

CS 754 : Advanced Image Processing Assignment 4

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April 4, 2017

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 last 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 equaiton 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 to each of the elements of the matrix H . We denote the element at row a and column b by h_{ab}

$$[\nabla H]_{ab} = \frac{dE}{dh_{ab}}$$

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

We consider the three terms separately. For the first term we will need to consider only the b^{th} column of H .

$$\frac{d}{dh_{ab}} \sum_{j=1}^n -y_{jb} \log \left(\sum_{l=1}^K W_{jl} h_{lb} \right) = \sum_{j=1}^n -\frac{y_{jb}}{W h_{b,j}} W_{ja}$$

For second term also we consider only b^{th} column

$$\sum_{j=1}^n \frac{d}{dh_{ab}} \sum_{l=1}^K W_{jl} h_{lb} = \sum_{j=1}^n W_{ja}$$

The third term will simply be λ Therefore

$$\frac{dE}{dh_{ab}} = \left(\sum_{j=1}^n -\frac{y_{jb}}{W h_{b,j}} W_{ja} + W_{ja} \right) + \lambda$$

We vectorize the equation for matlab as follows

$$[\nabla H] = W' * (1 - Y ./ WH) + \text{lambda}$$

Results

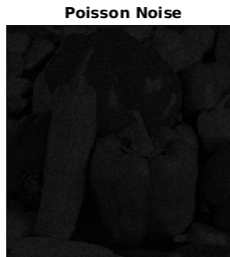
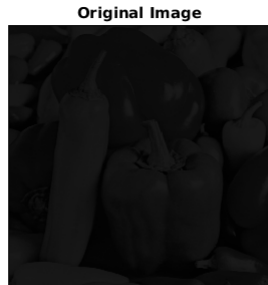
Parameters:

$\lambda = 20$

Initial step size = 0.1

Iterations = 200

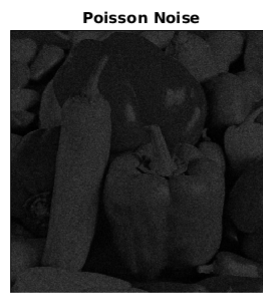
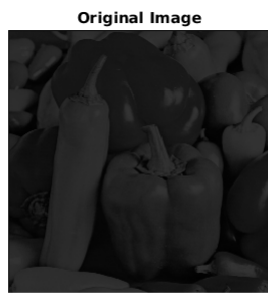
Peak = 30



$\text{PSNR}_{\text{denoised}} = 18.097 \text{ dB}$

$\text{PSNR}_{\text{noisy}} = 13.920 \text{ dB}$

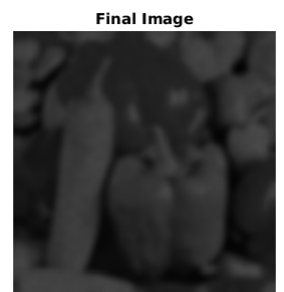
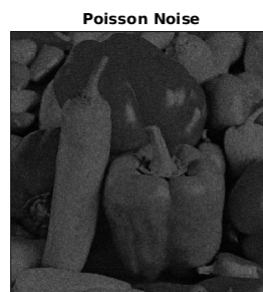
Peak = 60



$\text{PSNR}_{\text{denoised}} = 20.162 \text{ dB}$

$\text{PSNR}_{\text{noisy}} = 16.825 \text{ dB}$

Peak = 100



$\text{PSNR}_{\text{denoised}} = 20.295 \text{ dB}$

$\text{PSNR}_{\text{noisy}} = 19.093 \text{ dB}$