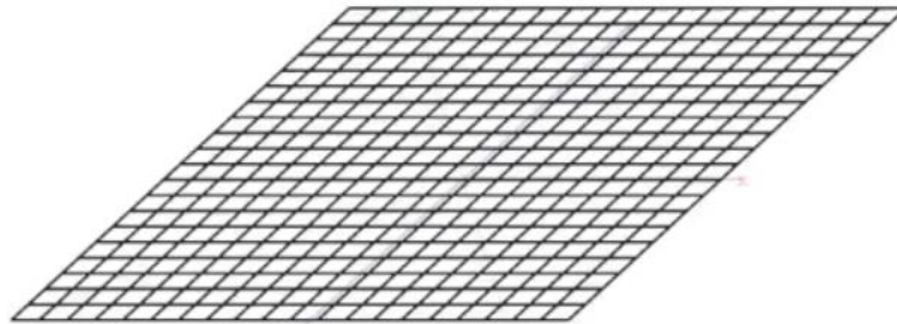
`sigmoid(w1·length + w2·width + b)`

db slope

db shift

$$w_1 x_1 + w_2 x_2 + b = 0$$

db rotation



slope

rotation

shift

`surface(f(x,z)=sig(w1·x+w2·z+b))`

`w1 = 0.00`

`w2 = 0.00`

`b = 0.00`

Machine Learning

```
import tensorflow as tf
#----- training data
x_data = [-2, -1, 1, 2]
y_data = [0, 0, 1, 1]
#----- a neuron
w = tf.Variable(tf.random_normal([1])) ①
hypo = tf.sigmoid(x_data * w)
#----- learning
cost = -tf.reduce_mean(y_data * tf.log(hypo) +
                        tf.subtract(1., y_data) * tf.log(tf.subtract(1., hypo)))
train =
tf.train.GradientDescentOptimizer(learning_rate=0.01).
minimize(cost)
```

Machine Learning

```
sess = tf.Session()  
sess.run(tf.global_variables_initializer()) ①
```

```
②  
for i in range(1001):  
    sess.run(train) ③④⑤
```

```
        if i % 100 == 0:  
            print( ' w: ' , sess.run(w), ' cost: ' , sess.run(cost))
```

Learning finished after
1001 times updates

```
#----- test (classification)  
x_data = [-2, 4]  
print(sess.run(hypo))
```

*hypo = tf.sigmoid(x_data * w)*

Testing new data

- After learning,
- a neuron can classify new input data correctly.

```
#----- test (classification)
x_data = [-2, 4]
print(sess.run(hypo))
```

- When the computation graph is created, data is set only one time.
- Still **old data** in the computation graph.
- No feeding the new data[-2,4] into the precreated computational graph
- **Therefore, not working!**

Place Holder

- **Marking** certain place holders in a computational graph
- and then **replace** it with new data when it runs(evaluation).

`sess.run ()`

Place Holder

```
import tensorflow as tf
```

```
#----- training data
```

```
x_data = [[-2], [-1], [1], [2]]
```

```
y_data = [[0], [0], [1], [1]]
```

```
X = tf.placeholder(tf.float32)
```

```
Y = tf.placeholder(tf.float32)
```

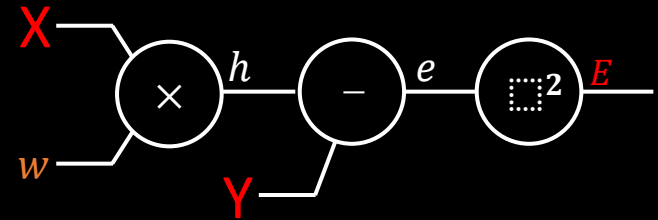
```
#----- a neuron
```

```
w = tf.Variable(tf.random_normal([1]))
```

```
hypo = tf.sigmoid(X * w)
```

1. Place holders : X, Y
2. $x_data \rightarrow X$, $y_data \rightarrow Y$
3. Feeding real data when hypo, cost, and train operations are executed.

Place Holder



```
#----- learning
```

```
cost = -tf.reduce_mean(Y * tf.log(hypo) +  
    tf.subtract(1., Y) * tf.log(tf.subtract(1., hypo)))
```

```
train =
```

```
tf.train.GradientDescentOptimizer(learning_rate=0.01).minimize(cost)
```

```
sess = tf.Session()
```

```
sess.run(tf.global_variables_initializer())
```

```
for i in range(1001):
```

```
    sess.run(train, feed_dict={X:x_data, Y:y_data})
```

```
    if i % 100 == 0:
```

```
        print(sess.run(w), sess.run(cost, feed_dict={X:x_data, Y:y_data}))
```

Place Holder

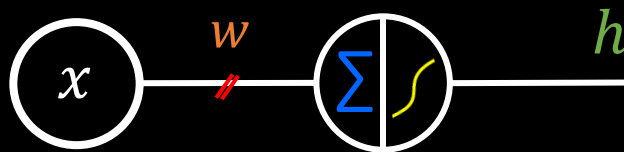
```
#----- testing(classification)
x_data = [-2, 4]
result = sess.run(hypo, feed_dict={X: x_data})
print(result)
```



Lab 15.py

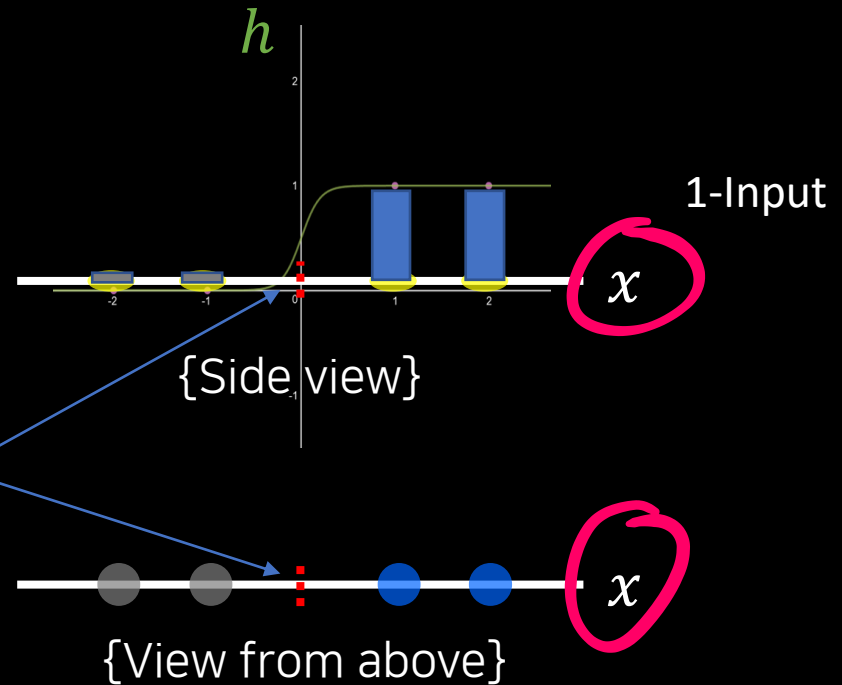
- Classification into one of **four classes**
- Using placeholders

