

CSE377 Spring2023 Homework 7: Image Restoration (10 pts)

Due April 18 2023, 11:59PM, via Brightspace

We derived a linear computational model to approximate the Phase Contrast or DIC microscopy imaging process:

$$\underset{N \times 1}{\mathbf{g}} = \underset{N \times N}{\mathbf{H}} \underset{N \times 1}{\mathbf{f}}$$

where \mathbf{g} is a microscopy image, \mathbf{H} is a matrix related to the image formation process and \mathbf{f} is the image to be restored.

We formulate the following sparsity-constrained quadratic optimization to restore \mathbf{f} :

$$\mathcal{O}(\mathbf{f}) = \|\mathbf{H}\mathbf{f} - \mathbf{g}\|_2^2 + \omega_s \mathbf{f}^T \mathbf{L} \mathbf{f} + \omega_r \|\mathbf{A}\mathbf{f}\|_1$$

$$\mathbf{A} = \begin{bmatrix} \lambda_{11} & \lambda_{12} & 0 \\ 0 & \dots & \lambda_{nn} \end{bmatrix}$$

which can be solved by the following algorithm:

Algorithm I: restoring artifact-free microscopy images

Initialize $\mathbf{f} = \mathbf{f}^{init}$ and $\mathbf{A} = \mathbf{A}^{init}$.

Repeat the following steps for all pixel j

$$\mathbf{A}^{init} = \begin{bmatrix} 1 & 0 & \dots \\ 0 & 1 & \dots \\ \vdots & \vdots & \ddots \end{bmatrix} \text{ or } \mathbf{A}^{init} = \begin{bmatrix} \lambda_{11} & \lambda_{12} & 0 \\ 0 & \dots & \lambda_{nn} \end{bmatrix}$$

$$\underset{N \times 1}{\mathbf{b}} = \underset{N \times 1}{-\mathbf{H}^T \mathbf{g}} + \omega_r \text{diag}(\mathbf{A})/2 \quad (1)$$

$$\mathbf{f}_j \leftarrow \mathbf{f}_j \frac{-\mathbf{b}_j + \sqrt{\mathbf{b}_j^2 + 4(\mathbf{Q}^+ \mathbf{f})_j (\mathbf{Q}^- \mathbf{f})_j}}{2(\mathbf{Q}^+ \mathbf{f})_j} \quad (2)$$

$$\mathbf{A}_{jj} \leftarrow \frac{\mathbf{A}_{jj}^{init}}{\mathbf{f}_j + \gamma} \quad (3)$$

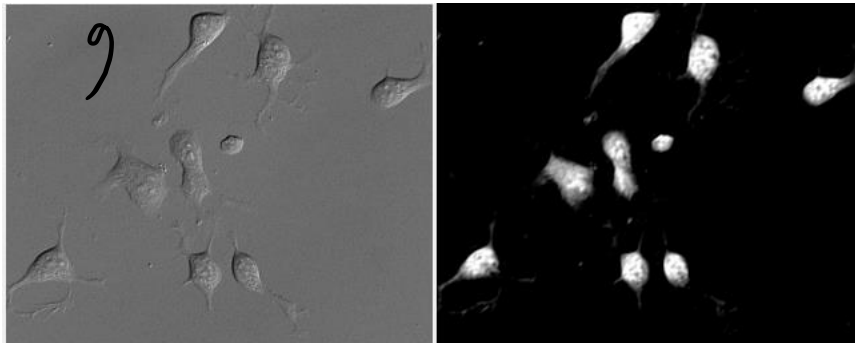
Until convergence.

$$\|\mathbf{f}^{(t)} - \mathbf{f}^{(t-1)}\| < \epsilon \text{ \& iter > maxiter}$$

$$\text{where } \mathbf{Q}_{uv}^+ = \begin{cases} \mathbf{Q}_{uv} & \text{if } \mathbf{Q}_{uv} > 0 \\ 0 & \text{otherwise} \end{cases} \text{ and } \mathbf{Q}_{uv}^- = \begin{cases} |\mathbf{Q}_{uv}| & \text{if } \mathbf{Q}_{uv} < 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\mathbf{Q} = \mathbf{H}^T \mathbf{H} + \omega_s \mathbf{L}$$

Starting codes are provided, including the computation of the \mathbf{H} matrix, the procedure to flatten the image, and the procedure to compute the Laplacian matrix \mathbf{L} of a regular image grid. A testing image is given. Implement Algorithm I and you are expected to achieve the following result:



$$\omega_s = 0, \omega_r = 0$$

Try different hyper parameters and see how the constraints affect the restoration processes and results.

Submit your Jupyter notebook, or .py code with your resultant image in a brief report.