## **Digital Halftoning**

- Many image rendering technologies only have binary output. For example, printers can either "fire a dot" or not.
- Halftoning is a method for creating the illusion of continuous tone output with a binary device.
- Effective digital halftoning can substantially improve the quality of rendered images at minimal cost.

## **Thresholding**

- Assume that the image falls in the range of 0 to 255.
- Apply a space varying threshold, T(i, j).

$$b(i,j) = \begin{cases} 255 & \text{if } X(i,j) > T(i,j) \\ 0 & \text{otherwise} \end{cases}.$$

- What is X(i, j)?
- Lightness
  - Larger  $\Rightarrow$  lighter
  - Used for display
- Absorptance
  - Larger  $\Rightarrow$  darker
  - Used for printing
- X(i, j) will generally be in units of absorptance.

#### **Constant Threshold**

- Assume that the image falls in the range of 0 to 255.
- $255 \Rightarrow Black \text{ and } 0 \Rightarrow White$
- The minimum squared error quantizer is a simple threshold

$$b(i,j) = \begin{cases} 255 & \text{if } X(i,j) > T \\ 0 & \text{otherwise} \end{cases}.$$

where T = 127.

• This produces a poor quality rendering of a continuous tone image.

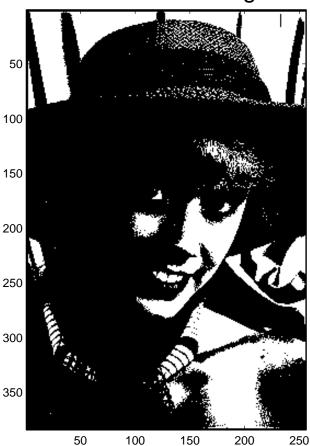
## **The Minimum Squared Error Solution**

- Threshold each pixel
  - Pixel> 127 Fire ink
  - Pixel≤ 127 do nothing

#### Original Image

## 50 100 150 200 250 300 350 50 100 150 200 250

#### Thresholded Image

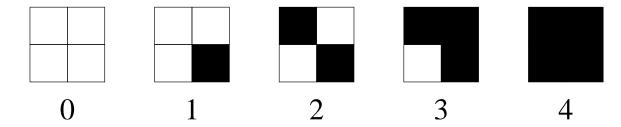


#### **Ordered Dither**

- For a constant gray level patch, turn the pixel "on" in a specified order.
- This creates the perception of continuous variations of gray.
- $\bullet$  An  $N \times N$  index matrix specifies what order to use.

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

• Pixels are turned on in the following order.



# Implementation of Ordered Dither via Thresholding

• The index matrix can be converted to a "threshold matrix" or "screen" using the following operation.

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

ullet The N imes N matrix can then be "tiled" over the image using periodic replication.

$$T(i \bmod N, j \bmod N)$$

• The ordered dither algorithm is then applied via thresholding.

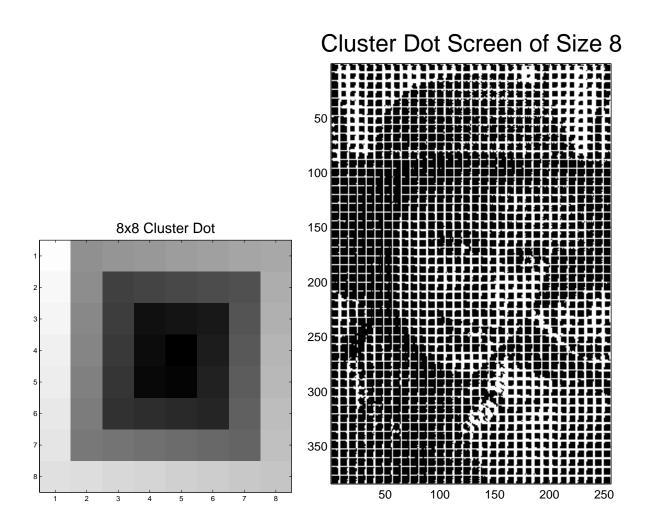
$$b(i,j) = \begin{cases} 255 & \text{if } X(i,j) > T(i \, \text{mod} N, j \, \text{mod} N) \\ 0 & \text{otherwise} \end{cases}$$

#### **Clustered Dot Screens**

- Definition: If the consecutive thresholds are located in spatial proximity, then this is called a "clustered dot screen.
- Example for  $8 \times 8$  matrix:

