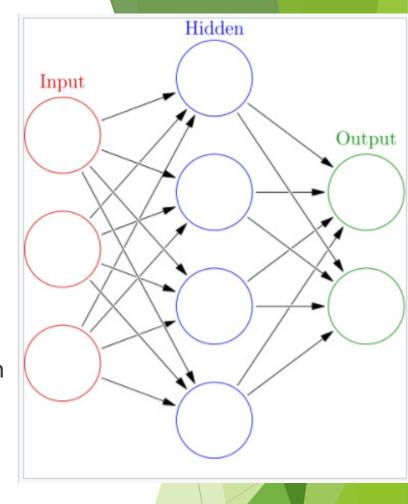
Back propagation

Outline

- Gradient descent
- Forward propagate and Back propagation
- Easy Example by tensorflow
- Homework

Gradient descent

- Neural network have many neural node
- Each neural node have its own weight and bias to compute its output
- ► Neural network output and Label(real answer) compute Loss function
- ► Know how far between output and Label(real answer) by Loss function



- Cost function is computed by output and real answer
- Output is computed by weight and bias
- Cost function is structured by weight and bias

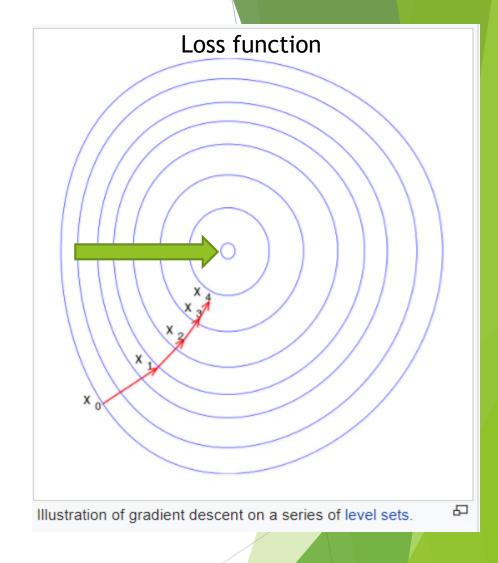
direction of the negative gradient

$$\mathbf{a}_{n+1} = \mathbf{a}_n - \gamma \nabla F(\mathbf{a}_n)$$

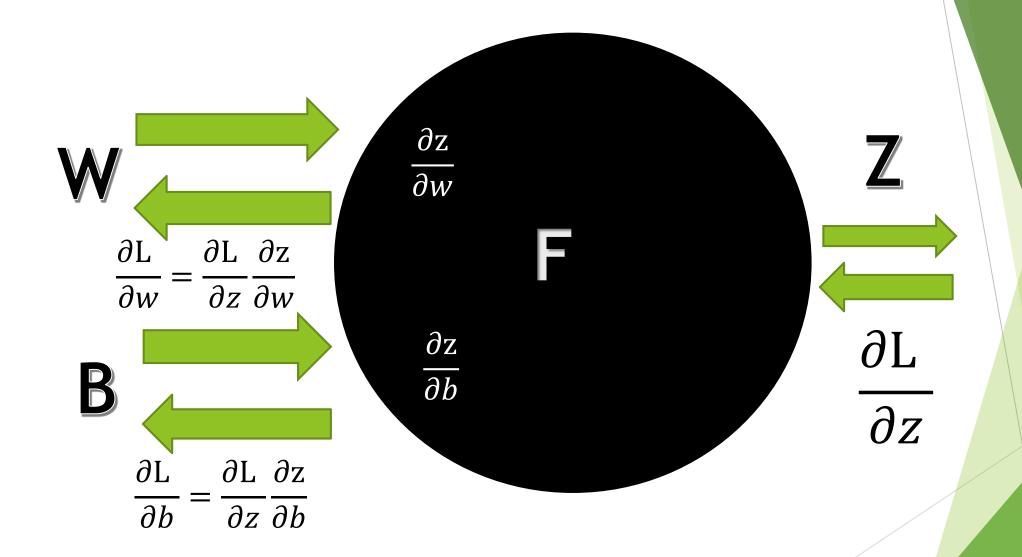
$$\mathbf{x}_{n+1} = \mathbf{x}_n - \gamma_n \nabla F(\mathbf{x}_n), \ n \geq 0.$$

