# Image Processing I 

15-463: Rendering and Image
Processing
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...with most slides shamelessly stolen from Steve Seitz and Gonzalez \& Woods

## Today

Image Formation
Range Transformations

- Point Processing


## Programming Assignment \#1 OUT

Reading for this week:

- Gonzalez \& Woods, Ch. 3
- Forsyth \& Ponce, Ch. 7

Image Formation

a b de
FIGURE 2.15 An example of the digital image acquisition process. (a) Energy ("illumination") source. (b) An element of a scene. (c) Imaging system. (d) Projection of the scene onto the image plane. (e) Digitized image.
$f(x, y)=$ reflectance $(x, y)$ *illumination $(x, y)$
Reflectance in [0,1], illumination in [0,inf]

Sampling and Quantization

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FIGURE 2.16 Generating a digital image. (a) Continuous image. (b) A scan line from $A$ to $B$ in the continuous image. used to illustrate the concepts of sampling and quantization. (c) Sampling and quantization. (d) Digital scan line.

## Sampling and Quantization


a b
FIGURE 2.17 (a) Continuos image projected onto a sensor array. (b) Result of image sampling and quantization.

## What is an image?

We can think of an image as a function, $f$, from $\mathrm{R}^{2}$ to R:

- $f(x, y)$ gives the intensity at position ( $x, y$ )
- Realistically, we expect the image only to be defined over a rectangle, with a finite range:
$-f:[a, b] \times[c, d] \rightarrow[0,1]$

A color image is just three functions pasted together. We can write this as a "vector-valued" function:

$$
f(x, y)=\left[\begin{array}{l}
r(x, y) \\
g(x, y) \\
b(x, y)
\end{array}\right]
$$

## Images as functions



## What is a digital image?

We usually operate on digital (discrete) images:

- Sample the 2D space on a regular grid
- Quantize each sample (round to nearest integer)

If our samples are $\Delta$ apart, we can write this as:
$f i, j]=$ Quantize $\{f(i \Delta, j \Delta)\}$
The image can now be represented as a matrix of inteaer values

| $\dot{\mathbf{z}}$ | 62 | 79 | 23 | 119 | 120 | 105 | 4 | 0 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 10 | 10 | 9 | 62 | 12 | 78 | 34 | 0 |
|  | 10 | 58 | 197 | 46 | 46 | 0 | 0 | 48 |
|  | 176 | 135 | 5 | 188 | 191 | 68 | 0 | 49 |
|  | 2 | 1 | 1 | 29 | 26 | 37 | 0 | 77 |
|  | 0 | 89 | 144 | 147 | 187 | 102 | 62 | 208 |
|  | 255 | 252 | 0 | 166 | 123 | 62 | 0 | 31 |
|  | 166 | 63 | 127 | 17 | 1 | 0 | 99 | 30 |

## Image processing

An image processing operation typically defines a new image $g$ in terms of an existing image $f$. We can transform either the range of $f$.

$$
g(x, y)=t(f(x, y))
$$

Or the domain of $f$ :

$$
g(x, y)=f\left(t_{x}(x, y), t_{y}(x, y)\right)
$$

What kinds of operations can each perform?

## Point Processing

The simplest kind of range transformations are these independent of position $x, y$ :

$$
g=t(f)
$$

This is called point processing.

What can they do?
What's the form of $t$ ?

Important: every pixel for himself - spatial information completely lost!

## Basic Point Processing

FIGURE 3.3 Some
basic gray-level
transformation
functions used for
enhancement.


Negative


## Log

a b
FIGURE 3.5
(a) Fourier
spectrum.
(b) Result
applying the log
transformation
transform
given in
Eq. (3.2-2) with
$c=1$.


Power-law transformations

of the equation
of the equatio
$s=c r^{\gamma}$ for
$s=c r^{\gamma}$ for
various values of
various values o
$\gamma(c=1$ in all
cases).

## Gamma Correction



Gamma Measuring Applet:
http://www.cs.berkeley.edu/~efros/java/gamma/gamma.html

## Image Enhancement

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.)


## Contrast Streching



## Image Histograms


a b
FIGURE 3.15 Four basic image types. dark, light, low con trast, high contrast, and their corof Biological Sciences, Australian National University, Canberra, Australia.)

## Cumulative Histograms

FIGURE 3.18
Transformation
functions (1)
obtained from the
histograms of the
images in
Fig.3.17(a), using
Eq. (3.3-8).


Histogram Equalization


## Histogram Matching

a b
FIGURE 3.19
(a) Graphical
(a) Graphical
interpretation of
mapping from $r_{k}$
mapping from $r_{k}$
(b) Mapping of $z_{q}$
(b) Mapping of $z_{q}$
to its
corresponding
value $v_{q}$ via $G(z)$.
(c) Inverse
mapping from $s_{k}$
o its
corresponding
value of $z_{k}$


## Match-histogram code

```
Match-histogram (im1,im2)
    im1-cdf = Make-cdf (im1)
    im2-cdf = Make-cdf(im2)
    inv-im2-cdf = Make-inverse-lookup-table(im2-cdf)
    Loop for each pixel do
        im1[pixel] =
            Lookup(inv-im2-cdf
            Lookup(im1-cdf,im1[pixel]))
```


## Neighborhood Processing (filtering)

Q: What happens if I reshuffle all pixels within the image?


A: It's histogram won't change. No point processing will be affected...

Need spatial information to capture this.

## Programming Assignment \#1

Easy stuff to get you started with Matlab

- James will hold tutorial this week

Distance Functions

- SSD
- Normalized Correlation

Bells and Whistles

- Point Processing (color?)
- Neighborhood Processing
- Using your data (3 copies!)
- Using your data (other images)


