

21.08.2020

# Digital Image Processing (CSE/ECE 478)

3

## Lecture-3: Recap/Discussion

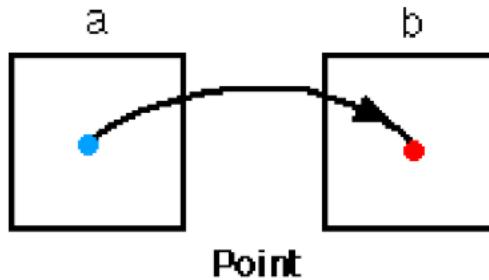
Ravi Kiran



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# Spatial Domain Processing

- ▶ Manipulating Pixels Directly in Spatial Domain
- ▶ 3 approaches
- ▶ 1. Point to Point



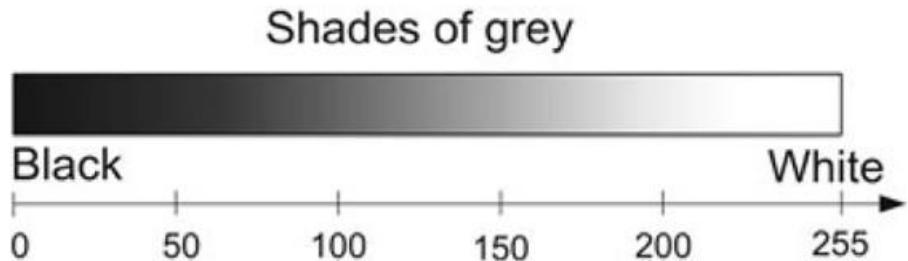
# Linear Intensity Transforms

- $T(z) = z + K$

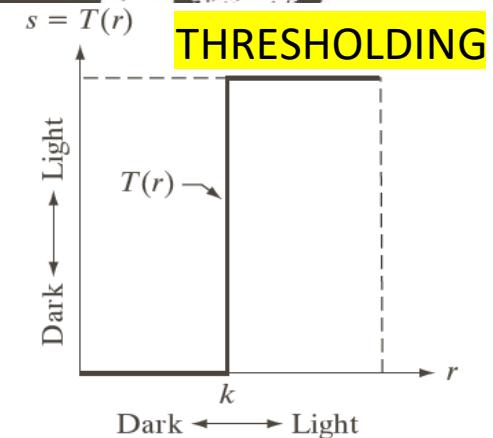
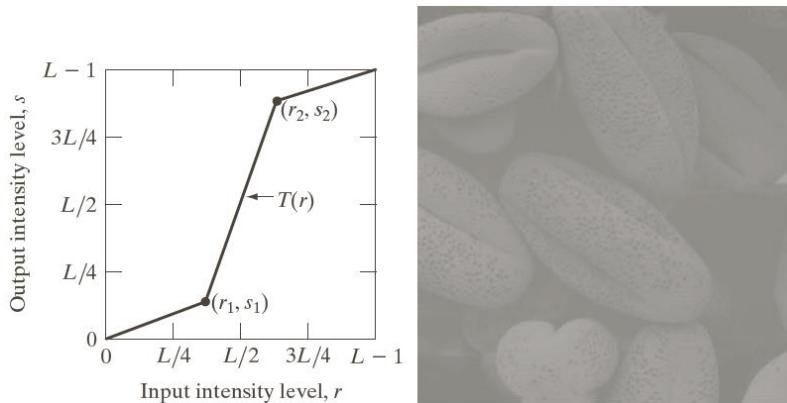
- $T(z) = z - K$

- $T(z) = Kz$

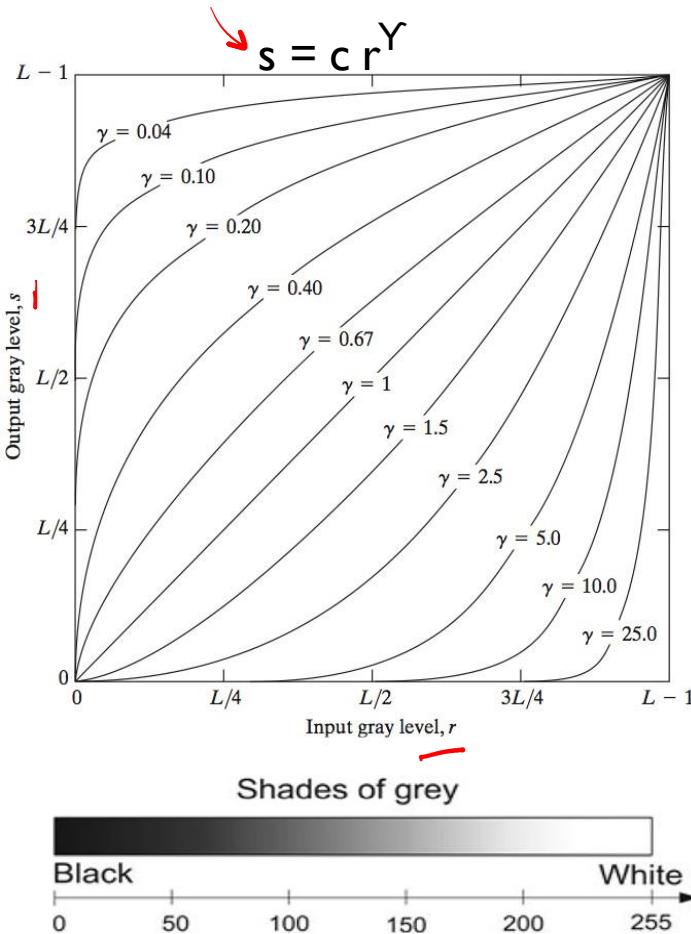
- $T(z) = K_1z + K_2$



# Piecewise-Linear Transformations



# Power-Law Transformations

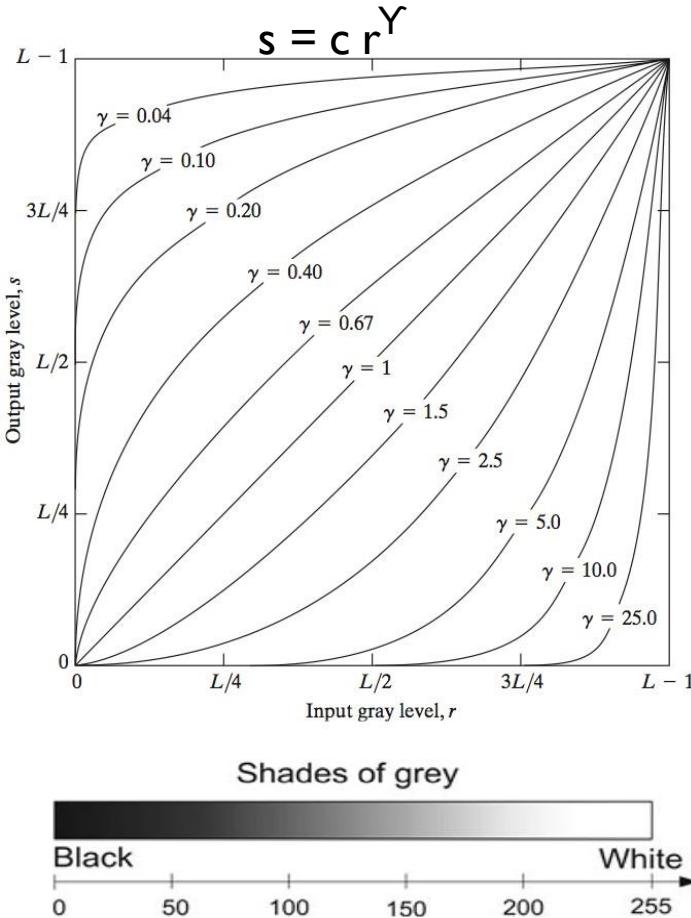


a  
b  
c  
d

**FIGURE 3.9**  
(a) Aerial image.  
(b)–(d) Results of applying the transformation in Eq. (3.2-3) with  $c = 1$  and  $\gamma = 3.0, 4.0$ , and  $5.0$ , respectively.  
(Original image for this example courtesy of NASA.)