1-Input Neuron

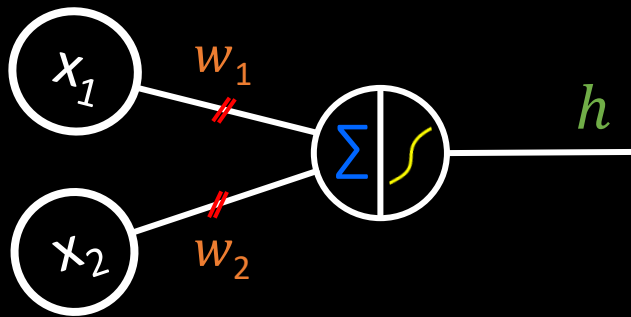


Decision boundary : Value

$$w \cdot x = 0$$
$$x = 0$$

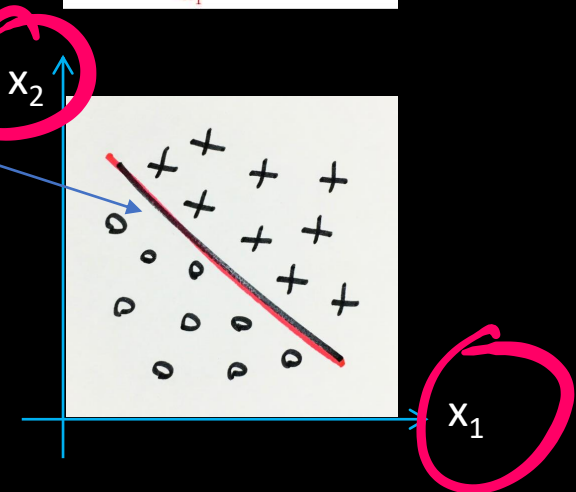
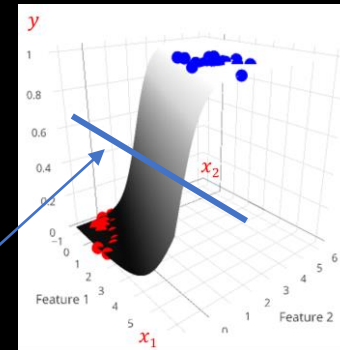


2-Input Neuron

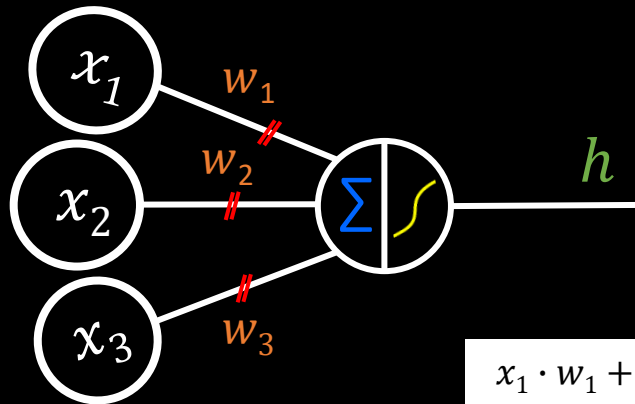


Decision boundary : Line

$$x_1 \cdot w_1 + x_2 \cdot w_2 = 0$$

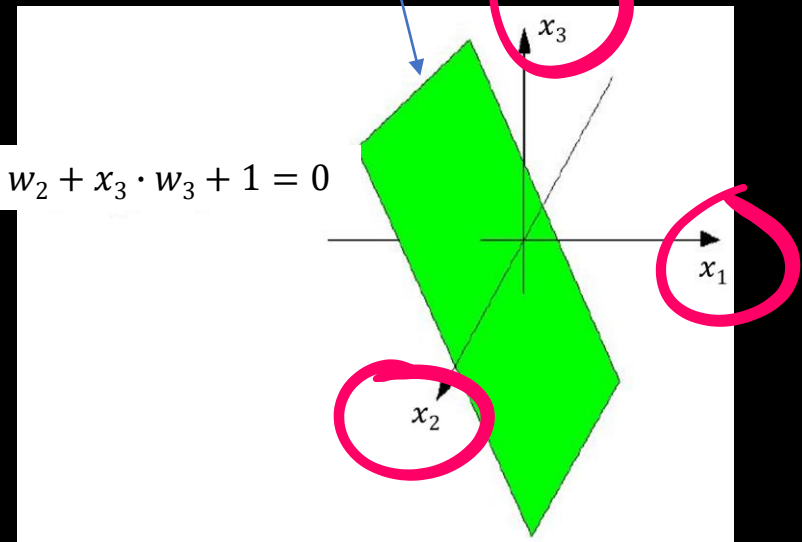


3-Input Neuron



Decision boundary : Plane

$$x_1 \cdot w_1 + x_2 \cdot w_2 + x_3 \cdot w_3 + 1 = 0$$



More than 4 inputs?

$$x_1w_1 + x_2w_2 + x_3w_3 + x_4w_4 + b = 0$$

→ hyperplane

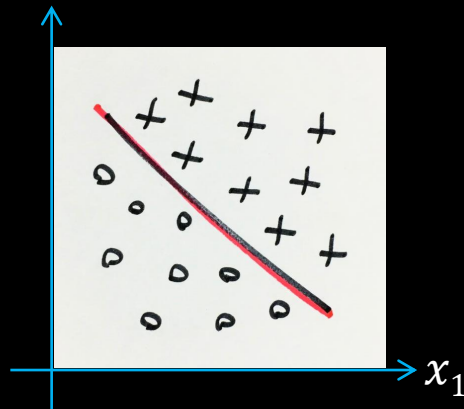
초평면

Linear Decision Boundary

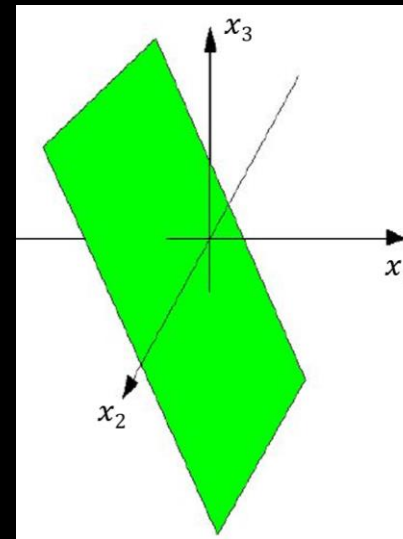
value



line



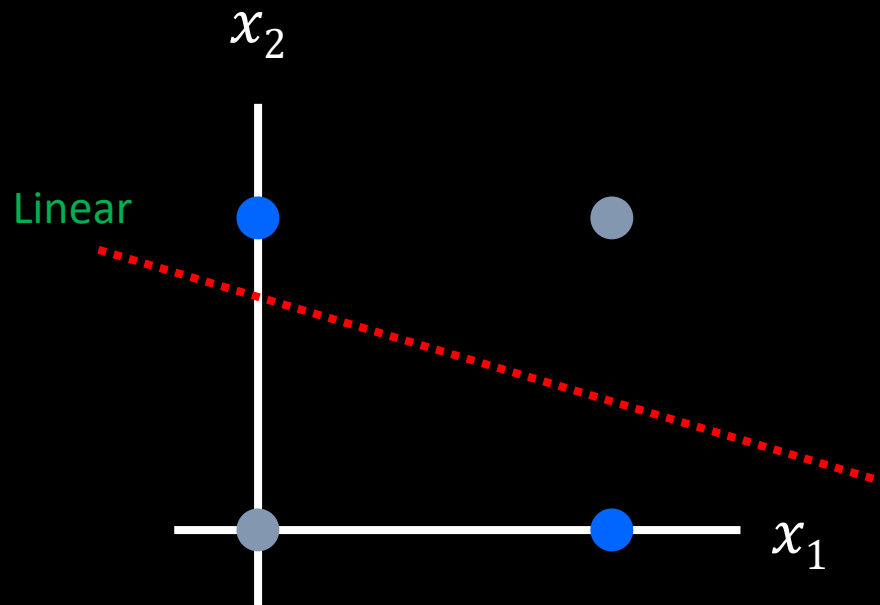
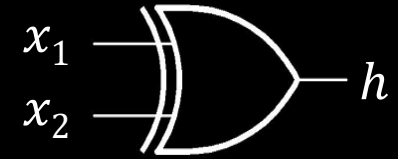
plane