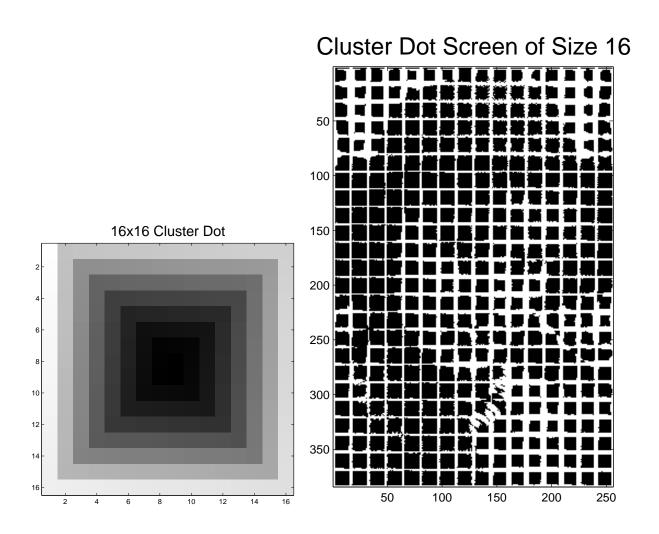
62	57	48	36	37	49	58	63
56	47	35	21	22	38	50	59
46	34	20	10	11	23	39	51
33	19	9	3	0	4	12	24
32	18	8	2	1	5	13	25
45	31	17	7	6	14	26	40
55	44	30	16	15	27	41	52
61	54	43	29	28	42	53	60

## **Example:** $8 \times 8$ **Clustered Dot Screening**



• Only supports 65 gray levels.

## **Example:** $16 \times 16$ **Clustered Dot Screening**



• Support a full 257 gray levels, but has half the resolution.

## **Properties of Clustered Dot Screens**

- Requires a trade-off between number of gray levels and resolution.
- Relatively visible texture
- Relatively poor detail rendition
- Uniform texture across entire gray scale.
- Robust performance with non-ideal output devices
  - Non-additive spot overlap
  - Spot-to-spot variability
  - Noise

## **Dispersed Dot Screens**

• Bayer's optimum index Matrix (1973) can be defined recursively.

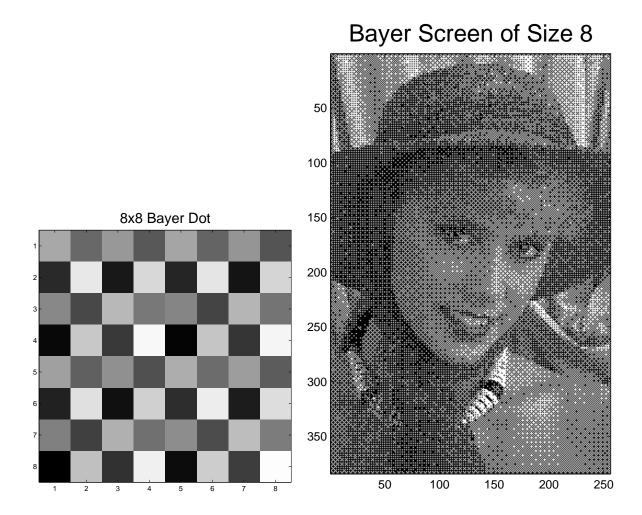
$$I_{2}(i,j) = \begin{bmatrix} 1 & 2 \\ 3 & 0 \end{bmatrix}$$

$$I_{2n} = \begin{bmatrix} 4 * I_n + 1 & 4 * I_n + 2 \\ 4 * I_n + 3 & 4 * I_n \end{bmatrix}$$

Examples

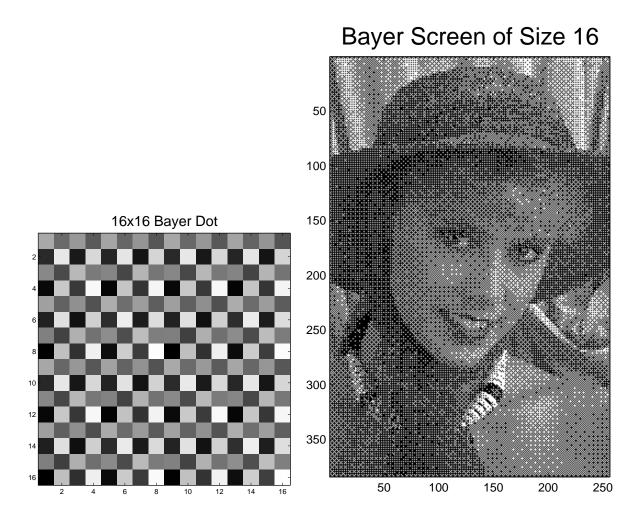
- Yields finer amplitude quantization over larger area.
- Retains good detail rendition within smaller area.

