6.3000: Signal Processing

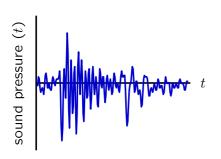
2D Fourier Transforms 1

- Introduction to 2D Signal Processing
- 2D Fourier Representations

Signals

Signals are functions that are used to convey information. – may have 1 or 2 or 3 or even more **independent variables**

U



brightness (x,y)



x

A 1D signal has a one-dimensional domain.

We have usually thought of the domain as time t or discrete time n.

A 2D signal has a two-dimensional domain. We will usually think of the domain as x and y or n_x and n_y .

Fourier Representations

From "Continuous Time" to "Continuous Space."

One dimensional CTFT:

$$F(\omega) = \int_{-\infty}^{\infty} f(t) e^{-j\omega t} dt$$
$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} F(\omega) e^{j\omega t} d\omega$$

Two dimensional CTFT:

$$F(\omega_x, \omega_y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j(\omega_x x + \omega_y y)} dx dy$$
$$f(x, y) = \frac{1}{4\pi^2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(\omega_x, \omega_y) e^{j(\omega_x x + \omega_y y)} d\omega_x d\omega_y$$

integrals \rightarrow double integrals; sum of x and y exponents in kernel function.

Fourier Representations

From "Discrete Time" to "Discrete Space."

One dimensional DTFT:

$$F(\Omega) = \sum_{n=-\infty}^{\infty} f[n] e^{-j\Omega n}$$
$$f[n] = \frac{1}{2\pi} \int_{2\pi} F(\Omega) e^{j\Omega n} d\Omega$$

Two dimensional DTFT:

$$F(\Omega_x, \Omega_y) = \sum_{n_x = -\infty}^{\infty} \sum_{n_y = -\infty}^{\infty} f[n_x, n_y] e^{-j(\Omega_x n_x + \Omega_y n_y)}$$
$$f[n_x, n_y] = \frac{1}{4\pi^2} \int_{2\pi} \int_{2\pi} F(\Omega_x, \Omega_y) e^{j(\Omega_x n_x + \Omega_y n_y)} d\Omega_x d\Omega_y$$

double integrals; double sums; sum of x and y exponents in kernel function.

 \mathcal{D}_{u}

Fourier Representations

From 1D DFT to 2D DFT.

One dimensional DFT:

$$F[k] = \frac{1}{N} \sum_{n=0}^{N-1} f[n] e^{-j\frac{2\pi k}{N}n}$$
$$f[n] = \sum_{k=0}^{N-1} F[k] e^{j\frac{2\pi k}{N}n}$$

Two dimensional DFT:

$$F[k_x, k_y] = \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x - 1} \sum_{n_y=0}^{N_y - 1} f[n_x, n_y] e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} n_y\right)}$$

$$f[n_x, n_y] = \sum_{k_x=0}^{N_x-1} \sum_{k_y=0}^{N_y-1} F[k_x, k_y] e^{j\left(\frac{2\pi k_x}{N_x}n_x + \frac{2\pi k_y}{N_y}n_y\right)}$$

double sums; sum of x and y exponents in kernel function.

Importance of Orthogonality

Fourier series represent periodic signals as weighted sum of **basis functions**.

$$f[n] = \sum_{k=0}^{N-1} F[k] e^{j\frac{2\pi}{N}kn}$$

We "sifted" out the $l^{\rm th}$ component by multiplying both sides by $e^{-j\frac{2\pi}{N}ln}$ and summing over a period.

$$\sum_{n=0}^{N-1} f[n]e^{-j\frac{2\pi}{N}ln} = \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} F[k]e^{j\frac{2\pi}{N}kn}e^{-j\frac{2\pi}{N}ln} = \sum_{k=0}^{N-1} F[k] \sum_{n=0}^{N-1} e^{j\frac{2\pi}{N}(k-l)n}$$
$$= \sum_{k=0}^{N-1} F[k]N\delta\Big[(k-l) \mod N\Big] = NF[l]$$

This sifting provided an explicit "analysis" formula for the coefficients:

$$F[k] = \frac{1}{N} \sum_{n=0}^{N-1} f[n] e^{-j\frac{2\pi}{N}kn}$$

Orthogonality of the basis functions is key to Fourier decomposition.

Orthogonality

The form of the 2D Fourier kernel preserves orthogonality.

