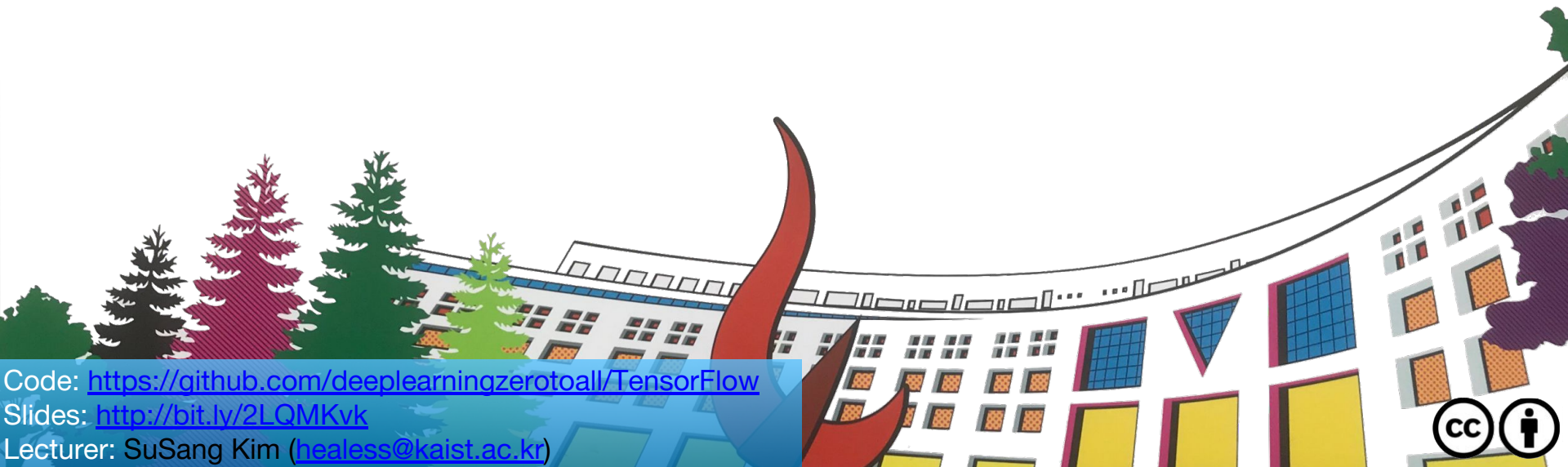


ML/DL for Everyone Season2

with  TensorFlow

Lab 09-1 Neural Nets for XOR



Code: <https://github.com/deeplearningzerotoall/TensorFlow>

Slides: <http://bit.ly/2LQMKvk>

Lecturer: SuSang Kim (healess@kaist.ac.kr)

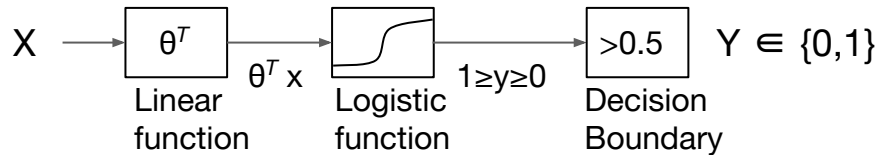
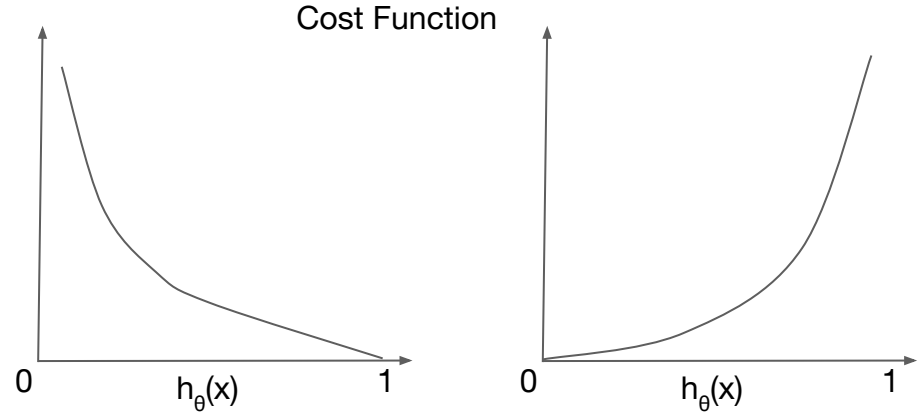
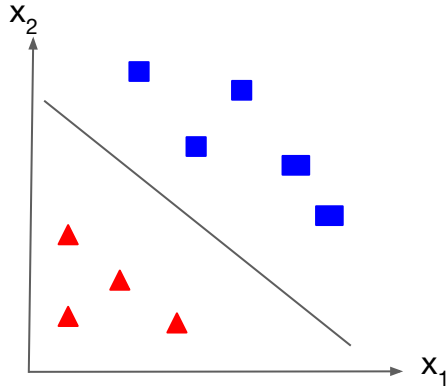