# Power-Law Transformations



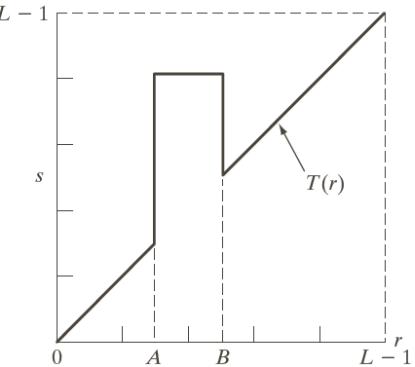
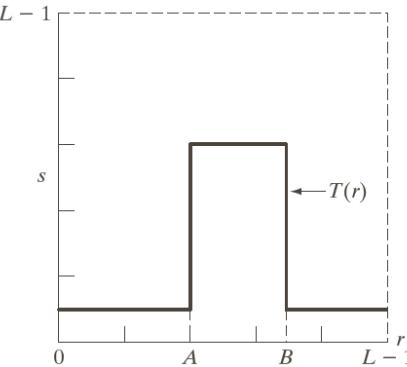
Demo:

<https://colab.research.google.com/drive/11qlLOVKleZnONtPuxAryAf9WkUC7kEMI#scrollTo=eaU5WQaqOpSCr&line=12&uniquifier=1>

# Intensity Slicing

a | b

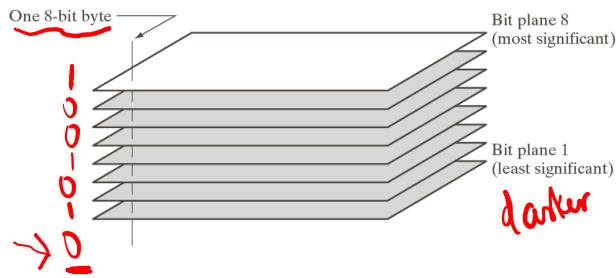
**FIGURE 3.11** (a) This transformation highlights intensity range  $[A, B]$  and reduces all other intensities to a lower level. (b) This transformation highlights range  $[A, B]$  and preserves all other intensity levels.



a | b | c

# Bit plane slicing

## 8-bit

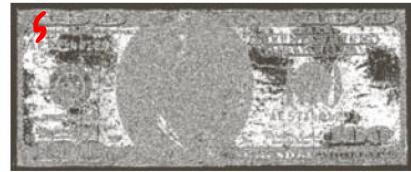
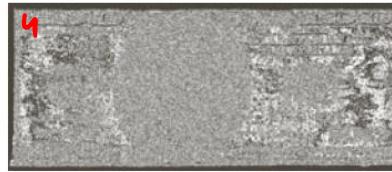
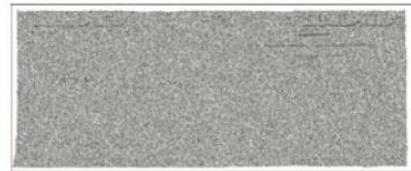
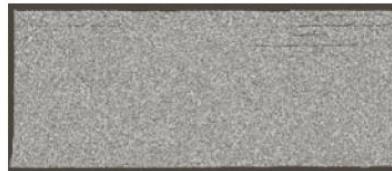


brighter

Bit plane 8  
(most significant)

↗ Bit plane 1  
↗ (least significant)

darker



0 0 0 0 0 0 0 ↗

0 | 0 0 0 0

1 0 0 0 0

১ ২ ৩ ৪ ৫ ৬ ৭

a	b	c
d	e	f
g	h	i

178

1 bis

256

**FIGURE 3.14** (a) An 8-bit gray-scale image of size  $500 \times 1192$  pixels. (b) through (i) Bit planes 1 through 8, with bit plane 1 corresponding to the least significant bit. Each bit plane is a binary image.

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# Digital Image Processing (CSE/ECE 478)

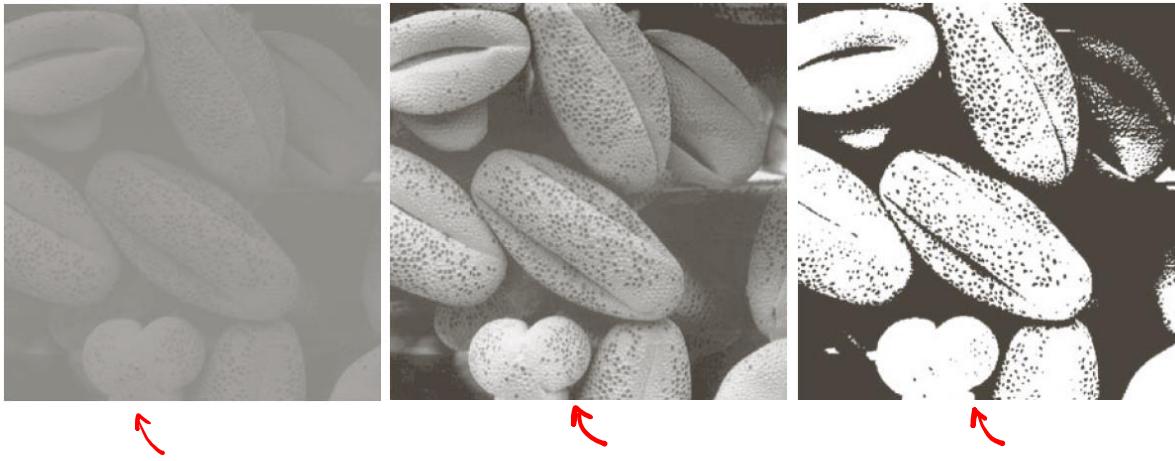
## Lecture-4: Histogram Processing

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# Piecewise-Linear Transformations



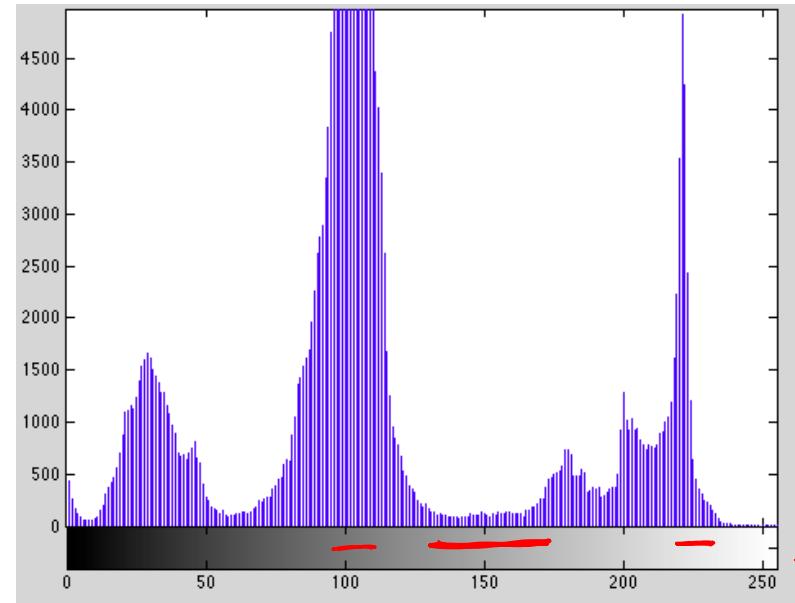
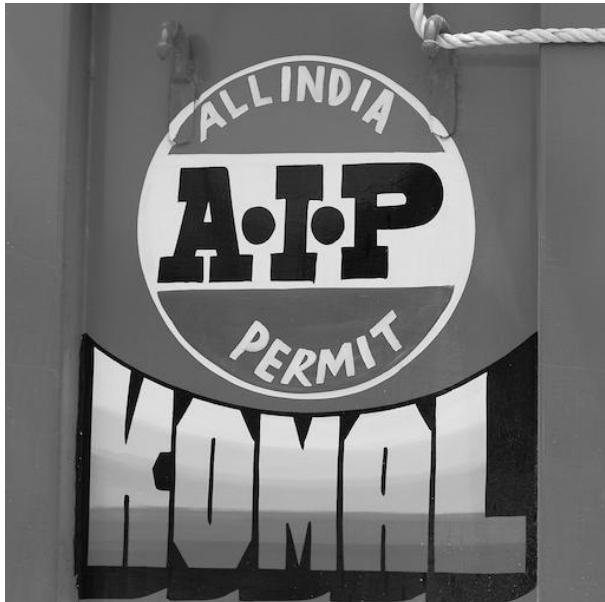
# Histogram: An image representation + visualization

