

2-D MATH OPERATIONS

CONTINUOUS 2-D FUNCTIONS

2-D MATH OPERATIONS

2-D CONVOLUTION

Definition

- **Extension of 1-D convolution**

$$g(x, y) = f(x, y) * * h(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(x - \alpha, y - \beta) d\alpha d\beta$$

Properties

- **generally same as 1-D convolution (commutative, distributive, associative)**

Define commutative, distributive, associative operations

2-D MATH OPERATIONS

Simplifies for separable functions

- **General case** $g(x, y) = f(x, y) * * h(x, y)$
- **Case I: *h* (or *f*) separable**
$$g(x, y) = f(x, y) * * [h_1(x)h_2(y)]$$
$$= f(x, y) * h_1(x) * h_2(y)$$



two 1-D
convolutions

2-D MATH OPERATIONS

- **Case II: f and h separable**

$$\begin{aligned}g(x, y) &= [f_1(x) * h_1(x)][f_2(y) * h_2(y)] \\ &= g_1(x)g_2(y)\end{aligned}$$

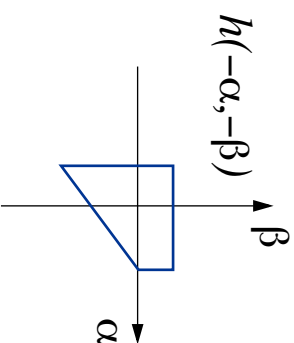
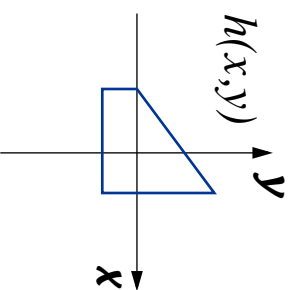
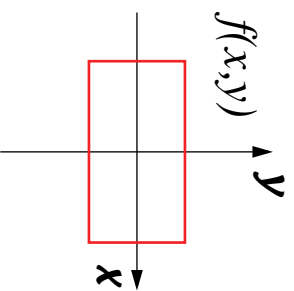


$g(x, y)$ is also
separable

2-D MATH OPERATIONS

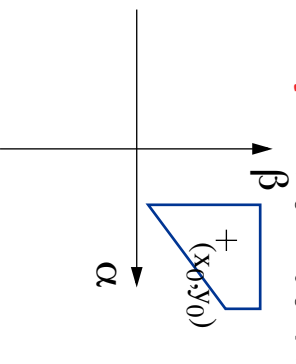
Example

- For simple graphics, let f and h be binary “mask” functions, equal to one within the boundary and zero elsewhere

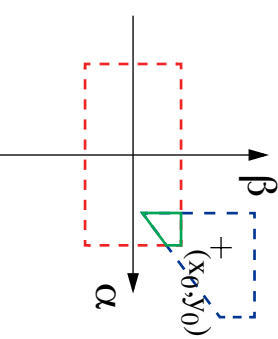


flip (rotate -180°)

shift $h(x_0 - \alpha, y_0 - \beta)$

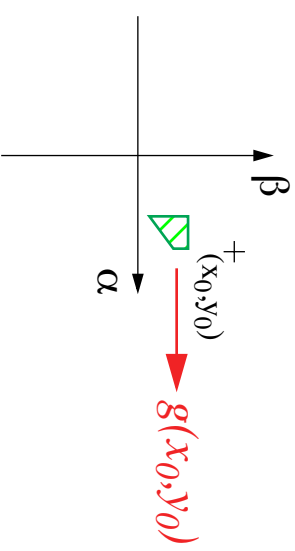


multiply $f(\alpha, \beta)h(x_0 - \alpha, y_0 - \beta)$



integrate

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta)h(x_0 - \alpha, y_0 - \beta) d\alpha d\beta$$



2-D MATH OPERATIONS

2-D CORRELATION

Identical to 2-D convolution, except neither function is “flipped”

$$g(x, y) = f(x, y) \star\star h(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(\alpha - x, \beta - y) d\alpha d\beta$$

Important difference

- **not commutative**

$$f \star\star h \neq h \star\star f$$

2-D MATH OPERATIONS

2-D FOURIER TRANSFORMS

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

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

Properties

- 1-D transform properties generally also apply to 2-D transform

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

- Useful property:

2-D MATH OPERATIONS

2-D Fourier Transform of Separable Functions

$$f(x, y) = f_1(x)f_2(y)$$

$$F_1(u) = \int_{-\infty}^{\infty} f_1(x)e^{-j2\pi xu} dx, \quad F_2(v) = \int_{-\infty}^{\infty} f_2(y)e^{-j2\pi yv} dy$$

$$\begin{aligned} F(u, v) &= \int_{-\infty}^{\infty} e^{-j2\pi yv} \left(\int_{-\infty}^{\infty} f(x, y)e^{-j2\pi xu} dx \right) dy \\ &= \int_{-\infty}^{\infty} f_2(y)e^{-j2\pi yv} \left(\int_{-\infty}^{\infty} f_1(x)e^{-j2\pi xu} dx \right) dy = F_1(u) \int_{-\infty}^{\infty} f_2(y)e^{-j2\pi yv} dy \\ &= F_1(u)F_2(v) \end{aligned}$$

- Product of two 1-D transforms
- Note the **difference from convolution theorem!**

2-D MATH OPERATIONS

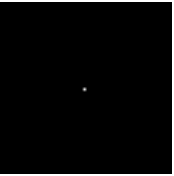
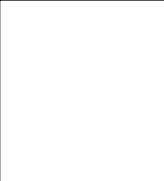

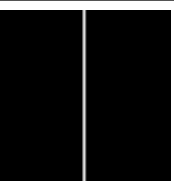

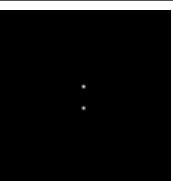
2-D Fourier Transform Properties

name	$f(x,y)$	$F(u,v)$
separability	$f_1(x)f_2(y)$	$F_1(u)F_2(v)$
scaling	$f(x/a, y/b)$	$ a b F(au, bv)$
shifting	$f(x \pm a, y \pm b)$	$e^{\pm j2\pi(au + bv)}F(u, v)$
linearity (superposition)	$af_1(x, y) + bf_2(x, y)$	$aF_1(u, v) + bF_2(u, v)$
convolution	$f_1(x, y) * * f_2(x, y)$	$F_1(u, v) \cdot F_2(u, v)$
	$f_1(x, y) \cdot f_2(x, y)$	$F_1(u, v) * * F_2(u, v)$
affine	$f(ax + by + c, dx + ey + f)$	$\frac{1}{ \Delta } e^{-j\left(\frac{2\pi}{\Delta}\right)[(bf - ec)u + (dc - af)v]}$ $F\left(\frac{eu - dv - bu + av}{\Delta}, \frac{eu - dv - bu + av}{\Delta}\right)$, where $\Delta = ae - db$

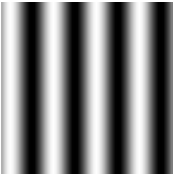
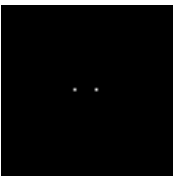

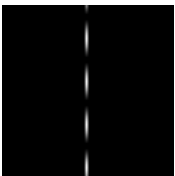

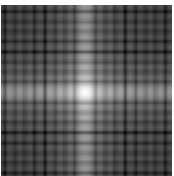
Show that the affine property reduces to the shifting property or the scaling property with appropriate parameters

2-D MATH OPERATIONS

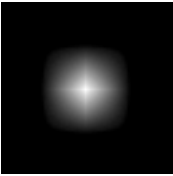
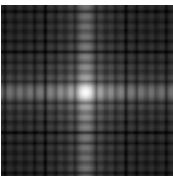

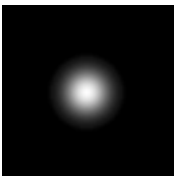
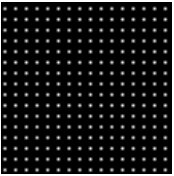