Neural Nets for XOR

- XOR Problem
 - Logistic Regression (Recap)
 - Neural Network
 - Chart
 - Codes (Eager Execution)
 - Summary

Logistic Regression

Recap

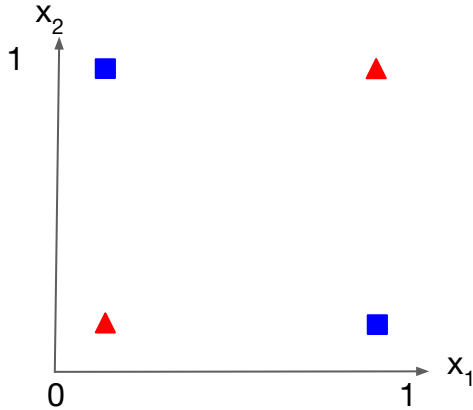


$$\text{Cost}(h_{\theta}(x), y) = \begin{cases} -\log(h_{\theta}(x)) & \text{if } y = 1 \\ -\log(1 - h_{\theta}(x)) & \text{if } y = 0 \end{cases}$$

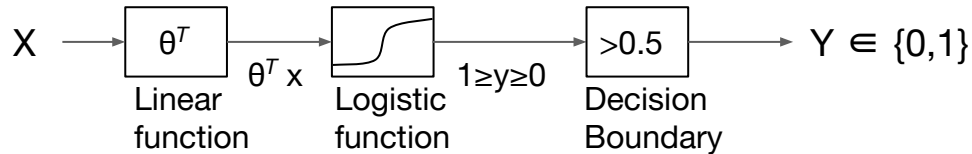
$$\text{cost}(h_{\theta}(x), y) = -y \log(h_{\theta}(x)) - (1-y) \log(1 - h_{\theta}(x))$$

Logistic Regression

XOR

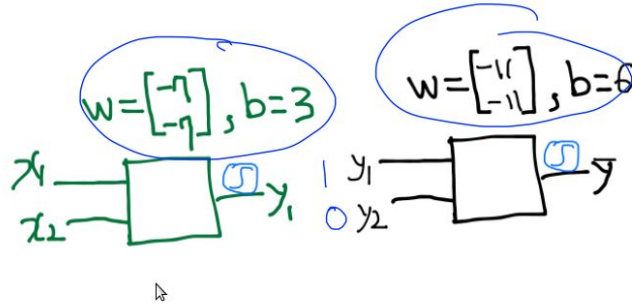
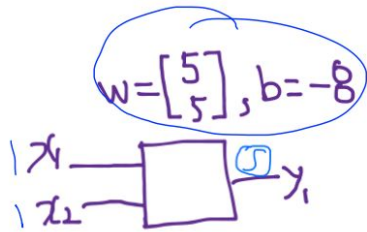


x_1	x_2	XOR
0	0	0
0	1	1
1	0	1
1	1	0



Data sets

Forward propagation

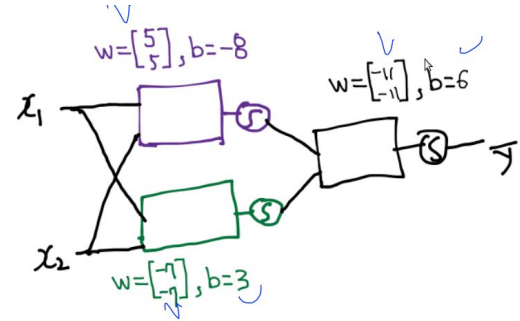


$$[1 \ 1] \begin{bmatrix} 5 \\ 5 \end{bmatrix} - 8 = 5 + 5 - 8 = 2, \text{Sigmoid}(2) = 1$$