$$h_r(i) = n_i$$

$i \rightarrow$  intensity value, range  $[0, L-1]$

$n_i \rightarrow$  number of pixels with intensity  $i$

I



← 256  
int

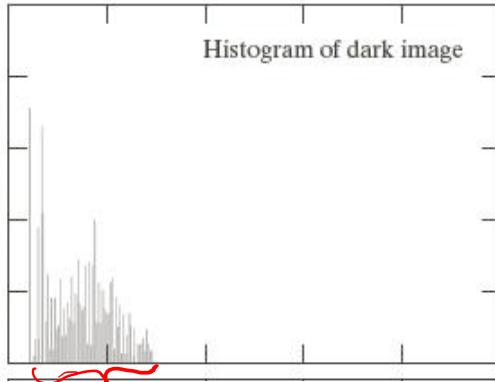
# Histograms

- ▶ What can we infer from histograms?



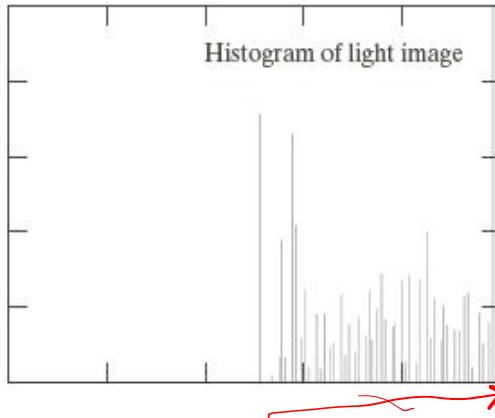
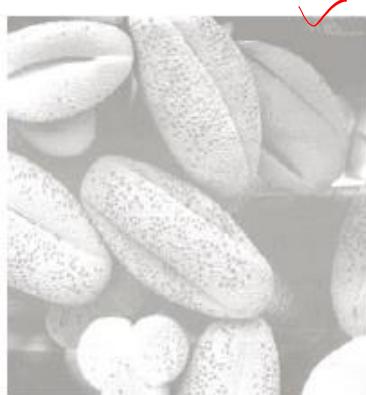
Histogram viewing standard in most DSLR cameras

# Histograms and Contrast

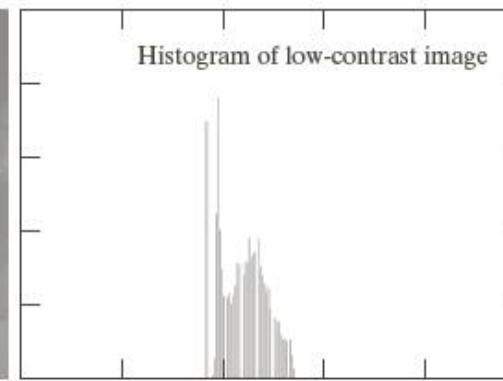
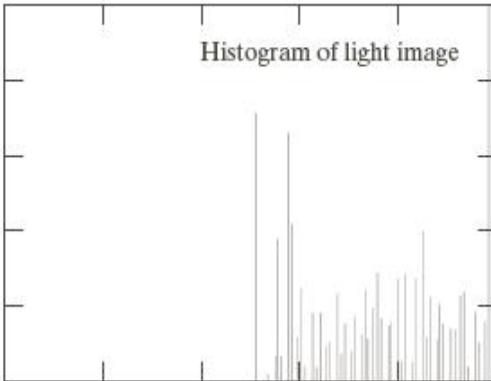
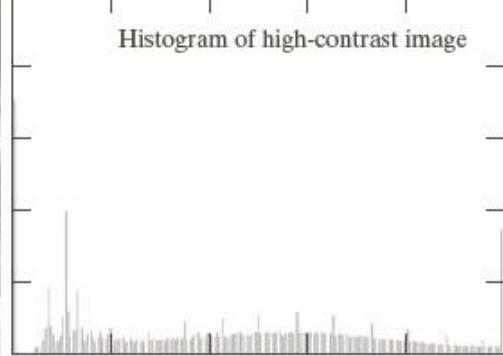
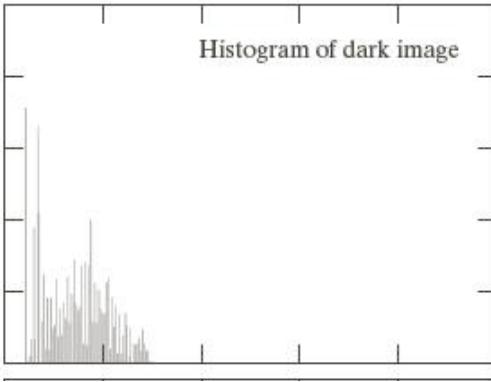


Contrast = 
$$\frac{I_{max} - I_{min}}{I_{max} + I_{min}}$$

Red annotations: A red bracket groups the terms  $I_{max} - I_{min}$  and  $I_{max} + I_{min}$ . Red arrows point from the top of the bracket to the  $I_{max}$  term and from the bottom of the bracket to the  $I_{min}$  term.