2-D Fourier Transform Pairs

<i>descriptive name of $f(x,y)$</i>	<i>picture of $f(x,y)$</i>	$f(x,y)$	$F(u,v)$	<i>descriptive name of $F(u,v)$</i>	<i>picture of $F(u,v)$</i>
delta		$\delta(x, y)$	1	constant	
blade		$\delta(x)$	$\delta(v)$	blade	
ripple		$\cos(2\pi ax)$	$\frac{1}{2 a } \delta\left(\frac{u}{a}\right) \delta(v)$	double delta	

2-D MATH OPERATIONS

<i>descriptive name of $f(x,y)$</i>	<i>picture of $f(x,y)$</i>	<i>$f(x,y)$</i>	<i>$F(u,v)$</i>	<i>descriptive name of $F(u,v)$</i>	<i>picture of $F(u,v)$</i>
<i>ripple</i>		$\cos(2\pi ay)$	$\frac{1}{2 a } \delta\delta\left(\frac{v}{a}\right) \delta(u)$	<i>double delta</i>	
<i>double blade</i>		$\frac{1}{2 a } \delta\delta\left(\frac{x}{a}\right)$	$\cos(2\pi au) \delta(v)$		
<i>box</i>		$rect(x, y)$	$sinc(u, v)$		

2-D MATH OPERATIONS

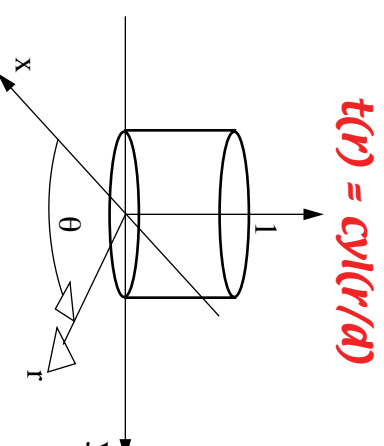
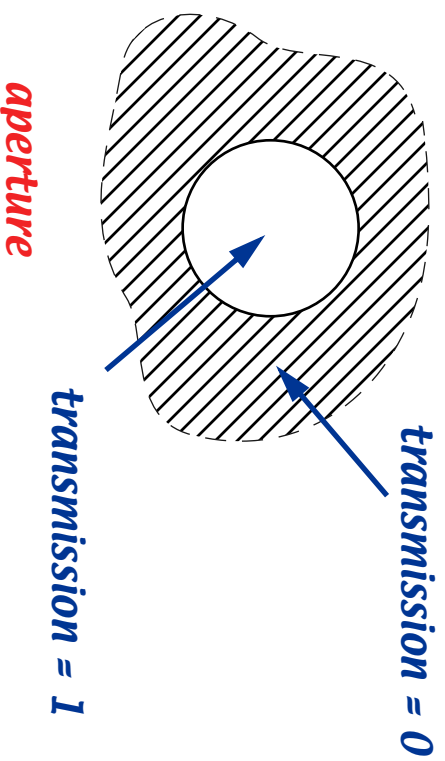
<i>descriptive name of $f(x,y)$</i>	<i>picture of $f(x,y)$</i>	<i>$f(x,y)$</i>	<i>$F(u,v)$</i>	<i>descriptive name of $F(u,v)$</i>	<i>picture of $F(u,v)$</i>
		$tri(x, y)$	$\text{sinc}^2(u, v)$		
		$gaus(x, y)$	$gaus(u, v)$		
<i>grid or bed-of-nails</i>		$comb(x, y)$	$comb(u, v)$	<i>grid or bed-of-nails</i>	

2-D MATH OPERATIONS

2-D FOURIER TRANSFORMS IN POLAR COORDINATES

Useful for radial functions

- $\text{cyl}(r/d)$ describes the transmission function of a circular optical aperture



2-D MATH OPERATIONS

Derive a special Fourier transform for such functions

$$\begin{aligned} F(u, v) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi(xu + yv)} dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\sqrt{x^2 + y^2}) e^{-j2\pi(xu + yv)} dx dy \end{aligned}$$

- **Substitute:**

$$\begin{array}{ll} x = r \cos \theta & u = \rho \cos \phi \\ y = r \sin \theta & v = \rho \sin \phi \end{array}$$

space domain *spatial frequency domain*

$$\begin{array}{ll} \text{where } r = \sqrt{x^2 + y^2} & \text{and } \theta = \text{atan}(y/x) \\ \rho = \sqrt{u^2 + v^2} & \phi = \text{atan}(v/u) \end{array}$$

