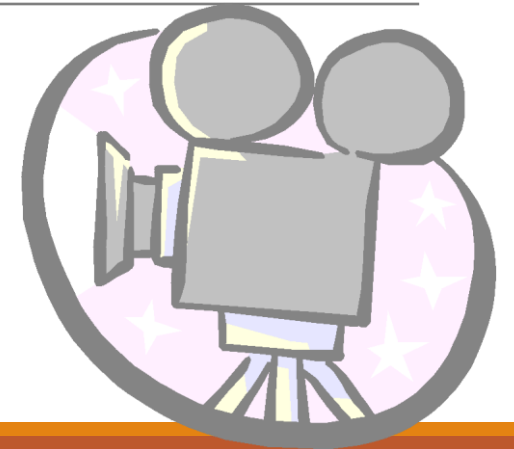


Image Processing

CS-317/CS-341



Outline

- Image Enhancement in the Frequency Domain

 - Sharpening Filters

- Image Restoration

Laplacian in the Frequency domain

The Laplacian in the Frequency domain can be implemented using the filter

$$H(u, v) = -4\pi^2(u^2 + v^2)$$

or, with respect to the center of the frequency rectangle, using the filter

$$\begin{aligned} H(u, v) &= -4\pi^2[(u - M/2)^2 + (v - N/2)^2] \\ &= -4\pi^2 D^2(u, v) \end{aligned} \quad \begin{matrix} M \\ N \end{matrix}$$

Where, $D(u, v)$ is the distance function

Laplacian in the Frequency domain: Image Enhancement

The Laplacian image is obtained as:

$$\nabla^2 f(x, y) = \mathfrak{Z}^{-1}\{H(u, v)F(u, v)\} = IFT\{H(u, v)F(u, v)\}$$

Where, $F(u, v)$ is DFT of $f(x, y)$, Now image enhancement is achieved using the equation (in spatial domain):

$$g(x, y) = f(x, y) + c\nabla^2 f(x, y)$$

Here, $c=-1$, because $H(u, v)$ is negative.

Laplacian in the Frequency domain: Image Enhancement

In frequency domain

$$\begin{aligned}g(x, y) &= \mathfrak{Z}^{-1}\{F(u, v) - H(u, v)F(u, v)\} \\&= \mathfrak{Z}^{-1}\{[1 - H(u, v)]F(u, v)\} \\&= \mathfrak{Z}^{-1}\{[1 + 4\pi^2 D^2(u, v)]F(u, v)\}\end{aligned}$$

As,

$$H(u, v) = -4\pi^2(u^2 + v^2)$$

Example: Laplacian filtered image

a	b
c	d

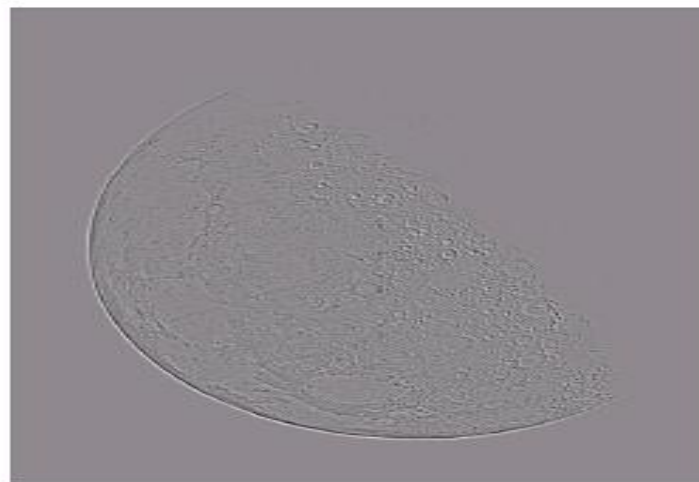
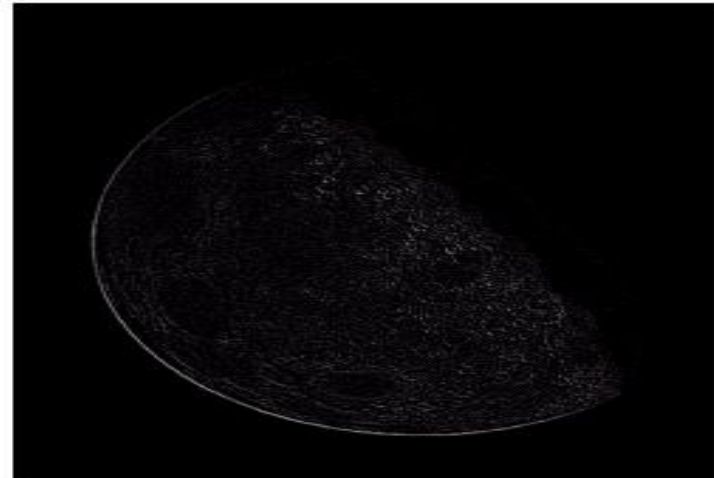
FIGURE 4.28

(a) Image of the North Pole of the moon.

