

EEL 6562
Image Processing & Computer Vision
Notes on Frequency Domain Theory

1 Fourier Transform of Continuous Functions

1.1 Definitions

Forward transform:

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

Reverse transform:

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

$F(u, v)$ is, in general, a complex function, so it has a rectangular form:

$$F(u, v) = R(u, v) + jI(u, v)$$

where $R(u, v)$ and $I(u, v)$ are real functions of u and v . $F(u, v)$ also has a polar form:

$$F(u, v) = |F(u, v)| e^{-j\phi(u, v)}$$

where

$$|F(u, v)| = (R^2(u, v) + I^2(u, v))^{\frac{1}{2}}$$

and

$$\phi(u, v) = \tan^{-1} \left(\frac{I(u, v)}{R(u, v)} \right)$$

Power spectrum:

$$P(u, v) = |F(u, v)|^2 = F(u, v) F^*(u, v) = R^2(u, v) + I^2(u, v)$$

1.2 Properties

Linearity:

$$af(x, y) + bg(x, y) \xleftrightarrow{\mathcal{F}} aF(u, v) + bG(u, v)$$

Spatial translation:

$$f(x - x_0, y - y_0) \xleftrightarrow{\mathcal{F}} F(u, v) e^{-j2\pi(u x_0 + v y_0)}$$

Frequency translation:

$$f(x, y) e^{j2\pi(u_0 x + v_0 y)} \xleftrightarrow{\mathcal{F}} F(u - u_0, v - v_0)$$

Differentiation:

$$\frac{\partial^n}{\partial x^n} f(x, y) \xleftrightarrow{\mathcal{F}} (j2\pi u)^n F(u, v)$$

Proof:

$$\begin{aligned} \frac{\partial^n}{\partial x^n} f(x, y) &= \frac{\partial^n}{\partial x^n} \left\{ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{j2\pi(u x + v y)} du dv \right\} \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) \left\{ \frac{\partial^n}{\partial x^n} e^{j2\pi(u x + v y)} \right\} du dv \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) \left\{ (j2\pi u)^n e^{j2\pi(u x + v y)} \right\} du dv \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left\{ (j2\pi u)^n F(u, v) \right\} e^{j2\pi(u x + v y)} du dv \end{aligned}$$

$$(-j2\pi x)^n f(x, y) \xleftrightarrow{\mathcal{F}} \frac{\partial^n}{\partial u^n} F(u, v)$$

Laplacian:

$$\nabla^2 f(x, y) \xleftrightarrow{\mathcal{F}} -(2\pi)^2 (u^2 + v^2) F(u, v)$$

Scaling:

$$f(ax, by) \xleftrightarrow{\mathcal{F}} \frac{1}{|ab|} F(u/a, v/b)$$

Separable functions:

$$f_x(x) f_y(y) = f(x, y) \xleftrightarrow{\mathcal{F}} F(u, v) = F_x(u) F_y(v)$$

Proof:

$$\begin{aligned} F(u, v) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_x(x) f_y(y) e^{-j2\pi(u x + v y)} dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left\{ f_x(x) e^{-j2\pi u x} \right\} \left\{ f_y(y) e^{-j2\pi v y} \right\} dx dy \\ &= \int_{-\infty}^{\infty} f_y(y) e^{-j2\pi v y} \left\{ \int_{-\infty}^{\infty} f_x(x) e^{-j2\pi u x} dx \right\} dy \\ &= \left\{ \int_{-\infty}^{\infty} f_x(x) e^{-j2\pi u x} dx \right\} \left\{ \int_{-\infty}^{\infty} f_y(y) e^{-j2\pi v y} dy \right\} \\ &= F_x(u) F_y(v) \end{aligned}$$

1.3 Convolution

Linear convolution in continuous spatial domain is defined as

$$f(x, y) * g(x, y) \triangleq \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(s, t) g(x-s, y-t) ds dt$$

Convolution is commutative and distributive over addition:

$$f(x, y) * g(x, y) = g(x, y) * f(x, y)$$

$$f(x, y) * (g(x, y) + h(x, y)) = f(x, y) * g(x, y) + f(x, y) * h(x, y)$$

The Dirac delta function is defined as the function $\delta(x, y)$ that satisfies:

$$\iint_R \delta(x, y) dA \triangleq \begin{cases} 1 & \text{if } (0, 0) \in R; \\ 0 & \text{otherwise.} \end{cases}$$

The Dirac delta function is the identity element for convolution:

$$f(x, y) * \delta(x, y) = \delta(x, y) * f(x, y) = f(x, y)$$

The convolution theorem:

$$f(x, y) * g(x, y) \xrightarrow{\mathcal{F}} F(u, v) G(u, v)$$

$$f(x, y) g(x, y) \xrightarrow{\mathcal{F}} F(u, v) * G(u, v)$$

1.4 Common transform pairs

Some common Fourier transform pairs:

$$\delta(x, y) \xrightarrow{\mathcal{F}} 1$$

$$2\pi\sigma^2 e^{-2\pi^2\sigma^2(x^2+y^2)} \xrightarrow{\mathcal{F}} e^{-(u^2+v^2)/(2\sigma^2)}$$