2-D MATH OPERATIONS

Hankel Transform

- 2-D Fourier transform in polar coordinates

$$\begin{aligned} F(\rho \cos \phi, \rho \sin \phi) &= \int_0^{\infty} r dr \int_0^{2\pi} d\theta f(r) e^{-j2\pi\rho r \cos(\theta - \phi)} \\ &= \int_0^{\infty} f(r) r dr [2\pi J_0(2\pi\rho r)] \end{aligned}$$

where $J_0(\cdot)$ is the zero-order Bessel function of the first kind.

- Simplifying, we have the **zero-order Hankel transform**,

$$F(\rho \cos \phi, \rho \sin \phi) = 2\pi \int_0^{\infty} f(r) J_0(2\pi\rho r) r dr$$

- If $f(r, \theta) = f(r)$ (**circularly symmetric**), then $F(\rho, \phi) = F(\rho)$

2-D MATH OPERATIONS

Inverse Hankel transform

$$f(r) = 2\pi \int_0^{\infty} F(\rho) J_0(2\pi r \rho) \rho d\rho$$

2-D MATH OPERATIONS

Hankel Transform Properties

<i>name</i>	<i>$f(r)$</i>	<i>$F(\rho)$</i>
<i>scaling</i>	$f\left(\frac{r}{b}\right)$	$ b ^2 F(b\rho)$
<i>convolution</i>	$f(r) * h(r)$	$F(\rho)H(\rho)$

2-D MATH OPERATIONS

Hankel Transform Pairs

$f(r)$	$F(\rho)$
$\frac{\delta(r)}{\pi r}$	1
$1/r$	$1/\rho$
$cyl(r)$	$\frac{\pi}{4} somb(\rho)$
e^{-r}	$\frac{2\pi}{(4\pi^2 \rho^2 + 1)^{3/2}}$
$gaus(r)$	$gaus(\rho)$

2-D MATH OPERATIONS

- **Summary - continuous 2-D functions**

case	convolution	Fourier transform
general	$f(x, y) * * h(x, y)$	$F(u, v)H(u, v)$
h separable	$f(x, y) * h_1(x) * h_2(y)$	$F(u, v)H_1(u)H_2(v)$
both separable	$[f_1(x) * h_1(x)][f_2(y) * h_2(y)]$	$F_1(u)H_1(u)F_2(v)H_2(v)$
radial functions	$f(r) * * h(r)$	$F(\rho)H(\rho)$

2-D MATH OPERATIONS

DISCRETE 2-D FUNCTIONS

2-D MATH OPERATIONS

2-D CONVOLUTION

Definition

- **Extension of 1-D convolution**

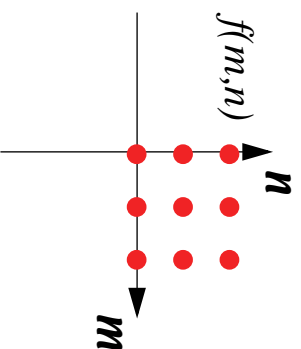
$$g(m, n) = f(m, n) * * h(m, n) = \sum_k \sum_l f(k, l)h(m-k, n-l)$$

Properties

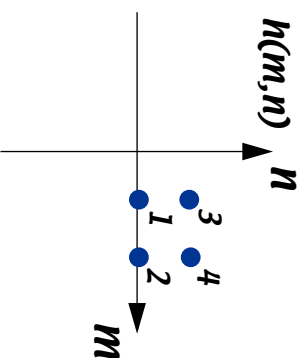
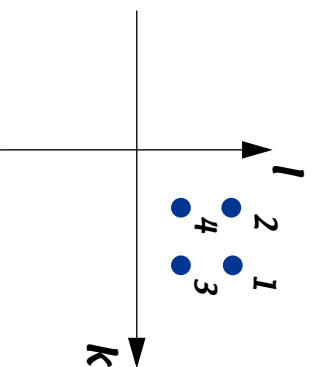
- **generally same as 1-D convolution (commutative, distributive, associative)**