(b) Laplacian filtered image.

(c) Laplacian image scaled.

(d) Image enhanced by using Eq. (4.4-12). (Original image courtesy of NASA.)



Unsharp masking

The unsharp masking is expressed as:

$$g_{mask}(x, y) = f(x, y) - f_{lp}(x, y)$$

$$\text{With, } f_{lp} = \nabla^2 f(x, y) = \mathfrak{Z}^{-1}\{H_{LP}(u, v)F(u, v)\}$$

Here, $H_{LP}(u, v)$ is a low pass filter and $F(u, v)$ is DFT of $f(x, y)$,

Then we add a weighted portion of the mask back to the original image:

$$g(x, y) = f(x, y) + k * g_{mask}(x, y)$$

Here, we included a weight k ($k \geq 0$), for generality.

When $k=1$, we have **unsharp masking**.

When $k>1$, the Process is referred to as **highboost filtering**.

Lowpass

$$g(x, y) = IFT \{ (1 + k * (1 - H_{LP}(u, v))) F(u, v) \}$$

Highpass, filter

$$g(x, y) = IFT \{ [1 + k * H_{HP}(u, v)] F(u, v) \}$$

General..Formula

$$g(x, y) = IFT \{ [k_1 + k_2 * H_{HP}(u, v)] F(u, v) \}$$

Where, $k_1 \geq 0$, offsets from origin

$k_2 \geq 0$, contribute to High frequency filtering

Homomorphic Filtering

The illumination and reflectance model can be used to develop a frequency domain procedure for improving the appearance of an image by simultaneous intensity range compression and contrast enhancement.

An image $f(x, y)$ can be expressed as the product of illumination $i(x, y)$ and reflectance $r(x, y)$ components as:

$$f(x, y) = i(x, y)r(x, y)$$

This equation can not be used directly to operate on the frequency components of illumination and reflectance because Fourier Transformation of a product is not the product of the transformations

$$\mathcal{F}[f(x, y)] \neq \mathcal{F}[i(x, y)] \mathcal{F}[r(x, y)]$$

Homomorphic Filtering

Suppose we define,

$$Z(x, y) = \ln f(x, y)$$

$$Z(x, y) = \ln i(x, y) + \ln r(x, y)$$

Now, in frequency domain FT of $z(x, y)$

$$FT(z(x, y)) = FT(\ln i(x, y) + \ln r(x, y))$$

$$FT(z(x, y)) = FT(\ln i(x, y)) + FT(\ln r(x, y))$$

$$z(u, v) = F_i(u, v) + F_r(u, v)$$

where $F_i(u, v)$ and $F_r(u, v)$ are the Fourier transforms of $\ln i(x, y)$ and $\ln r(x, y)$, respectively.

Now image is in Frequency domain, so we can apply filter

Homomorphic Filtering

We can filter $Z(u, v)$ using a filter $H(u, v)$ so that

$$\begin{aligned} S(u, v) &= H(u, v)Z(u, v) \\ &= H(u, v)F_i(u, v) + H(u, v)F_r(u, v) \end{aligned}$$

The filtered image in the spatial domain is

$$\begin{aligned} s(x, y) &= \mathfrak{Z}^{-1}\{S(u, v)\} \\ &= \mathfrak{Z}^{-1}\{H(u, v)F_i(u, v)\} + \mathfrak{Z}^{-1}\{H(u, v)F_r(u, v)\} \end{aligned}$$

Where, \mathfrak{Z}^{-1} is the inverse FT.

By defining

$$i'(x, y) = \mathfrak{Z}^{-1}\{H(u, v)F_i(u, v)\}$$

and

$$r'(x, y) = \mathfrak{Z}^{-1}\{H(u, v)F_r(u, v)\}$$

we can express Eq. (4.9-23) in the form

$$s(x, y) = i'(x, y) + r'(x, y)$$

Finally, because $z(x, y)$ was formed by taking the natural logarithm of the input image, we reverse the process by taking the exponential of the filtered result to form the output image:

$$\begin{aligned}g(x, y) &= e^{s(x, y)} \\&= e^{i'(x, y)} e^{r'(x, y)} \\&= i_0(x, y) r_0(x, y)\end{aligned}$$