Proof:

$$\begin{aligned} F(u, v) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} 2\pi\sigma^2 e^{-2\pi^2\sigma^2(x^2+y^2)} e^{-j2\pi(ux+vy)} dx dy \\ &= e^{-\frac{u^2+v^2}{2\sigma^2}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} 2\pi\sigma^2 e^{-2\pi^2\sigma^2(x^2+y^2)} e^{-j2\pi(ux+vy)} e^{\frac{u^2+v^2}{2\sigma^2}} dx dy \\ &= e^{-\frac{u^2+v^2}{2\sigma^2}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} 2\pi\sigma^2 e^{-2\pi^2\sigma^2 x^2 - j2\pi ux + \frac{u^2}{2\sigma^2}} \\ &\quad \cdot e^{-2\pi^2\sigma^2 y^2 - j2\pi vy + \frac{v^2}{2\sigma^2}} dx dy \end{aligned}$$

Working with the first exponent after the integral signs:

$$\begin{aligned} -2\pi^2\sigma^2 x^2 - j2\pi ux + \frac{1}{2\sigma^2} u^2 &= -\frac{1}{2\sigma^2} (4\pi^2\sigma^4 x^2 + j4\pi\sigma^2 ux + u^2) \\ &= -\frac{1}{2\sigma^2} (2\pi\sigma^2 x + ju)^2 \end{aligned}$$

The second exponent after the integral signs is similarly manipulated:

$$F(u, v) = e^{-\frac{u^2+v^2}{2\sigma^2}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} 2\pi\sigma^2 e^{-\frac{1}{2\sigma^2}(2\pi\sigma^2x+ju)^2} e^{-\frac{1}{2\sigma^2}(2\pi\sigma^2y+jv)^2} dx dy$$

Perform the following substitution:

$$\begin{aligned} r &= 2\pi\sigma^2x + ju & s &= 2\pi\sigma^2y + jv \\ dr &= 2\pi\sigma^2dx & ds &= 2\pi\sigma^2dy \end{aligned}$$

$$F(u, v) = e^{-\frac{u^2+v^2}{2\sigma^2}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma^2} e^{-\frac{r^2+s^2}{2\sigma^2}} dr ds$$

The integral of a two dimensional Gaussian distribution is exactly unity:

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma^2} e^{-\frac{r^2+s^2}{2\sigma^2}} dr ds = 1$$

$$F(u, v) = e^{-\frac{u^2+v^2}{2\sigma^2}}$$

$$\text{Rect}_{a,b}(x, y) \xleftrightarrow{\mathcal{F}} ab \frac{\sin \pi ua}{\pi ua} \frac{\sin \pi vb}{\pi vb}$$

where

$$\text{Rect}_{a,b}(x, y) \triangleq \begin{cases} 1 & \text{if } |x| \leq a/2 \text{ and } |y| \leq b/2; \\ 0 & \text{otherwise.} \end{cases}$$

$$\cos(2\pi u_0 x + 2\pi v_0 y) \xleftrightarrow{\mathcal{F}} \frac{1}{2} (\delta(u + u_0, v + v_0) + \delta(u - u_0, v - v_0))$$

$$\sin(2\pi u_0 x + 2\pi v_0 y) \xleftrightarrow{\mathcal{F}} j \frac{1}{2} (\delta(u + u_0, v + v_0) - \delta(u - u_0, v - v_0))$$

2 Fourier Transform of Discrete Functions

2.1 Definitions

For functions that are either discrete and finite or discrete and periodic, we use the discrete Fourier transform (DFT):

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

Inverse transform:

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

Define:

$$W_N \triangleq e^{-j\frac{2\pi}{N}}$$

Forward transform:

$$F(u, v) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) W_N^{ux} W_M^{vy}$$

Inverse transform:

$$f(x, y) = \frac{1}{MN} \sum_{u=0}^{M-1} \sum_{v=0}^{N-1} F(u, v) W_N^{-ux} W_M^{-vy}$$