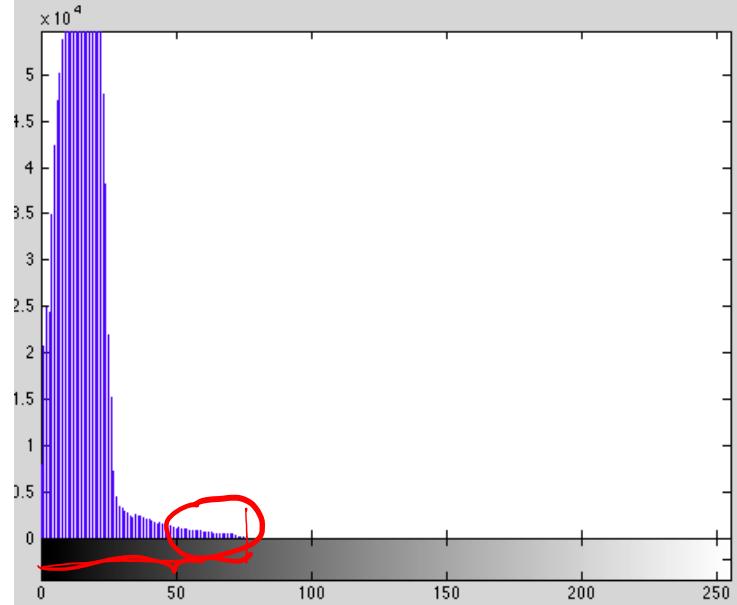
# Histograms and Contrast



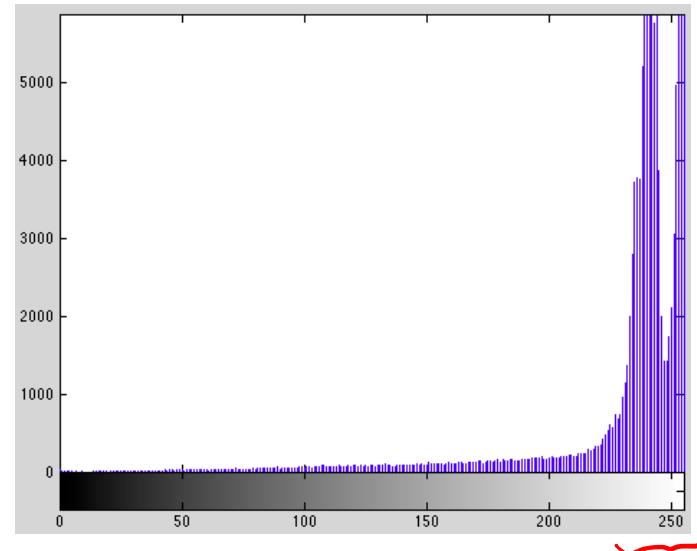
# Histograms



Under exposure

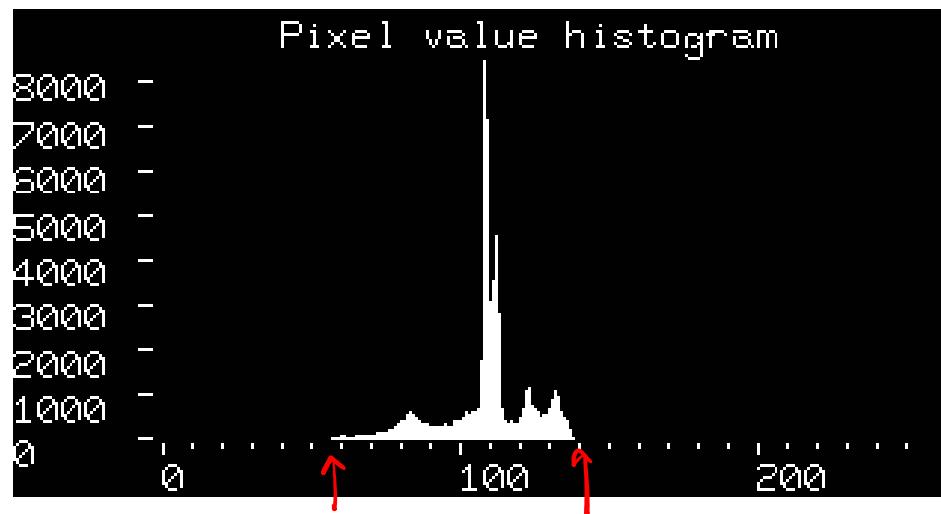


# Histograms

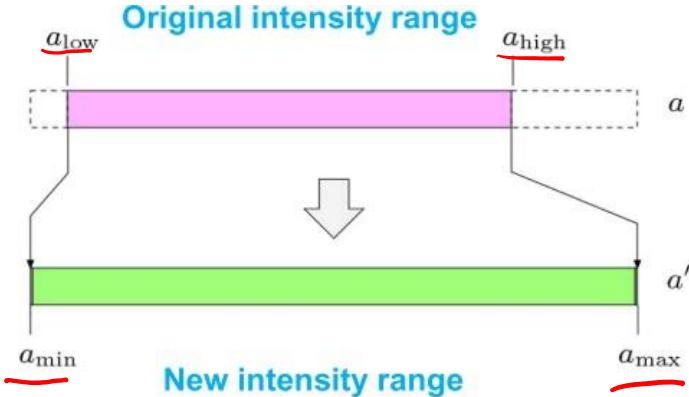


Over exposure

# A low-contrast image and its histogram

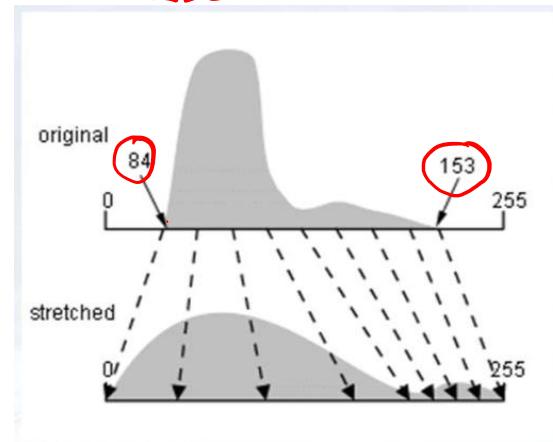
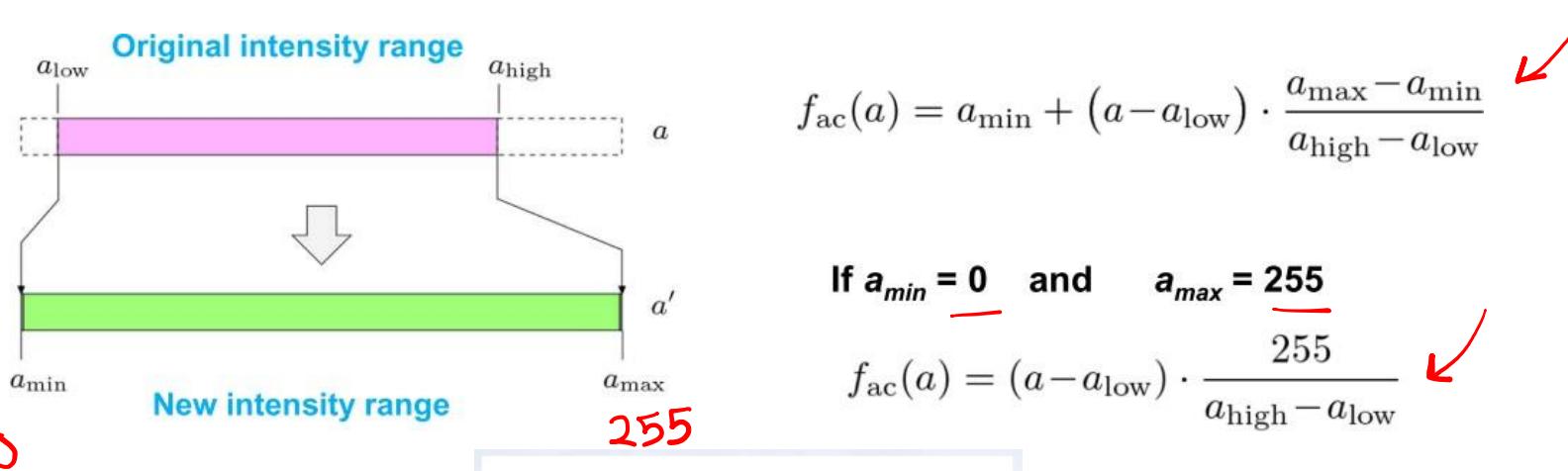


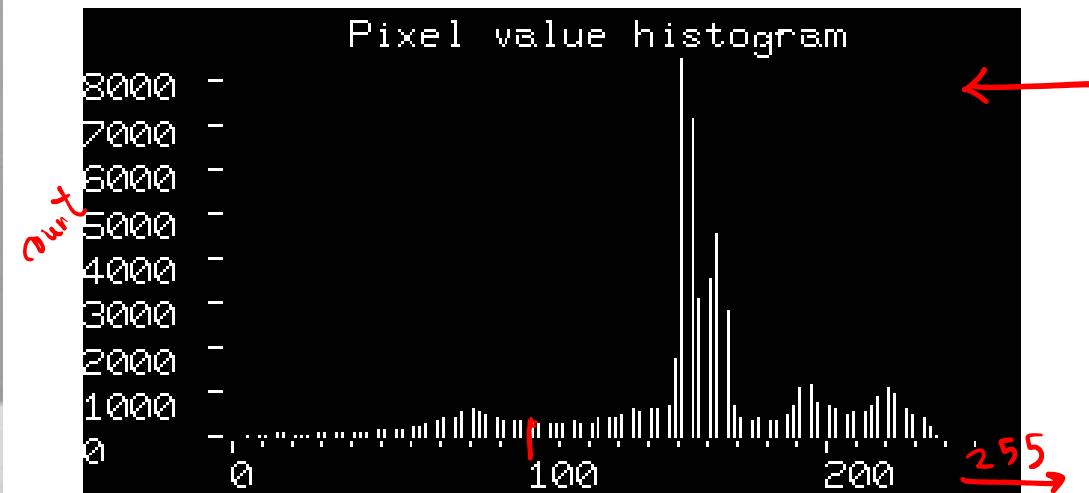
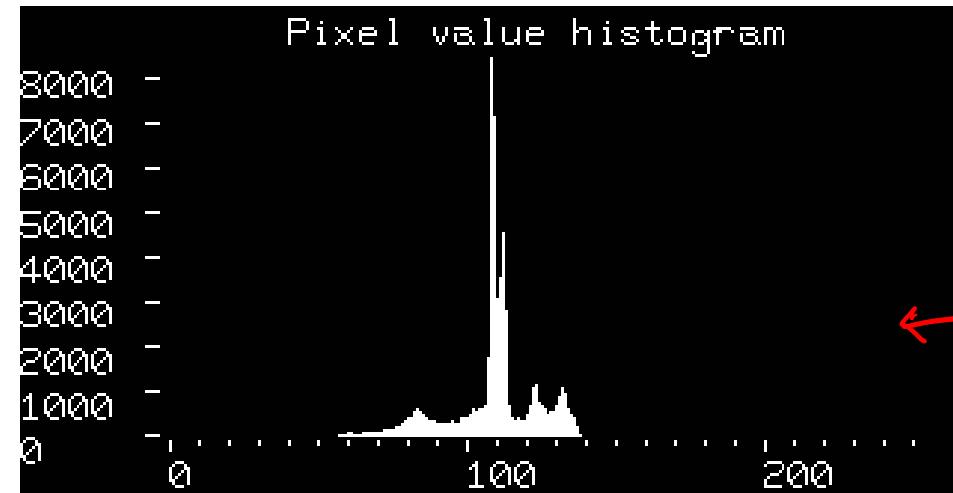
# Contrast Stretching