hyperplane



XOR



x_1	x_2	h
0	0	0
0	1	1
1	0	1
1	1	0

Truth table

View from above

XOR

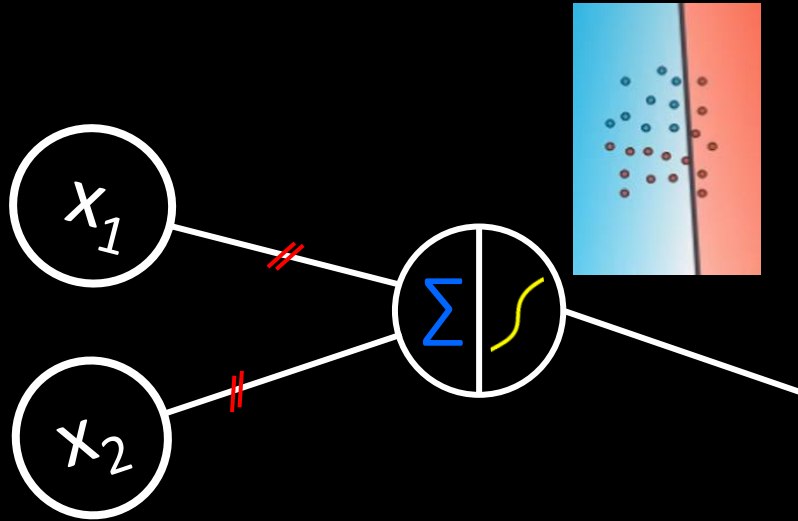
Binary Cross-Entropy Loss

- Number of class : 2 (0 or 1)
- 1-decision boundary for 2-class classification
- However, **impossible** to classify using a single linear decision boundary

Lab 16.py

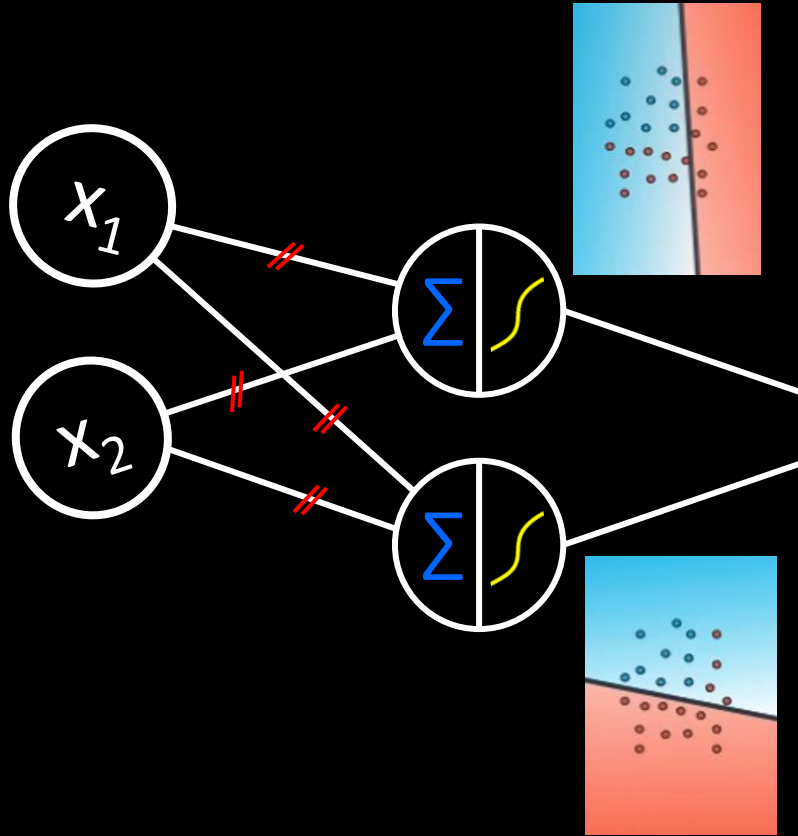
- XOR problem
- A neuron, 1 linear decision boundary
- Cannot be solved!

How to solve



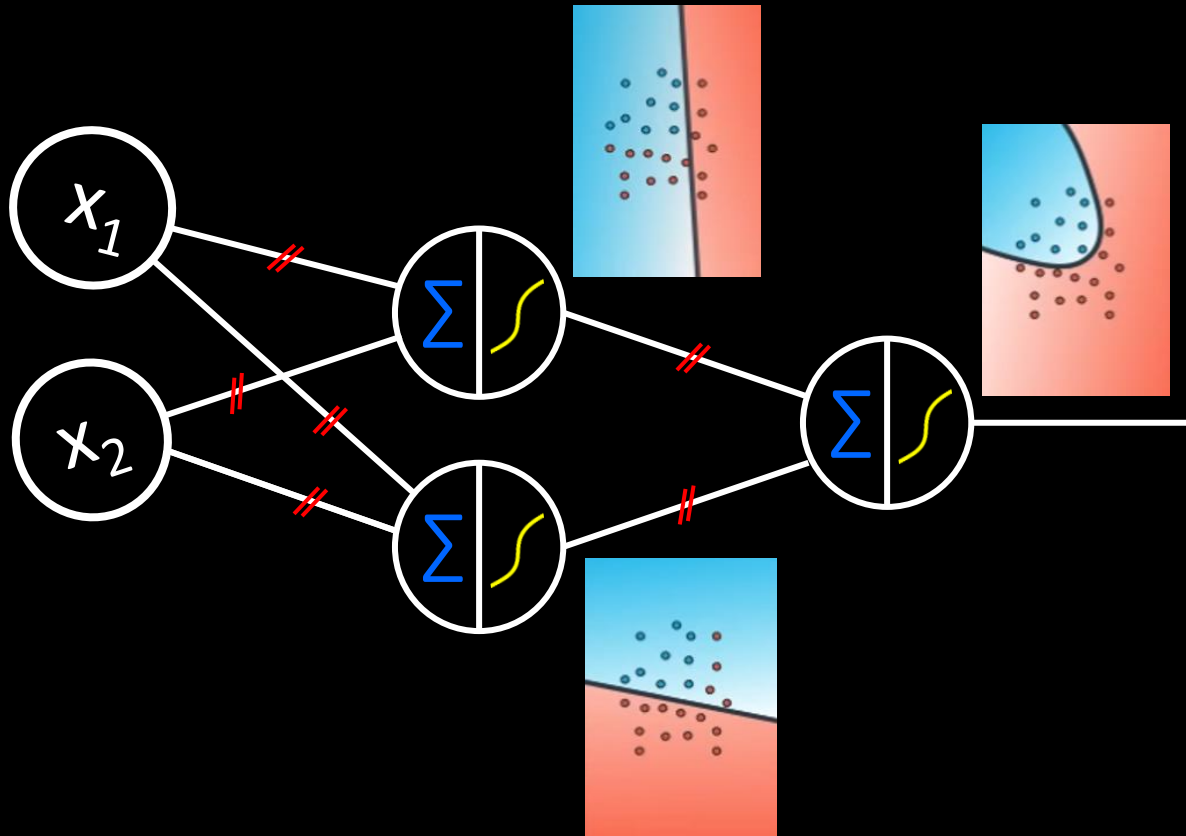
by Luis Serrano, A friendly introduction to Deep Learning, UDACITY