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

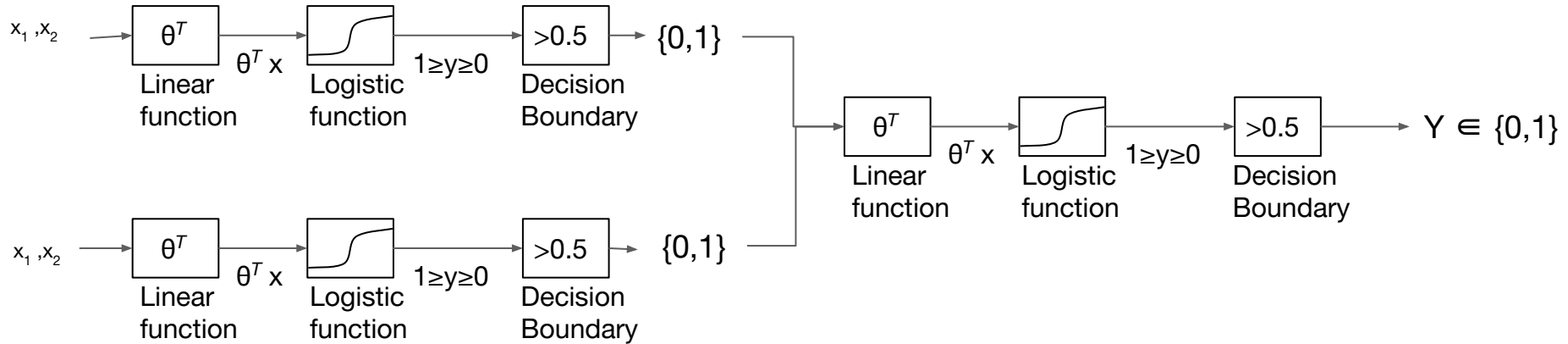
$$[1 \ 0] \begin{bmatrix} -11 \\ -11 \end{bmatrix} + 6 = -11 + 0 + 6 = -5, \text{Sigmoid}(-5) = 0$$

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 ✓



Neural Net

2 layer



[Tensorflow Code]

```
def neural_net(features):
```

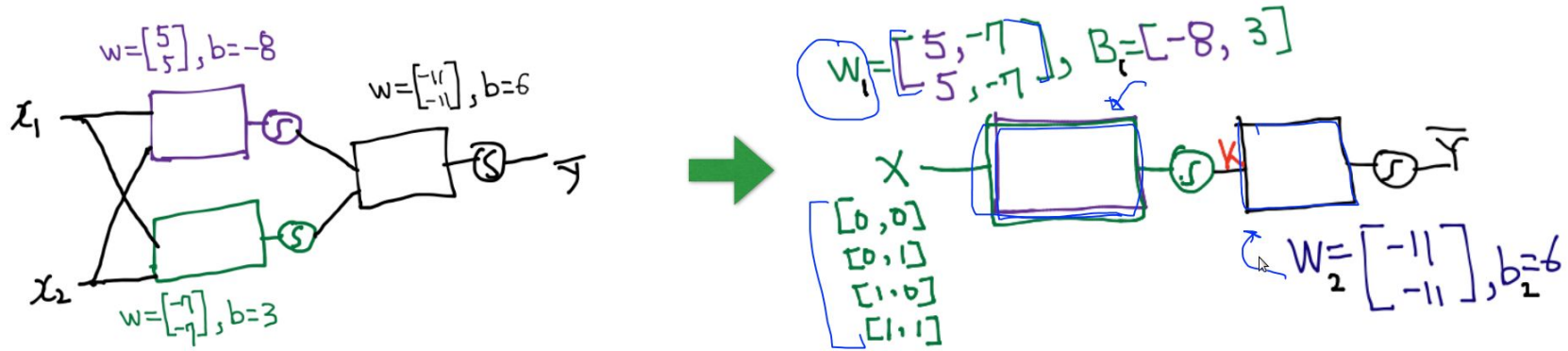
```
    layer1 = tf.sigmoid(tf.matmul(features, W1) + b1) # W1=[2,1], b1=[1]
```

```
    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
```

