

type of techniques

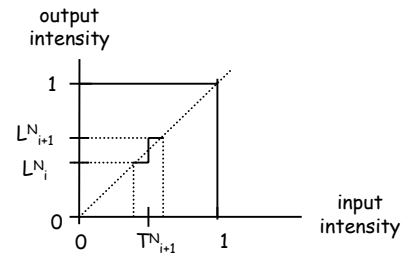
- simple pixel modification
- interpolation/extrapolation
- compositing
- convolution
- **dithering**
- warping
- morphing
- misc. effects

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N level uniform quantization

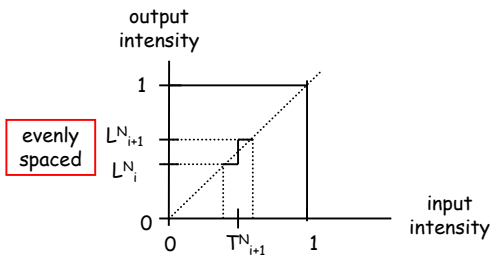


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N level uniform quantization

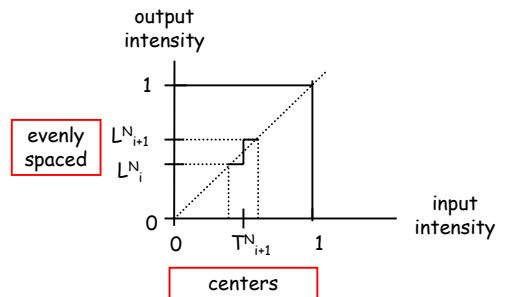


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N level uniform quantization



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quantization



8 bits per channel per pixel



1 bits per channel per pixel

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random dither

add noise to camouflage quantization artifacts



8 bits per channel per pixel



1 bits per channel per pixel



1 bits per channel per pixel noisy

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ordered dither



8 bits per channel per pixel

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1 bits per channel per pixel

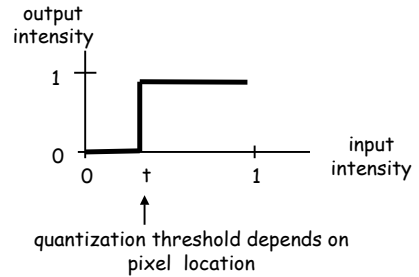
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1 bits per channel per pixel dithered

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ordered dither: 2 level output



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ordered dither

- ordered dither takes an input parameter $m \geq 1$, which controls the number of output levels we want to simulate
- to simulate $M = m^2 + 1$ levels
 - assign pixel locations to m^2 classes
 - use T_i^M as threshold to quantize pixel in i^{th} class

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ordered dither: $m=2, M=5$

assign pixel locations to $m^2=4$ classes

4	2	4	2	4	2	4
1	3	1	3	1	3	1
4	2	4	2	4	2	4
1	3	1	3	1	3	1
4	2	4	2	4	2	4

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ordered dither: $m=2, M=5$

compute thresholds $T_i^5, i=1, \dots, 4$:

$$T_1^5 = 1/8$$

$$T_2^5 = 3/8$$

$$T_3^5 = 5/8$$

$$T_4^5 = 7/8$$

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ordered dither: $m=2, M=5$

assign the thresholds to the pixel classes

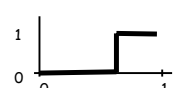
Class 1: $T_1^5 = 1/8$



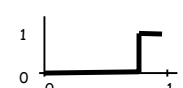
Class 2: $T_2^5 = 3/8$



Class 3: $T_3^5 = 5/8$



Class 4: $T_4^5 = 7/8$



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ordered dither: example

Input image

$\frac{1}{2}$	$\frac{1}{2}$
$\frac{1}{2}$	$\frac{1}{2}$



Output image

0	1
1	0

ordered dither: another view

add "noise" to each pixel depending on its class
then quantize as usual

How much noise?
for $m=2$ use $\frac{1}{2} - T^m_i$

ordered dither

To simulate $M=m^2+1$ intensity levels:

- assign pixel locations to m^2 classes
- use T^M_i as threshold to quantize pixel in i^{th} class