How to solve



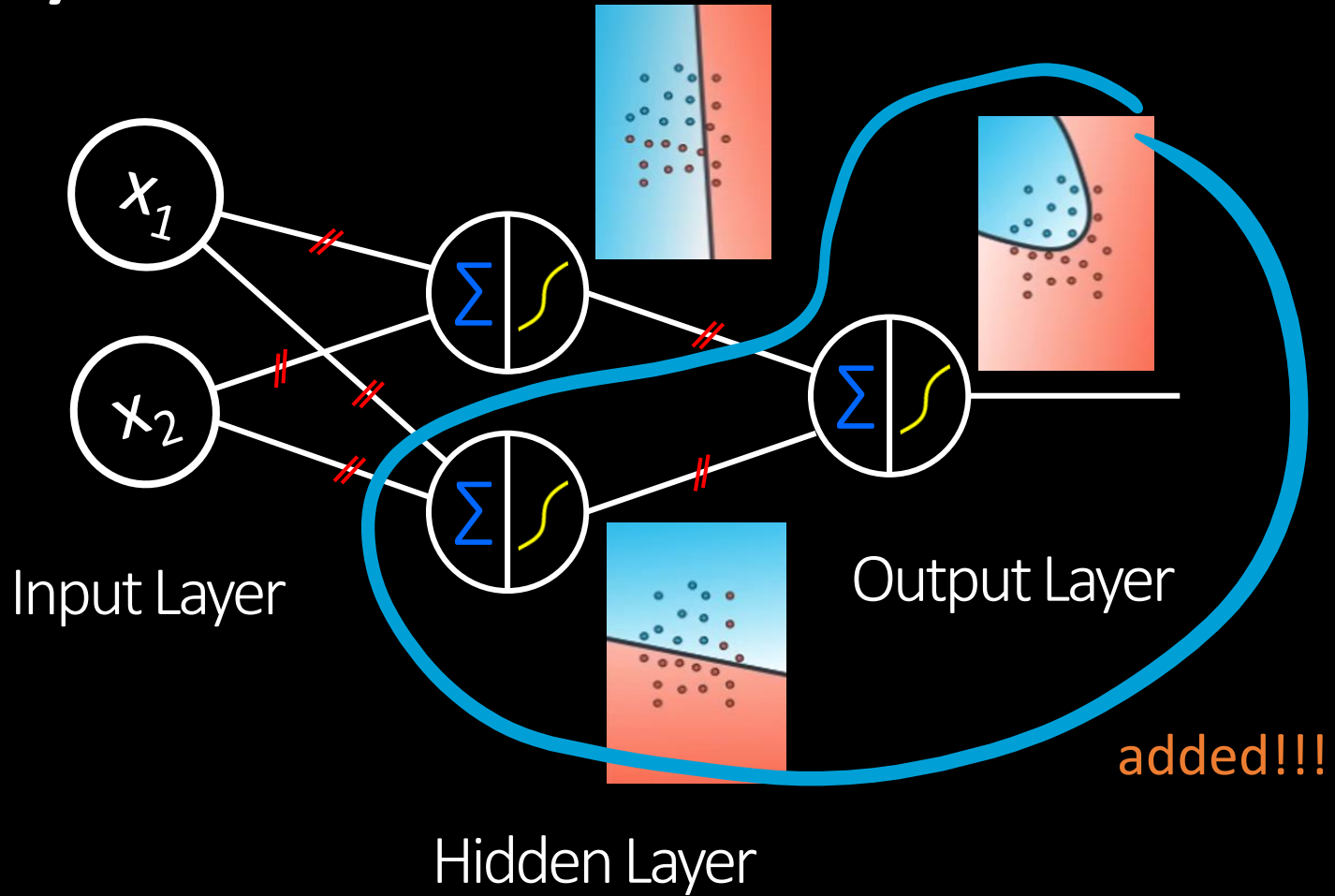
by Luis Serrano, A friendly introduction to Deep Learning, UDACITY

