### **Digital Image Processing**

Lecture #6 Ming-Sui (Amy) Lee

## Announcement

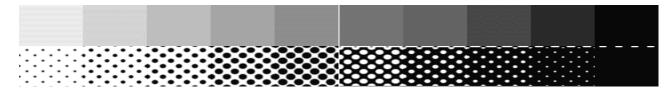
- Class Information
  - The following schedule

03/23	Lecture 5	05/11	proposal
03/30	Lecture 6	05/18	Lecture 9
04/06	Lecture 7	05/25	Lecture 10
04/13	RealSense	06/01	Lecture 11
04/20	midterm	06/08	Demo
04/27	RealSense	06/15	Demo
05/04	Lecture 8	06/22	Final Package Due



#### Goal

 Render the illusion of a continuous-tone image based on two-tone (half-tone) display



- Applications
  - Computer hardcopies
    - Laser printers/dot-matrix printers/color printers
    - Fax machine
- Implementation
  - Thresholding at 1/2?





**Gray-level image** 

Half-toned images

#### Color Printer

Continuous Image





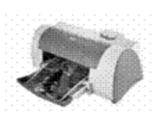
**Binary Image** 











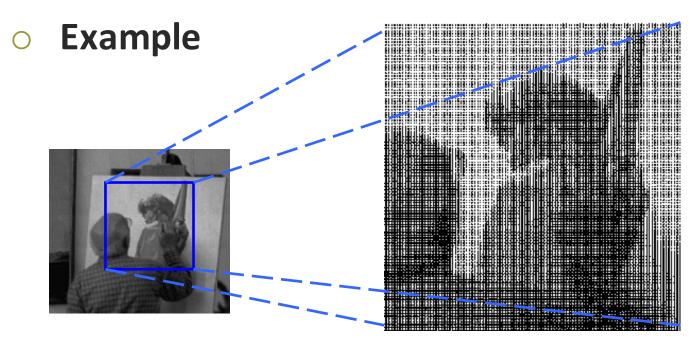


- Basic idea
  - Spatial modulation
  - Three approaches
    - Patterning
    - Dithering
    - Error Diffusion

**Patterning** p If p=4→ 16 binary pixels → 17 levels (0~16) → 256 gray levels → Quantization 1 Dot pattern 1 Gray-level pixel Rylander's recursive

- Patterning
  - Four steps
    - Read in the given grey-level image
    - Quantization
    - Design the patterning table
    - Map each pixel to its corresponding pattern
  - Simplest way
  - Generates image with higher spatial resolution than the source image

Patterning



Original gray-level image

Half-toned image: patterning

- Dithering
  - Create an image with the same number of dots as the number of pixels in the source image
  - Idea

$$N(j,k) \rightarrow Why??$$

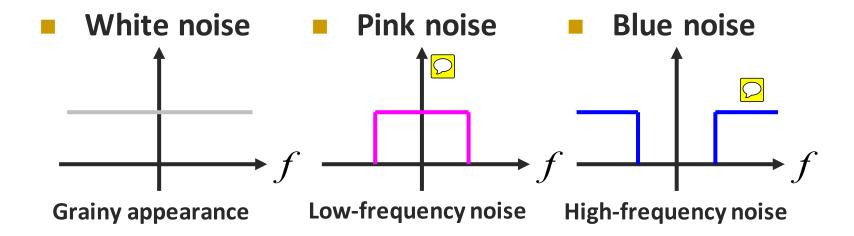
$$F(j,k) \longrightarrow H(j,k)$$
Threshold  $G(j,k)$ 

- Dithering
  - Owner of the control of the contr
    - Under fixed thresholding → taking MSB
      - E.g. before and after adding noise

≥128 <127

- To break the monotonicity of accumulated error in the area of constant (nearly constant) gray level
- White noise, pink noise, blue noise and green noise

- Dithering
  - Noise Type
    - Power spectral density



- Robert Ulichney, "Digital Halftoning"
  - http://www.hpl.hp.com/people/u/

#### Dithering

- Adaptive thresholding
  - Generate a threshold matrix according to a dither matrix
  - Whenever the pixel value of the image is greater than the value in the threshold matrix, the pixel is turned on

#### Notes

- No randomness
- Region-to-region mapping
- Recursive definition allowed

- Dithering
  - Dither matrix

$$I_2(i,j) = \begin{bmatrix} 1 & 2 \\ 3 & 0 \end{bmatrix}; \quad I_2(i,j) = \begin{bmatrix} 3 & 1 \\ 0 & 2 \end{bmatrix}$$

- 0 → lowest threshold
- 3 → highest threshold

#### Dithering

- The general form of the NxN dither matrix
  - = 2x2  $\rightarrow$  4x4  $\rightarrow$  8x8  $\rightarrow$  16x16...

$$I_{2n}(i,j) = \begin{bmatrix} 4I_n(i,j) + 1 & 4I_n(i,j) + 2 \\ 4I_n(i,j) + 3 & 4I_n(i,j) + 0 \end{bmatrix}$$

Eg. What is 
$$I_4(i,j)$$
 if  $I_2(i,j) = \begin{bmatrix} 1 & 2 \\ 3 & 0 \end{bmatrix}$ ?

