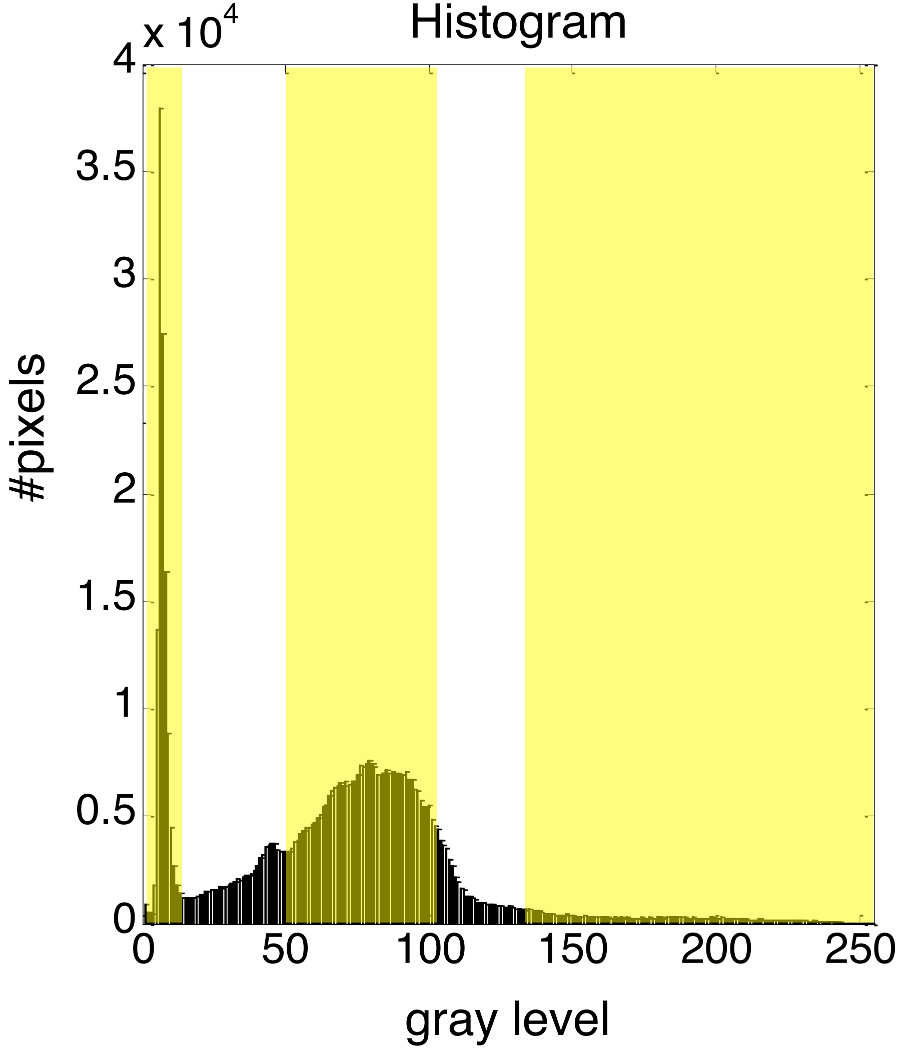
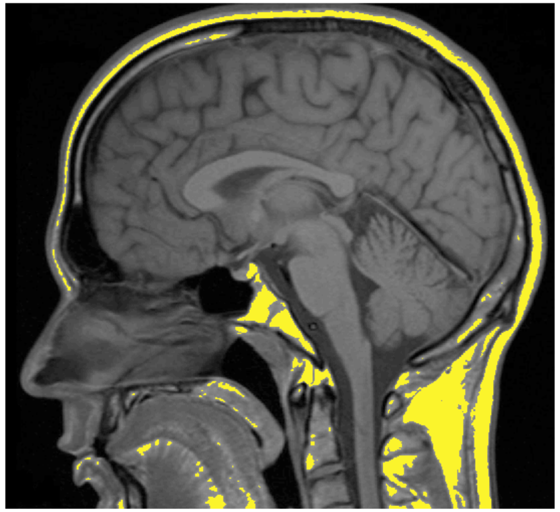
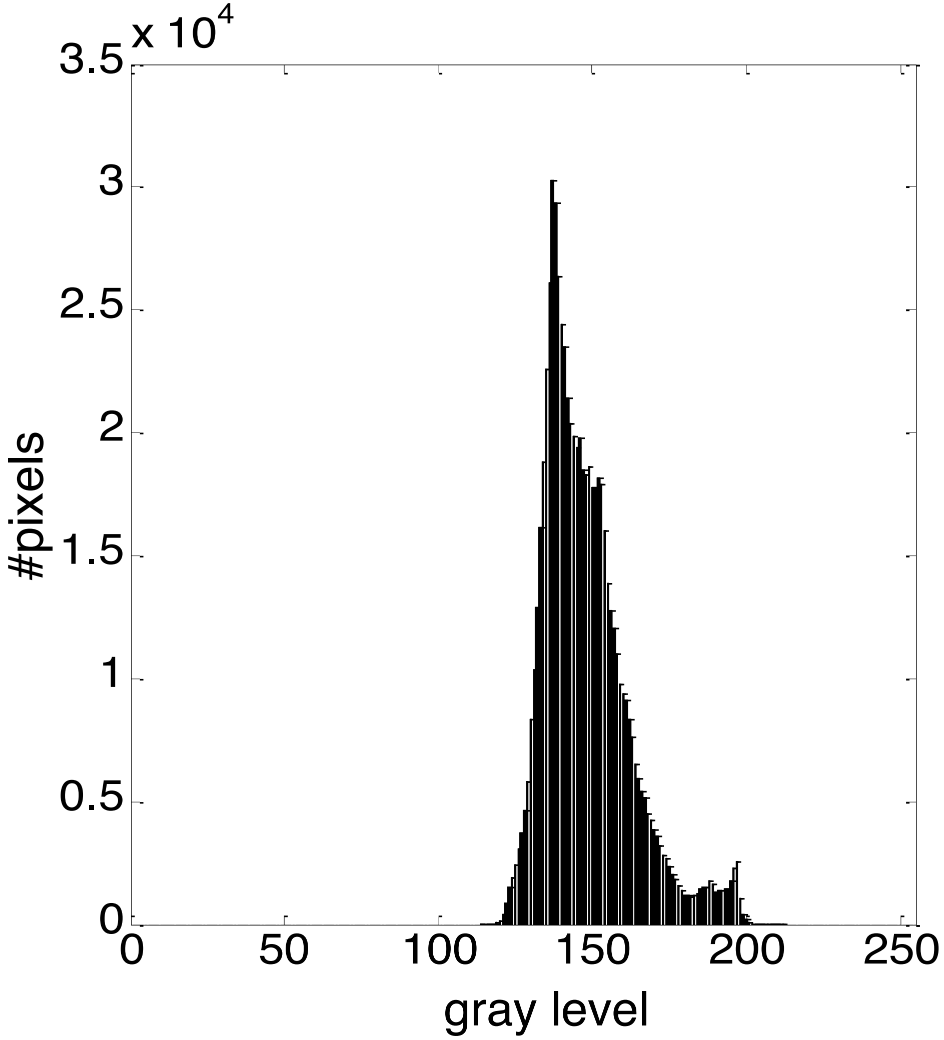
Gray level histograms