How do you assign pixel locations to the classes? answer: carefully

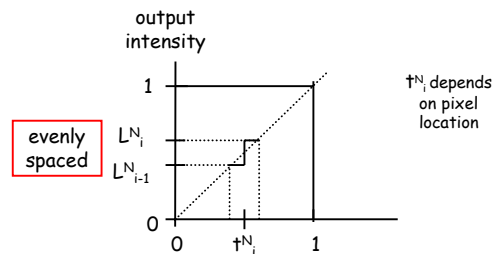
bayer's 2x2 ordered dither

4	2
1	3

bayer's 4x4 ordered dither matrix

16	8	14	6
4	12	2	10
13	5	15	7
1	9	3	11

Ordered dither: n level output



ordered dither: $m=2, M=5, N$ levels

t_i^N for pixel class j is $L_{i-1}^N + T_j^5/2^{N-1}$

intuition: scale the 2 level case for each input interval

error-diffusion dither



8 bits per channel per pixel



1 bits per channel per pixel



1 bits per channel per pixel dithered

comparison



original



ordered



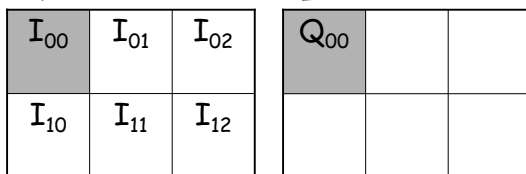
error-diffusion

error-diffusion dither

intuition

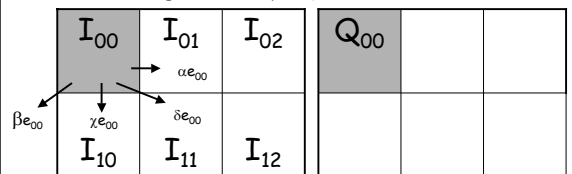
error diffusion dither

quantize I_{00} using uniform quantization



error diffusion dither

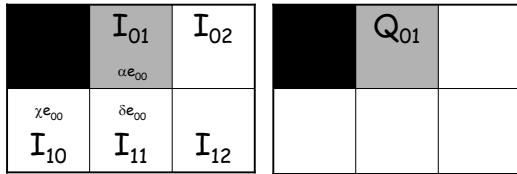
distribute error $e_{00} = I_{00} - Q_{00}$ to neighbors not yet quantized



$$\alpha + \beta + \gamma + \delta = 1$$

error diffusion dither

quantize $I_{01} + \alpha e_{00}$



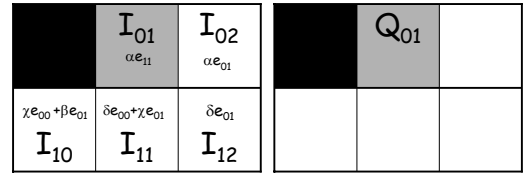
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error diffusion dither

distribute error: $e_{01} = I_{01} + \alpha e_{00} - Q_{01}$



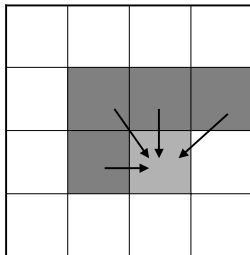
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error diffusion dither

error contributions by upper & left neighbors



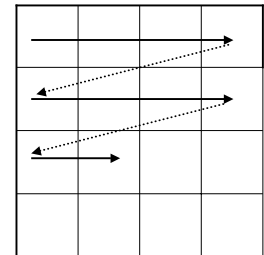
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error diffusion dither

order of quantization is important



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floyd-steinberg

$$\alpha = 7/16$$

$$\beta = 3/16$$

$$\chi = 5/16$$

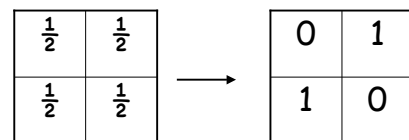
$$\delta = 1/16$$

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floyd-steinberg: example



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types of techniques

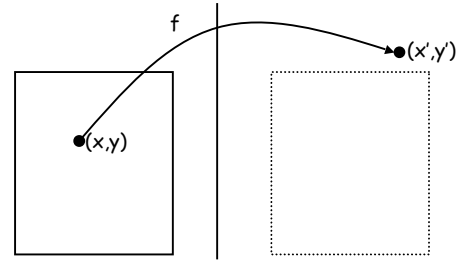
- simple pixel modification
- interpolation/extrapolation
- compositing
- convolution
- dithering
- **warping**
- morphing
- misc. effects

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forward warp



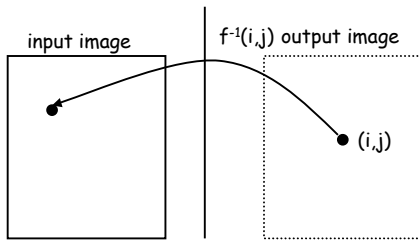
f maps points in input image to the plane

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forward warp



pixel at (i,j) in output image is assigned the value at location $f^{-1}(i,j)$ in input image

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forward warp: problems

if f is not bijective

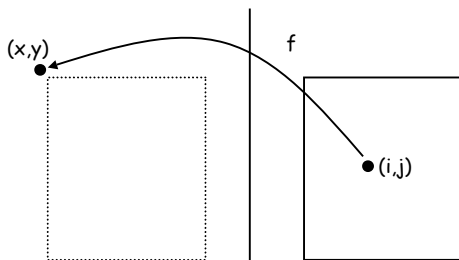
1. $f^{-1}(i,j)$ may not be defined
2. $f^{-1}(i,j)$ may not be unique