1D DFT basis functions:
$$\phi_k[n] = e^{jrac{2\pi}{N}kn}$$

"Inner product" of 1D basis functions:

$$\sum_{n} \phi_{k}^{*}[n]\phi_{l}[n] = \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}kn} e^{j\frac{2\pi}{N}ln} = \sum_{n=0}^{N-1} e^{-j\frac{2\pi}{N}(k-l)n} = N\delta\Big[(k-l) \mod N\Big]$$

Orthogonality

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2D DFT basis functions:
$$\phi_{kx,ky}[n_x,n_y] = e^{j\frac{2\pi}{N_x}k_xn_x}e^{j\frac{2\pi}{N_y}k_yn_y}$$

"Inner product" of 2D basis functions:

$$\sum_{n_x, n_y} \phi_{kx, ky}^*[n_x, n_y] \phi_{lx, ly}[n_x, n_y] = \sum_{n_x, n_y} e^{-j\left(\frac{2\pi}{N_x}k_x n_x + \frac{2\pi}{N_y}k_y n_y\right)} e^{j\left(\frac{2\pi}{N_x}l_x n_x + \frac{2\pi}{N_y}l_y n_y\right)}$$

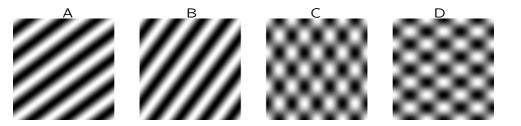
$$= \left(\sum_{n_x} e^{-j\frac{2\pi}{N_x}(k_x - l_x)n_x}\right) \left(\sum_{n_y} e^{-j\frac{2\pi}{N_y}(k_y - l_y)n_y}\right)$$
$$= N_x N_y \delta\left[(k_x - l_x) \mod N_x\right] \delta\left[(k_y - l_y) \mod N_y\right]$$

Check Yourself

The 2D Fourier basis functions have the following form.

$$\phi_{kx,ky}[n_x,n_y] = e^{j\left(\frac{2\pi kx}{Nx}n_x + \frac{2\pi ky}{Ny}n_y\right)}$$

Which (if any) of the following images show the real part of one of the basis functions $\phi_{k_x,k_y}[n_x, n_y]$?



What values of k_x and k_y correspond to basis function?

Check Yourself

The 2D Fourier basis functions have the following form.

$$\phi_{kx,ky}[n_x, n_y] = e^{j\left(\frac{2\pi k_x}{N_x}n_x + \frac{2\pi k_y}{N_y}n_y\right)}$$
$$= \cos\left(\frac{2\pi k_x}{N_x}n_x + \frac{2\pi k_y}{N_y}n_y\right) + j\sin\left(\frac{2\pi k_x}{N_x}n_x + \frac{2\pi k_y}{N_y}n_y\right)$$

If $\frac{2\pi k_x}{N_x}n_x + \frac{2\pi ky}{N_y}n_y$ is constant, the real part (i.e., cosine term) will be constant.

This constraint defines a straight line: $n_y = mn_x + b$.

$$n_y = -\frac{\frac{2\pi kx}{N_x}}{\frac{2\pi ky}{N_y}}n_x + \frac{b}{\frac{2\pi ky}{N_y}} = -\frac{k_x/N_x}{k_y/N_y}n_x + b'$$

For each row in panel A, $k_x = \pm 3$.

Similarly, for each column in panel A, $k_y = \pm 4$.

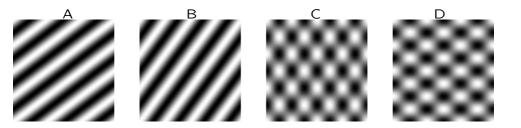
There are two possible solutions for panel A: $k_x = 3, k_y = -4$ and $k_x = -3, k_y = 4$. For panel B, $k_x = 4, k_y = -3$ and $k_x = -4, k_y = 3$. Panels C and D do not correspond to basis functions.

Check Yourself

The 2D Fourier basis functions have the following form.

$$\phi_{kx,ky}[n_x, n_y] = e^{j\left(\frac{2\pi kx}{N_x}n_x + \frac{2\pi ky}{N_y}n_y\right)}$$

Which (if any) of the following images show the real part of one of the basis functions $\phi_{k_x,k_y}[n_x,n_y]$? A and B



What values of k_x and k_y correspond to basis function?