Brain image



Gray level histograms



Bay image

Gray level histogram in viewfinder

Gray level histograms

- To measure a histogram:
  - For B-bit image, initialize 2^B counters with 0
  - Loop over all pixels x,y
  - When encountering gray level $f[x,y]=i$, increment counter $\#\iota$
- Normalized histogram can be thought of as an estimate of the probability distribution of the continuous signal amplitude
- Use fewer, larger bins to trade off amplitude resolution against sample size.

Histogram equalization

Idea:

Find a non-linear transformation

$$\boxed{g = T\left(f\right)}$$

that is applied to each pixel of the input image $f[x,y]$, such that a uniform distribution of gray levels results for the output image $g[x,y]$.

Histogram equalization

Analyse ideal, continuous case first ...

Assume

- Normalized input values $0 \le f \le 1$ and output values $0 \le g \le 1$
- $T(f)$ is differentiable, increasing, and invertible, i.e., there exists

$$f = T^{-1}(g) \qquad 0 \le g \le 1$$

Goal: pdf $p_g(g) = 1$ over the entire range $0 \le g \le 1$

Histogram equalization for continuous case

- From basic probability theory

$$p_f(f) \xrightarrow{f} \boxed{T(f)} \xrightarrow{g} \qquad p_g(g) = \left[p_f(f) \frac{df}{dg} \right]_{f=T^{-1}(g)}$$

- Consider the transformation function

$$\boxed{g = T(f) = \int_0^f p_f(\alpha)\, d\alpha \qquad 0 \le f \le 1}$$

- Then . . .

$$\frac{dg}{df} = p_f(f)$$

$$p_g(g) = \left[p_f(f) \frac{df}{dg} \right]_{f=T^{-1}(g)} = \left[p_f(f) \frac{1}{p_f(f)} \right]_{f=T^{-1}(g)} = 1 \qquad 0 \le g \le 1$$

Histogram equalization for discrete case

- Now, f only assumes discrete amplitude values $f_0, f_1, \cdots, f_{L-1}$ with empirical probabilities

pixel count for amplitude f_l

$$P_0 = \frac{n_0}{n} \qquad P_1 = \frac{n_1}{n} \qquad \cdots \qquad P_{L-1} = \frac{n_{L-1}}{n} \qquad \text{where } n = \sum_{l=0}^{L-1} n_l$$

- Discrete approximation of $g = T(f) = \int_0^f p_f(\alpha)\, d\alpha$

$$\boxed{g_k = T[f_k] = \sum_{i=0}^{k} P_i \qquad \text{for } k = 0,1\ldots,L-1}$$

- The resulting values g_k are in the range $[0,1]$ and might have to be scaled and rounded appropriately.