Neural Net Vector



[Tensorflow Code]

```
def neural_net(features):
    layer = tf.sigmoid(tf.matmul(features, W1) + b1) # W1=[2,2], b1=[2]
    hypothesis = tf.sigmoid(tf.matmul(layer, W2) + b2) # W2=[2,1], b2=[1]
    return hypothesis
```

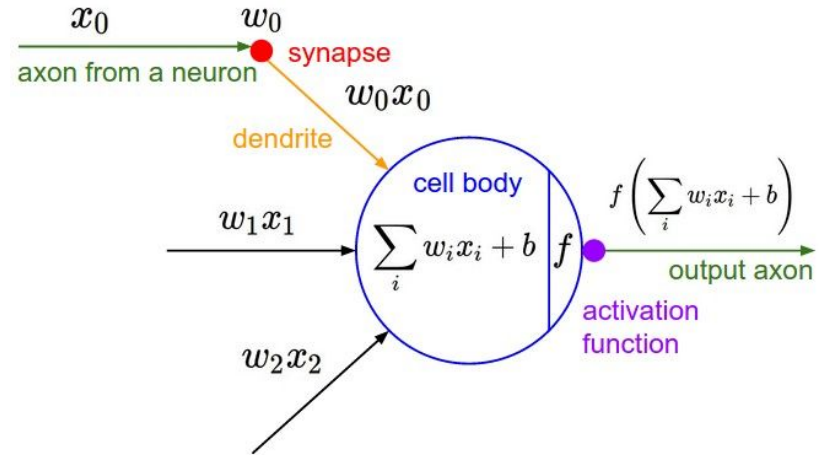
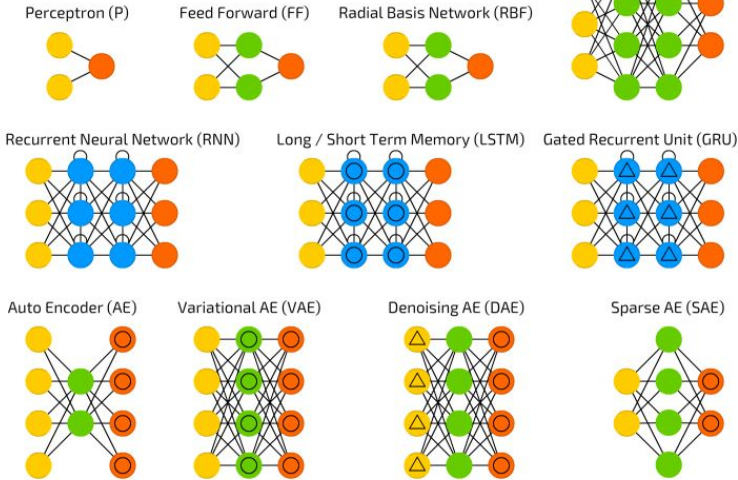
Neural Net Chart

A mostly complete chart of

Neural Networks

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-  Backfed Input Cell
-  Input Cell
-  Noisy Input Cell
-  Hidden Cell
-  Probabilistic Hidden Cell
-  Spiking Hidden Cell
-  Output Cell
-  Match Input Output Cell
-  Recurrent Cell
-  Memory Cell
-  Different Memory Cell
-  Kernel
-  Convolution or Pool