$$f_{\text{ac}}(a) = a_{\text{min}} + (a - a_{\text{low}}) \cdot \frac{a_{\text{max}} - a_{\text{min}}}{a_{\text{high}} - a_{\text{low}}}$$

# Contrast Stretching





# Contrast Stretching

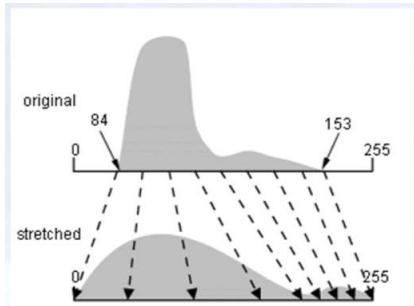


Suppose we have a single pixel with intensity 255 in the original intensity range.  
What happens ?

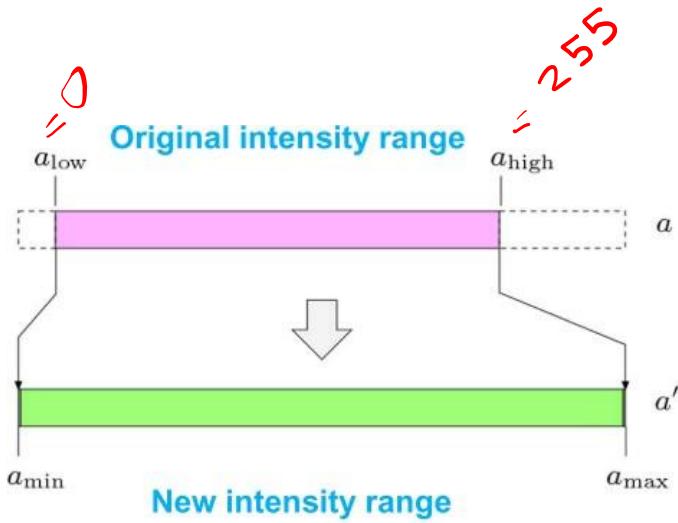
$$f_{ac}(a) = a_{\min} + (a - a_{\min}) \cdot \frac{a_{\max} - a_{\min}}{a_{\text{high}} - a_{\text{low}}}$$

If  $a_{\min} = 0$  and  $a_{\max} = 255$

$$f_{ac}(a) = (a - a_{\min}) \cdot \frac{255}{a_{\text{high}} - a_{\text{low}}}$$



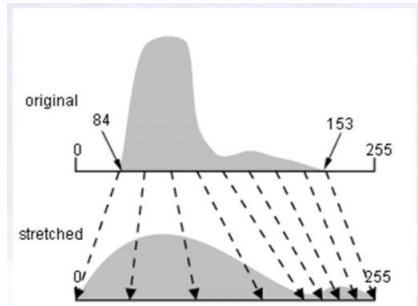
# Contrast Stretching



$$f_{ac}(a) = a_{\min} + (a - a_{\min}) \cdot \frac{a_{\max} - a_{\min}}{a_{\max} - a_{\min}}$$

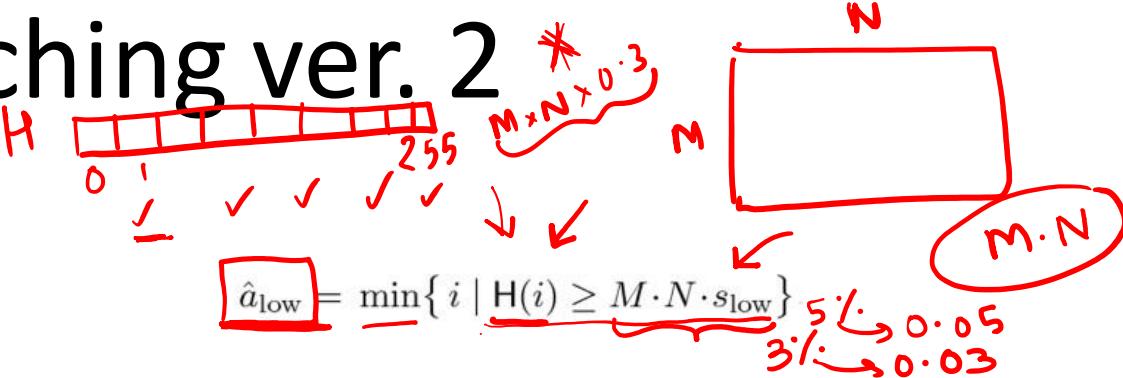
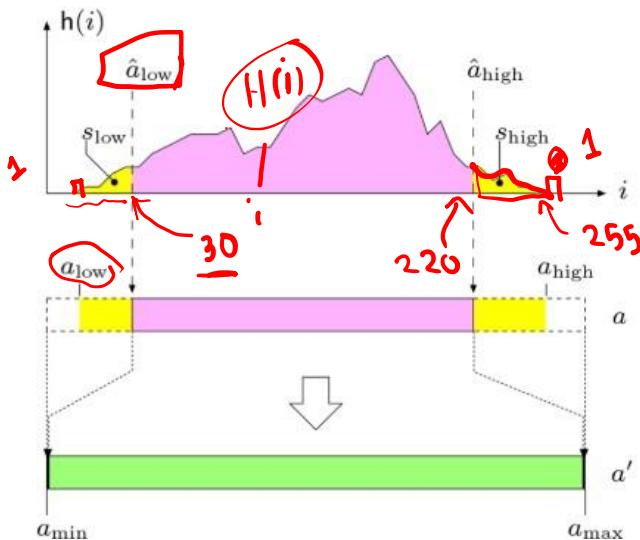
If  $a_{\min} = 0$  and  $a_{\max} = 255$

$$f_{ac}(a) = (a - a_{\min}) \cdot \frac{255}{a_{\max} - a_{\min}}$$



Suppose we have a single pixel with intensity 0 in the original intensity range.  
What happens ?