A: (3,-4) or (-3,4); B: (4,-3) or (-4,3); C: none; D: none

Finding a 2D DFT.

Example: Find the DFT of a **2D unit sample**.

$$f_0[n_x, n_y] = \delta[n_x]\delta[n_y] = \begin{cases} 1 & n_x = 0 \text{ and } n_y = 0\\ 0 & \text{otherwise} \end{cases}$$

$$F_0[k_x, k_y] = \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x-1} \sum_{n_y=0}^{N_y-1} \delta[n_x] \delta[n_y] e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} n_y\right)}$$
$$= \frac{1}{N_x N_y} \sum_{n_x=0} \sum_{n_y=0} e^{-j\left(\frac{2\pi k_x}{N_x} 0 + \frac{2\pi k_y}{N_y} 0\right)}$$
$$= \frac{1}{N_x N_y}$$
$$\delta[n_x] \delta[n_y] \stackrel{\text{DFT}}{\Longrightarrow} \frac{1}{N_x N_y}$$

This is a perfectly fine way to compute a Fourier Transform. But there are other methods that provide additional insights.

Alternatively, implement a 2D DFT as a sequence of 1D DFTs.

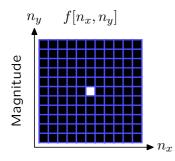
$$F[k_x, k_y] = \frac{1}{N_x N_y} \sum_{n_y=0}^{N_y-1} \sum_{n_x=0}^{N_x-1} f[n_x, n_y] e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} n_y\right)}$$
$$= \frac{1}{N_y} \sum_{n_y=0}^{N_y-1} \underbrace{\left(\frac{1}{N_x} \sum_{n_x=0}^{N_x-1} f[n_x, n_y] e^{-j\frac{2\pi k_x}{N_x} n_x}\right)}_{\text{first take DFTs of rows}} e^{-j\frac{2\pi k_y}{N_y} n_y}$$

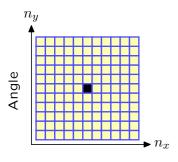
Start with a 2D function of space $f[n_x, n_y)$.

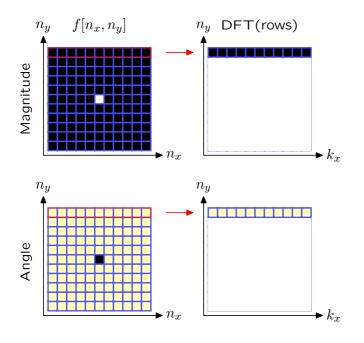
- Replace each row by the DFT of that row.
- Replace each column by the DFT of that column.

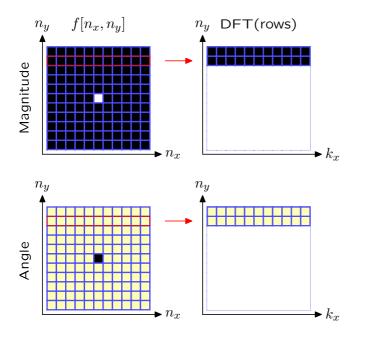
The result is $F[k_x, k_y]$, the 2D DFT of $f[n_x, n_y]$.

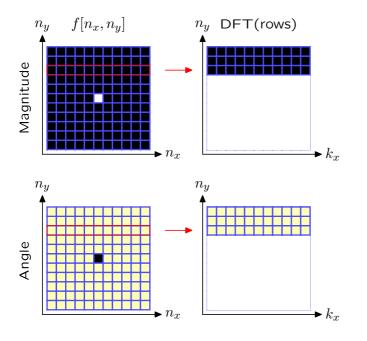
Could just as well start with columns and then do rows.