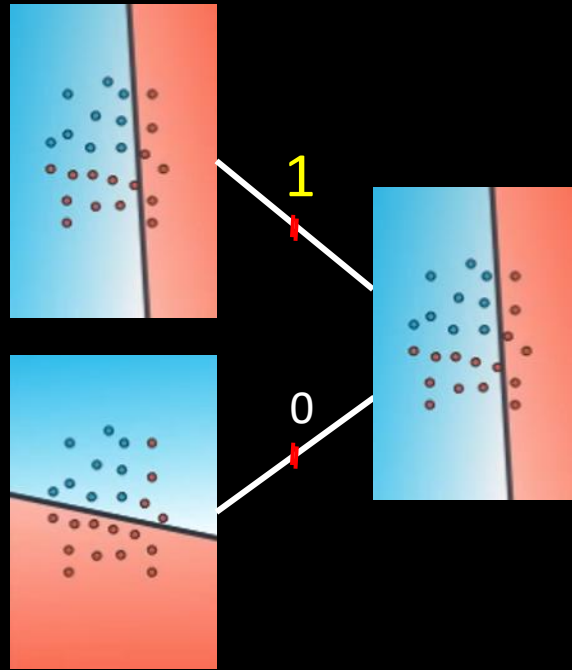
How to solve

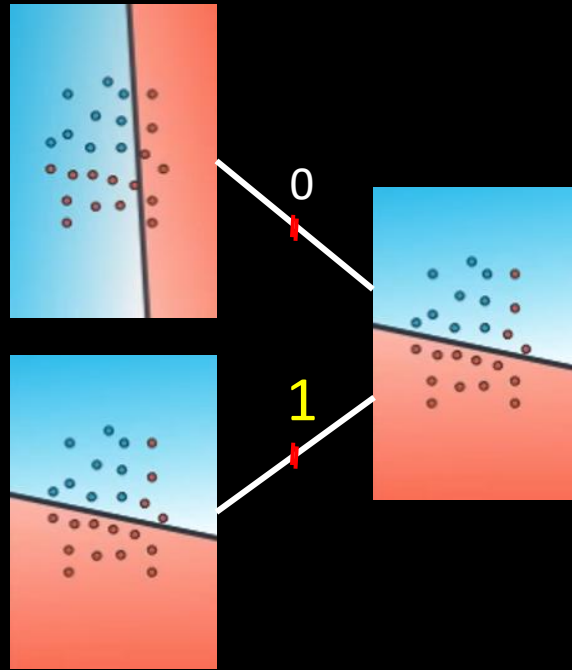


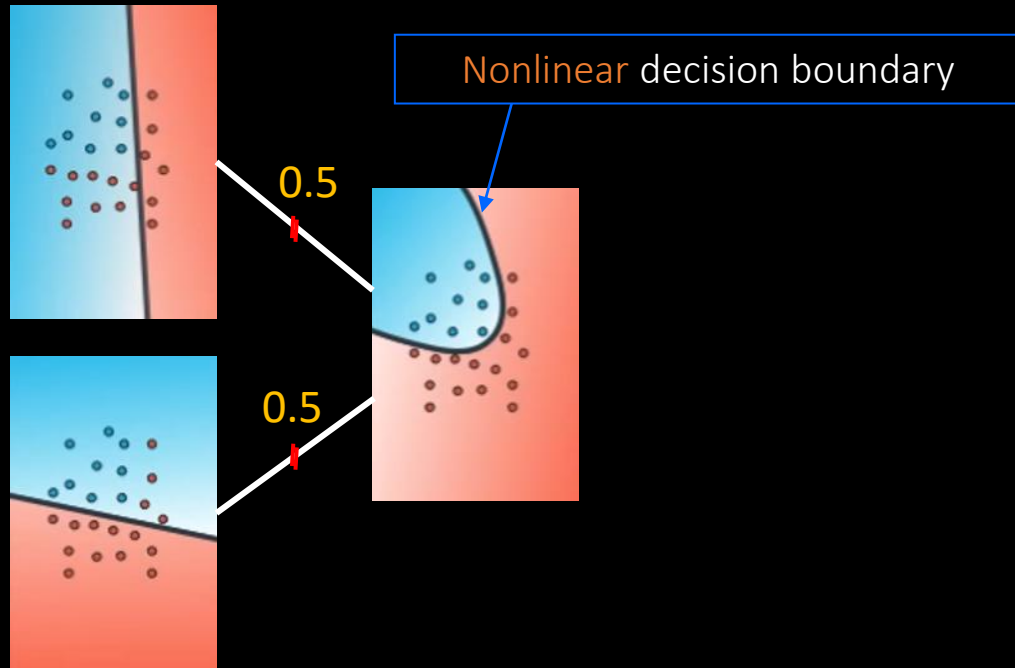
by Luis Serrano, A friendly introduction to Deep Learning, UDACITY

3-layer Neural Network





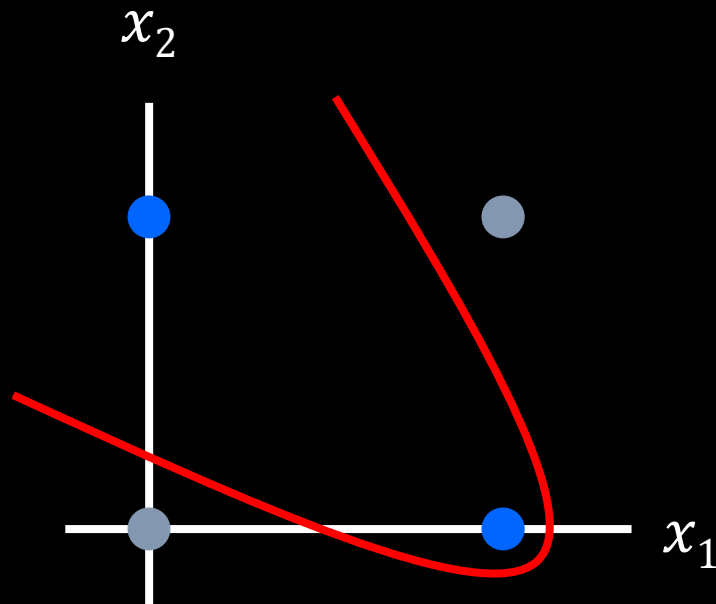




What is a **linear** decision boundary?

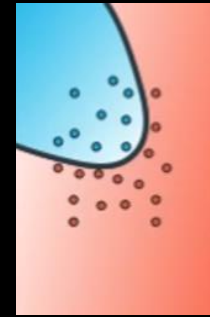
3-layer NN for nonlinear
decision boundary

