Computational photography techniques based on deconvolution

CS 178, Spring 2010

not responsible on exam for this "sidebar" lecture



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◆ Can you fix a blurry image by sharpening it in Photoshop?



simulated blurry image

◆ Can you fix a blurry image by sharpening it in Photoshop?



Photoshop Unsharp Mask

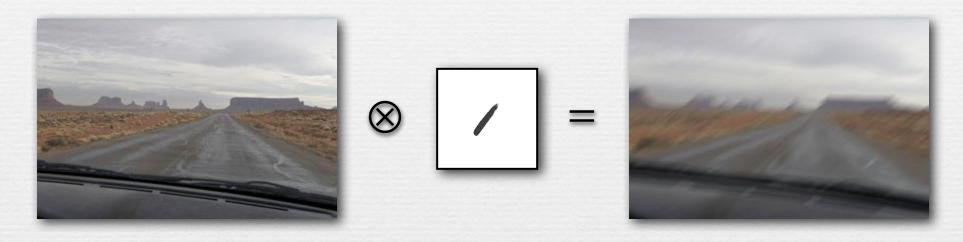
◆ Can you fix a blurry image by sharpening it in Photoshop?



more Unsharp Mask

Removing camera shake, 2nd try

→ camera shake can be modeled as a 2D convolution



- recall that discrete convolution replaces each pixel with a linear combination of nearby pixels
- → matrix-vector multiplication replaces each element in a vector with a linear combination of all other elements

Convolution as matrix multiplication

♦ let the sharp scene be represented by a vector

$$\mathbf{f} = [4 \ 7 \ 8 \ 4 \ 2 \ 5 \ 9 \ 6 \ 8 \ 4 \ 2]$$

♦ let the filter kernel be represented as a second vector

$$\mathbf{g} = [1 \ 2 \ 3 \ 2 \ 1]$$

 \bullet the convolution $\mathbf{f} \otimes \mathbf{g}$ becomes the matrix-vector product

$$\mathbf{Ax} = \begin{bmatrix} 3 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 2 & 3 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 2 & 3 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 3 & 2 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 2 & 3 & 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 2 & 3 & 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 2 & 3 & 2 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 2 & 3 & 2 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 2 & 3 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 2 & 3 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 2 & 3 & 2 \end{bmatrix} \begin{bmatrix} 4 \\ 7 \\ 8 \\ 4 \\ 2 \\ 5 \\ 9 \\ 6 \\ 8 \\ 4 \\ 2 \\ 2 \end{bmatrix}$$

where $\mathbf{x} = \mathbf{f}^{T}$ and \mathbf{A} is built from \mathbf{g} as shown

Convolution as matrix multiplication

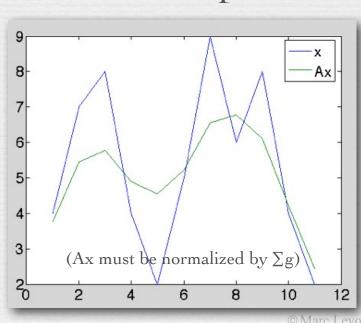
♦ let the sharp scene be represented by a vector

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Inverting convolution (deconvolution)

→ if the blurred image b is given by

$$Ax = b$$

◆ then the sharp scene x can be recovered by

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$$

where

 $A^{-1} =$

```
0.8571 -0.7857 0.0000 0.6429 -0.5714
                                   0 0.4286 -0.3571 -0.0000 0.2143 -0.1429
      1.5536 -0.7500 -0.5893 1.1071 -0.5000 -0.3929 0.6607 -0.2500 -0.1964 0.2143
0.0000 - 0.7500
             1.5000 -0.7500 -0.5000 1.0000 -0.5000 -0.2500 0.5000 -0.2500
0.6429 -0.5893 -0.7500 1.9821 -1.1786 -0.5000 1.3214 -0.7679 -0.2500 0.6607 -0.3571
      1.1071 -0.5000 -1.1786 2.2143 -1.0000 -0.7857 1.3214 -0.5000 -0.3929
             1.0000 -0.5000 -1.0000 2.0000 -1.0000 -0.5000 1.0000 -0.5000
    0 -0.5000
                    1.3214 -0.7857 -1.0000 2.2143 -1.1786 -0.5000
0.4286 -0.3929 -0.5000
       0.6607 -0.2500 -0.7679 1.3214 -0.5000 -1.1786 1.9821 -0.7500 -0.5893
-0.0000 -0.2500
              0.5000 -0.2500 -0.5000 1.0000 -0.5000 -0.7500 1.5000 -0.7500
0.2143 - 0.1964 - 0.2500 - 0.6607 - 0.3929 - 0.5000 - 1.1071 - 0.5893 - 0.7500
```

Why is deconvolution hard?

