

# Computer Programming

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**Session: Associative Arrays for Histogram Equalization**

# Quick Recap

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- We looked at the digital representation of images, and use of matrices to represent images
- We discussed
  - Histogram and Cumulative Distribution Function
  - Histogram equalization technique to improve contrast

# Overview

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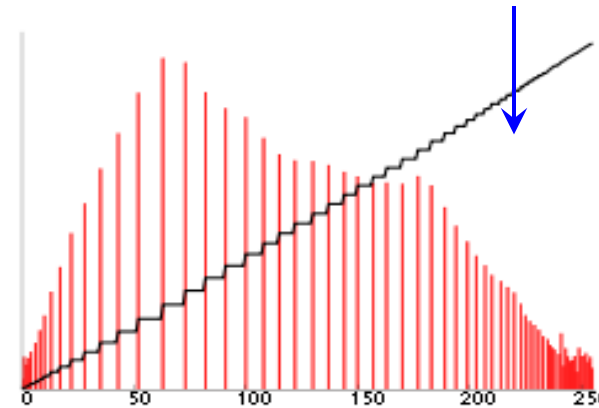
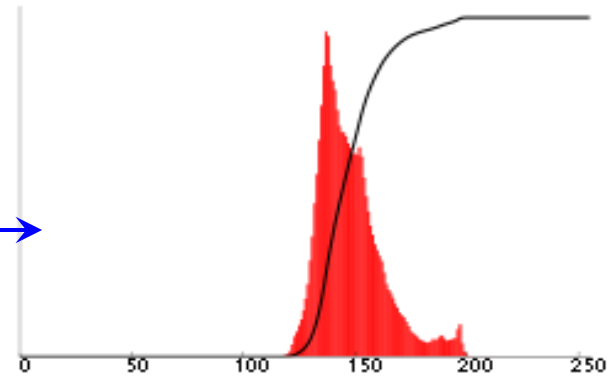


- In the context of histogram calculations, we will look at different computations to be performed
- We will discuss the important concept of “Associative Array” for use in our program, to be written for contrast enhancement