- Dithering
  - Determine the threshold matrix

$$T(i,j) = 255 \cdot \frac{I(i,j) + 0.5}{N^2}$$

Eg. N=4

Eg. N=4
$$I_4(i,j) = \begin{bmatrix} 5 & 9 & 6 & 10 \\ 13 & 1 & 14 & 2 \\ 7 & 11 & 4 & 8 \\ 15 & 3 & 12 & 0 \end{bmatrix}, \quad T_4(i,j) = ?$$

#### Dithering

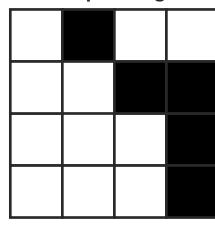
Input image

12	51	34	121
78	254	10	97
45	113	110	16
90	200	206	34

Repeated threshold matrix

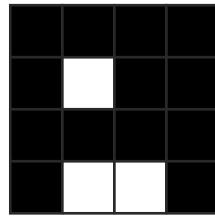
0	60	0	60
45	110	45	110
0	60	0	60
45	110	45	110

**Output image** 



#### Another repeated threshold matrix

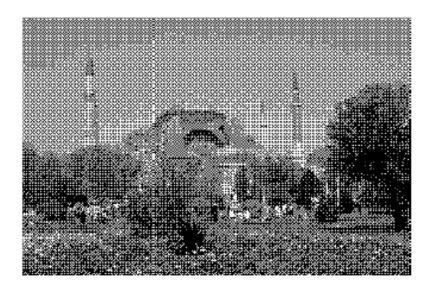
128	128	128	128
128	128	128	128
128	128	128	128
128	128	128	128



#### Experimental results



**Original Image** 



**Dithering** 

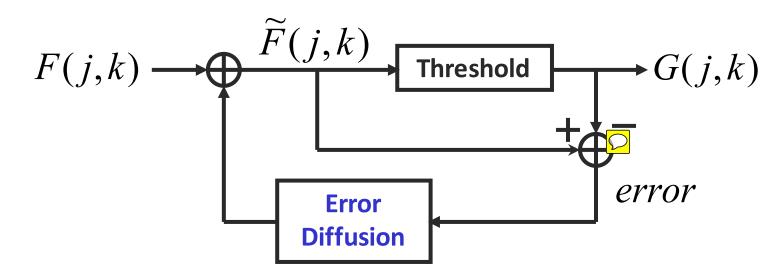
#### Experimental results



**Original Image** 

**Dithering** 

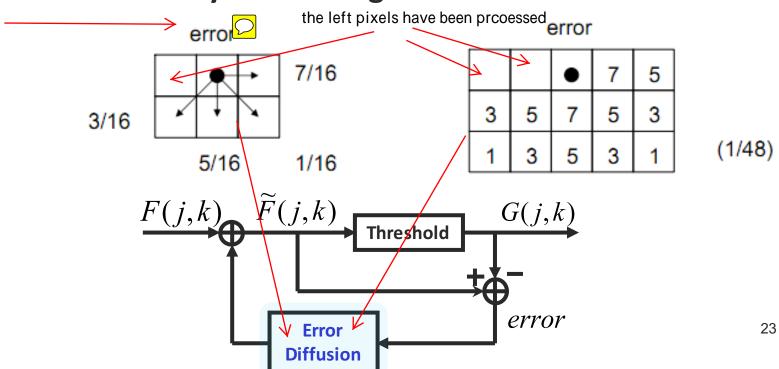
- Error diffusion
  - 1975 Floyd & Steinberg
    - A practical algorithm to implement blue noise dithering
    - Framework



- Error diffusion
  - Normalize F(j,k) to lie between [0,1]
  - Set threshold=0.5
  - Output image: 0 or 1

$$F(j,k) \xrightarrow{\widetilde{F}(j,k)} \text{Threshold} \qquad if \quad \widetilde{F}(j,k) \geq 0.5 \quad \Rightarrow \quad G(j,k) = 1$$
 
$$if \quad \widetilde{F}(j,k) < 0.5 \quad \Rightarrow \quad G(j,k) = 0$$
 
$$Define \quad E(j,k) = \widetilde{F}(j,k) - G(j,k)$$

- Error diffusion
  - Error diffusion filter masks
    - 1975 Floyd Steinberg: 1976 Jarvis et al:



- **Error diffusion** 
  - Error diffusion + serpentine scanning

$$\frac{1}{16} \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 7 \\ 3 & 5 & 1 \end{pmatrix}$$

Left to Right

$$\frac{1}{16} \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 7 \\ 3 & 5 & 1 \end{pmatrix} \qquad \frac{1}{16} \begin{pmatrix} 0 & 0 & 0 \\ 7 & 0 & 0 \\ 1 & 5 & 3 \end{pmatrix}$$

Right to Left

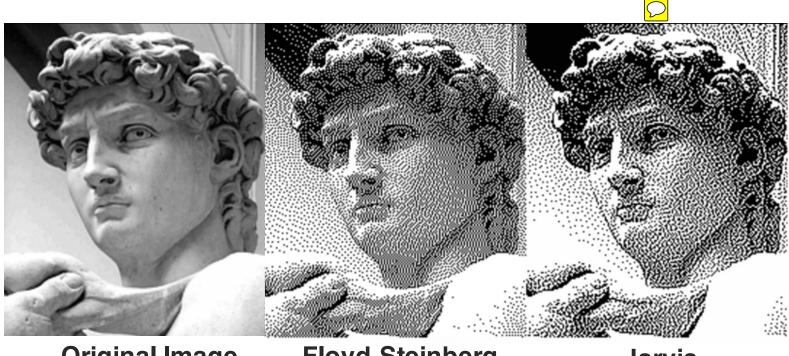
#### Experimental results



**Original Image** 

**Error Diffusion** 

**Experimental results** 



**Original Image** 

Floyd-Steinberg

**Jarvis** 

- Multi-scale Error diffusion
  - Several issues
    - Region-to-region mapping
      - Multi-resolution
    - Time series/causal error diffusion process
      - Easy to implement
      - Causalit pappears to be artificial in images
      - Is non-causal error diffusion possible?
    - Quality metrics of half-toned images

#### **Multi-scale Error diffusion**

"A multiscale error diffusion technique for digital halftoning" Ioannis Katsavounidis and C. -C. Jay Kuo

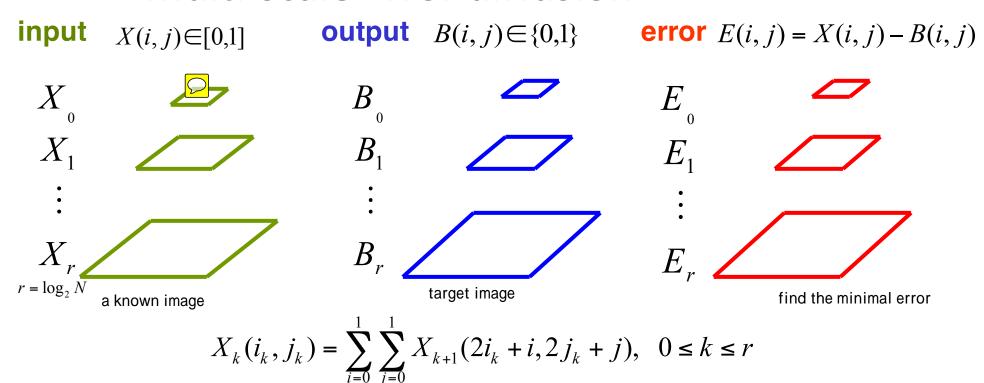
#### Problem set-up

- Input image  $\rightarrow X(i,j) \in [0,1]$ Output image  $\rightarrow B(i,j) \in \{0,1\}$
- **Error image**  $\rightarrow$  E(i, j) = X(i, j) B(i, j)
- Intermediate stage →

$$X_{k}(i_{k}, j_{k}), \quad 0 \le k \le r, \quad r = \log_{2} N$$

$$X_{k}(i_{k}, j_{k}) = \sum_{i=0}^{1} \sum_{j=0}^{1} X_{k+1}(2i_{k} + i, 2j_{k} + j)$$

#### Multi-scale Error diffusion



Goal: minimize the error pyramid in a certain way!

