

# ECE 468: Digital Image Processing

## Lecture 7

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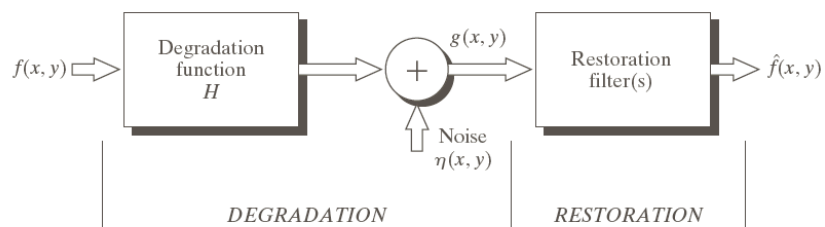
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### Model of Image Degradation/Restoration

**FIGURE 5.1**  
A model of the  
image  
degradation/  
restoration  
process.



$$g(x, y) = h(x, y) \star f(x, y) + \eta(x, y)$$

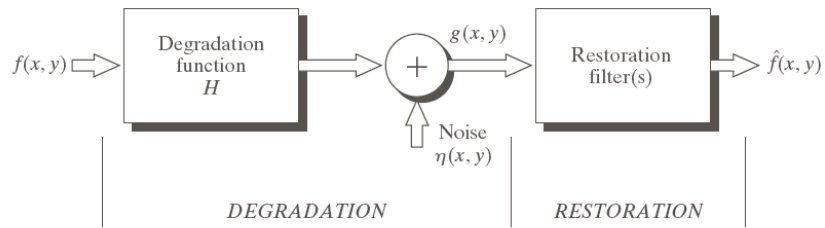
$$G(u, v) = H(u, v)F(u, v) + N(u, v)$$

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## Image Restoration by Inverse Filtering

**FIGURE 5.1**

A model of the image degradation/restoration process.



$$G(u, v) = H(u, v)F(u, v) + N(u, v)$$

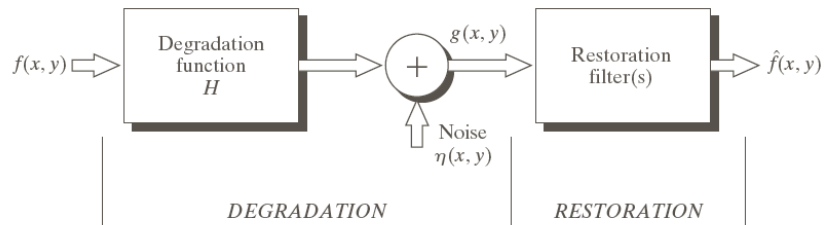
$$\hat{F}(u, v) = \frac{G(u, v)}{H(u, v)}$$

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## Image Restoration by Inverse Filtering

**FIGURE 5.1**

A model of the image degradation/restoration process.



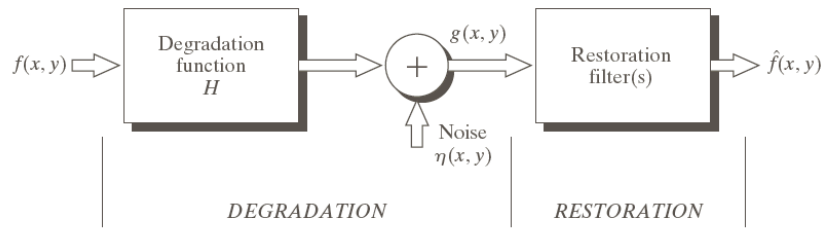
$$G(u, v) = H(u, v)F(u, v) + N(u, v)$$

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## Image Restoration by Inverse Filtering

**FIGURE 5.1**

A model of the image degradation/restoration process.



$$G(u, v) = H(u, v)F(u, v) + N(u, v)$$

$$\hat{F}(u, v) = F(u, v) + \frac{N(u, v)}{H(u, v)}$$

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## Inverse Filtering

$$\hat{F}(u, v) = F(u, v) + \frac{N(u, v)}{H(u, v)}$$

Bad news:

- Even when  $H(u, v)$  is known, there is always unknown noise
- Often  $H(u, v)$  has values close to zero

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## Example: Inverse Filtering



Atmospheric turbulence effect

$$H(u, v) = \exp \left\{ -k \left[ (u - M/2)^2 + (v - N/2)^2 \right]^{5/6} \right\}$$