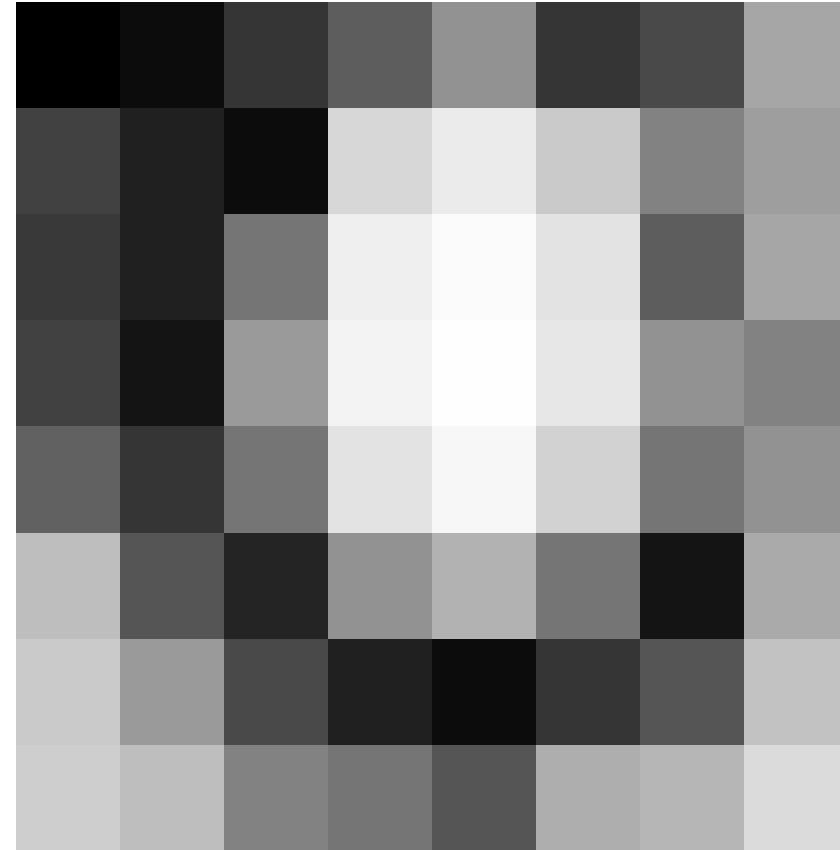
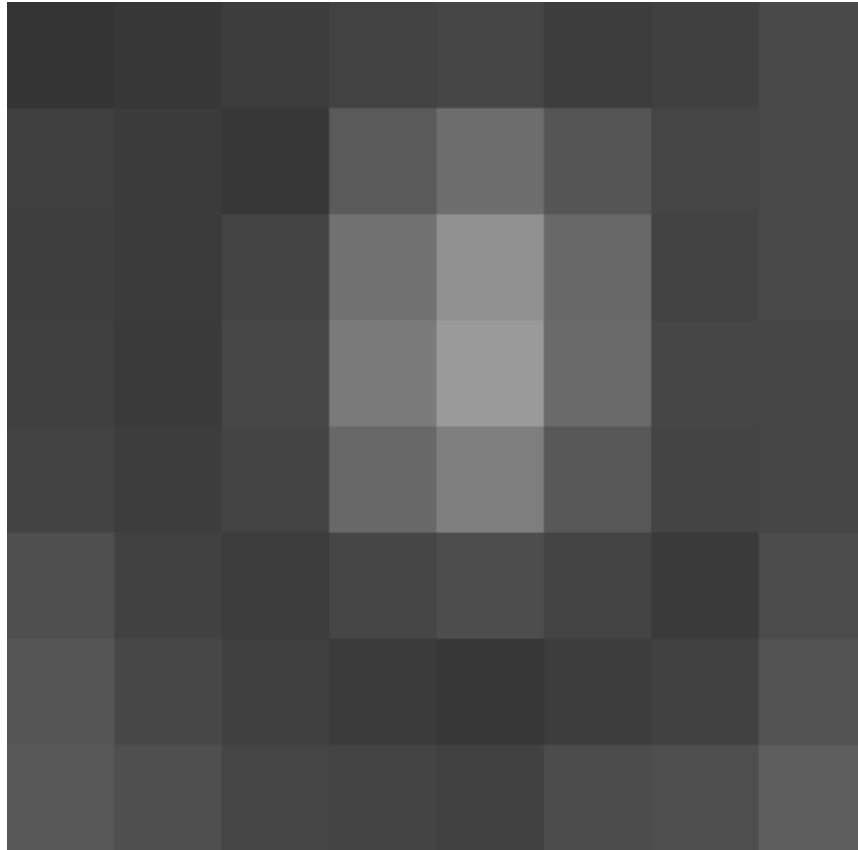
[Note: The histogram equalization technique described here, and the digital images used, are directly based on a wikipedia article:

[http://en.wikipedia.org/wiki/Histogram\\_equalization](http://en.wikipedia.org/wiki/Histogram_equalization)]

# Original picture, equalization, modified picture



# Original and contrast-enhanced pictures



# Pixel values for the image

|    |    |    |     |     |     |    |    |
|----|----|----|-----|-----|-----|----|----|
| 52 | 55 | 61 | 66  | 70  | 61  | 64 | 73 |
| 63 | 59 | 55 | 90  | 109 | 85  | 69 | 72 |
| 62 | 59 | 68 | 113 | 144 | 104 | 66 | 73 |
| 63 | 58 | 71 | 122 | 154 | 106 | 70 | 69 |
| 67 | 61 | 68 | 104 | 126 | 88  | 68 | 70 |
| 79 | 65 | 60 | 70  | 77  | 68  | 58 | 75 |
| 85 | 71 | 64 | 59  | 55  | 61  | 65 | 83 |
| 87 | 79 | 69 | 68  | 65  | 76  | 78 | 94 |