 $E_k(i_k, j_k) = X_k(i_k, j_k) - B_k(i_k, j_k), \quad 0 \le k \le r$ 

#### **Multi-scale Error diffusion**

- //Step 1// Initialization
  - Set the entire output image pyramid to "0"

//Step 2// Dot assignment



- Find the largest error from top to bottom level
- 1 parent node distributes its dots (integer numbers) to 4 children
- //Step 3// Error diffusion proces

error diffuse 
$$\frac{1}{12} \begin{pmatrix} 1 & 2 & 1 \\ 2 & -12 & 2 \\ 1 & 2 & 1 \end{pmatrix} \qquad \frac{1}{8} \begin{pmatrix} 0 & 0 & 0 \\ 2 & -8 & 2 \\ 1 & 2 & 1 \end{pmatrix} \qquad \frac{1}{5} \begin{pmatrix} 0 & 0 & 0 \\ 0 & -5 & 2 \\ 0 & 2 & 1 \end{pmatrix}$$
center
side
corner

$$\frac{1}{8} \begin{pmatrix} 0 & 0 & 0 \\ 2 & -8 & 2 \\ 1 & 2 & 1 \end{pmatrix}$$
side

$$\frac{1}{5} \begin{pmatrix} 0 & 0 & 0 \\ 0 & -5 & 2 \\ 0 & 2 & 1 \end{pmatrix}$$

- Multi-scale Error diffusion
  - Quality management
    - MSE vector

$$MSEV = \begin{pmatrix} MSE_0 \\ MSE_1 \\ \vdots \\ MSE_r \end{pmatrix} \qquad MSE_k = \frac{1}{N^2} \sum_{i=0}^{2^k - 1} \sum_{j=0}^{2^k - 1} E_k^2(i, j)$$

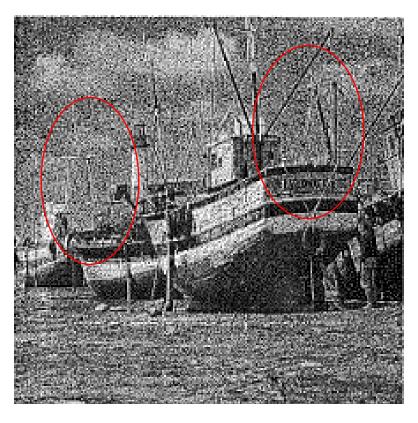
- Notes
  - Preserve contrast of the original image
  - Does not over-smooth the image

### Experimental results

better detail and contrast



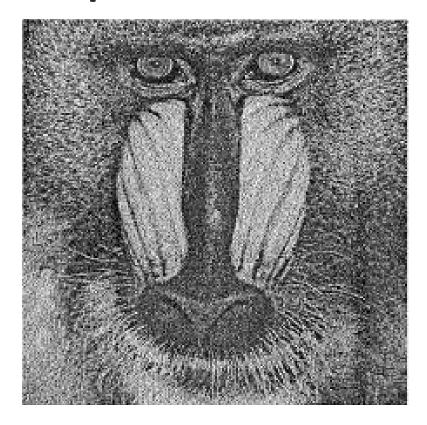
**Error Diffusion** 

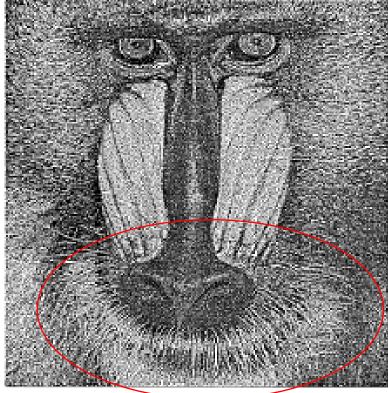


**Multi-Scale Error Diffusion** 

Experimental results







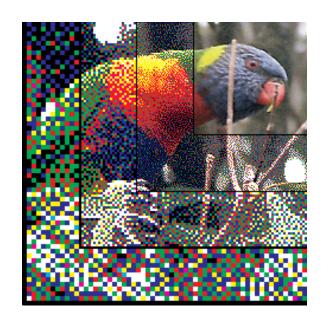
**Error Diffusion**worse contrast and detail

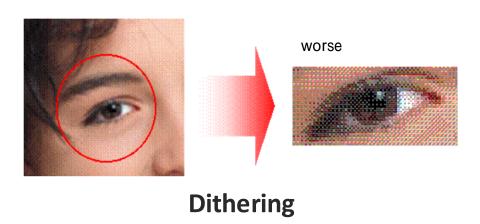
**Multi-Scale Error Diffusion** 

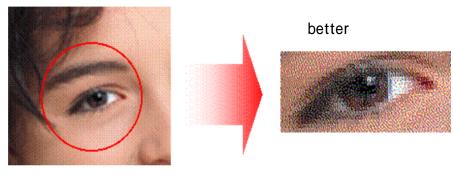
**Color image** R,G,B is half-tone individually and then combined again.



#### Examples







**Error Diffusion** 

- Application
  - Visual cryptography

"visual cryptography based on void-and-cluster halftoning technique" E. Myodo, S. Sakazawa and Y. Takishima