Code(Eager)

```
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import tensorflow as tf
import tensorflow.contrib.eager as tfe

tf.enable_eager_execution()
tf.set_random_seed(777) # for reproducibility

print(tf.__version__)

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

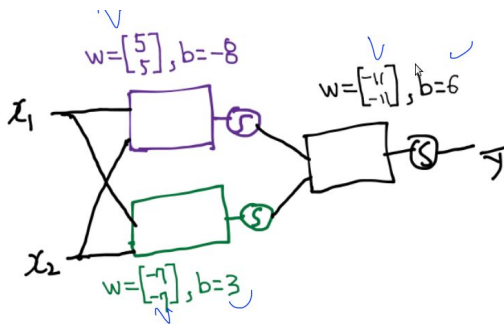
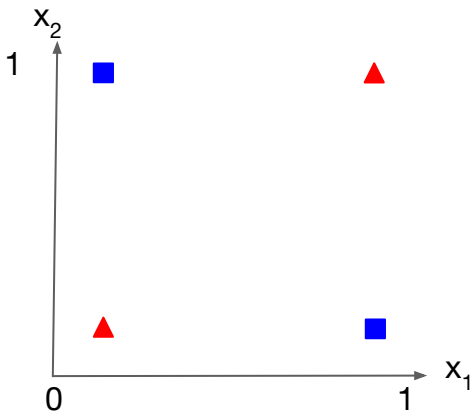
```
dataset = tf.data.Dataset.from_tensor_slices((x_data, y_data)).batch(len(x_data))
```

```
def preprocess_data(features, labels):
    features = tf.cast(features, tf.float32)
    labels = tf.cast(labels, tf.float32)
    return features, labels
```

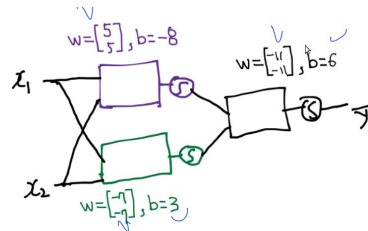
```
W1 = tf.Variable(tf.random_normal([2, 1]), name='weight1')
b1 = tf.Variable(tf.random_normal([1]), name='bias1')
```

```
W2 = tf.Variable(tf.random_normal([2, 1]), name='weight2')
b2 = tf.Variable(tf.random_normal([1]), name='bias2')
```

```
W3 = tf.Variable(tf.random_normal([2, 1]), name='weight3')
b3 = tf.Variable(tf.random_normal([1]), name='bias3')
```



```
def neural_net(features):
    layer1 = tf.sigmoid(tf.matmul(features, W1) + b1)
    layer2 = tf.sigmoid(tf.matmul(features, W2) + b2)
    layer3 = tf.concat([layer1, layer2], -1)
    layer3 = tf.reshape(layer3, shape = [-1, 2])
    hypothesis = tf.sigmoid(tf.matmul(layer3, W3) + b3)
    return hypothesis
```



Code(Eager)

```
def loss_fn(hypothesis, features, labels):
    cost = -tf.reduce_mean(labels * tf.log(hypothesis) + (1 - labels) * tf.log(1 - hypothesis))
    return cost
```

```
optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.01)
```

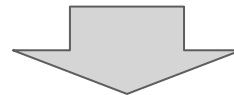
```
def accuracy_fn(hypothesis, labels):
    predicted = tf.cast(hypothesis > 0.5, dtype=tf.float32)
    accuracy = tf.reduce_mean(tf.cast(tf.equal(predicted, labels), dtype=tf.float32))
    return accuracy
```

