

## Computer Vision and Image Processing

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## Frequency Domain: Frequency Transform

$$F(u) = \int_{-\infty}^{\infty} f(x) e^{-j2\pi ux} dx \quad f(x) = \int_{-\infty}^{\infty} F(u) e^{j2\pi ux} du$$

$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi(ux+vy)} dx dy$$

$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{j2\pi(ux+vy)} du dv$$

- discrete Fourier transform

$$F(u) = \frac{1}{M} \sum_{x=0}^{M-1} f(x) e^{-j2\pi ux/M} \quad x, u = 0, 1, 2, \dots, M-1$$

$$f(x) = \sum_{u=0}^{M-1} F(u) e^{j2\pi ux/M} \quad x, u = 0, 1, 2, \dots, M-1$$

both equations are multiplied by  $1/\sqrt{M}$



## Frequency Domain Presentation and Processing

- French mathematician Jean Baptiste Joseph Fourier
- any function that periodically repeats itself, can be expressed as
- sum of sines / cosines of different frequencies
- known as Fourier series
- functions that are not periodic, can be expressed as
- integral of sines / cosines multiplied by a weighing function
- known as Fourier transform
- inverse process; reconstruction with no loss of information



## Fourier transform

- $e^{j\theta} = \cos \theta + j \sin \theta$

$$F(u) = \frac{1}{M} \sum_{x=0}^{M-1} f(x) [\cos 2\pi ux/M - j \sin 2\pi ux/M]$$

- domain over which the values of  $F(u)$  range - called the frequency domain
- $M$  terms of  $F(u)$  - called frequency components
- $F(u) = |F(u)| e^{-j\phi(u)}$  where  $|F(u)| = [R^2(u) + I^2(u)]^{1/2}$  called spectrum of Fourier transform
- phase spectrum of transform  $\phi(u) = \tan^{-1} \left[ \frac{I(u)}{R(u)} \right]$
- power spectrum  $P(u) = |F(u)|^2$  - called spectral density



## Discrete Fourier transform

- spectrum centered at  $u = 0$ ; multiplying  $f(x)$  by  $(-1)^x$  before taking the transform
- $K = 8$  points,  $M = 1024$  vs.  $K = 16$  points and  $M = 1024$
- height of spectrum doubled as area under the curve is doubled
- the number of zeros in the spectrum in the same interval doubled as the length of function doubled
- $f(x)$  having  $M$  samples, not necessarily taken at integer value of  $x$  in the interval  $[0, M - 1]$ , but equally spaced, first point  $x_0; f(x_0)$
- next sample  $f(x_0 + \Delta x)$ ,  $k$ th sample  $f(x_0 + k \Delta x)$  and final sample  $f(x_0 + [M - 1] \Delta x)$



## DFT

- $u$  and  $v$  are transform or frequency variables
- $x$  and  $y$  are spatial or image variables
- multiply input image function  $(-1)^{x+y}$  prior to computing Fourier transform
- shifting origin of  $F(u, v)$  to frequency coordinates  $(M/2, N/2)$
- $F(0, 0)$  is located at  $u = M/2$  and  $v = N/2$

$$\mathcal{F}[f(x, y)(-1)^{x+y}] = F(u - M/2, v - N/2)$$

$$F(0, 0) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y)$$

dc component of the spectrum, Fourier transform at the origin is equal to the average gray level of the image



## Discrete Fourier transform

- $f(k)$  notation for  $f(x_0 + k \Delta x)$ , i.e.,  $f(x) \triangleq f(x_0 + x \Delta x)$
- $u$  always starts at 0 frequency, values of  $u$ :  
 $0, \Delta u, 2 \Delta u, \dots, [M - 1] \Delta u$   $F(u) \triangleq F(u \Delta u)$
- $\Delta x$  and  $\Delta u$  are inversely related  $\Delta u = \frac{1}{M \Delta x}$
- two dimensional DFT  $f(x, y)$  of size  $M \times N$

$$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(ux/M + vy/N)}$$

$$f(x, y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi(ux/M + vy/N)}$$

$x = 0, 1, 2, \dots, M - 1$  and  $y = 0, 1, 2, \dots, N - 1$



## DFT

- if  $f(x, y)$  is real, Fourier transform is conjugate symmetric  
 $F(u, v) = F^*(-u, -v)$
- $|F(u, v)| = |F(-u, -v)|$  spectrum of Fourier transform is symmetric
- centering property and conjugate symmetry: circularly symmetric filters in the frequency domain
- $\Delta u = \frac{1}{M \Delta x}$  and  $\Delta v = \frac{1}{N \Delta y}$
- log transform,  $c = 0.5$ , plotting Fourier transform



## Filtering in Frequency Domain

- each term  $F(u, v)$  contains all values of  $f(x, y)$ , modified by values of exponential terms
- frequency related to rate of change
- frequencies in Fourier transform associated with patterns of intensity variations in an image
- low frequencies correspond to slowly varying components of an image
- e.g. correspond to smooth variations
- high frequencies correspond to faster gray level changes in the image
- e.g. edges, abrupt changes in gray level such as noise
- $H(u, v)$  filter suppresses certain frequencies and leaving other unchanged

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



## Filtering in Frequency Domain

- relationship between spatial and frequency domain
- convolution theorem

$$f(x, y) * h(x, y) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f(m, n)h(x - m, y - n)$$

