

Assignment 1.3: Image histograms

A1:

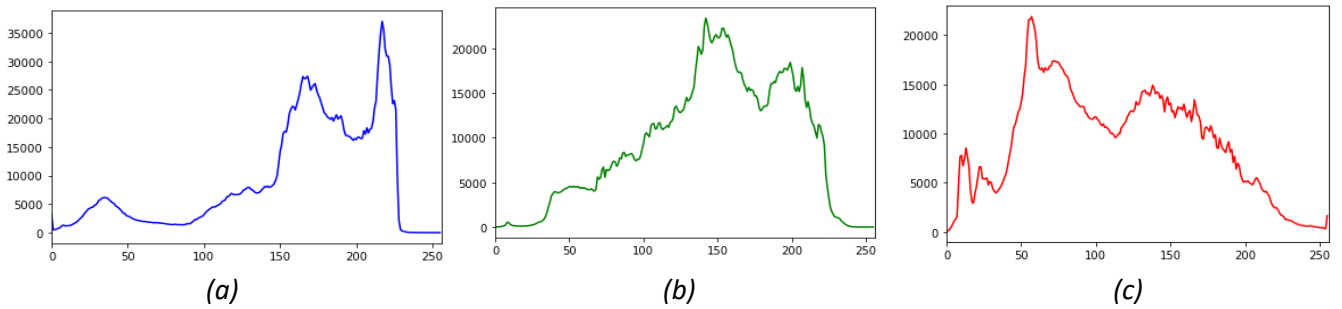
a: Histograms for the blue, green and red channel of img1.jpg from Task 1

Fig 1: a) Histogram for Blue channel, b) Histogram for Green Channel, c) Histogram for Red channel

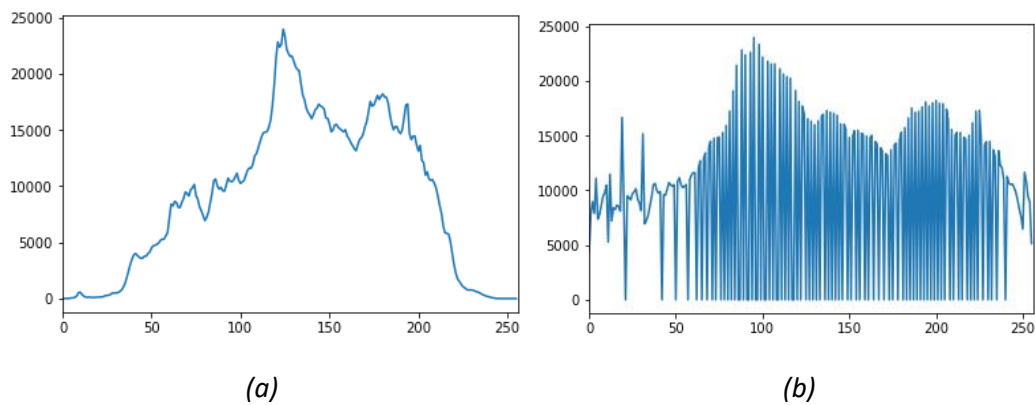
b: Intensity histograms of img_gray and img_eq from Task 2

Fig 2: (a) Intensity Histogram for grayscale image, (b) intensity histogram for equalised image

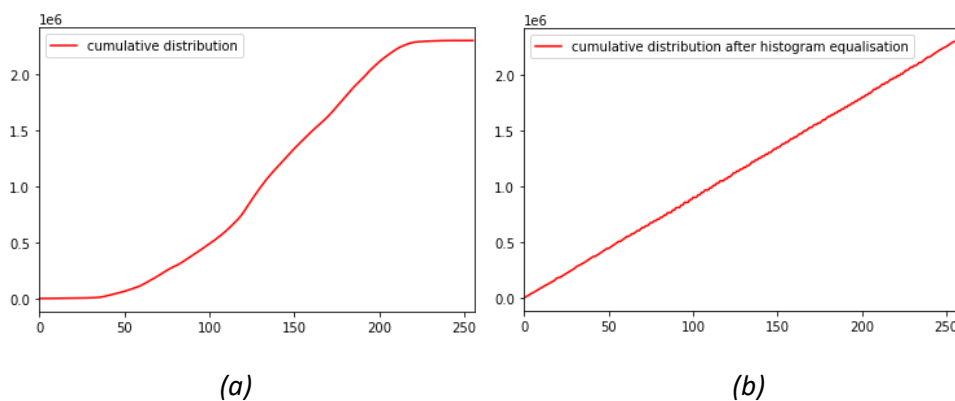
c. Cumulative distributions of intensity of img_gray and img_eq from Task 2.

Fig 3: (a) Cumulative distribution of img_gray, (b) cumulative distribution of img_eq

A2:



(a)

(b)

Fig 4: (a) Grayscale image, (b) equalised grayscale image

A3:

The distance values using χ^2 distance:

- $\text{Img1} - \text{img2}$: 1567300 (with normalised hist: 0.68)
- $\text{Img1} - \text{img3}$: 1434813 (with normalised hist: 0.62)
- $\text{Img2} - \text{img3}$: 837586 (with normalised hist: 0.36)

This shows that using χ^2 distance **Img2 – Img3** pair is most similar, as it has least distance (0.36).

Note: The normalised hist values denote when χ^2 distance is calculated using normalised histogram.

The KL Divergence values:

- $\text{Img1} - \text{img2}$: 1.73
- $\text{Img1} - \text{img3}$: 1.66
- $\text{Img2} - \text{img3}$: 1.91

This shows that using KL Divergence, **img1 – img3** pair is most similar as it has least distance (1.66).

Note: Here, KL Divergence is calculated as $KL(h1, h2) + KL(h2, h1)$ for $h1$ and $h2$. The workings are shown in the code.