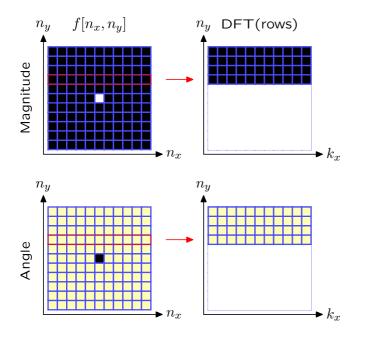


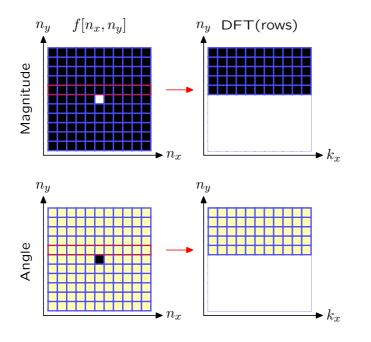


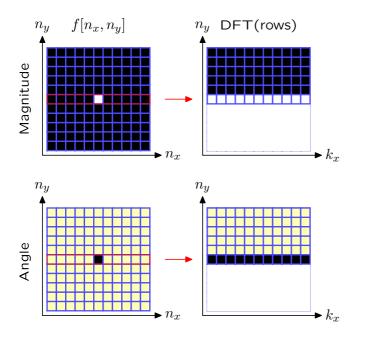


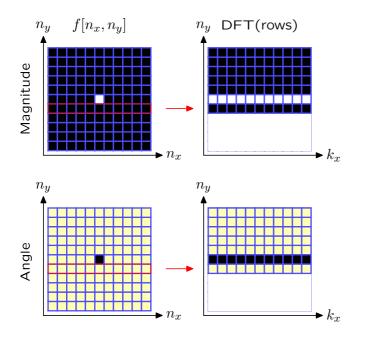


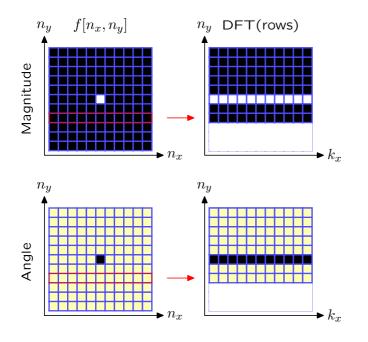


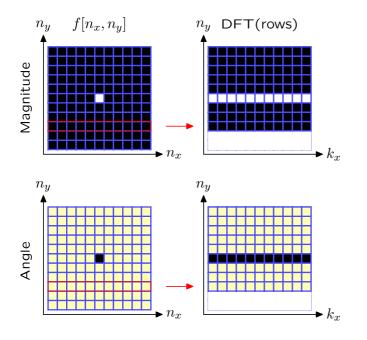


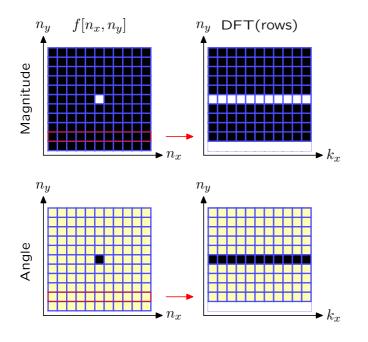


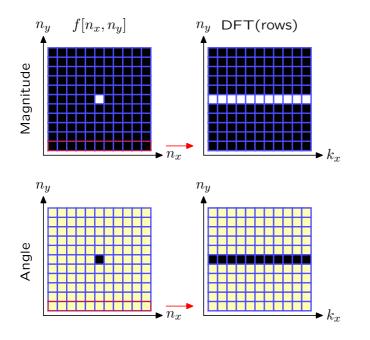


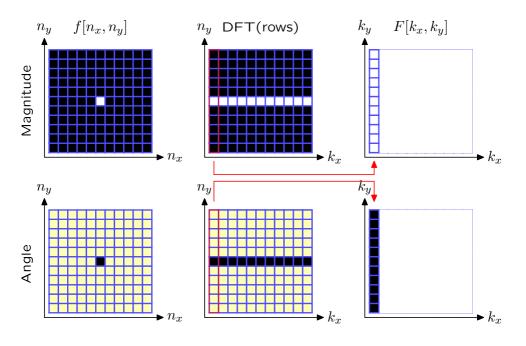


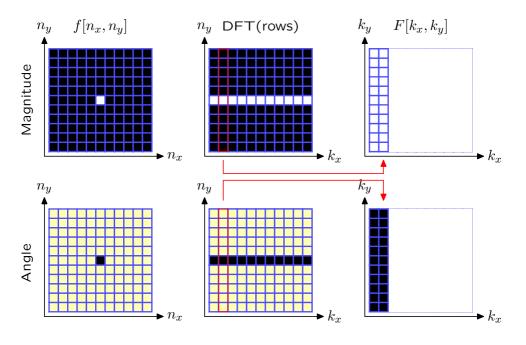