where

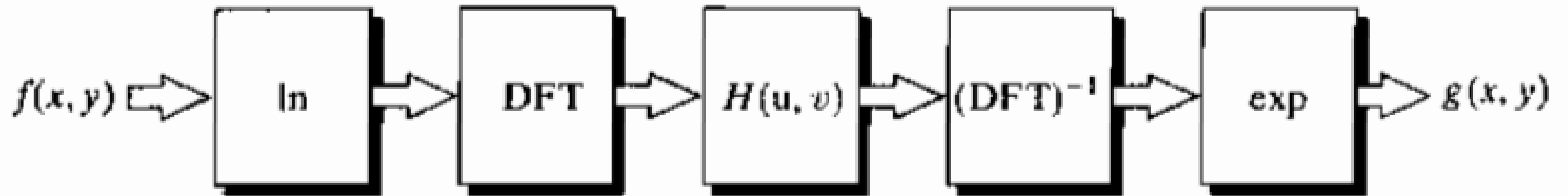
$$i_0(x, y) = e^{i'(x, y)}$$

and

$$r_0(x, y) = e^{r'(x, y)}$$

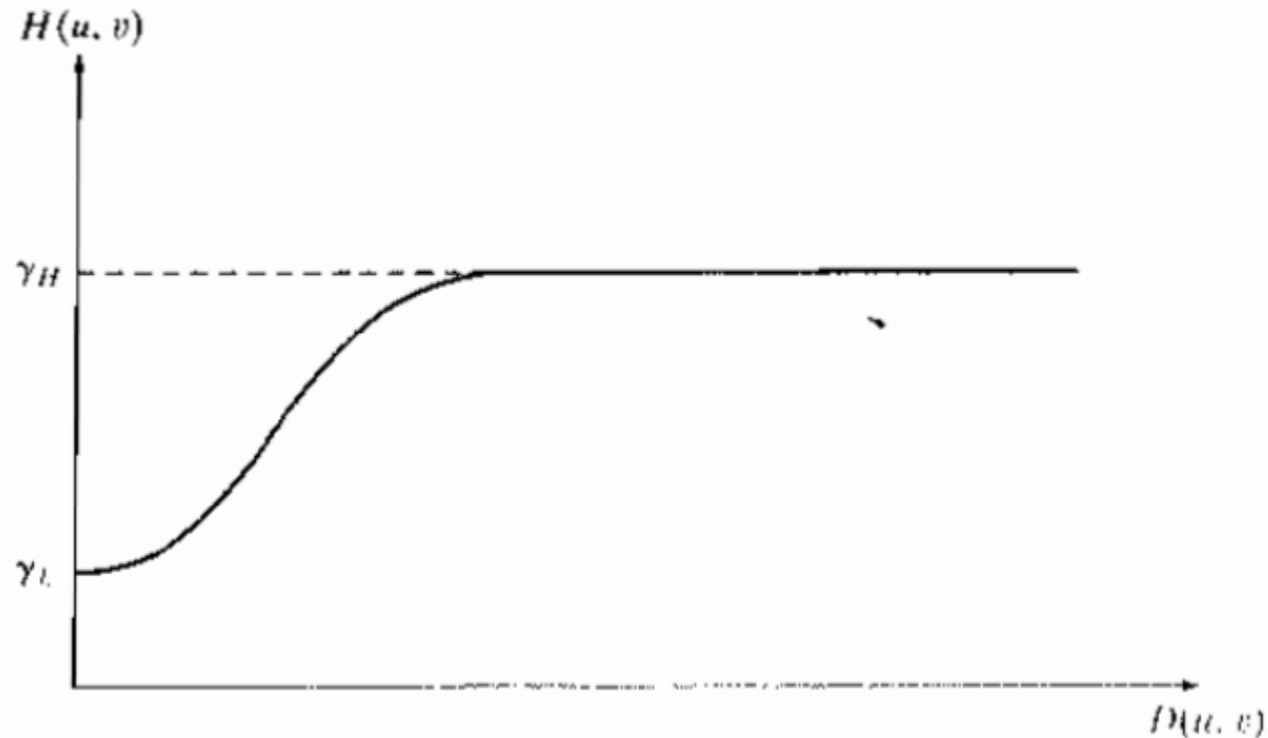
are the illumination and reflectance components of the output (processed) image.

Steps of Homomorphic Filtering



$\gamma_L < 1$, it compresses low frequency components
 $\gamma_H > 1$, contrast enhancement

$$H(u, v) = (\gamma_H - \gamma_L) \left[1 - e^{-c[D^2(u, v)/D_0^2]} \right] + \gamma_L$$



Suggested Readings

- ❑ **Digital Image Processing by Rafael Gonzalez, Richard Woods, Pearson Education India, 2017.**
- ❑ **Fundamental of Digital image processing by A. K Jain, Pearson Education India, 2015.**

Thank you