2-D MATH OPERATIONS

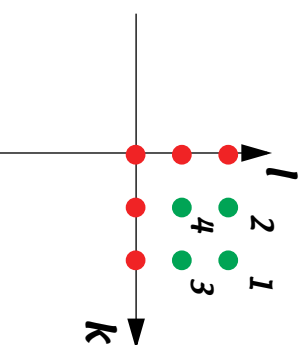
Example



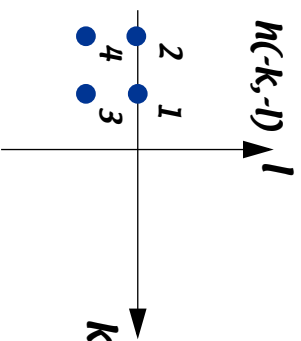
shift $h(3-k, 2-l)$



multiply $f(k,l)h(3-k, 2-l)$

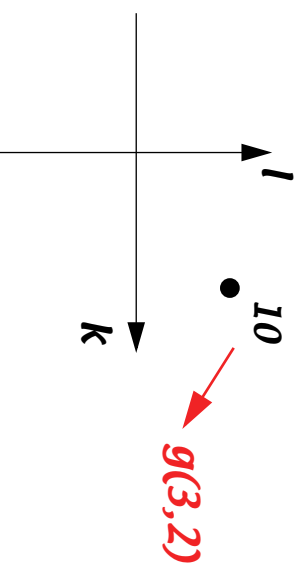


Flip (rotate -180°)



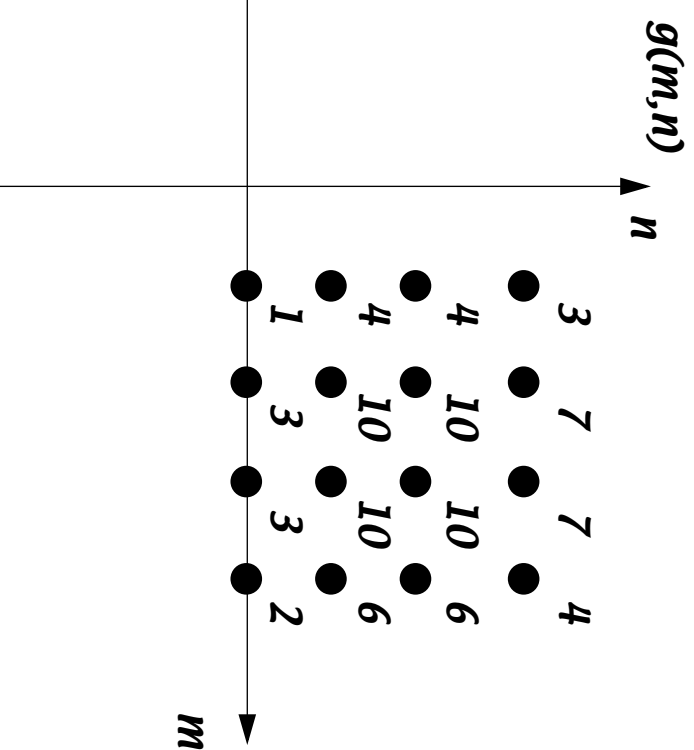
sum

$$\sum_k \sum_l f(k, l)h(3-k, 2-l)$$



2-D MATH OPERATIONS

Final result



- If f is $M \times N$ and h is $K \times L$, then g is $(M + K - 1) \times (N + L - 1)$
- Note phase shift in m in $g(m,n)$ due to linear phase in $h(m,n)$

2-D MATH OPERATIONS

Convolution with separable h

$$\begin{aligned}g(m, n) &= \sum_k \sum_l f(k, l)h(m-k, n-l) \\ &= \sum_k \sum_l f(k, l)h_1(m-k)h_2(n-l) \\ &= \sum_k h_1(m-k) \sum_l f(k, l)h_2(n-l)\end{aligned}$$

- **inner sum is 1-D convolution between f and h_2**

array size $\rightarrow M \times (N + L - 1)$

- **outer sum is 1-D convolution between result of inner sum and h_1**

array size $\rightarrow (M + K - 1) \times (N + L - 1)$

For $N \times N$ image and $M \times M$ filter, show that computational advantage of the separable filter is $M/2$