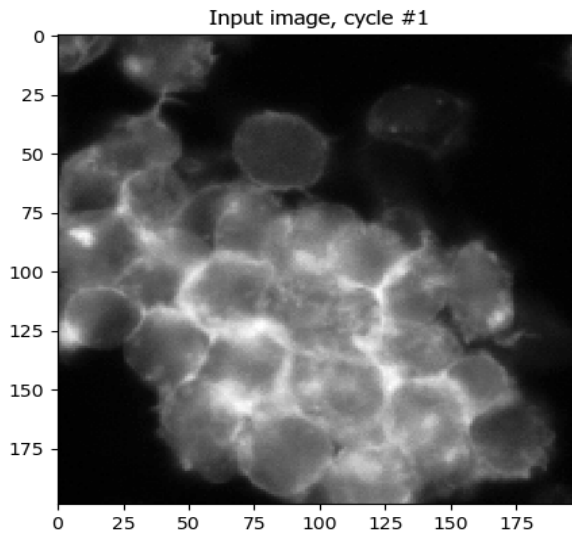


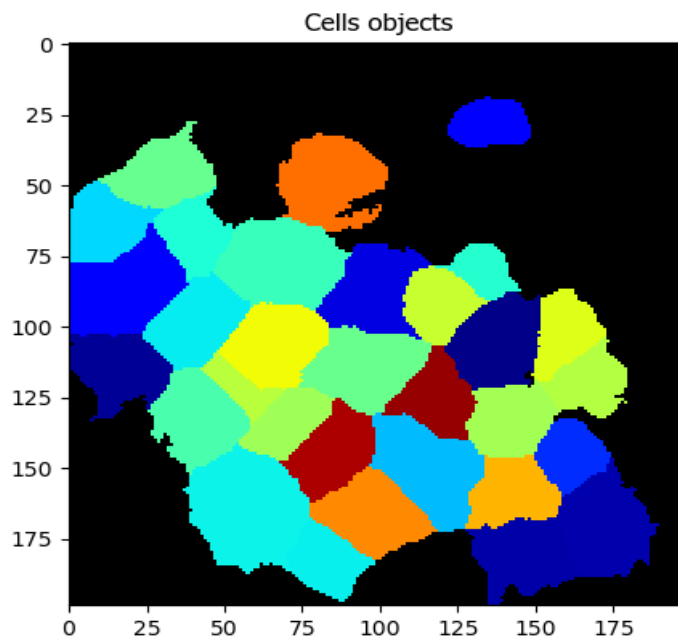
Software Name- Cellprofiler

Website reference- <https://cellprofiler.org/>

Input image -



Output image-



Explanation - It uses the watershed algorithm for cells separate from each other

- o Input: image

- o convert I into inverted grey: $\text{inv}I = L - \text{grey}(I)$

- o compute the negative distance transform of $\text{inv}I$

- o non-max suppression over shallow minima

- o fill basins & get watersheds

- o update each segmented mask with watersheds

o Initialize:

A *grey* I having: [Hmin, Hmax]

Minima points: M1, ..., MR

Thresholded set: $T_h = \{p \in I \mid I(p) \leq h\}$, where p is a pixel in I and h is some intensity level.

Let's define the Influence set

$C(M)$ = cluster associated with M

$I_{Zh+1}(M_i) = \{p \in T_{h+1} \mid d(p, C(M_i)) < d(p, C(M_j))\}$

$\forall j, j \neq i$

Run:

$h = h_{min}$

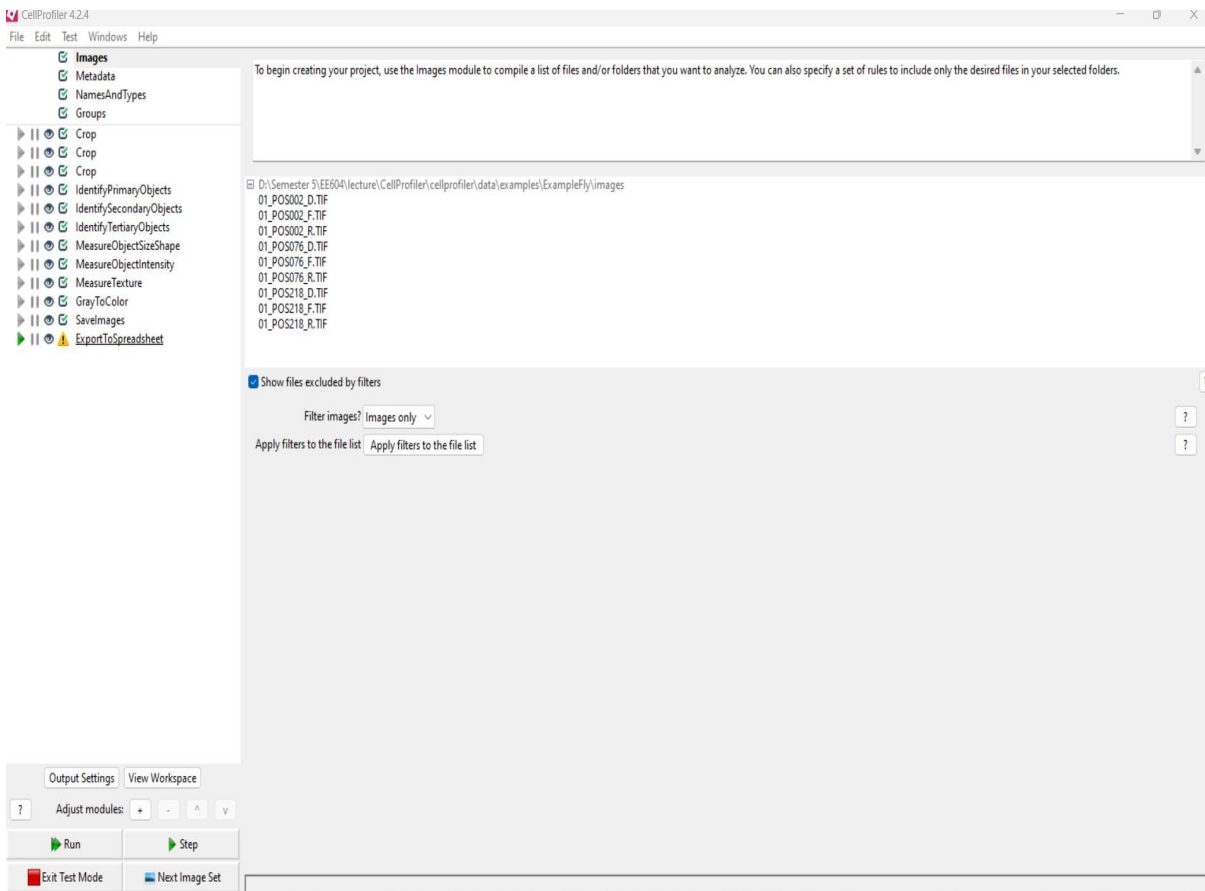
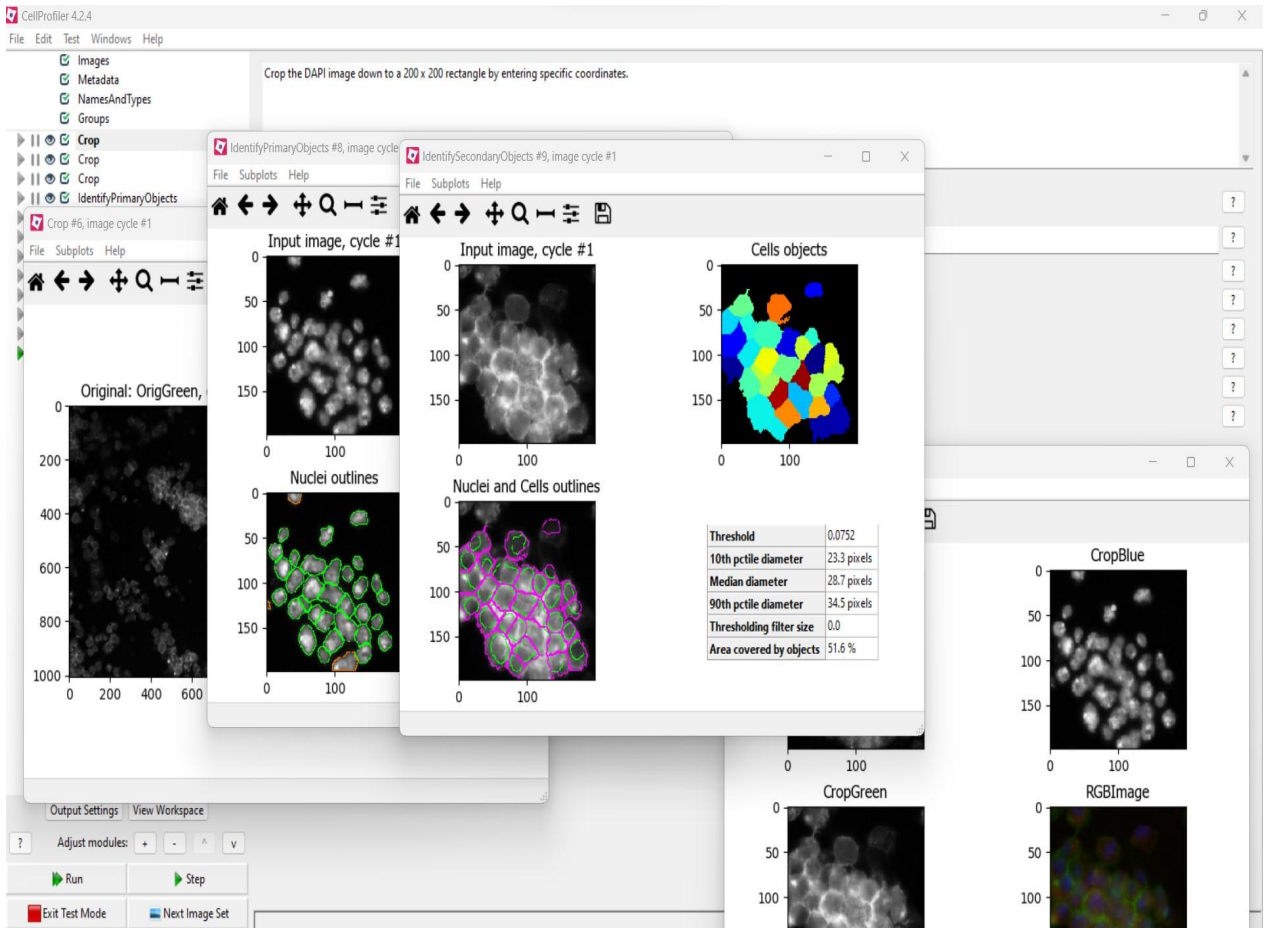
Immersed set: $X_h = X_{hmin} = T_{hmin} = \{p \in I \mid I(p) \leq h_{min}\}$