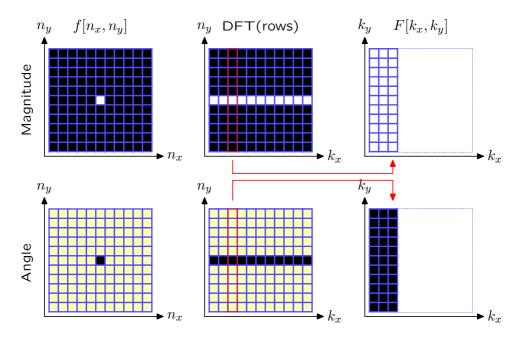


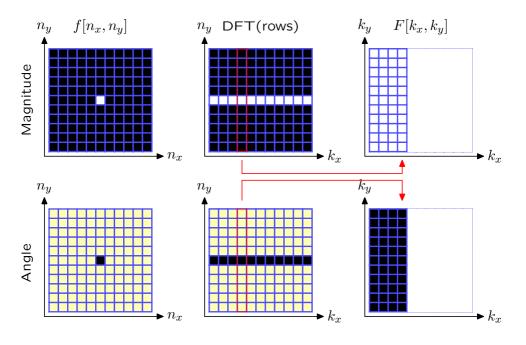


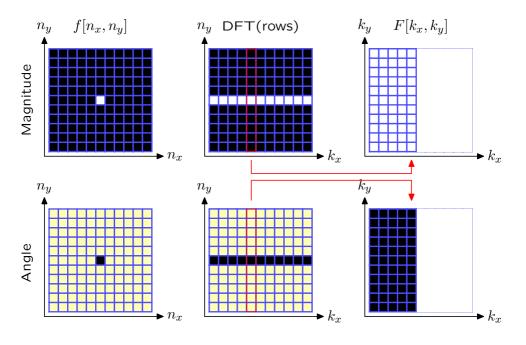


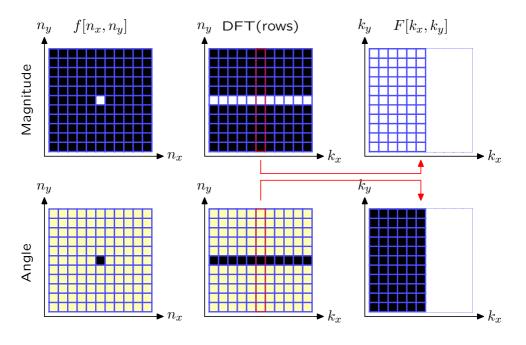


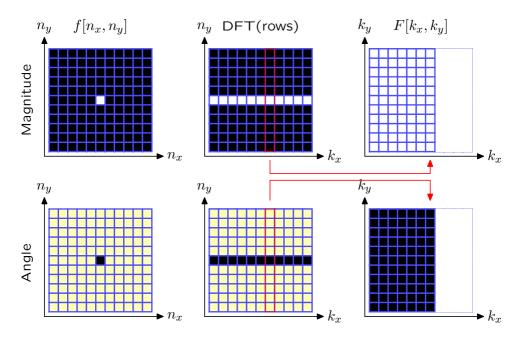


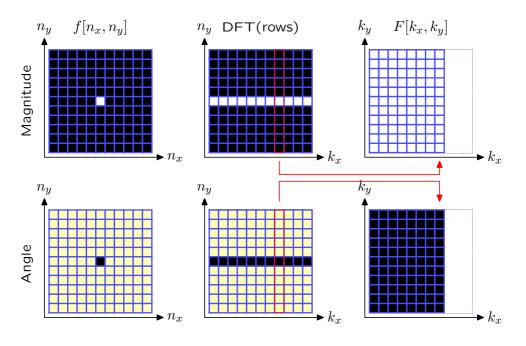


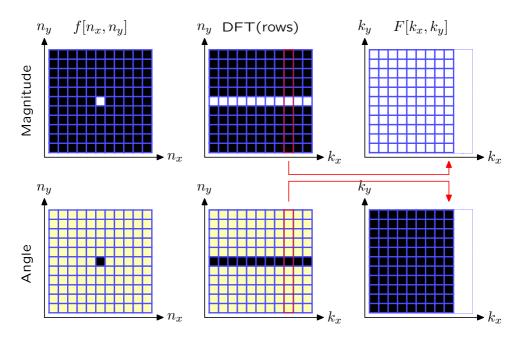


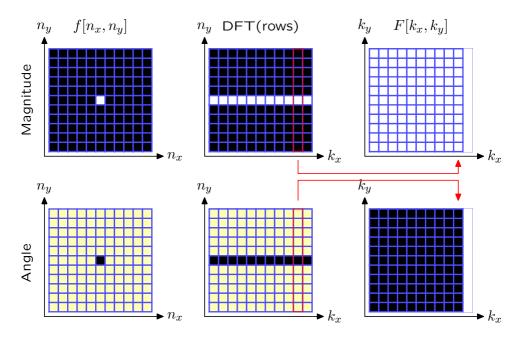


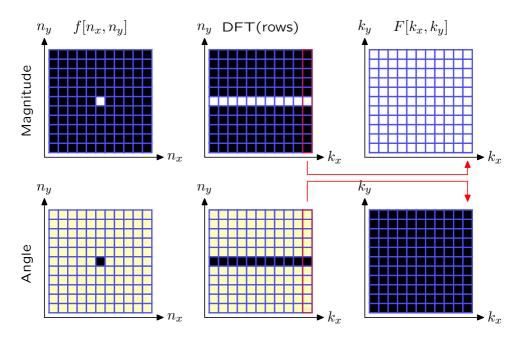




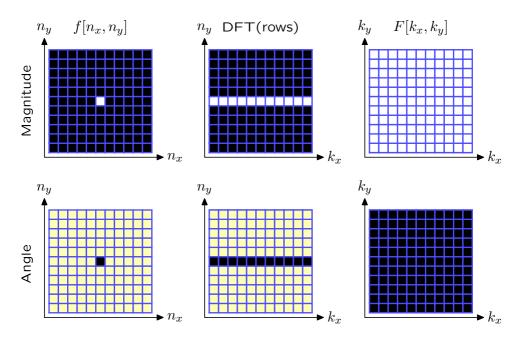




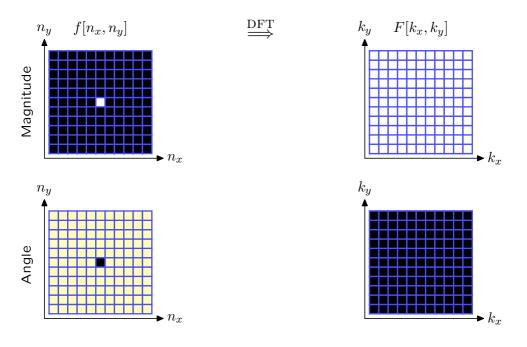