Histogram equalization for discrete case

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- Discrete approximation of $g = T(f) = \int_0^f p_f(\alpha)\, d\alpha$

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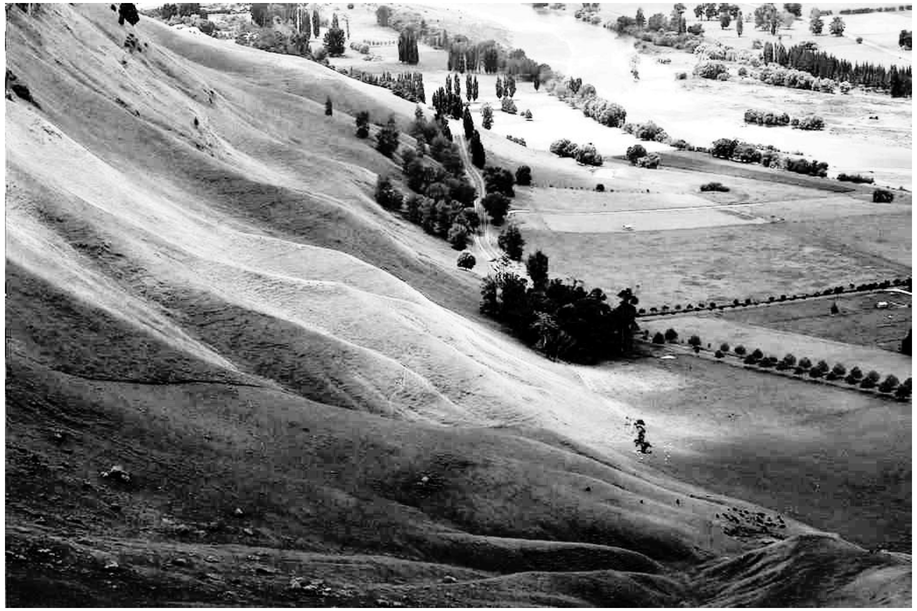
Histogram equalization example