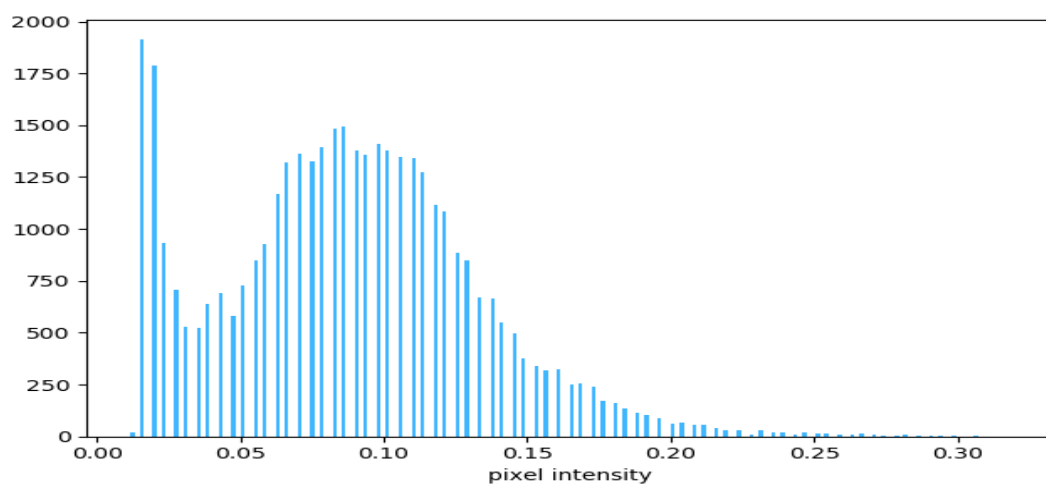
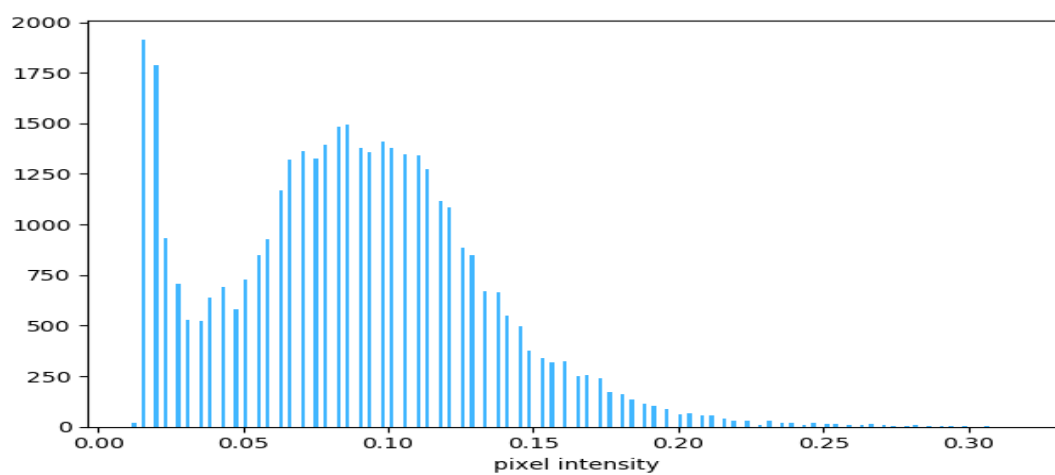
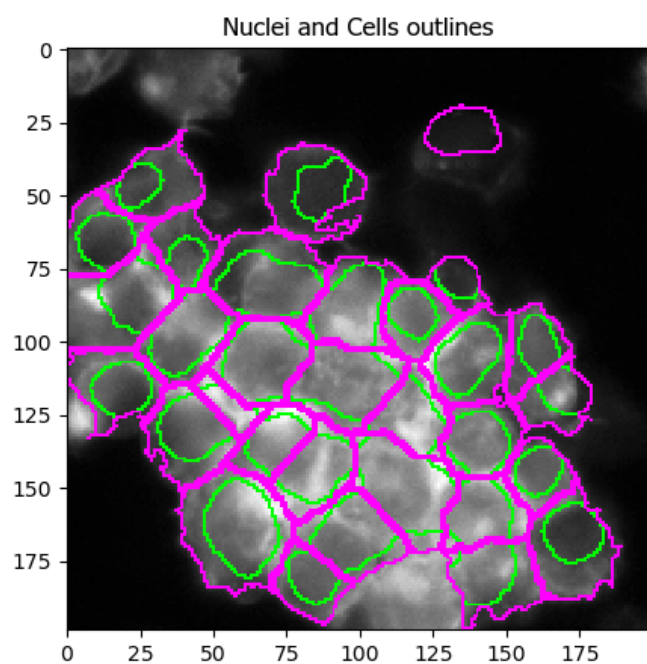
Loop until Hmax