XOR



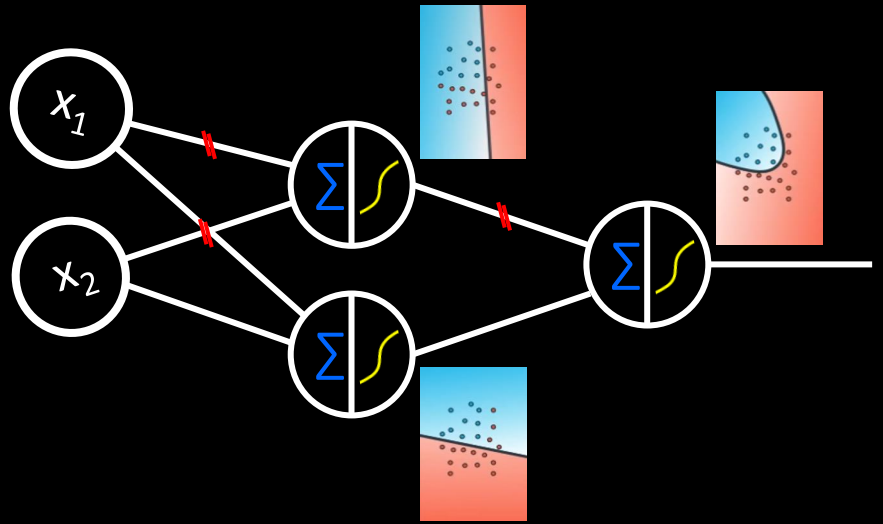
View from above

View from above

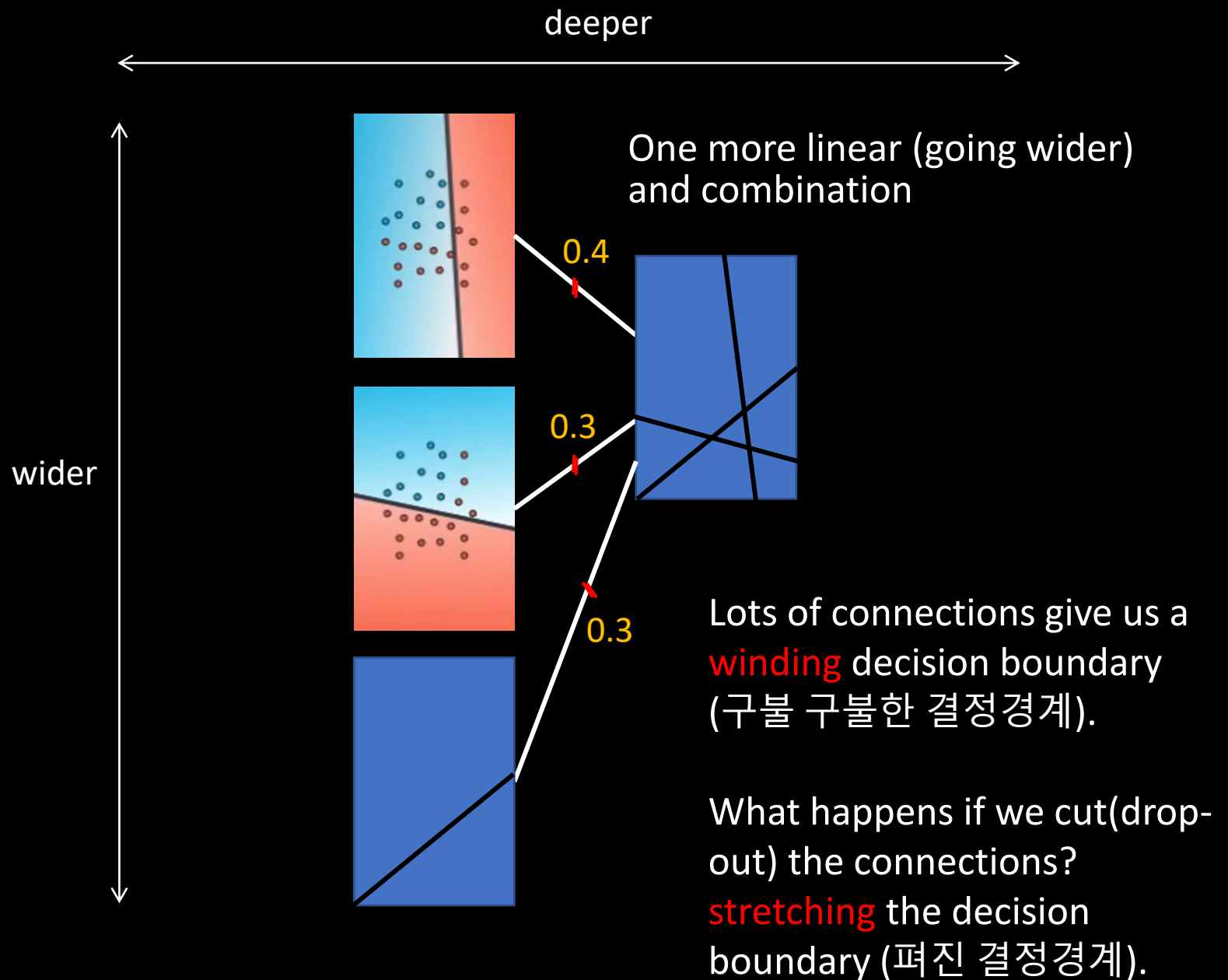


Side view

Lab 17.py

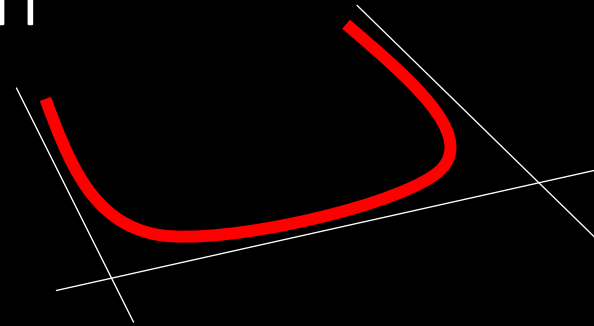


- Solving XOR gate problem using 3-layer neural network
- The way to create non-linear decision boundary



Nonlinear Decision Boundary

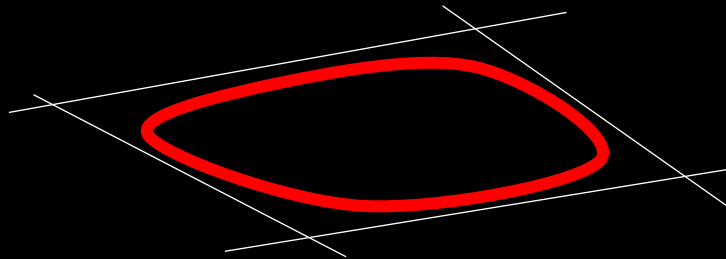
- Combination of **three** linear decision boundaries