# Contrast Stretching ver. 2



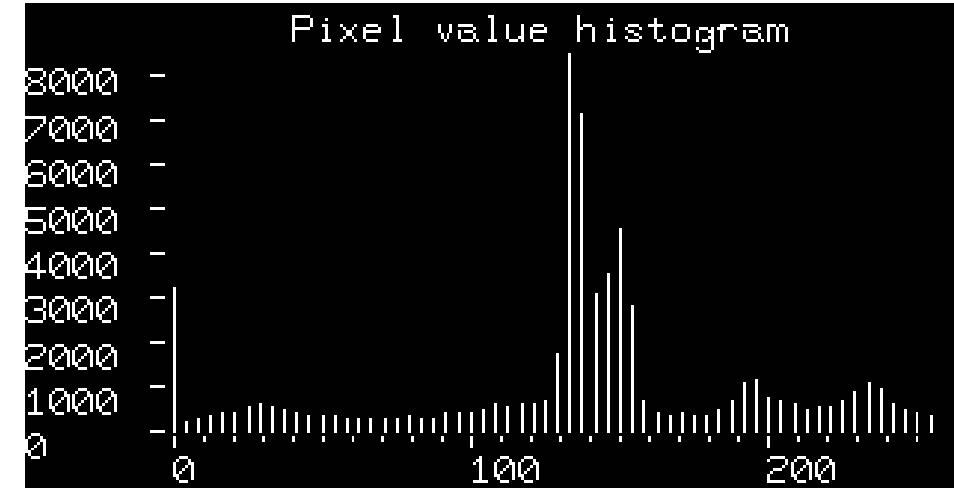
$$\hat{a}_{\text{low}} = \min\{ i \mid H(i) \geq M \cdot N \cdot s_{\text{low}} \}$$

$$\hat{a}_{\text{high}} = \max\{ i \mid H(i) \leq M \cdot N \cdot (1 - s_{\text{high}}) \}$$

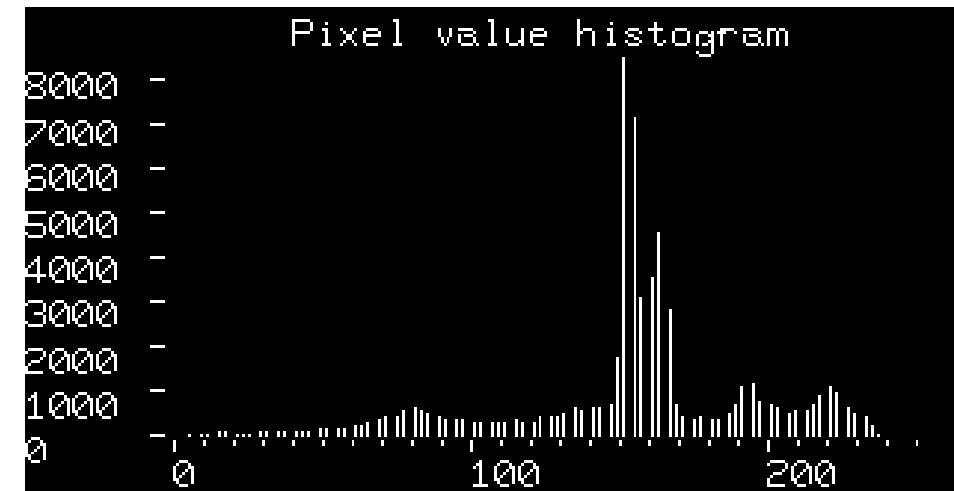
$$f_{\text{mac}}(a) = \begin{cases} a_{\min} & \text{for } a \leq \hat{a}_{\text{low}} \\ a_{\min} + (a - \hat{a}_{\text{low}}) \cdot \frac{a_{\max} - a_{\min}}{\hat{a}_{\text{high}} - \hat{a}_{\text{low}}} & \text{for } \hat{a}_{\text{low}} < a < \hat{a}_{\text{high}} \\ a_{\max} & \text{for } a \geq \hat{a}_{\text{high}} \end{cases}$$

$s_{\text{low}}$      $s_{\text{high}}$

Ver. 2

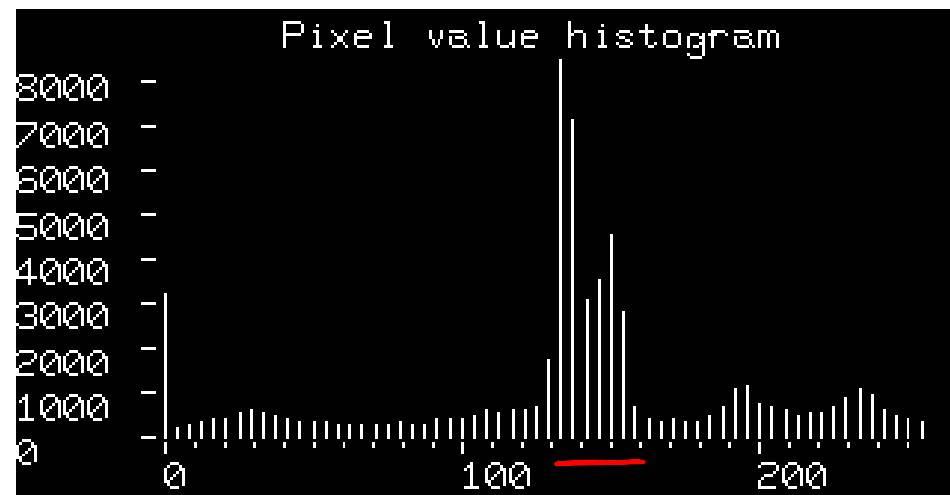


Ver. 1



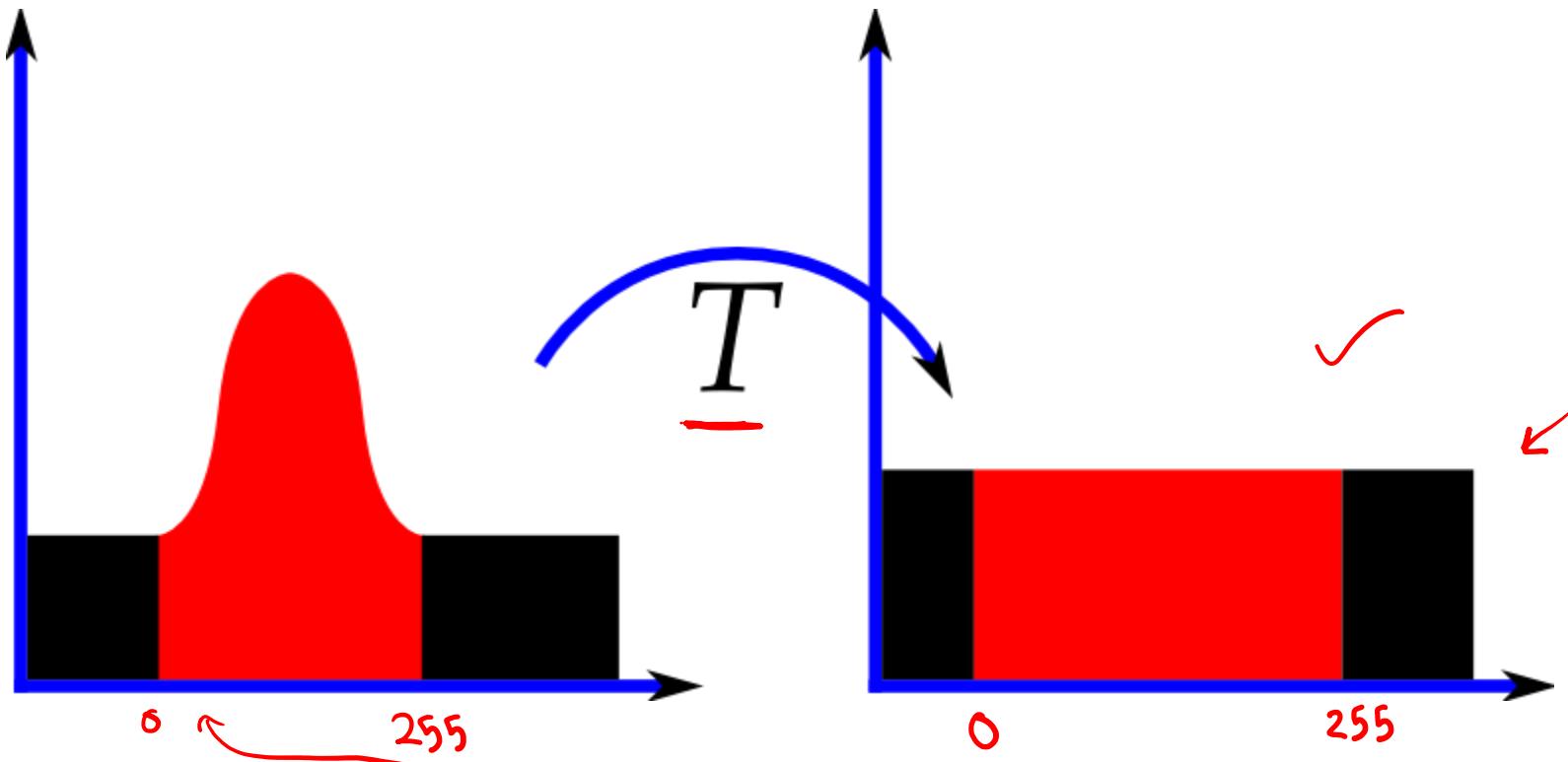
# Are all intensities well represented ?