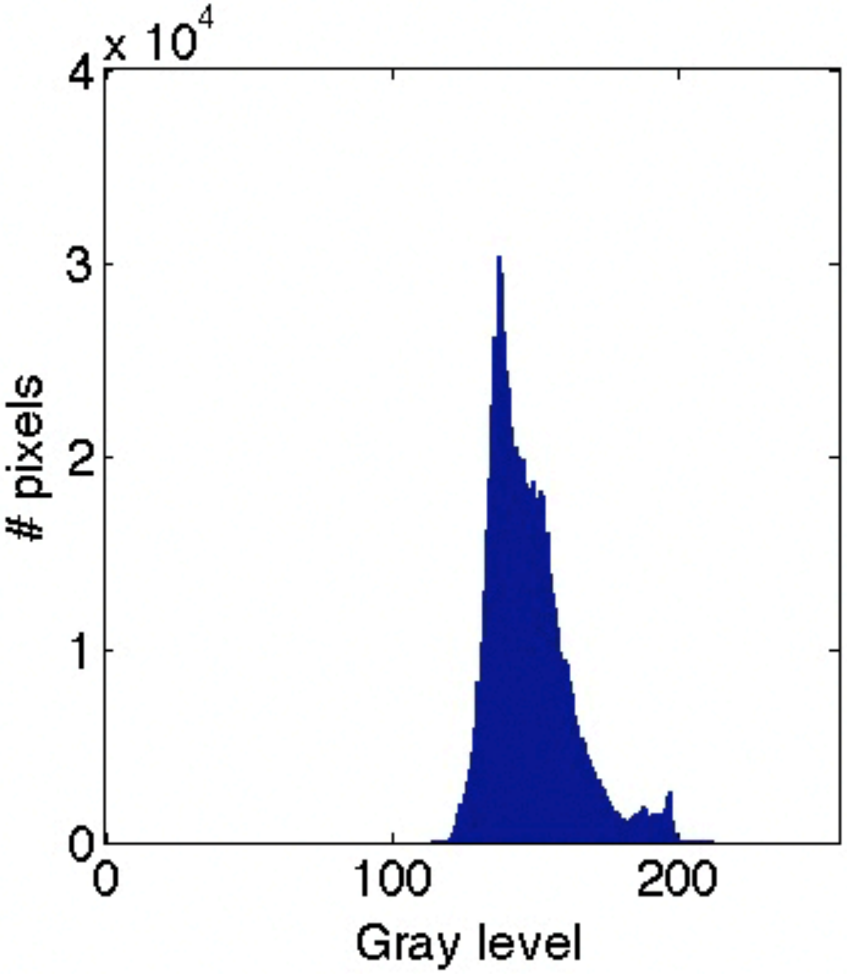
Original image *Bay*



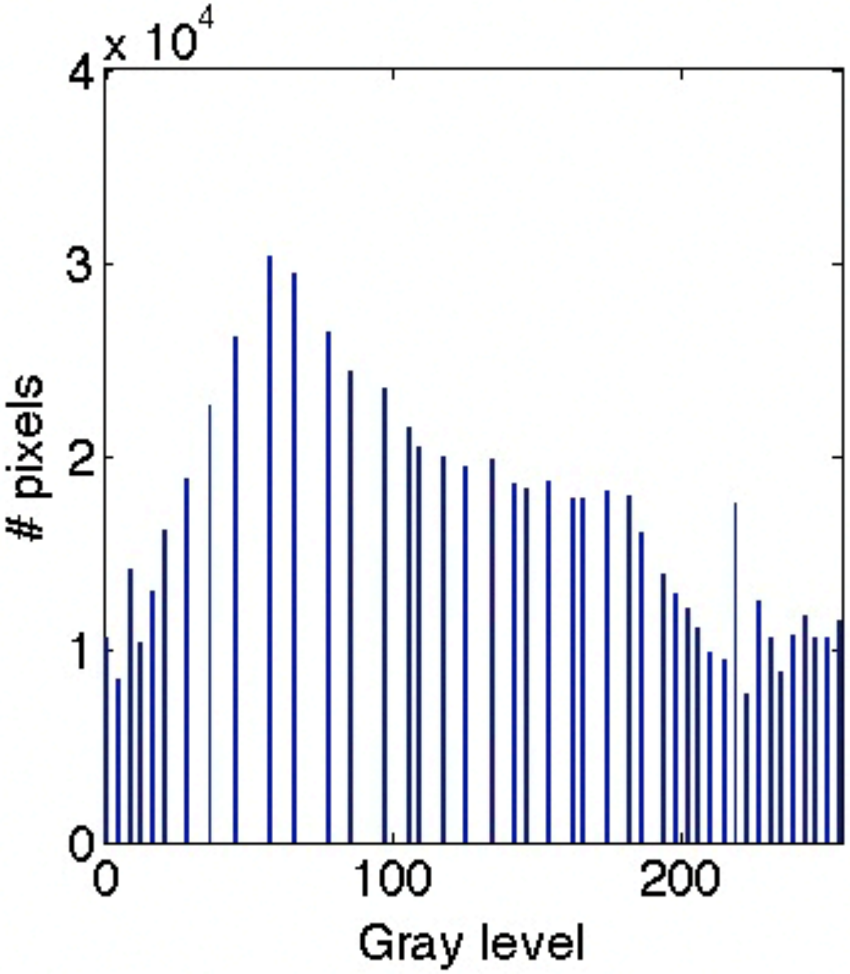
... after histogram equalization

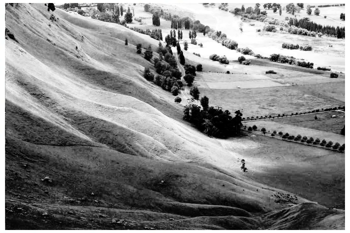
Histogram equalization example

Original image *Bay*

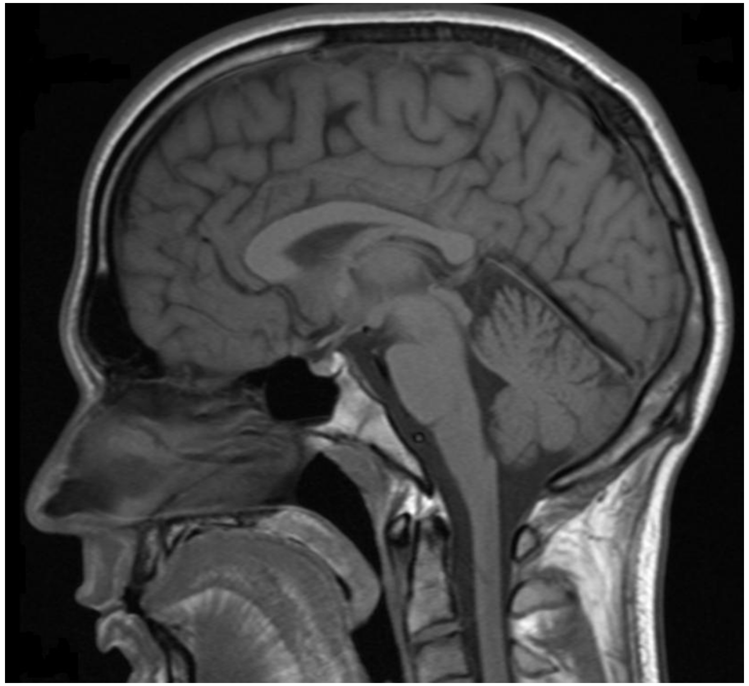


. . . after histogram equalization

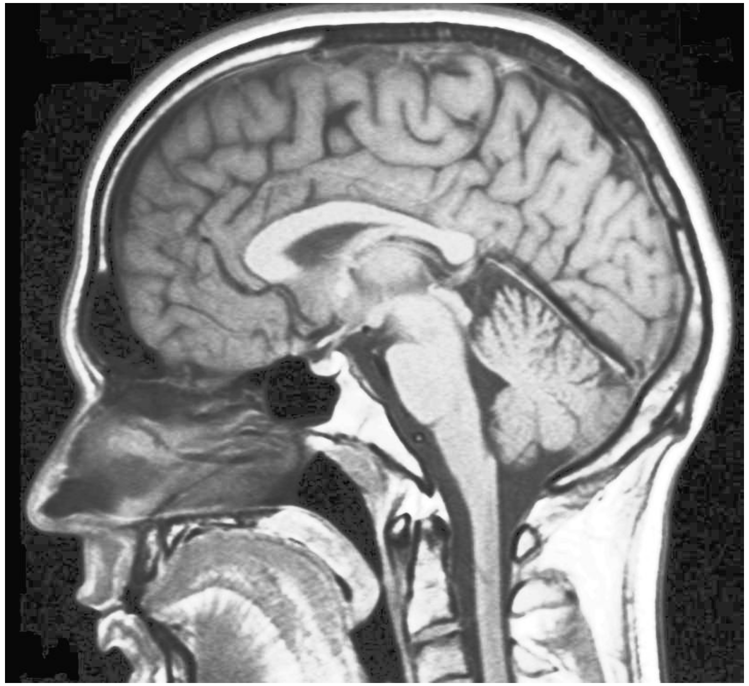




Histogram equalization example



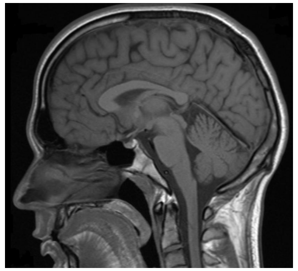