## **Example:** $8 \times 8$ **Bayer Dot Screening**



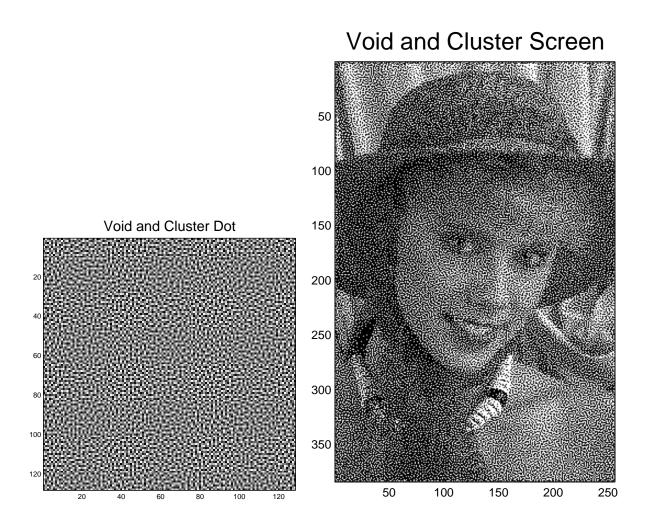
• Again, only 65 gray levels.

## **Example:** $16 \times 16$ **Bayer Dot Screening**



- Doesn't look much different than the  $8 \times 8$  case.
- No trade-off between resolution and number of gray levels.

# Example: $128 \times 128$ Void and Cluster Screen (1989)



• Substantially improved quality over Bayer screen.

## **Properties of Dispersed Dot Screens**

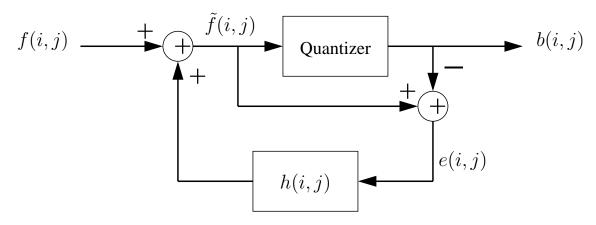
- Eliminate the trade-off between number of gray levels and resolution.
- Within any region containing K dots, the K thresholds should be distributed as uniformly as possible.
- Textures used to represent individual gray levels have low visibility.
- Improved detail rendition.
- Transitions between textures corresponding to different gray levels may be more visible.
- Not robust to non-ideal output devices
  - Requires stable formation of isolated single dots.

#### **Error Diffusion**

#### • Error Diffusion

- Quantizes each pixel using a neighborhood operation, rather than a simple pointwise operation.
- Moves through image in raster order, quantizing the result, and "pushing" the error forward.
- Can produce better quality images than is possible with screens.

## **Filter View of Error Diffusion**



#### • Equations are

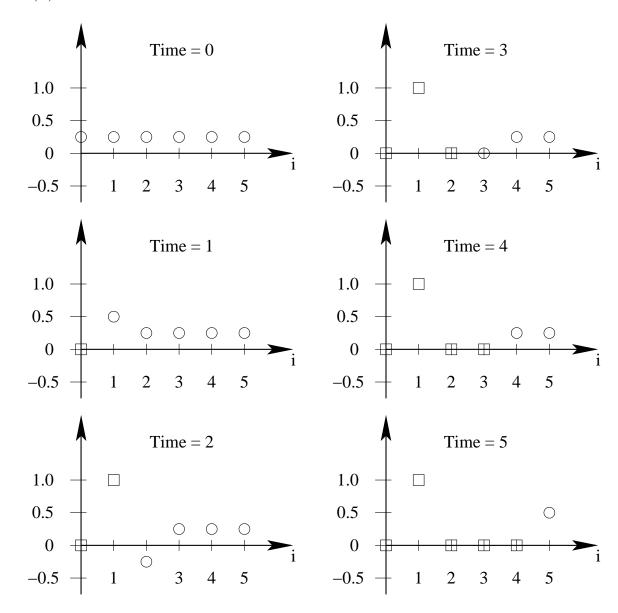
$$\begin{split} b(i,j) &= \begin{cases} 255 & \text{if } \tilde{f}(i,j) > T \\ 0 & \text{otherwise} \end{cases} \\ e(i,j) &= \tilde{f}(i,j) - b(i,j) \\ \tilde{f}(i,j) &= f(i,j) + \sum_{k,l \in S} h(k,l) e(i-k,j-l) \end{split}$$