# Histogram values (shown for non-zero pixels)



| Val | n | Val | n | Val | n | Val | n | Val | n |
|-----|---|-----|---|-----|---|-----|---|-----|---|
| 52  | 1 | 64  | 2 | 72  | 1 | 85  | 2 | 113 | 1 |
| 55  | 3 | 65  | 3 | 73  | 2 | 87  | 1 | 122 | 1 |
| 58  | 2 | 66  | 2 | 75  | 1 | 88  | 1 | 126 | 1 |
| 59  | 3 | 67  | 1 | 76  | 1 | 90  | 1 | 144 | 1 |
| 60  | 1 | 68  | 5 | 77  | 1 | 94  | 1 | 154 | 1 |
| 61  | 4 | 69  | 3 | 78  | 1 | 104 | 2 |     |   |
| 62  | 1 | 70  | 4 | 79  | 2 | 106 | 1 |     |   |
| 63  | 2 | 71  | 2 | 83  | 1 | 109 | 1 |     |   |

# Histogram Calculations

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- Histogram for a particular pixel value, is the total number of pixels in the image which have the same value
- Best way to represent the histogram, is to use a one dimensional array of 256 elements
  - a pixel can only be between 0 to 255
- We need to accumulate the count of each pixel value found in the image, in the corresponding element of the histogram array



# Histogram Calculations

| Index | Initial Value | Pixel values in the image |    |    |     |     |     |    |    |
|-------|---------------|---------------------------|----|----|-----|-----|-----|----|----|
| 0     | 0             | 52                        | 55 | 61 | 66  | 70  | 61  | 64 | 73 |
| 1     | 0             | 63                        | 59 | 55 | 90  | 109 | 85  | 69 | 72 |
| .     | .             | 62                        | 59 | 68 | 113 | 144 | 104 | 66 | 73 |
| 68    | 0             | 63                        | 58 | 71 | 122 | 154 | 106 | 70 | 69 |
| 69    | 0             | 67                        | 61 | 68 | 104 | 126 | 88  | 68 | 70 |
| .     | .             | 79                        | 65 | 60 | 70  | 77  | 68  | 58 | 75 |
| 255   | 0             | 85                        | 71 | 64 | 59  | 55  | 61  | 65 | 83 |
|       |               | 87                        | 79 | 69 | 68  | 65  | 76  | 78 | 94 |

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| 1     | 0             | 63                        | 59 | 55 | 90  | 109 | 85  | 69 | 72 |
| .     | .             | 62                        | 59 | 68 | 113 | 144 | 104 | 66 | 73 |
| 68    | 5             | 63                        | 58 | 71 | 122 | 154 | 106 | 70 | 69 |
| 69    | 0             | 67                        | 61 | 68 | 104 | 126 | 88  | 68 | 70 |
| .     | .             | 79                        | 65 | 60 | 70  | 77  | 68  | 58 | 75 |
| 255   | 0             | 85                        | 71 | 64 | 59  | 55  | 61  | 65 | 83 |
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|       |               | 52                        | 55 | 61 | 66  | 70  | 61  | 64 | 73 |
| 0     | 0             | 63                        | 59 | 55 | 90  | 109 | 85  | 69 | 72 |
| 1     | 0             | 62                        | 59 | 68 | 113 | 144 | 104 | 66 | 73 |
| .     | .             | 63                        | 58 | 71 | 122 | 154 | 106 | 70 | 69 |
| 68    | 5             | 67                        | 61 | 68 | 104 | 126 | 88  | 68 | 70 |
| 69    | 3             | 79                        | 65 | 60 | 70  | 77  | 68  | 58 | 75 |
| .     | .             | 85                        | 71 | 64 | 59  | 55  | 61  | 65 | 83 |
| 255   | 0             | 87                        | 79 | 69 | 68  | 65  | 76  | 78 | 94 |

# Cost of Computing Histogram Values



- If this logic is used to compute histogram entries, what is the number of operations the computer has to perform?
- Assume array name `IM[8][8]` for image, and `H[256]` for histogram
  - For each pixel value  $p$ , we scan all  $8 \times 8$  ( $= 64$ ) pixel values in `IM`, and compare each with  $p$ .
  - We assign the total count of those pixel values which match, to the appropriate element of `H`
- We repeat this for each of the 256 possible values ( $256 \times 64$  comparisons)
- In a large image of size 500 by 300, there will be 1,50,000 pixels!