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backward warp



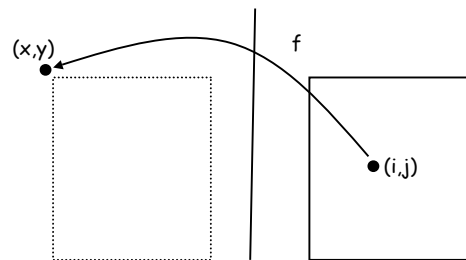
f maps points in output image to the plane

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backward warp



pixel (i,j) in output image is assigned value of input image at location $f(i,j)$

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backward warp: problems

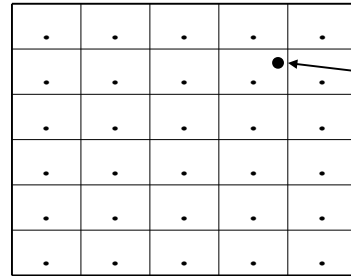
1. $f(i,j)$ may lie outside the input image area
solution: give image an infinite, black (or other default) border
2. $f(i,j)$ may not lie on a sample of the input image
solution: resample input

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re-sample



what is image value here?

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re-sample

interpolate based on nearby samples

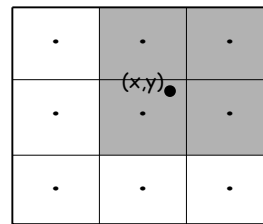
- nearest
- bilinear
- bicubic
- gaussian

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which way is up?



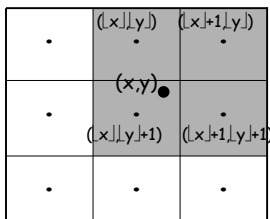
what are the coordinates of the pixels surrounding (x,y) ?

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which way is up?



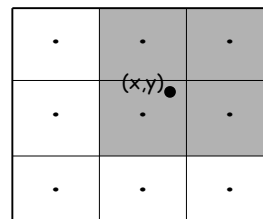
what are the coordinates of the pixels surrounding (x,y) ?

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nearest



- compute distance between x,y and the locations of the neighboring samples
- set value at x,y to the value of the closest neighbor

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re-sample

interpolate based on nearby samples

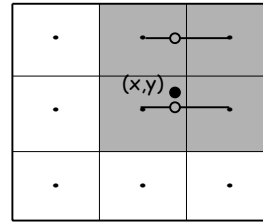
- nearest
- **bilinear**
- bicubic
- gaussian

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bilinear interpolation



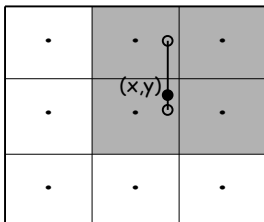
1. interpolate to find values at $(x, \lfloor y \rfloor)$ and $(x, \lfloor y \rfloor + 1)$

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bilinear interpolation



1. interpolate to find values at $(x, \lfloor y \rfloor)$ and $(x, \lfloor y \rfloor + 1)$
2. interpolate to find value at x, y

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re-sample

interpolate based on nearby samples

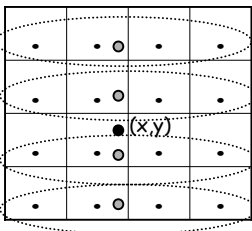
- nearest
- bilinear
- **bicubic**
- gaussian

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bicubic



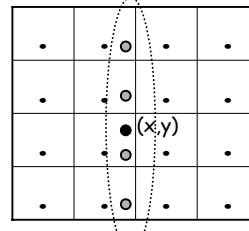
1. interpolate to find values at $(x, \lfloor y \rfloor + i)$

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bicubic



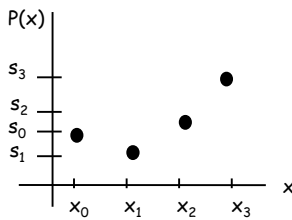
1. interpolate to find values at $(x, \lfloor y \rfloor + i)$
2. interpolate to find value at (x, y)

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bicubic: lagrangian



there is a unique cubic polynomial through any four distinct sample point

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lagrange cubic polynomial

$$P(x) = \sum_{i=0,1,2,3} s_i \prod_{j=0,1,2,3, j \neq i} (x-x_j)/(x_i-x_j)$$

exercise: what is the value of $P(x_i)$ $i=0,1,2,3$

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re-sample

interpolate based on nearby samples

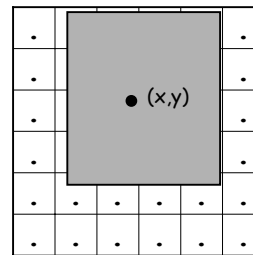
- nearest
- bilinear
- bicubic
- **gaussian**

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gaussian



interpolate nearby samples using normalized gaussian weights

unnormalized weight at (i,j) in window is $\exp[-((x-i)^2+(y-j)^2)/\sigma^2]$

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types of techniques

- simple pixel modification
- interpolation/extrapolation
- compositing
- convolution
- dithering
- warping
- **morphing**
- misc. effects

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