$X_{h+1} = \bigcup_{i=1}^R I_{Zh+1}(M_i)$

Influence set of minima M, at level h+1

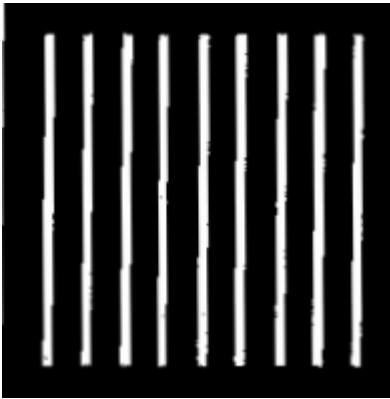
Watershed (1) = Set of all pixels in $I \setminus X_{hmax}$





2-

(i) The white bars in the test pattern shown are 7 pixels wide and 210 pixels high. The separation between bars is 17 pixels. What would this image look like after the application of



1-

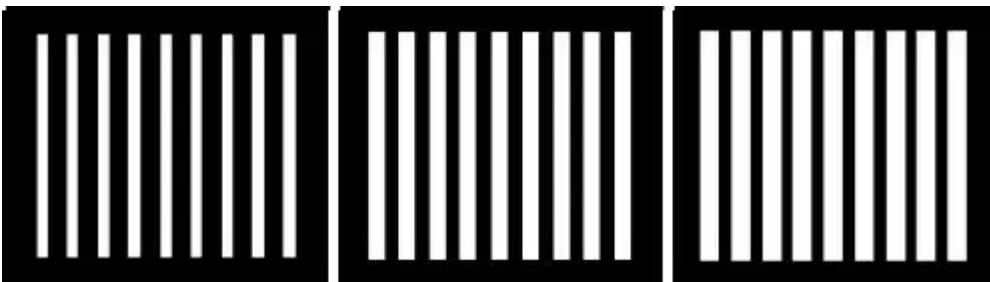
- (a) 3×3 Harmonic mean filter
- (b) 5×5 Harmonic mean filter
- (c) 9×9 Harmonic mean filter