Original image *Brain*

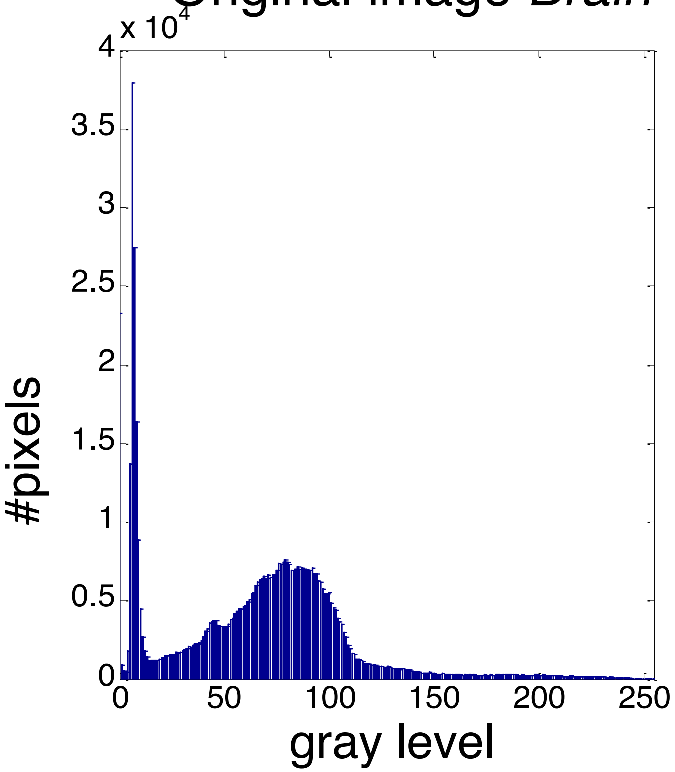


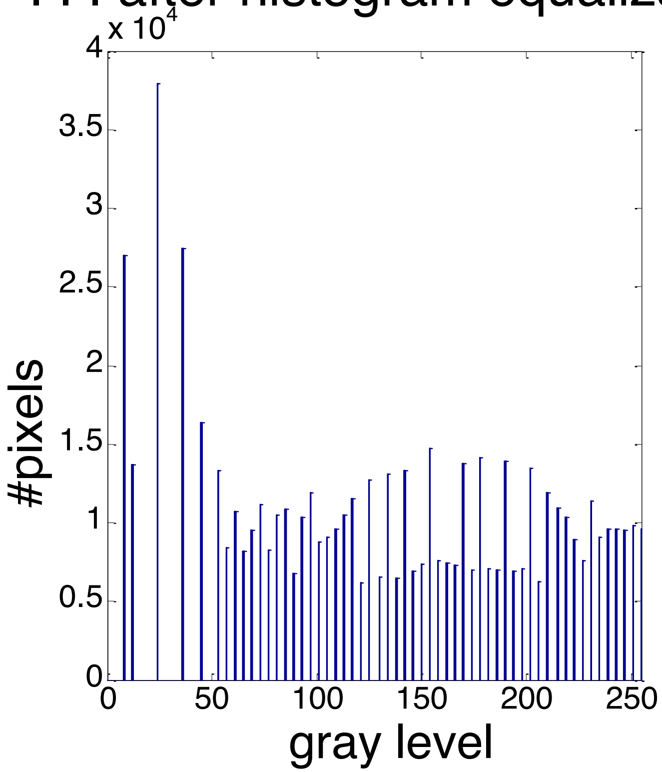
... after histogram equalization

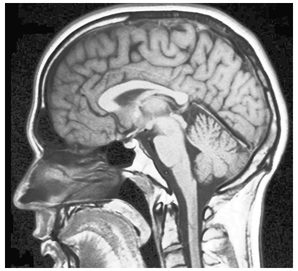
Histogram equalization example

Original image *Brain* . . . after histogram equalization

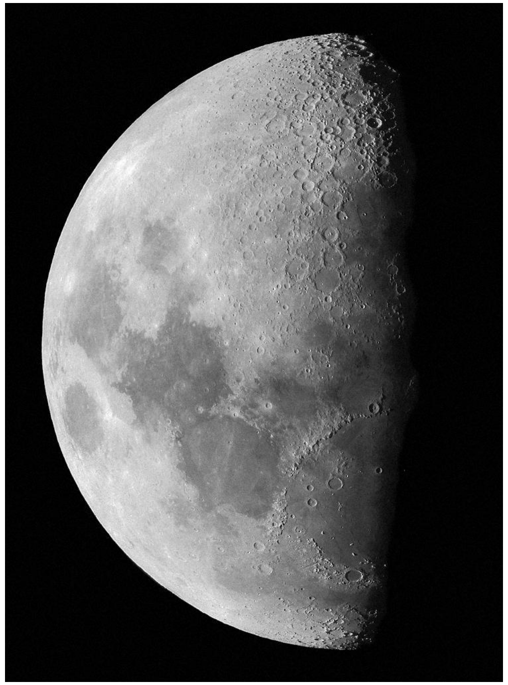




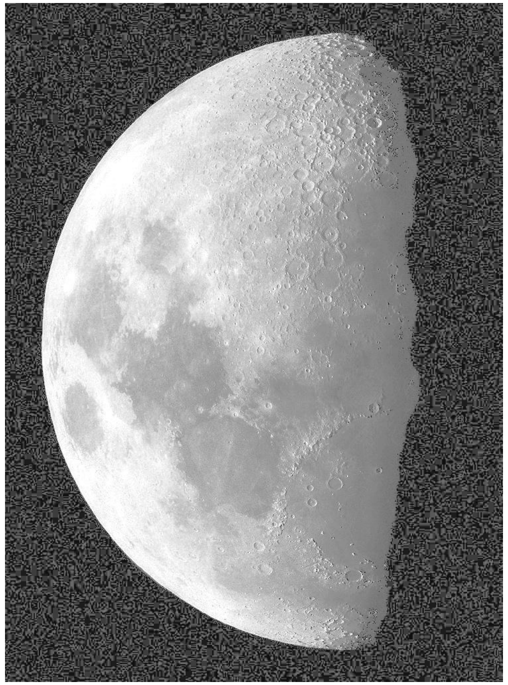




Histogram equalization example



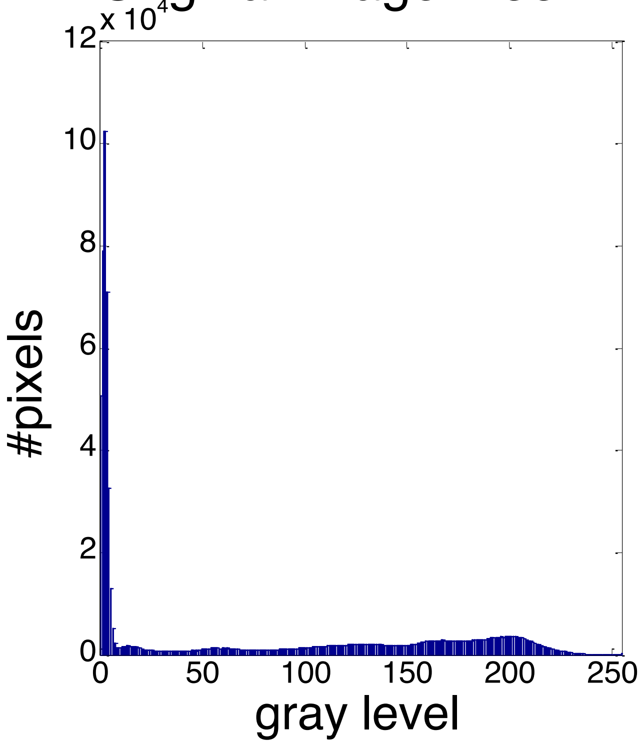
Original image *Moon*