$F(u, v)$ is, in general, a complex function, so it has a rectangular form:

$$F(u, v) = R(u, v) + jI(u, v)$$

where $R(u, v)$ and $I(u, v)$ are real functions of u and v . $F(u, v)$ also has a polar form:

$$F(u, v) = |F(u, v)|e^{-j\phi(u, v)}$$

where

$$|F(u, v)| = (R^2(u, v) + I^2(u, v))^{\frac{1}{2}}$$

and

$$\phi(u, v) = \tan^{-1} \left(\frac{I(u, v)}{R(u, v)} \right)$$

2.2 Properties

Power spectrum:

$$P(u, v) = |F(u, v)|^2 = F(u, v) F^*(u, v) = R^2(u, v) + I^2(u, v)$$

Periodicity:

$$F(u, v) = F(u + pM, v + qN) \text{ for all integers } p \text{ and } q$$

DFT of the complex conjugate:

$$\boxed{f^*(x, y) \xleftrightarrow{\text{DFT}} F^*(-u, -v)}$$

The above DFT pair implies that for real $f(x, y)$, the DFT is conjugate symmetric:

$$\boxed{F(u, v) = F^*(-u, -v)}$$

Conjugation in frequency domain:

$$\boxed{f^*(-x, -y) \xleftrightarrow{\text{DFT}} F^*(u, v)}$$

DC term:

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

Also:

$$F(M/2, N/2) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) (-1)^{x+y}$$

$$F(M/2, 0) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) (-1)^x$$

$$F(0, N/2) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) (-1)^y$$

Linearity:

$$\boxed{af(x, y) + bg(x, y) \xleftrightarrow{\text{DFT}} aF(u, v) + bG(u, v)}$$

So far, we have tacitly assumed that $f(x, y)$ is a spatially finite image and that its region of support is $0 \leq x \leq M - 1$, $0 \leq y \leq N - 1$. Define the periodic extension of $f(x, y)$:

$$\tilde{f}(x, y) \triangleq f(x \bmod M, y \bmod N)$$

The DFT has a dual interpretation:

- The DFT consists of frequency samples of the continuous Fourier transform of a finite image.
- The DFT corresponds to the Fourier series expansion of a periodic signal.

Both interpretations are useful, and we can switch back and forth as long as we are careful.

Spatial translation:

$$\boxed{\tilde{f}(x - x_0, y - y_0) \xleftrightarrow{\text{DFT}} F(u, v) W_M^{ux_0} W_N^{vy_0}}$$

Frequency translation:

$$\boxed{f(x, y) W_M^{-u_0x} W_N^{-v_0y} \xleftrightarrow{\text{DFT}} F(u - u_0, v - v_0)}$$

Proof:

$$F(r, s) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) W_N^{rx} W_M^{sy}$$

Let $r = u - u_0$ and $s = v - v_0$:

$$\begin{aligned} F(u - u_0, v - v_0) &= \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) W_N^{(u-u_0)x} W_M^{(v-v_0)y} \\ &= \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) W_N^{ux} W_N^{-u_0x} W_M^{vy} W_M^{-v_0y} \\ &= \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} \{f(x, y) W_N^{-u_0x} W_M^{-v_0y}\} W_N^{ux} W_M^{vy} \end{aligned}$$

A special case of frequency translation:

$$f(x, y) (-1)^{x+y} \xleftrightarrow{\text{DFT}} F(u - M/2, v - N/2)$$

Separable functions:

$$f_x(x) f_y(y) = f(x, y) \xleftrightarrow{\text{DFT}} F(u, v) = F_x(u) F_y(v)$$

2.3 Convolution and correlation

Linear discrete convolution:

$$f(x, y) * g(x, y) \triangleq \sum_{s=-\infty}^{\infty} \sum_{t=-\infty}^{\infty} f(s, t) g(x - s, y - t)$$

Circular discrete convolution:

$$f(x, y) \circledast g(x, y) = \sum_{s=0}^{M-1} \sum_{t=0}^{N-1} \tilde{f}(s, t) \tilde{g}(x - s, y - t)$$

Both forms of convolution defined above are commutative and distribute over addition:

$$f(x, y) * g(x, y) = g(x, y) * f(x, y)$$

$$f(x, y) * (g(x, y) + h(x, y)) = f(x, y) * g(x, y) + f(x, y) * h(x, y)$$

and

$$f(x, y) \circledast g(x, y) = g(x, y) \circledast f(x, y)$$

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

The unit sample function, $\delta(x, y)$, is defined as:

$$\delta(x, y) \triangleq \begin{cases} 1 & \text{if } x = y = 0; \\ 0 & \text{otherwise.} \end{cases}$$

The unit sample function is the identity element for both forms of discrete convolution:

$$f(x, y) * \delta(x, y) = \delta(x, y) * f(x, y) = f(x, y)$$

and

$$f(x, y) \circledast \delta(x, y) = \delta(x, y) \circledast f(x, y) = f(x, y)$$

The discrete form of the convolution theorem applies only to circular convolution:

$$f(x, y) \circledast g(x, y) \xleftrightarrow{\text{DFT}} F(u, v) G(u, v)$$

$$f(x, y) g(x, y) \xleftrightarrow{\text{DFT}} \frac{1}{MN} F(u, v) \circledast G(u, v)$$