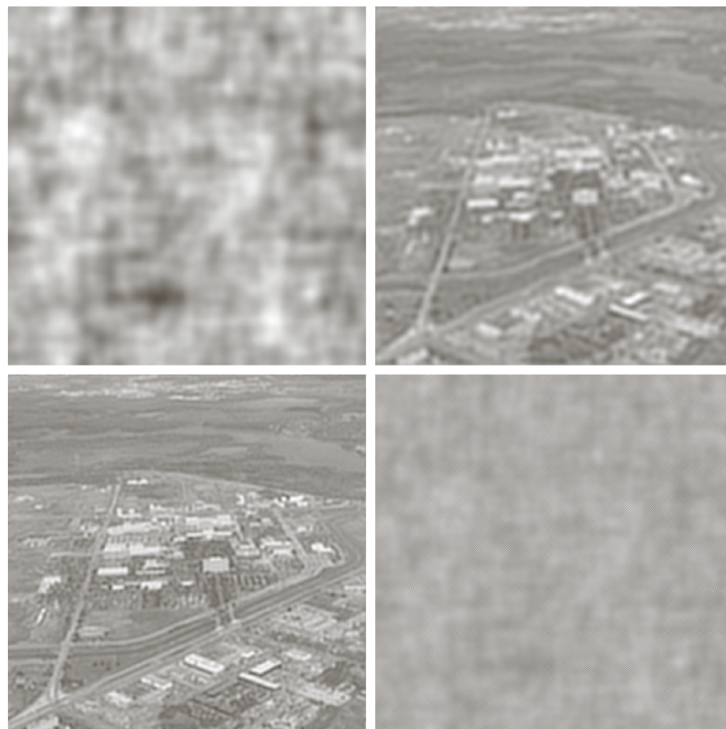
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## Example: Inverse Filtering

a b  
c d

**FIGURE 5.27**

Restoring  
Fig. 5.25(b) with  
Eq. (5.7-1).  
(a) Result of  
using the full  
filter. (b) Result  
with  $H$  cut off  
outside a radius of  
40; (c) outside a  
radius of 70; and  
(d) outside a  
radius of 85.



$$\frac{G(u, v)}{H(u, v)}$$

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## Wiener Filtering = Mean Square Error Filtering

- Incorporates both:
  - Degradation function
  - Statistical characteristics of noise
- Assumption: noise and the image are uncorrelated
- Optimizes the filter so that MSE is minimized

$$e^2 = E \left\{ [f(x, y) - \hat{f}(x, y)]^2 \right\}$$

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## Wiener Filtering

$$\underset{\text{degraded output}}{G(u, v)} = H(u, v) \underset{\text{input}}{F(u, v)} + N(u, v)$$

$$\underset{\text{restored}}{\hat{F}(u, v)} = H_W(u, v) G(u, v)$$

$$H_W(u, v) = \frac{1}{H(u, v)} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{|N(u, v)|^2}{|F(u, v)|^2}}$$

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## Signal to Noise Ratio

$$H_W(u, v) = \frac{1}{H(u, v)} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{|N(u, v)|^2}{|F(u, v)|^2}}$$

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## Signal to Noise Ratio

$$H_W(u, v) = \underbrace{\frac{1}{H(u, v)}}_{\text{inverse filter}} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{|N(u, v)|^2}{|F(u, v)|^2}}$$

10

## Signal to Noise Ratio

$$H_W(u, v) = \frac{1}{H(u, v)} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{|N(u, v)|^2}{|F(u, v)|^2}}$$

inverse filter                      unknown

10

## Signal to Noise Ratio

$$H_W(u, v) = \frac{1}{H(u, v)} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{|N(u, v)|^2}{|F(u, v)|^2}}$$

inverse filter                      unknown

$$\text{SNR} = \frac{\sum_{u=0}^{M-1} \sum_{v=0}^{N-1} |F(u, v)|^2}{\sum_{u=0}^{M-1} \sum_{v=0}^{N-1} |N(u, v)|^2}$$

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## Approximation

$$H_W(u, v) = \frac{1}{H(u, v)} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{|N(u, v)|^2}{|F(u, v)|^2}} \approx \frac{1}{H(u, v)} \frac{|H(u, v)|^2}{|H(u, v)|^2 + \frac{1}{\text{SNR}}}$$

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## Example: Wiener Filtering



a b c

**FIGURE 5.28** Comparison of inverse and Wiener filtering. (a) Result of full inverse filtering of Fig. 5.25(b). (b) Radially limited inverse filter result. (c) Wiener filter result.

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## Example: Wiener Filtering



**FIGURE 5.29** (a) 8-bit image corrupted by motion blur and additive noise. (b) Result of inverse filtering. (c) Result of Wiener filtering. (d)–(f) Same sequence, but with noise variance one order of magnitude less. (g)–(i) Same sequence, but noise variance reduced by five orders of magnitude from (a). Note in (h) how the deblurred image is quite visible through a “curtain” of noise.

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## Midterm Exam Review

- Exam:
  - Average: 66.8
  - Median: 70.5
  - STD: 14.2
  
- Projected Total:
  - Average: 85.5
  - Median: 87.1
  - STD: 6.1

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