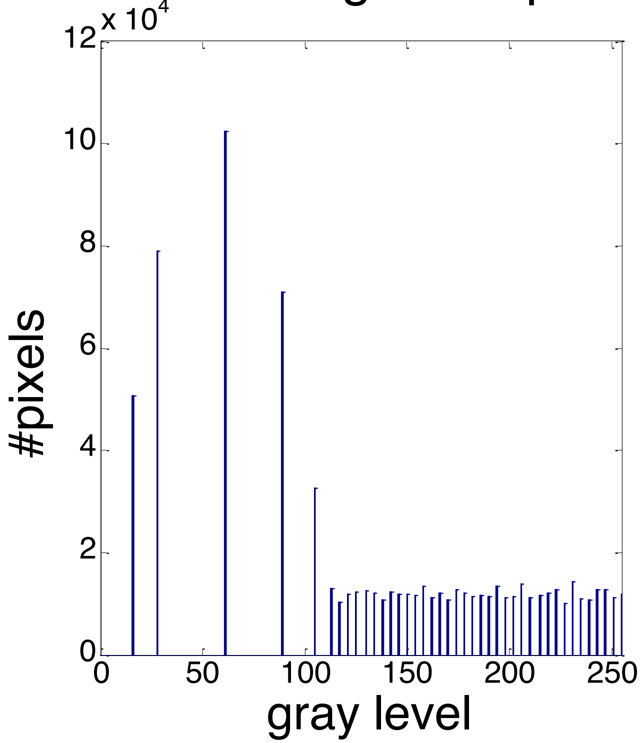
... after histogram equalization

Histogram equalization example

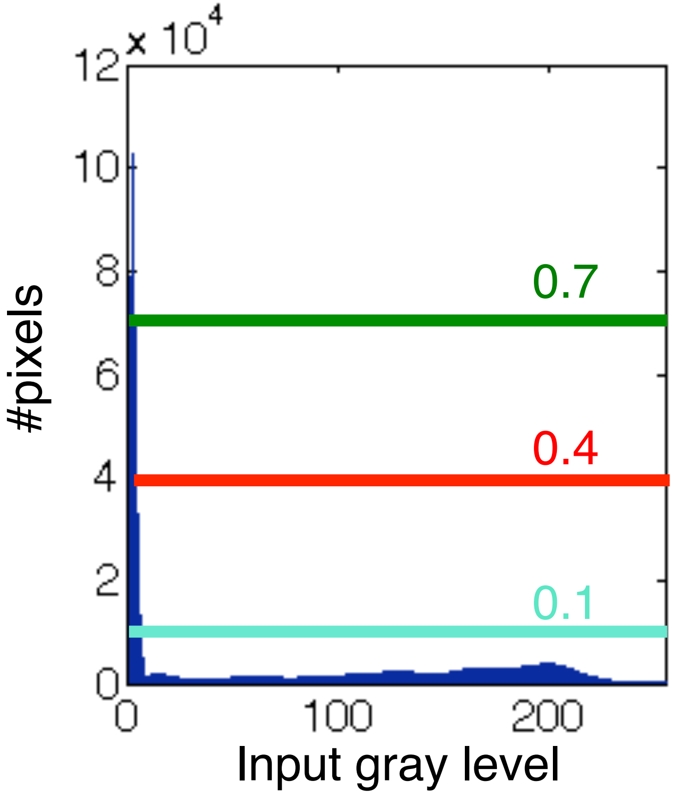
Original image *Moon*

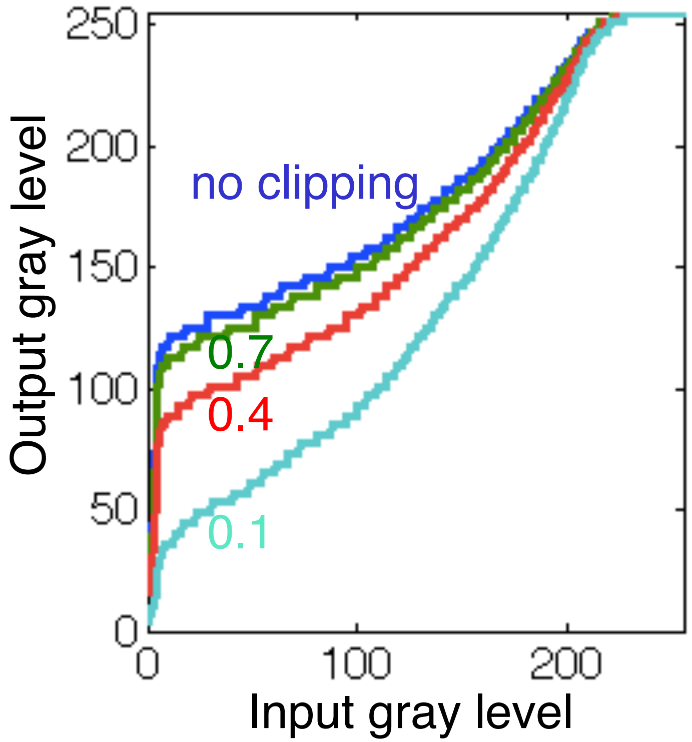


. . . after histogram equalization



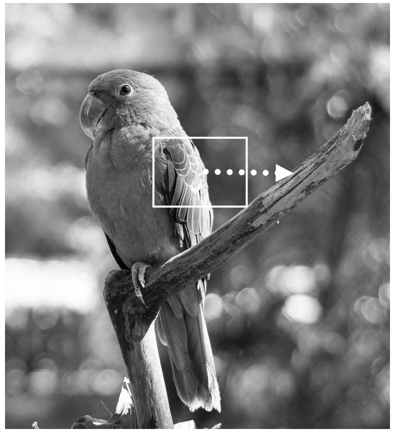
Contrast-limited histogram equalization



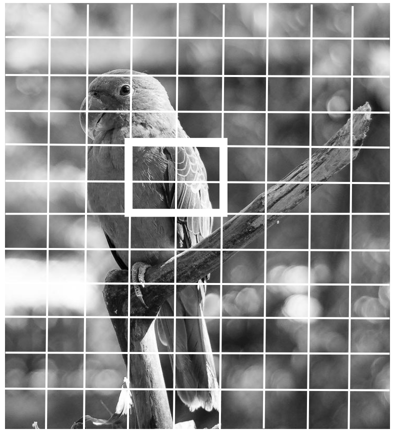


Adaptive histogram equalization

- Histogram equalization based on a histogram obtained from a portion of the image



Sliding window approach:
different histogram (and mapping) for every pixel



Tiling approach:
subdivide into overlapping regions, mitigate blocking effect by smooth blending between neighboring tiles

- Limit contrast expansion in flat regions of the image, e.g., by clipping histogram values.
 ("Contrast-limited adaptive histogram equalization")

[Pizer, Amburn et al. 1987]