- * matrix A and blurred image b are typically very big
- → for a 10 megapixel image
 - A has 10 million rows and 10 million columns
 - b has 10 million entries
- → matrix A is typically <u>very sparse</u>
 - mostly zeros
- methods for solving big sparse systems of equations
 - conjugate gradient descent
 - etc.

Another reason deconvolution is hard

- → matrix A may be poorly conditioned
 - a small change in b causes a large change in x
 - 2nd row is poorly conditioned, producing ringing and noise (simulated)



Another reason deconvolution is hard

- → matrix A may be poorly conditioned
 - a small change in b causes a large change in x
- + equivalently, b's Fourier transform may contain zeros
 - sinusoids of some frequencies will be missing from b
- ◆ to be well conditioned, the filter shouldn't be smooth
 - bad:



better:



- convolution by first throws away detail, creating zeros in b
- convolution by the second makes many sharp copies, ok
- → inverting an ill-conditioned A produces a noisy result

Blind deconvolution

- → sometimes you don't know x or A
 - i.e. you don't know the sharp scene or the blur filter
- solving blind deconvolution problems
 - use a *prior* assumption about what the unknown sharp scene **x** should look like
- + this is hard, and we're not very good at it
 - solutions typically contain ringing, or worse...

[Fergus SIGGRAPH 2006]

deconvolve blurred image,
 using the statistics of natural scenes as a prior



blurred image

Photoshop Unsharp Mask

deconvolution

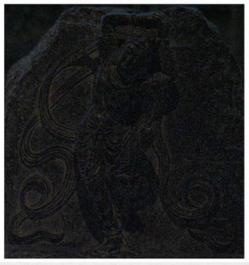


[Yuan SIGGRAPH 2007]

→ deconvolve long-exposure (blurred) image, using short-exposure (noisy) image as a prior



long exposure (blurry)



short exposure (dark)



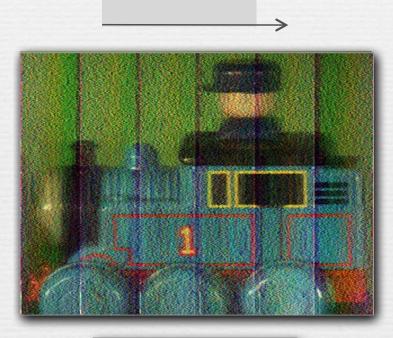
(noisy)

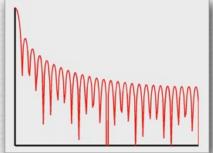


same, scaled up joint deconvolution

Removing motion blur [Raskar SIGGRAPH 2006]

continuous shutter



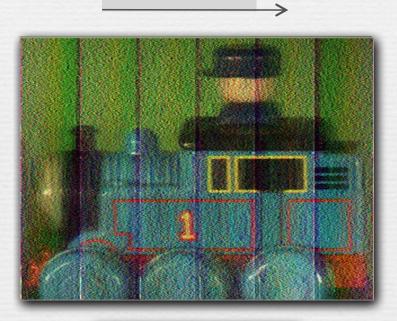


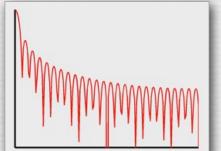
Removing motion blur

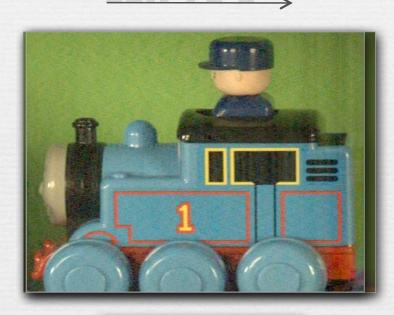
[Raskar SIGGRAPH 2006]

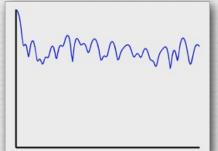
continuous shutter

fluttered shutter









Slide credits

◆ Andrew Adams