However, the theorem given below gives us a way to calculate the linear convolution using circular convolution:

Suppose $f(x, y)$ is $A \times B$ and $h(x, y)$ is $C \times D$. Suppose further that we zero pad $f(x, y)$ and $h(x, y)$ out to $P \times Q$ and call the new functions $f_e(x, y)$ and $h_e(x, y)$, respectively. Then if $P \geq A + C - 1$ and $Q \geq B + D - 1$, the results of circular and linear convolution match:

$$f_e(x, y) * g_e(x, y) = f_e(x, y) \circledast g_e(x, y) \text{ for } 0 \leq x \leq P - 1 \text{ and } 0 \leq y \leq Q - 1.$$

Correlation:

$$f(x, y) \circ g(x, y) \triangleq \sum_{s=0}^{M-1} \sum_{t=0}^{N-1} f^*(s, t) g(x + s, y + t)$$

Relationship between convolution and correlation:

$$f(x, y) \circ g(x, y) = \frac{1}{MN} f^*(-x, -y) * g(x, y)$$

If $f(x, y)$ and $g(x, y)$ are sufficiently zero padded so that linear and circular convolution correspond, then

$$f(x, y) \circ g(x, y) \xleftrightarrow{\text{DFT}} \frac{1}{MN} F^*(u, v) G(u, v),$$

$$f^*(x, y) g(x, y) \xleftrightarrow{\text{DFT}} F(u, v) \circ G(u, v).$$

2.4 Example 1

Find the frequency response of the filter with the following mask:

1	0	0
0	0	0
0	0	1

$$g(x, y) = f(x + 1, y + 1) + f(x - 1, y - 1)$$

$$G(u, v) = F(u, v) W_M^{-u} W_N^{-v} + F(u, v) W_M^u W_N^v$$

Frequency response:

$$\begin{aligned} H(u, v) &= \frac{G(u, v)}{F(u, v)} \\ &= W_M^{-u} W_N^{-v} + W_M^u W_N^v \\ &= e^{j2\pi(u/M+v/N)} + e^{-j2\pi(u/M+v/N)} \\ &= 2 \cos(2\pi(u/M + v/N)) \end{aligned}$$

Magnitude and phase of frequency response:

$$|H(u, v)| = 2 |\cos(2\pi(u/M + v/N))|$$

$$\phi(u, v) = 0$$

2.5 Example 2

Find the frequency response of the filter with the following mask:

-1	0	0
0	0	0
0	0	1

$$g(x, y) = f(x + 1, y + 1) - f(x - 1, y - 1)$$

$$G(u, v) = F(u, v)W_M^{-u}W_N^{-v} - F(u, v)W_M^uW_N^v$$

Frequency response:

$$\begin{aligned} H(u, v) &= \frac{G(u, v)}{F(u, v)} \\ &= W_M^{-u}W_N^{-v} - W_M^uW_N^v \\ &= e^{j2\pi(u/M+v/N)} - e^{-j2\pi(u/M+v/N)} \\ &= j2 \sin(2\pi(u/M + v/N)) \end{aligned}$$

Magnitude and phase of frequency response:

$$|H(u, v)| = 2|\sin(2\pi(u/M + v/N))|$$

$$\phi(u, v) = \begin{cases} \pi/2 & \text{if } \sin(2\pi(u/M + v/N)) > 0 \\ -\pi/2 & \text{if } \sin(2\pi(u/M + v/N)) < 0 \end{cases}$$

3 Relationship Between Discrete and Continuous Fourier Transforms

Define the following transform pairs:

$$f_c(x, y) \xleftrightarrow{\mathcal{F}} F_c(u_c, v_c)$$

$$f(x, y) \xleftrightarrow{\mathcal{DFT}} F(u, v)$$

To simplify this discussion, let us assume that the sampling rate is exactly unity; i.e. $f_s = T_s = 1$. Suppose $f_c(x, y)$ has a finite region of support; specifically, suppose $f_c(x, y) = 0$ for $x \leq -1$, $x \geq M$, $y \leq -1$, and $y \geq M$.

Let $f(x, y) = f_c(x, y)$ for integers x and y in the intervals $0 \leq x \leq M - 1$ and $0 \leq y \leq N - 1$. Then

$$F(u, v) = \sum_{k=-\infty}^{\infty} \sum_{l=-\infty}^{\infty} F_c(u/M + k, v/N + l).$$

Suppose $F_c(u, v) \approx 0$ for $|u| > 1/2$, and $|v| > 1/2$, then for integers u and v in the intervals $-M/2 < u < M/2$ and $-N/2 < v < N/2$,

$$F(u, v) \approx F_c(u/M, v/N).$$

Since a function cannot be both spatially finite and band limited, the above expression can only be approximately satisfied.