Adaptive histogram equalization

Original image
Parrot

Global histogram equalization

Adaptive histogram equalization, 8x8 tiles

Adaptive histogram equalization, 16x16 tiles

Adaptive histogram equalization

Original image
Dental Xray

Global histogram equalization

Adaptive histogram equalization, 8x8 tiles

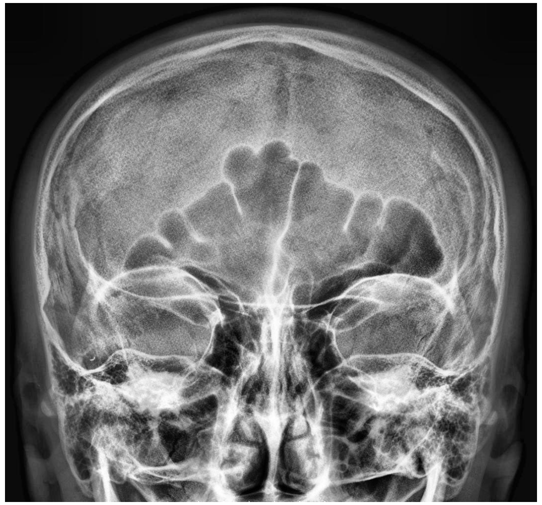
Adaptive histogram equalization, 16x16 tiles

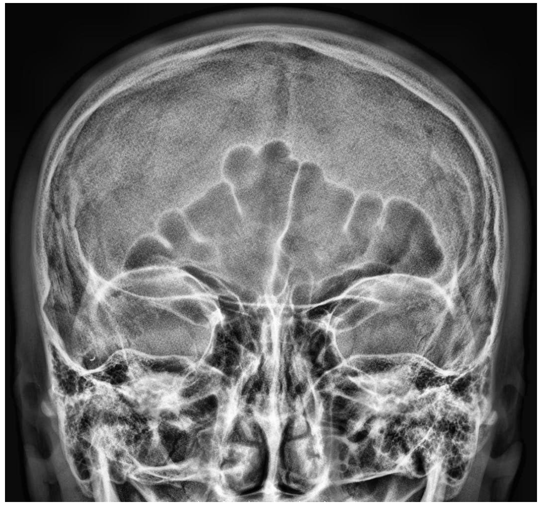
Adaptive histogram equalization

Original image
Skull Xray

Global histogram
equalization

Adaptive histogram
equalization, 8x8 tiles



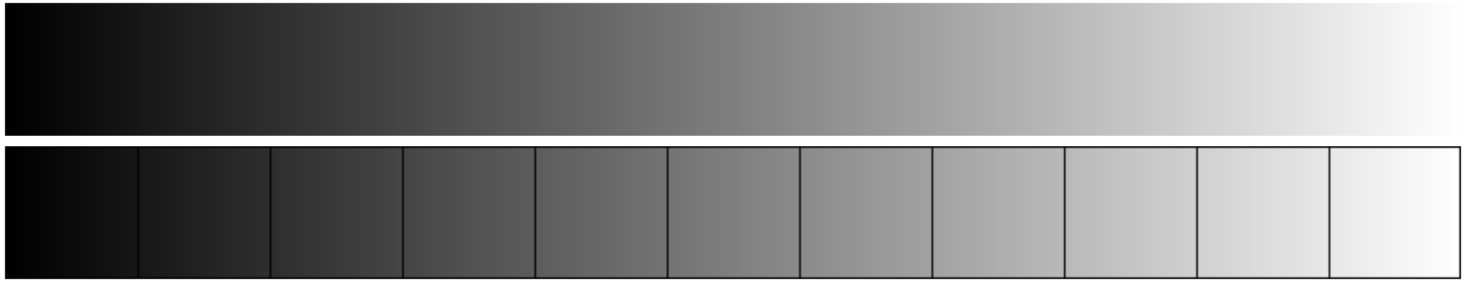


Adaptive histogram
equalization, 16x16 tiles

Ansel Adam's Zone System (1939)

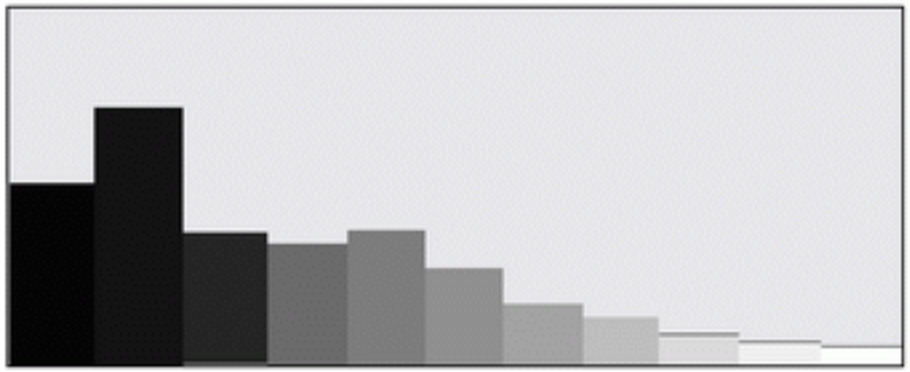
full tonal gradation

11-step gradation



Zone	Description
0	Pure black
I	Near black, with slight tonality but no texture
II	Textured black; the darkest part of the image in which slight detail is recorded
III	Average dark materials and low values showing adequate texture
IV	Average dark foliage, dark stone, or landscape shadows
V	Middle gray: clear north sky; dark skin, average weathered wood
VI	Average Caucasian skin; light stone; shadows on snow in sunlit landscapes
VII	Very light skin; shadows in snow with acute side lighting
VIII	Lightest tone with texture: textured snow
IX	Slight tone without texture; glaring snow
X	Pure white: light sources and specular reflections

histogram of 11 zones → not quite flat



Ansel Adams, The Tetons and Snake River, 1942

Photoshop Demo