#### Parameters

- Threshold is typically T = 127.
- -h(k, l) are typically chosen to be positive and sum to 1

## 1-D Error Diffusion Example

- $\tilde{f}(i) \Rightarrow$  circles
- $b(i) \Rightarrow boxes$



#### **Two Views of Error Diffusion**

- Two mathematically equivalent views of error diffusion
  - Pulling errors forward
  - Pushing errors ahead
- Pulling errors forward
  - More similar to common view of IIR filter
  - Has advantages for analysis
- Pushing errors ahead
  - Original view of error diffusion
  - Can be more easily extended to important cases when weights area time/space varying

## **ED: Pulling Errors Forward**

- 1. For each pixel in the image (in raster order)
  - (a) Pull error forward

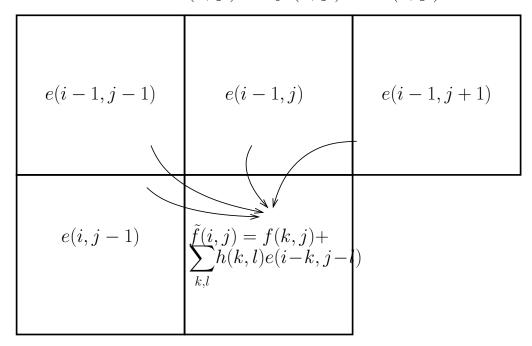
$$\tilde{f}(i,j) = f(i,j) + \sum_{k,l \in S} h(k,l)e(i-k,j-l)$$

(b) Compute binary output

$$b(i,j) = \begin{cases} 255 & \text{if } \tilde{f}(i,j) > T \\ 0 & \text{otherwise} \end{cases}$$

(c) Compute pixel's error

$$e(i,j) = \tilde{f}(i,j) - b(i,j)$$



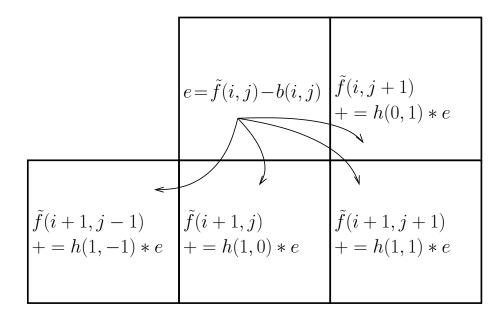
2. Display binary image b(i, j)

## **ED: Pushing Errors Ahead**

- 1. Initialize  $\tilde{f}(i,j) \leftarrow f(i,j)$
- 2. For each pixel in the image (in raster order)
  - (a) Compute

$$b(i,j) \ = \ \left\{ \begin{array}{ll} 255 \ \ \text{if} \ \tilde{f}(i,j) > T \\ 0 \ \ \ \text{otherwise} \end{array} \right.$$

(b) Diffuse error forward using the following scheme



3. Display binary image b(i, j)

## **Commonly Used Error Diffusion Weights**

• Floyd and Steinberg (1976)

		7/16
3/16	5/16	1/16

• Jarvis, Judice, and Ninke (1976)

			7/48	5/48
3/48	5/48	7/48	5/48	3/48
1/48	3/48	5/48	3/48	1/48

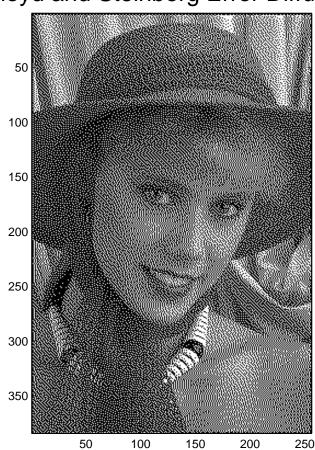
## Floyd Steinberg Error Diffusion (1976)

• Process pixels in neighborhoods by "diffusing error" and quantizing.