Ver. 2

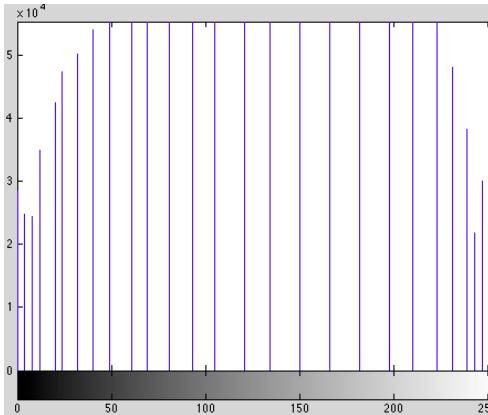
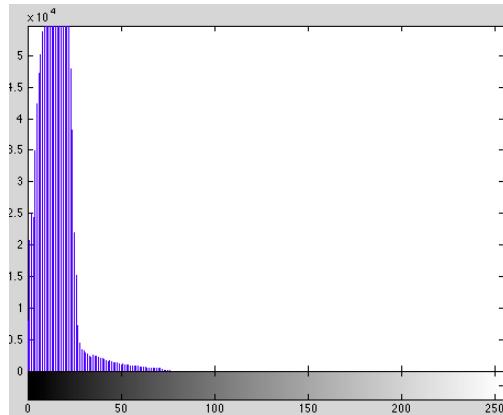
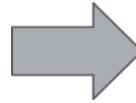




# Histogram Equalization



# Histogram Equalization



# The issue with contrast stretching

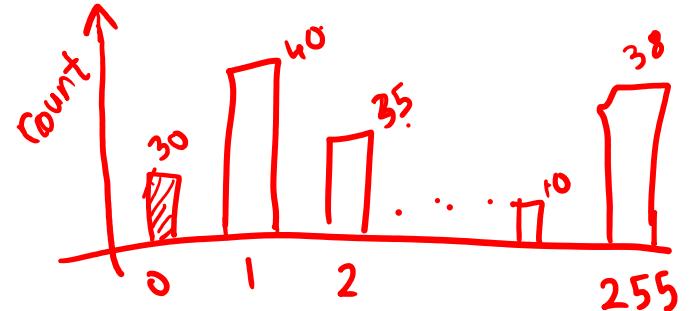


$$f_{ac}(a) = a_{\min} + (a - a_{\min}) \cdot \frac{a_{\max} - a_{\min}}{a_{\text{high}} - a_{\text{low}}}$$

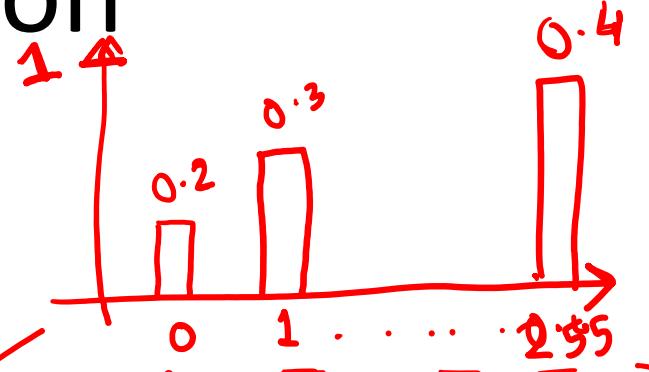
If  $a_{\min} = 0$  and  $a_{\max} = 255$

$$f_{ac}(a) = (a - a_{\min}) \cdot \frac{255}{a_{\text{high}} - a_{\text{low}}}$$

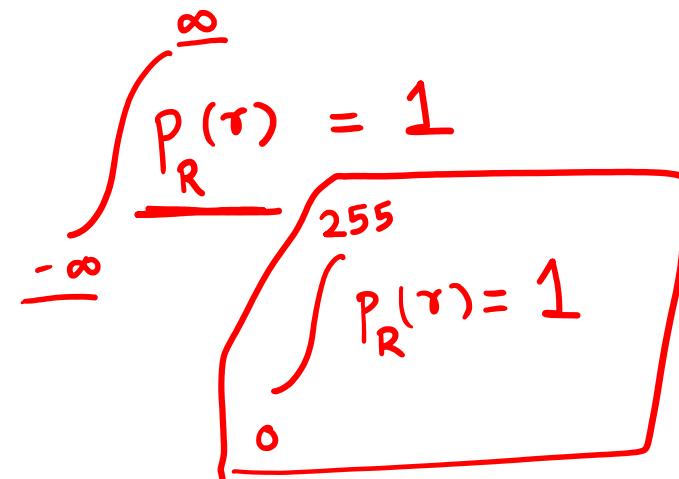
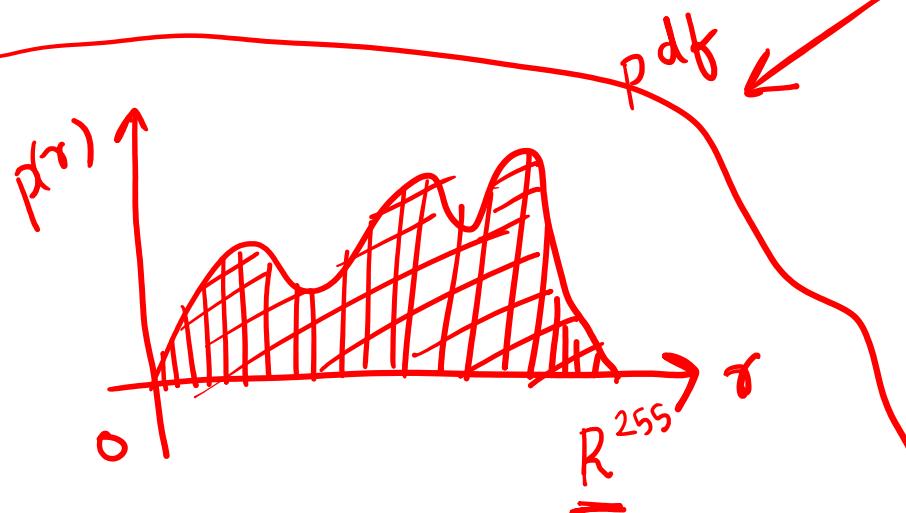
# Histogram Equalization