# A list of students



- A list is available containing roll numbers of students attending a MOOC, and the cities they belong to

|       |           |       |            |       |        |
|-------|-----------|-------|------------|-------|--------|
| 10001 | Mumbai    | 10009 | Tokyo      | 10017 | Paris  |
| 10002 | Delhi     | 10010 | Kathmandu  | 10018 | Delhi  |
| 10003 | Tokyo     | 10011 | Delhi      | 10019 | Mumbai |
| 10004 | Karachi   | 10012 | Dhaka      | 10020 | Dhaka  |
| 10005 | Delhi     | 10013 | Mumbai     | 10021 | Tokyo  |
| 10006 | Kathmandu | 10014 | Delhi      | 10022 | Delhi  |
| 10007 | Dhaka     | 10015 | Karachi    | ---   | ---    |
| 10008 | Mumabi    | 10016 | Washington | ---   | ---    |

# City Statistics

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- We can count the number of students from each city, and store these counts in an array (table) which looks like

| City       | Count |
|------------|-------|
| Delhi      | 4024  |
| Dhaka      | 1729  |
| Dubai      | 572   |
| Karachi    | 984   |
| Kathmandu  | 431   |
| Mumbai     | 5102  |
| Paris      | 301   |
| Tokyo      | 659   |
| Washington | 850   |

# Cost of Calculating City Statistics



- If we use our approach to do such counting
  - Take a city, say Mumbai; scan the complete list to count students who belong to that city, and put the count value against that city in our count table
  - Repeat this for each city
- If there are 50,000 students taking a MOOC, and they belong to 200 different cities
  - We need to look at each of 50,000 entries, 200 times
  - We need to perform a total of 1,00,00,000 comparisons
- A very time consuming exercise

**This is not how statisticians do their counting!**

# Associative Array



- It will be useful, if name of a city (which is a value in our list) , can be used as a key (as an index) to directly access the associated count, which is a value in our table
- An associative array is a set of key-value pairs.
  - The organization is such that the 'key' is the index of the array
- Assume that we have such an associative array for our example of city names and corresponding student population
  - If the key “Mumbai” is given, using that as an index of the array, we should directly get 5102
  - If the key “Kathmandu” is given, we should directly get 431
- We will like to use the value (name of the city) in our list of students, directly as a key for the table, to access the count for the city



# Associative Array ...

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- In C++, the arrays we know are indexed by 'keys' which are only integer numbers, in the range of 0 to  $N-1$  (and not strings like city names)
- For the Histogram array, the key range is 0 to 255
- Pixel values in the image array are also integer numbers, and are precisely in this range
- Thus, the pixel value itself can be used as a 'key' or index of the Histogram array
  - This array can be treated as an associative array

# Associative Array 'H'



- Suppose the histogram count for the image array IM, is stored in array H[256], counting process will require a single scan of the image matrix
- For each pixel value 'p' we look at, we know which element of the array H needs to be updated. It is the  $p^{\text{th}}$  element!
  - “Association” between the pixel value p, and the index of array H
- The count for a pixel in the image array element IM[i][j], which has value p, can simply be incremented by the assignment:
$$p = \text{IM}[i][j]; \quad H[p] = H[p] + 1;$$
or more simply  $H[\text{IM}[i][j]] = H[\text{IM}[i][j]] + 1$
- Multiple scans of the image array, are not required.

# Summary

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- In the context of contrast enhancement of digital images, we studied the computation of the histogram
- We discussed the concept of an “associative array”, and found it to be a useful way to do efficient calculations