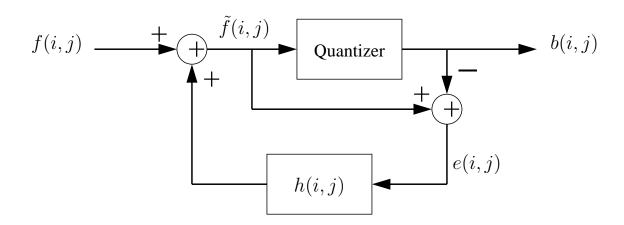
Original Image

50 100 200 250 300 350 50 100 150 200 250

Floyd and Steinberg Error Diffusion



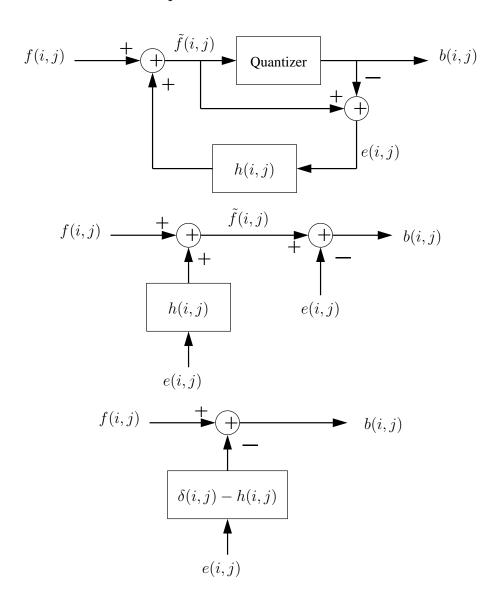
## Quantization Error Modeling for Error Diffusion



- Quantization error is commonly assumed to be:
  - Uniformly distributed on [-0.5, 0.5]
  - Uncorrelated in space
  - Independent of signal  $\tilde{f}(i,j)$
  - $-E\left[e(i,j)\right] = 0$
  - $-E[e(i,j)e(i+k,j+l)] = \frac{\delta(k,l)}{12}$

## **Modified Error Diffusion Block Diagram**

• The error diffusion block diagram can be rearranged to facilitate error analysis



## **Error Diffusion Spectral Analysis**

• So we see that

$$b(i,j) = f(i,j) - (\delta(i,j) - h(i,j)) * e(i,j)$$

rewriting ...

$$f(i,j) - b(i,j) = \underbrace{(\delta(i,j) - h(i,j))}_{\mbox{high pass filter}} * \underbrace{e(i,j)}_{\mbox{quantization}} *$$

- Display error is f(i, j) b(i, j)
- Quantization error is e(i, j)
- Display error is a high pass version of quantization error
- Human visual system is less sensitive to high spatial frequencies

## **Error Image in Floyd Steinberg Error Diffusion**

• Process pixels in neighborhoods by "diffusing error" and quantizing.



## **Correlation of Quantization Error and Image**

- Quantizer error spectrum is unknown
- Quantizer error model

$$\begin{split} E(\mu,\nu) &= \rho F(\mu,\nu) + R(\mu,\nu) \\ &= \rho(\mathrm{Image}) + (\mathrm{Residual}) \end{split}$$

–  $\rho$  represents correlation between quantizer error and image

Weight	$\rho$
1-D	0.0
Floyd and Steinberg	0.55
Jarvis, Judice, and Ninke	0.8

• Using this model, we have

$$\begin{split} B(\mu,\nu) \ = \ F(\mu,\nu) - (1-H(\mu,\nu)) \, E(\mu,\nu) \\ \\ = \ \left[1-\rho \left(1-H(\mu,\nu)\right)\right] F(\mu,\nu) + \text{noise} \end{split}$$

• This is unsharp masking

## **Additional Topics**

- Pattern Printing
- Dot Profiles
- Halftone quality metrics
  - Radially averaged power spectrum (RAPS)
  - Weighted least squares with HVS constrast sensitivity function
  - Blue noise dot patterns
- Error diffusion
  - Unsharp masking effects
  - Serpentine scan patterns
  - Threshold dithering
  - TDED
- Least squared halftoning
- Printing and display technologies
  - Electrophotographic
  - Inkjet