```
def grad(features, labels):
    with tf.GradientTape() as tape:
        loss_value = loss_fn(neural_net(features), features, labels)
    return tape.gradient(loss_value, [W1, W2, W3, b1, b2, b3])
```

EPOCHS = 50000

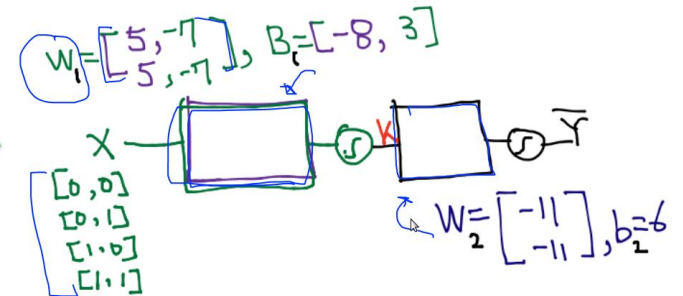
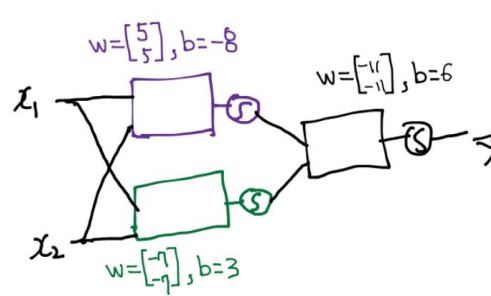
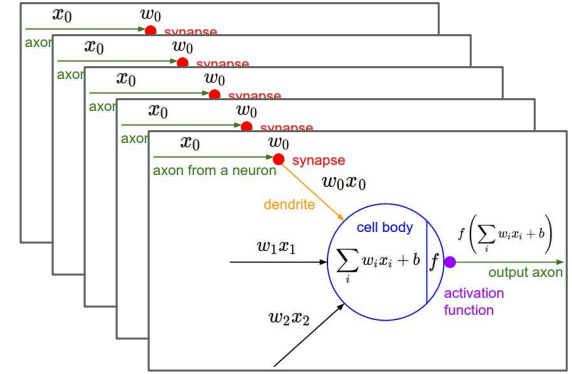
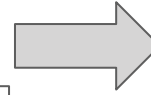
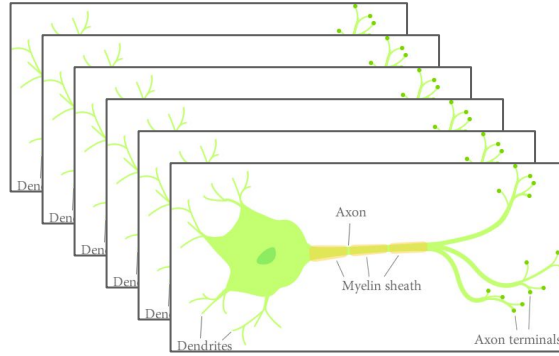
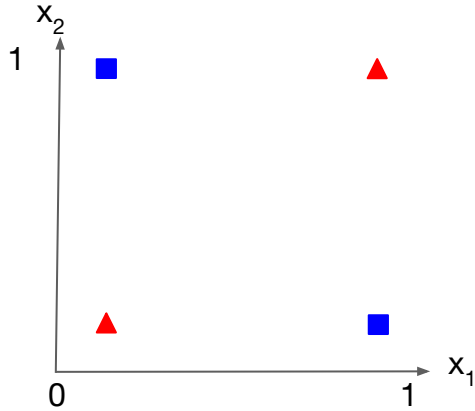
```
for step in range(EPOCHS):
    for features, labels in tfe.Iterator(dataset):
        features, labels = preprocess_data(features, labels)
        grads = grad(neural_net(features), labels)
        optimizer.apply_gradients(grads_and_vars=zip(grads, [W1, W2, W3, b1, b2, b3]))
        if step % 5000 == 0:
            print("Iter: {}, Loss: {:.4f}".format(step, loss_fn(neural_net(features), labels)))
    x_data, y_data = preprocess_data(x_data, y_data)
    test_acc = accuracy_fn(neural_net(x_data), y_data)
    print("Testset Accuracy: {:.4f}".format(test_acc))
```

```
Iter: 0, Loss: 0.6931
Iter: 100, Loss: 0.6931
Iter: 200, Loss: 0.6931
Iter: 300, Loss: 0.6931
Iter: 400, Loss: 0.6931
Iter: 500, Loss: 0.6931
Iter: 600, Loss: 0.6931
Iter: 700, Loss: 0.6931
Iter: 800, Loss: 0.6931
Iter: 900, Loss: 0.6931
Iter: 1000, Loss: 0.6931
Testset Accuracy: 0.5000
```



```
Iter: 0, Loss: 0.8487
Iter: 5000, Loss: 0.6847
Iter: 10000, Loss: 0.6610
Iter: 15000, Loss: 0.6154
Iter: 20000, Loss: 0.5722
Iter: 25000, Loss: 0.5433
Iter: 30000, Loss: 0.5211
Testset Accuracy: 1.0000
```

Summary



What's Next?

Sigmoid \rightarrow Relu