- flipping function about origin
- shift function w.r.t. other by changing  $(x, y)$
- computing sum of products over all values of  $m$  and  $n$

$$f(x, y) * h(x, y) \Leftrightarrow F(u, v)H(u, v)$$

$$f(x, y)h(x, y) \Leftrightarrow F(u, v) * H(u, v)$$

- convolution in frequency domain - multiplication in spatial domain



## Filtering in Frequency Domain

- $F$  complex quantity,  $H$  generally real, are called zero phase shift filters
- filters do not change phase of the transform
- filtered image  $\mathcal{F}^{-1}[G(u, v)]$ , is also complex
- image and filter function are real, imaginary components of inverse transform should all be zero
- due to round off errors inverse DFT has imaginary components
- $F(0, 0)$  set to zero, average value of an image becomes zero, notch filter

$$H(u, v) = \begin{cases} 0 & \text{if } (u, v) = (M/2, N/2) \\ 1 & \text{otherwise} \end{cases}$$



## Filtering in Frequency Domain

- impulse function, located at  $(x_0, y_0)$ :  $A\delta(x - x_0, y - y_0)$

$$\sum_{x=0}^{M-1} \sum_{y=0}^{N-1} s(x, y)A\delta(x - x_0, y - y_0) = As(x_0, y_0)$$

- sifting property of the impulse function
- Fourier transform of a unit impulse at the origin

$$\sum_{x=0}^{M-1} \sum_{y=0}^{N-1} s(x, y)\delta(x, y) = s(0, 0)$$

$$F(u, v) = \frac{1}{MN} \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} \delta(x, y)e^{-j2\pi(ux/M + vy/N)} = \frac{1}{MN}$$



## Filtering in Frequency Domain

- impulse at origin, Fourier transform is a real constant
- impulse located elsewhere, the transform has complex value
- $f(x, y) = \delta(x, y)$

$$f(x, y) * h(x, y) = \frac{1}{MN} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} \delta(m, n) h(x - m, y - n) = \frac{1}{MN} h(x, y)$$

$$\delta(x, y) * h(x, y) \Leftrightarrow \mathcal{F}[\delta(x, y)] H(u, v) \quad h(x, y) \Leftrightarrow H(u, v)$$

- given a filter in frequency domain; inverse Fourier transform corresponds to filter in spatial domain
- frequency domain filter size  $M \times N$ , much small filter in spatial domain



## Frequency Domain Filters

- $G(u, v) = H(u, v)F(u, v)$
- lowpass filters: ideal, Butterworth and Gaussian
- Butterworth filter, filter order as parameter, high value of it makes filter ideal

$$H(u, v) = \begin{cases} 1 & \text{if } D(u, v) \leq D_0 \\ 0 & \text{if } D(u, v) > D_0 \end{cases}$$

- $D_0$  is specified nonnegative quantity, cutoff frequency
- $D(u, v)$  is distance from point  $(u, v)$  to origin of frequency rectangle
- center is at  $(u, v) = (M/2, N/2)$

$$D(u, v) = [(u - M/2)^2 + (v - N/2)^2]^{1/2}$$



## Gaussian Filter

- forward and inverse Fourier transforms of Gaussian function are real Gaussian functions

$$H(u) = Ae^{-u^2/2\sigma^2} \quad h(x) = \sqrt{2\pi}\sigma Ae^{-2\pi^2\sigma^2 x^2}$$

- these functions reciprocal w.r.t. one another
- $H(u)$  broad profile (larger value of  $\sigma$ ),  $h(x)$  has narrow profile
- $\sigma$  approaches infinity,  $H(u)$  becomes a constant function and  $h(x)$  an impulse
- highpass filter as difference of Gaussians

$$H(u) = Ae^{-u^2/2\sigma_1^2} - Be^{-u^2/2\sigma_2^2} \quad (A \geq B, \sigma_1 > \sigma_2)$$

$$h(x) = \sqrt{2\pi}\sigma_1 Ae^{-2\pi^2\sigma_1^2 x^2} - \sqrt{2\pi}\sigma_2 Be^{-2\pi^2\sigma_2^2 x^2}$$



## Frequency Domain Filters

- cutoff frequency, compute radius that enclose specified amount of total image power
- as filter radius increases, less power is removed, reduces blurring
- "ringing" becomes finer as amount of high frequency content removed decreases
- "ringing" behavior is a characteristic of ideal filters

$$\alpha = 100 \left[ \sum_u \sum_v P(u, v) / P_T \right] \quad P_T = \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} P(u, v)$$

- $h(x, y)$  obtained by  $H(u, v)$  multiplied by  $(-1)^{u+v}$  for centering
- inverse DFT and real part of it multiplied by  $(-1)^{x+y}$
- center component responsible for blurring
- concentric components responsible for ringing characteristic



## Frequency Domain Filters

- Butterworth lowpass filter of order  $n$ , cutoff frequency at distance  $D_0$

$$H(u, v) = \frac{1}{1 + [D(u, v)/D_0]^{2n}}$$

- does not have sharp discontinuity
- Gaussian lowpass filter

$$H(u, v) = e^{-D^2(u, v)/2\sigma^2}$$

- $\sigma$  measure of spread of Gaussian curve ( $\sigma = D_0$ )
- highpass filter  $H_{hp}(u, v) = 1 - H_{lp}(u, v)$



## Frequency Domain Filters

- Ideal highpass filter

$$H(u, v) = \begin{cases} 0 & \text{if } D(u, v) \leq D_0 \\ 1 & \text{if } D(u, v) > D_0 \end{cases}$$

- Butterworth highpass filter

$$H(u, v) = \frac{1}{1 + [D_0/D(u, v)]^{2n}}$$

- Gaussian highpass filter

$$H(u, v) = 1 - e^{-D^2(u, v)/2D_0^2}$$