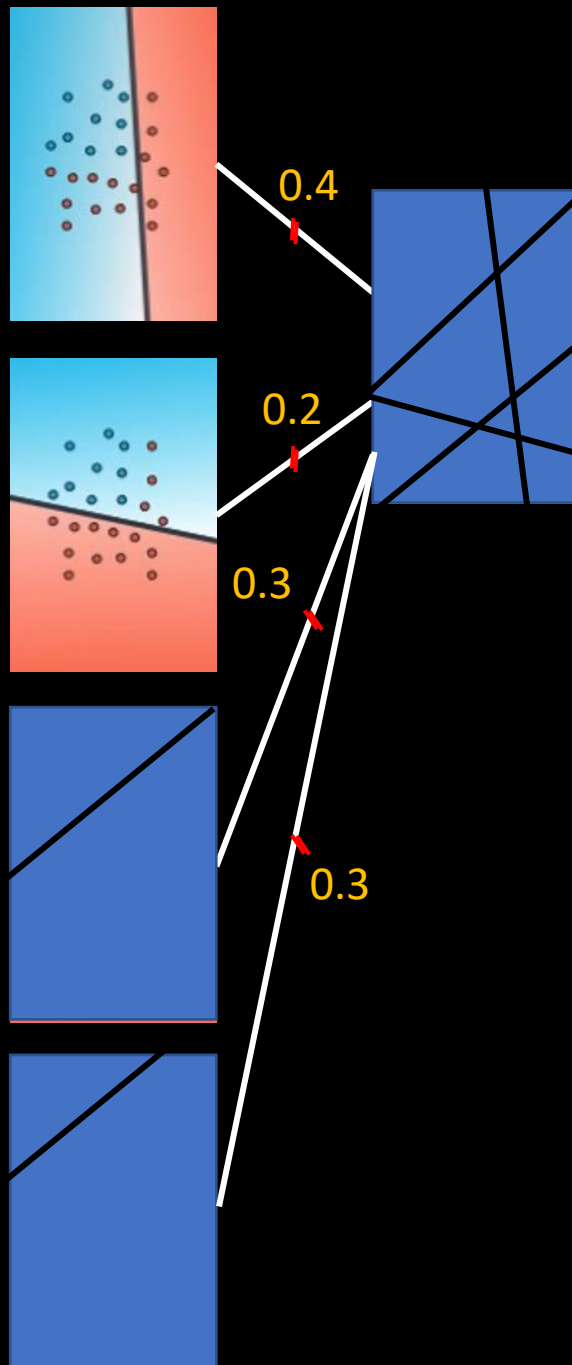


Nonlinear Decision Boundary

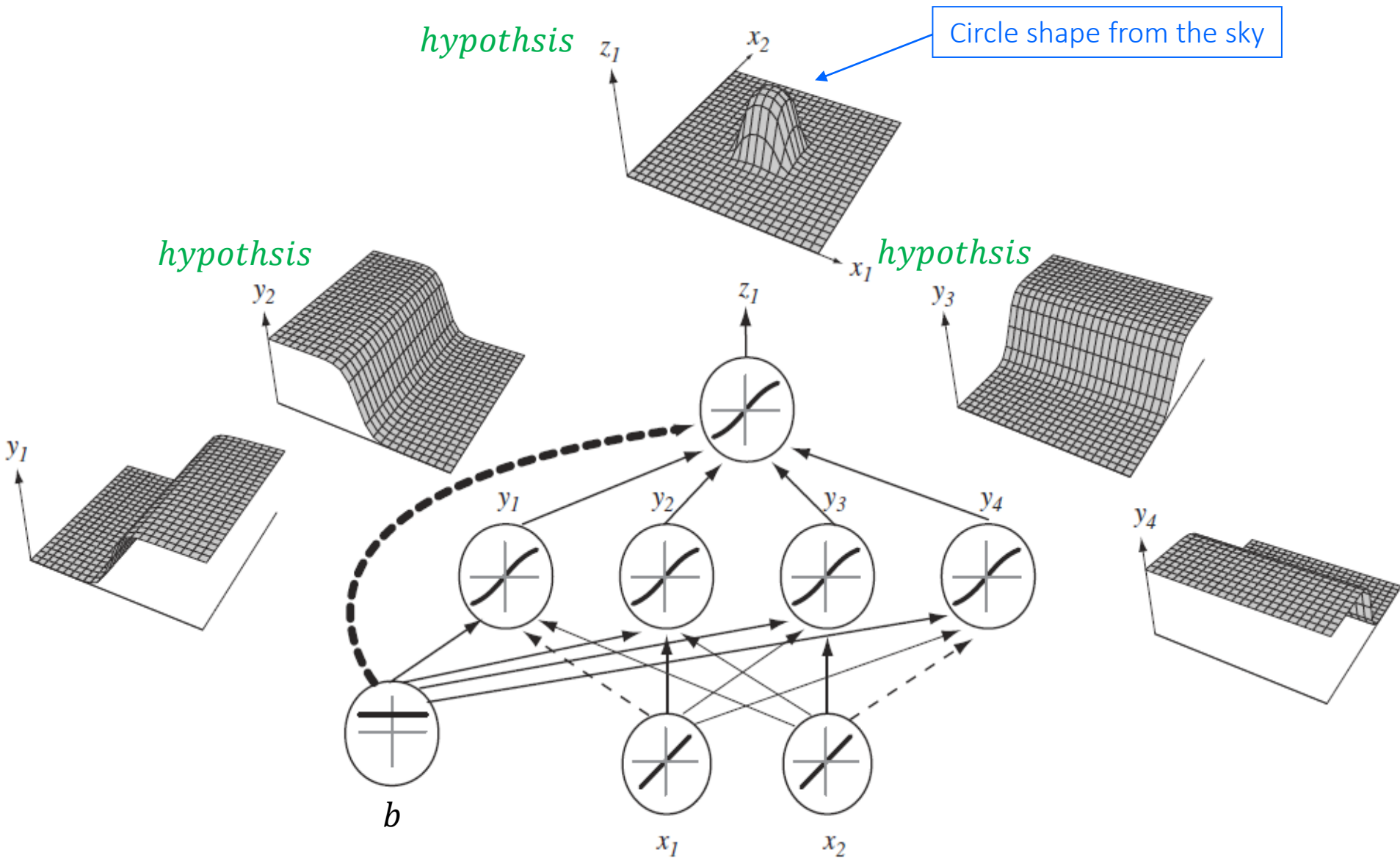
- Merging **four** linear decision boundaries



View from above



Nonlinear Decision Boundary



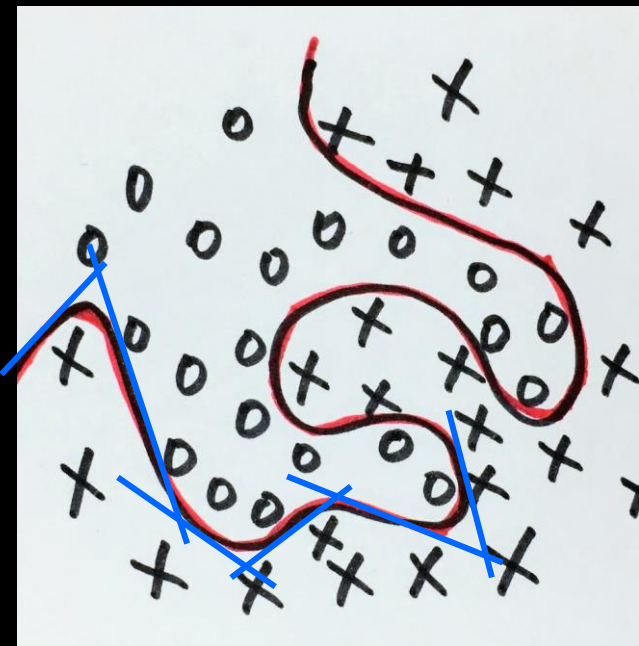
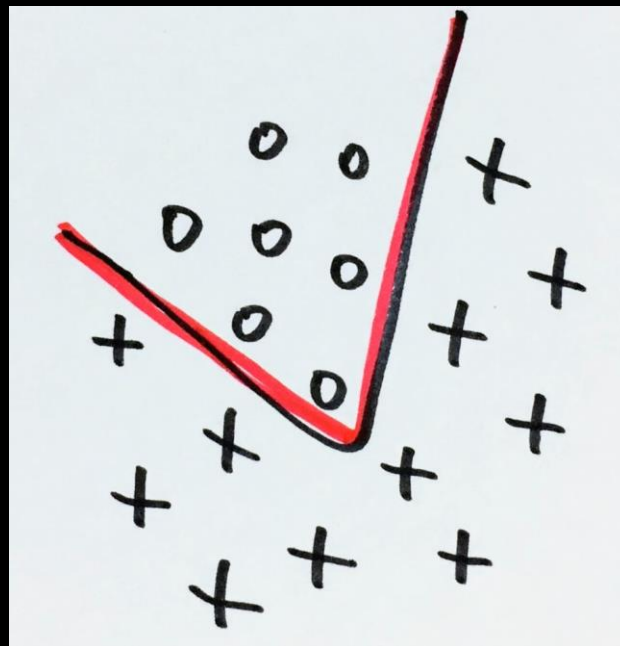
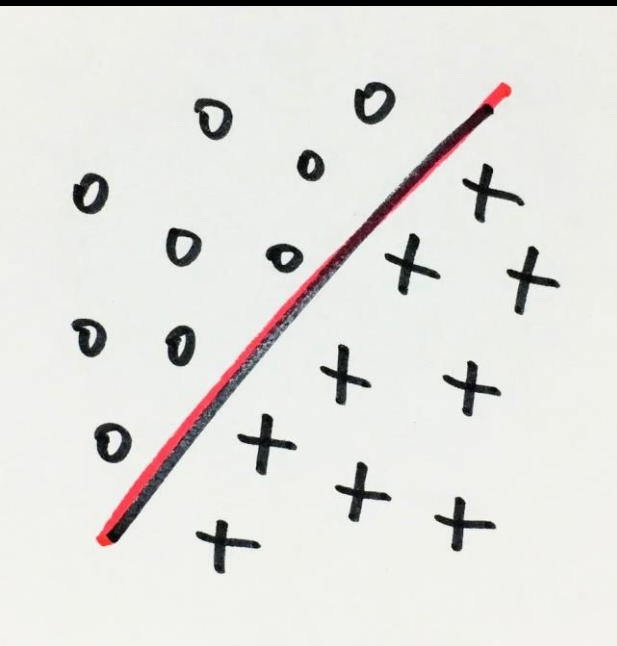


View from above

Side view

As you wish (2 classes)

