

# CS 754 : Advanced Image Processing Assignment 4

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## Q1

### A1.1 Derivatives for Calculation of Gradient Descent

The given cost function is

$$E(W, h_i) = \sum_{i=1}^N \sum_{j=1}^n (-y_{ji} \log(Wh_i)_j + (Wh_i)_j) + \lambda \sum_{k=1}^K \sum_{j=1}^N h_{kj} \quad (1)$$

- Gradient of W:

We note to get the gradient of a matrix, we need to differentiate with respect to every element of the matrix. The gradient matrix will simply be derivative to of the cost function with respect to the corresponding element of  $W$ .

Let us consider the derivative with respect to the  $a^{th}$  row and  $b^{th}$  column denoted by  $w_{ab}$ .

$$[\nabla W]_{ab} = \frac{dE}{dw_{ab}}$$

$$\frac{dE}{dw_{ab}} = \frac{d}{dw_{ab}} \left( \sum_{i=1}^N \sum_{j=1}^n (-y_{ji} \log(Wh_i)_j + (Wh_i)_j) + \lambda \sum_{k=1}^K \sum_{j=1}^N h_{kj} \right)$$

We note that

$$(Wh_i)_j = \sum_{l=1}^K W_{jl} h_{li}$$

Clearly the second term of the cost function is not dependent on  $W$  and hence goes to zero. For the first term we need to consider only the  $a^{th}$  row

$$\frac{dE}{dw_{ab}} = \frac{d}{dw_{ab}} \left( \sum_{i=1}^N (-y_{ai} \log(Wh_i)_a + (Wh_i)_a) \right)$$

Consider the first term

$$\frac{d}{dw_{ab}} \sum_{i=1}^N (-y_{ai} \log(Wh_i)_a) = \sum_{i=1}^N \frac{d}{dw_{ab}} (-y_{ai} \log(\sum_{l=1}^K W_{al} h_{li})) = \sum_{i=1}^N -\frac{y_{ai}}{(Wh_i)_a} (h_{bi})$$

Now consider the second term

$$\frac{d}{dw_{ab}} \sum_{i=1}^N (Wh_i)_a = \sum_{i=1}^N h_{bi}$$

Therefore

$$\frac{dE}{dw_{ab}} = \sum_{i=1}^N -\frac{y_{ai}}{(Wh_i)_a} (h_{bi}) + h_{bi} = \sum_{i=1}^N \left( 1 - \frac{y_{ai}}{(Wh_i)_a} \right) h_{bi}$$

We vectorize the equation for matlab as follows

$$[\nabla(W)] = (1 - Y ./ WH) H'$$

- Gradient of H:

We follow similar strategy to get the gradient of  $H$ . We differentiate the cost with respect