We have

$$F(\mathbf{x}_0) \geq F(\mathbf{x}_1) \geq F(\mathbf{x}_2) \geq \cdots,$$



https://en.wikipedia.org/wiki/Gradient_descent

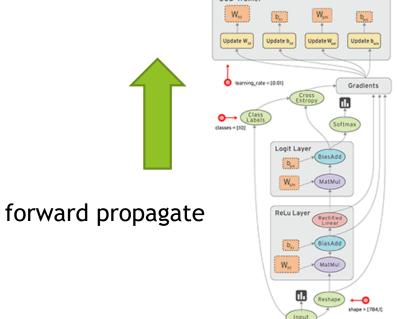


Forward propagate and back propagation

Forward propagate: know neural network behaving and get the neural networks output

Back propagation : derivative of error(Loss function) ,update weights to

minima for error value



Back propagation



Easy example by tensorflow

- 3 x [1,3] input, 3 x [1] real answer
- hidden layer_1 have 3 input and 2 output
- hidden layer_2 have 2 input and 1 output
- Activation function : sigmoid
- loss function : Mean squared error
- Optimizer : Gradient Descent

Prepare data

```
#training data
\existstrain_1 = np.array( [[1.,2.,3.],
       [3.,4.,5.],
       [8.,5.,7.],
       [7.,1.,8.]])
\existstrain_2 = np.array( [[1.],
             [0.],
             [0.],
             [1.]])
```

Prepare input

```
input_1 = tf.placeholder(tf.float32, shape = [None, 3])
input_2 = tf.placeholder(tf.float32, shape = [None, 1])
```

Prepare hidden layer

```
weight_1 = tf.get_variable(name='weight_1', shape = [3,2], dtype = tf.float32, initializer =
    tf.truncated_normal_initializer(mean=0.0, stddev=0.1) )
bias_1 = tf.get_variable(name='bias_1', shape = [2], dtype = tf.float32, initializer =
    tf.truncated_normal_initializer(mean=0.0, stddev=0.1) )
layer_1_output = tf.add(tf.matmul( input_1, weight_1 ), bias_1)

weight_2 = tf.get_variable(name='weight_2', shape = [2,1], dtype = tf.float32, initializer =
    tf.truncated_normal_initializer(mean=0.0, stddev=0.1) )
bias_2 = tf.get_variable(name='bias_2', shape = [1], dtype = tf.float32, initializer =
    tf.truncated_normal_initializer(mean=0.0, stddev=0.1) )
layer_2_output = tf.sigmoid( tf.add(tf.matmul( layer_1_output, weight_2 ), bias_2) )
```

Prepare loss function

```
loss = tf.losses.mean_squared_error(train_2, layer_2_output)
```

Prepare optimizer

```
optimizer = tf.train.GradientDescentOptimizer(0.1)
```

Prepare How to train

```
train = optimizer.minimize(loss)
```

Prepare session to start train

```
with tf.Session() as sess:
    #initial
    init = tf.global_variables_initializer()
    sess.run( init )
    #train
    for step in range(201) :
        if step % 20 == 0:
            print ('loss : ', sess.run(loss, feed_dict = {input_1: train_1, input_2: train_2}))
            print ('predict : ', sess.run(layer_2_output, feed_dict = {input_1: train_1}))
        sess.run(train, feed_dict = {input_1: train_1, input_2: train_2})
```

```
#training data
∃train_1 = np.array( [[1.,2.,3.],
       [3.,4.,5.],
       [8.,5.,7.],
       [7.,1.,8.]])
∃train_2 = np.array( [[1.],
            [0.],
            [0.],
            [1.]])
```

```
loss: 0.251956
predict : [[ 0.46995696]
  0.471921091
predict : [[ 0.47125581]
  0.465668591
predict : [[ 0.46862936]
loss : 0.0668569
predict : [[ 0.57679796]
 [ 0.270938371
loss: 0.0565343
predict : [[ 0.61291397]
  0.260891141
predict : [[ 0.64426953]
  0.24730685]
```

Homework

Data.txt

Answer.txt

1 0 0 1 0←

- Any number hidden layer you want
- Activation function : Anything you wanted
- loss function: Anything you wanted
- Optimizer : Anything you wanted
- Need demo

grade

```
0.005:1
```

- **0.001:2**
- **0.0003:3**
- **0.0006:4**
- Use add layer function : +1

```
loss: 0.000501916
predict: [[ 0.97440165]
[ 0.01172936]
[ 0.01904387]
[ 0.97416872]
[ 0.02620688]]
```