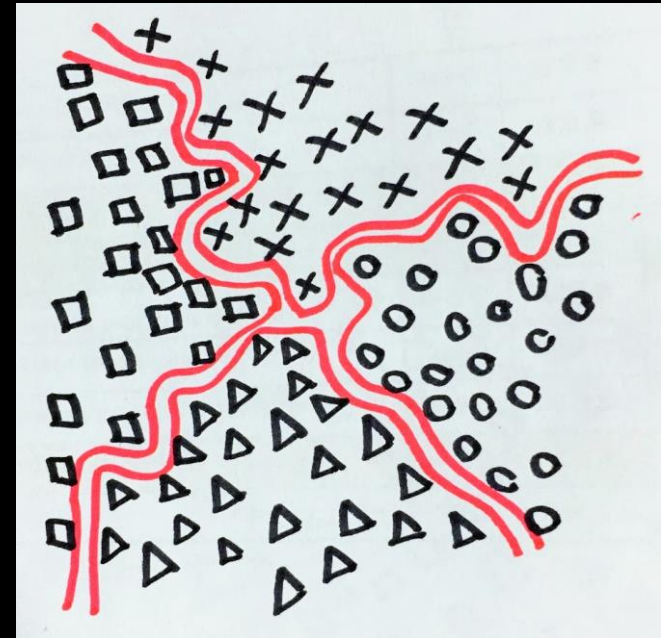
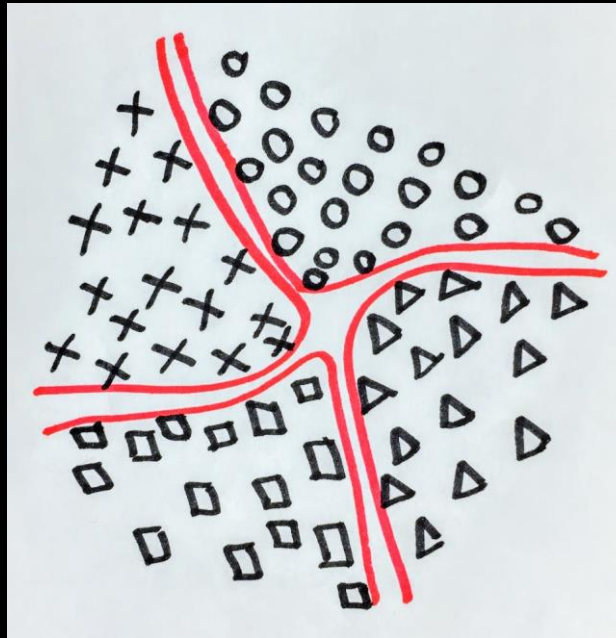
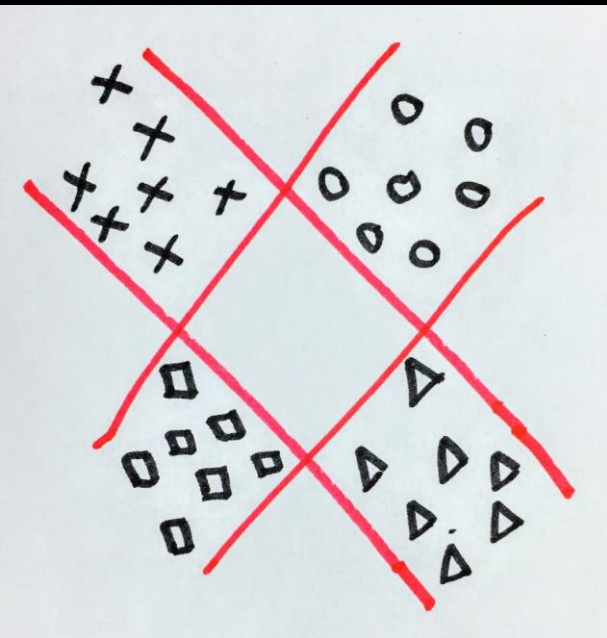
x^2



x_1

View from above

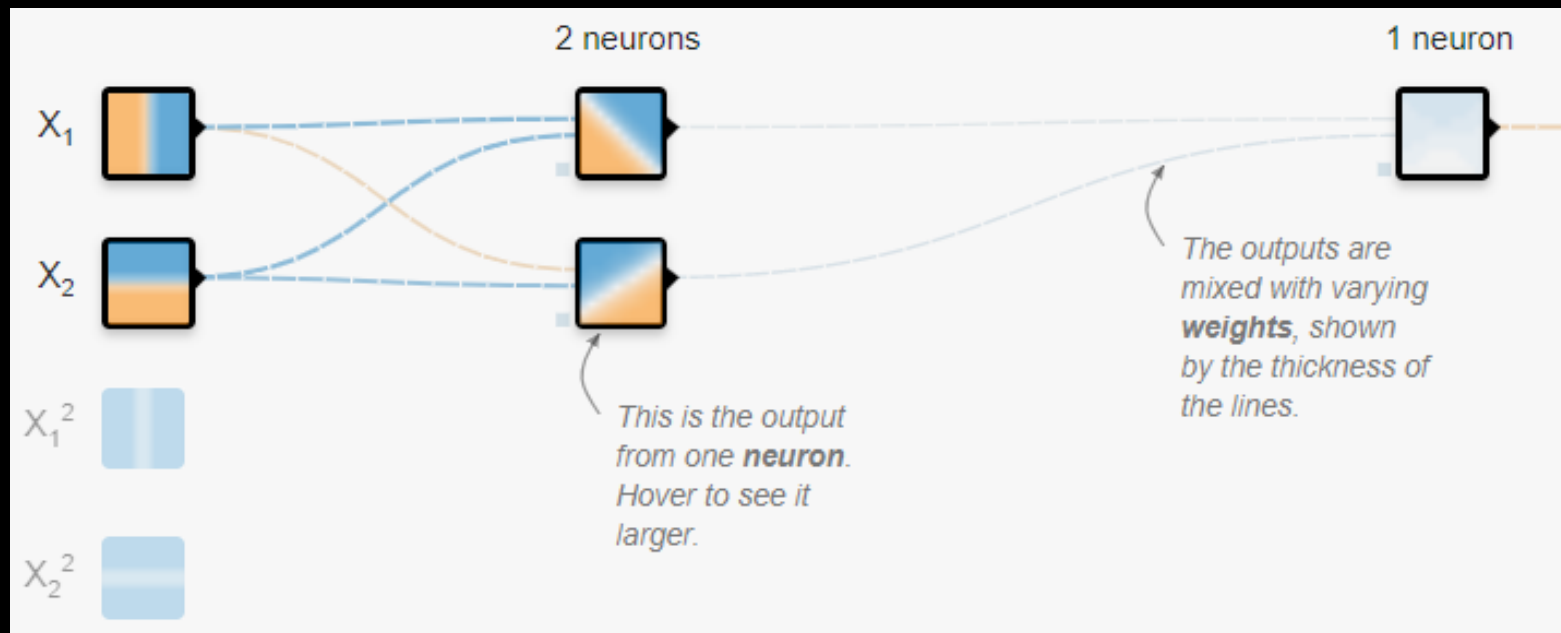
As you wish (4 classes)



From the above

More neurons (wider & deeper),
more complex(detail)
decision boundary

<http://playground.tensorflow.org>



As you wish

go wider & deeper

- to make more complex nonlinear decision boundaries.
- We can classify anything we imagine.

The way of machine learning

- Learning over and over again just like human being
- If it misrecognizes, just say 'Nope, you were wrong', which makes it update its weights to do better next time.
- Try it over and over again just like a child.

Learning or Programming?

“This (machine learning) is the next transformation...the programming paradigm is changing. Instead of programming a computer, you teach a computer to learn something and it does what you want”

— Eric Schmidt, Google



Change of Paradigm

Not programming,
but data-driven learning
(parameter tuning)