2-Repeat 1 for Cantraharmonic mean filter with $Q = -1.5$

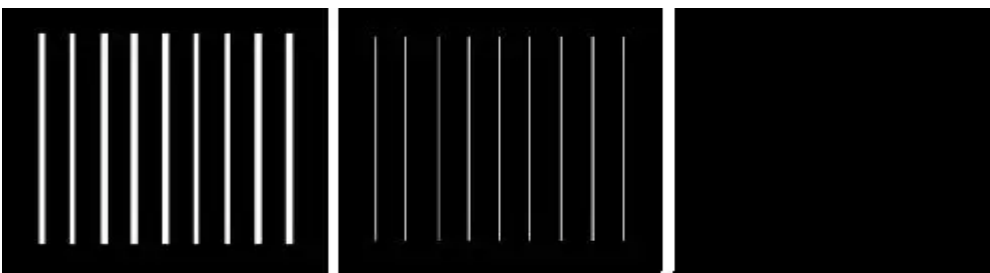
3-Repeat 2 for Max filter

Sol-1-

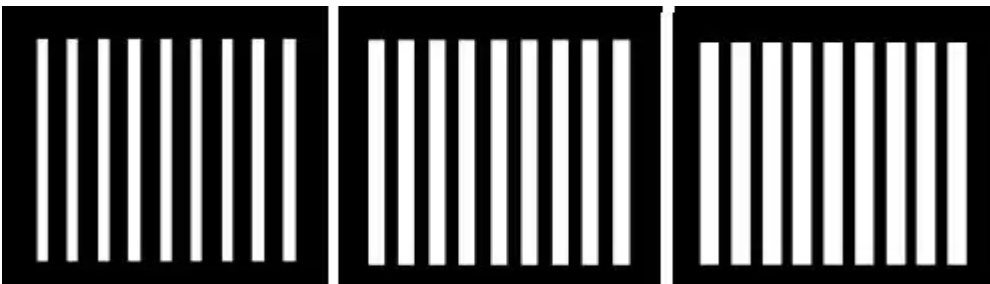
1-



2-



3-



(ii) Write an expression for 2-D discrete convolution.

ans-

convolution in one dimension:

$$f(x) * g(x) = \int_{-\infty}^{\infty} f(\alpha)g(x - \alpha)d\alpha.$$

The Fourier transform of this expression is

$$\begin{aligned}\mathfrak{F}[f(x) * g(x)] &= \int_{-\infty}^{\infty} \left[\int_{-\infty}^{\infty} f(\alpha)g(x - \alpha)d\alpha \right] e^{-j2\pi ux} dx \\ &= \int_{-\infty}^{\infty} f(\alpha) \left[\int_{-\infty}^{\infty} g(x - \alpha)e^{-j2\pi ux} dx \right] d\alpha.\end{aligned}$$

The term inside the inner brackets is the Fourier transform of $g(x - \alpha)$. But,

$$\mathfrak{F}[g(x - \alpha)] = G(u)e^{-j2\pi u\alpha}$$

so

$$\begin{aligned}\mathfrak{F}[f(x) * g(x)] &= \int_{-\infty}^{\infty} f(\alpha) [G(u)e^{-j2\pi u\alpha}] d\alpha \\ &= G(u) \int_{-\infty}^{\infty} f(\alpha)e^{-j2\pi u\alpha} d\alpha \\ &= G(u)F(u).\end{aligned}$$

This proves that multiplication in the frequency domain is equal to convolution in the spatial domain. The proof that multiplication in the spatial domain is equal to convolution in the spatial domain is done in similar way.

A- which are the following statements correct about filters-

- (1) The median filter uses to remove Salt & pepper noise
- (2) Burst filter uses to remove Impulse noise
- (3) Gaussian filter is a Nonlinear filter
- (4) In a Bilinear filter blur in the image increases as the sigma of the range kernel increases

Ans-1,4

B -which are the following statements correct about clustering -

- (1) intra-class variance is low(high similarity) in clustering
- (2) inter-class variance is low (high similarity) in clustering
- (3) K-means clustering unsupervised learning method requires labels but not data
- (4) K-means clustering guaranteed to converge in finite iterations
- (5) K-means clustering is useful for pattern recognition when we don't know what to look for

Ans- 1,4,5