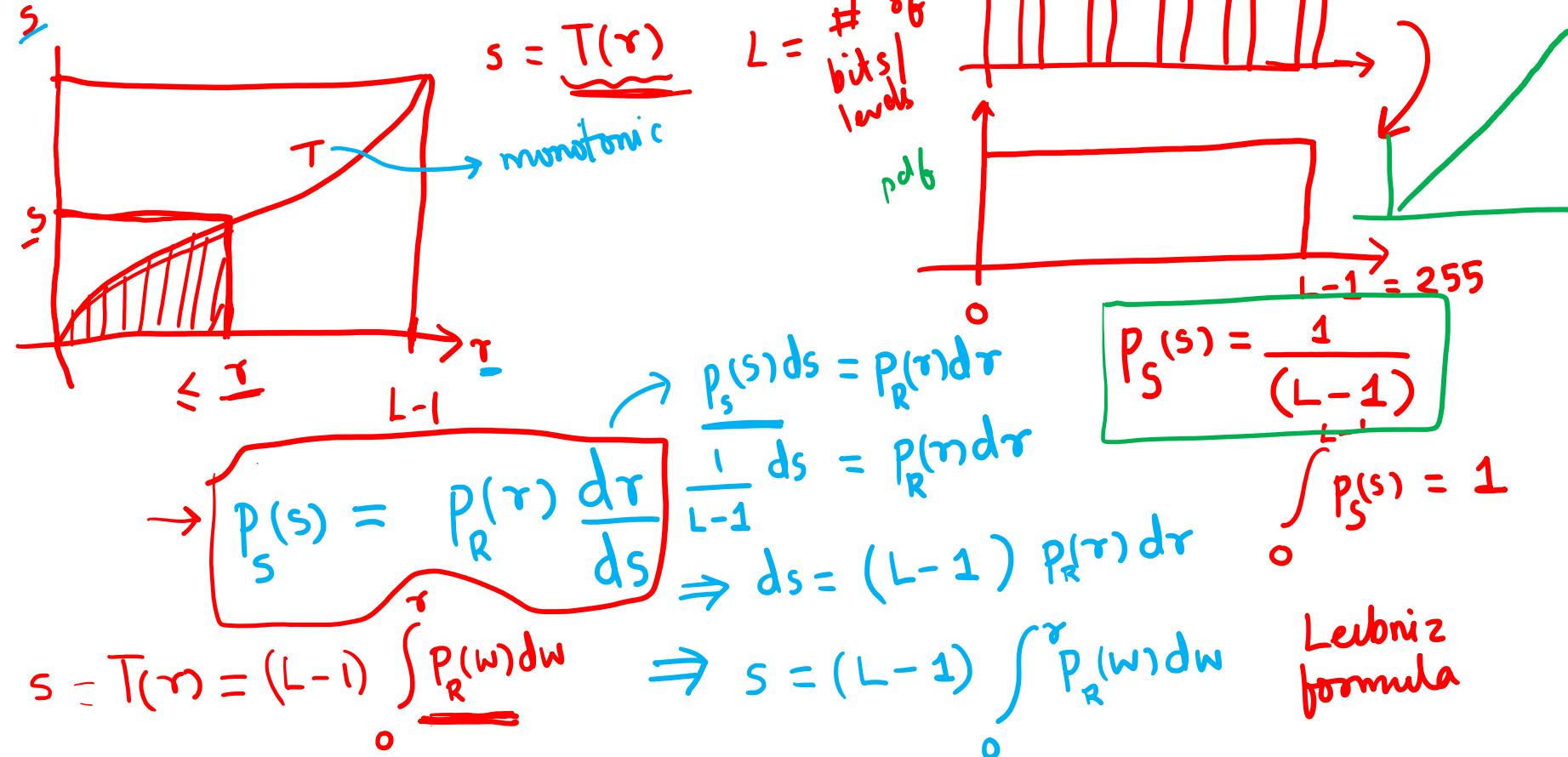
Histogram



Normalized  
 $P(R)$   
← probability distribution

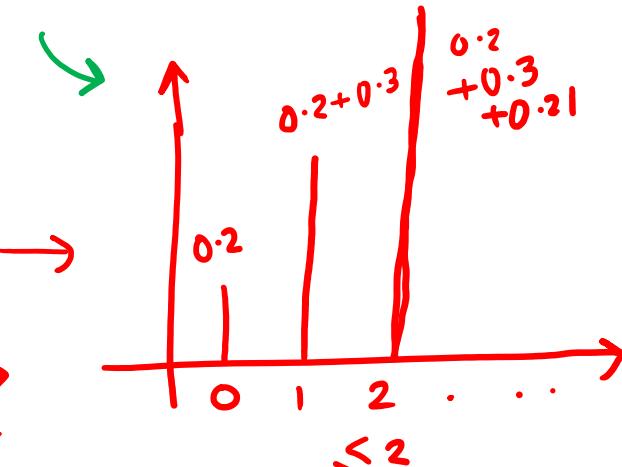


# Histogram Equalization



# Histogram Equalization

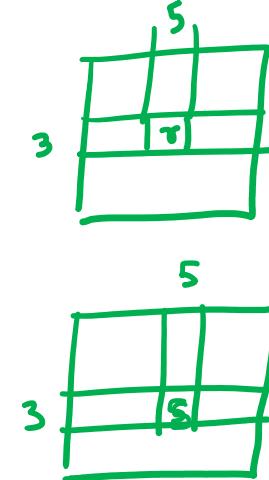
$$P_S(s) = \frac{1}{L-1}$$



$$s = T(\tau) = (L-1) \int_{w=0}^{\tau} P_S(w) dw$$

*digital equivalent*

$$s = \text{round} \left( (L-1) \sum_{w=0}^{\tau} P[w] \right)$$



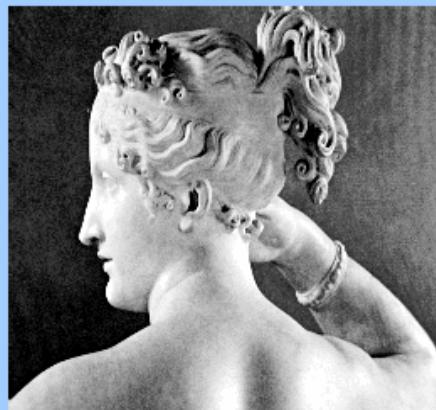
# Histogram Equalization



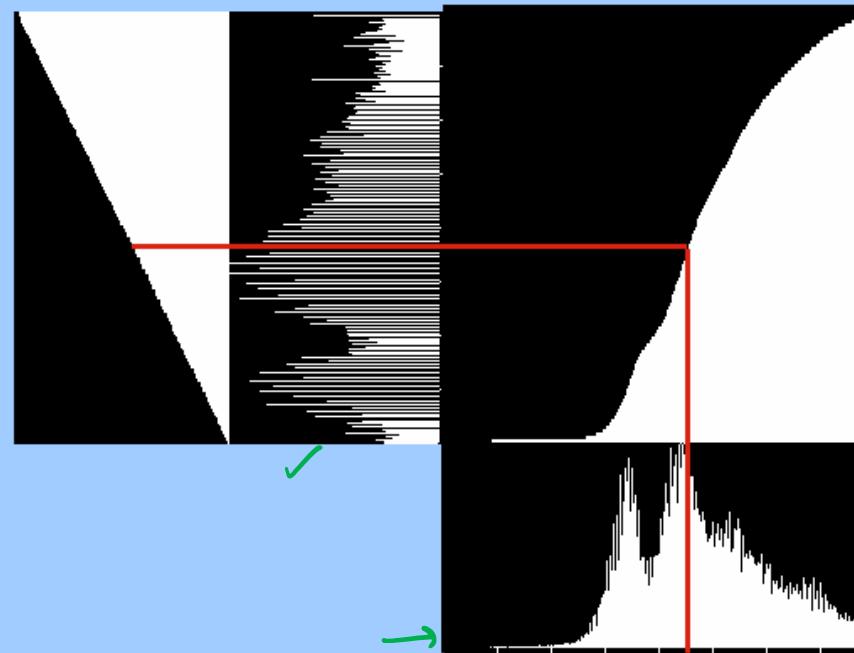
Contrast  
Stretching



Histogram  
Equalization



Equalized histogram



Histogram of original image

cdf

pdf

cdf



original image

# References

- ▶ GW Chapter – 3.3.1 to 3.3.3
- Transformations of Random Variables
  - <http://www.randomservices.org/random/dist/Transformations.html>
  - Section 1 of <http://www.cs.cmu.edu/~minx/transform.pdf>
  - Leibnitz Integration Rule :  
[https://en.wikipedia.org/wiki/Leibniz\\_integral\\_rule#Alternative\\_derivation](https://en.wikipedia.org/wiki/Leibniz_integral_rule#Alternative_derivation)
  - [Univariate transformation of a random variable](#)

# Scribe Group

20171172
20171205
20171208
2018101002
2018101003
2018101005

# Mini Quiz 1 Link

<https://forms.office.com/Pages/ResponsePage.aspx?id=vDsaA3zPK06W7IZ1VVQKHNFN1LYrWjxAktM68Sb-hiFUOEEdKVEIEOU8xTjNZTjNCUDFRTjhHQ09BNC4u>