#### Report of Homework 0

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#### 1. Introduction:

首先定義notation, $a^0$ 為input layer, $W^l$ 為每層的weight, $b^l$ 為每層的bias, $z^l$ 為每層的output,而 $a^l$ 則是每層 $z^l$ 經過activation function後的output,即為下一層的input, $C^r$ 為error,此notation參考自台大李宏毅的講義[1],並且實作也基於其上完成。

feedforward pass:

$$z^{1} = W^{1}a^{0} + b^{1}$$

$$z^{l} = W^{l}a^{l-1} + b^{l}, \ a^{l} = \sigma(z^{l})$$

$$a^{l} = \sigma(W^{l}a^{l-1} + b^{l})$$

backward pass:

$$\frac{\partial C^r}{\partial w_{ij}^l} = \frac{\partial z_i^l}{\partial w_{ij}^l} \frac{\partial C^r}{\partial z_i^l} = a_j^{l-1} \delta_i^l$$

$$\frac{\partial C^r}{\partial b_i^l} = \frac{\partial z_i^l}{\partial b_i^l} \frac{\partial C^r}{\partial z_i^l} = \delta_i^l$$

1. output layer

$$\delta^L = \sigma(z^L) \cdot \nabla C^r(y^r)$$

2. other layers

$$\delta^l = \sigma'(z^l) \cdot (W^{l+1})^T \delta^{l+1}$$

### 2. Experiment setups:

# A. sigmoid functions:

```
def sigmoid(self, x, derivative=False):
    # you already input the sigmoid function, so use x rather than sigmoid(x)
    if derivative:
        return x * (1-x)
    return np.exp(-np.logaddexp(0, -x))
    return 1 / (np.exp(-x))
```

## B. Neural network:(Bonus done)

實作n-layer NN,詳情見code,error function採用SE

## C. Backpropagation:

如上面的notation定義中的backward pass,實作見code

### 3. Experiment setups:

# A. Results of your testing

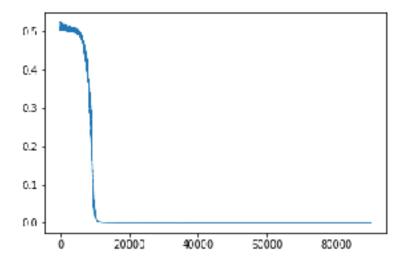
### (1) Screenshot

```
In [59]: MM.fit(X, Y, batch_size=1, epochs=90000, shuffle=True, learning_rate=0.05, decreasing_rate=1e-4
         epoch 77400:
         epoch 78300;
         epoch 79200:
         epoch 80100:
         epoch 81000;
         epoch 81900:
         epoch 82800:
         epoch 83700;
         epoch 84600:
         epoch 85500:
         epoch 86400;
         epoch 87300:
         epoch 88200;
         epoch 89100:
         epoch 90000:
         [0 0] 0 [[0.000272]] [[3.69928677e-08]]
         [0 1] 1 [[0.99938087]] [[1.91663513e-07]]
         [1 0] 1 [[0.99851592]] [[1.10124435e-06]]
         [1 1] 0 [[0.0014624]] [[1.06930902e-06]]
```

#### 4. Discussion:

- (1)實作上在sigmoid的部分由於有exponential項的關係,因此有可能造成數值問題[2],所以選擇了此種實作。
  - (2)在此例中應用decreasing rate和shuffle似乎並沒有明顯的進一步改善。
  - (3)收斂過程如圖,x軸為epoch數,y軸為error

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# 5. Reference:

[1] http://speech.ee.ntu.edu.tw/~tlkagk/courses/MLDS\_2015\_2/Lecture/
DNN%20backprop.pdf

[2] https://docs.scipy.org/doc/numpy-1.14.0/reference/generated/numpy.logad-dexp.html