Example: Find the DFT of a 2D unit sample.



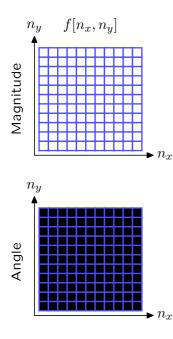
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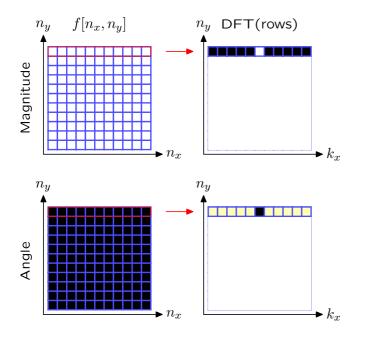


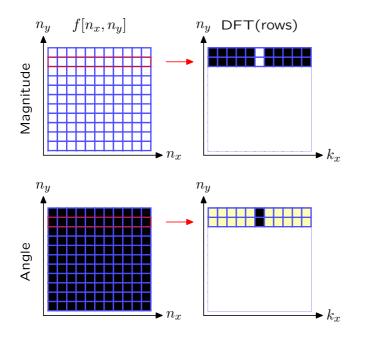
Example: Find the DFT of a constant.

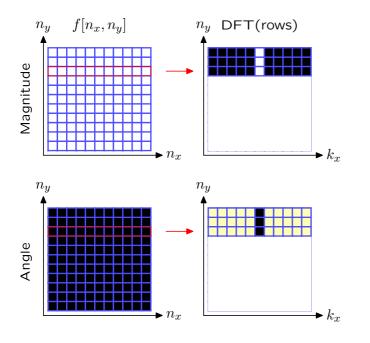
$$\begin{split} f_1[n_x, n_y] &= 1\\ F_1[k_x, k_y] &= \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x-1} \sum_{n_y=0}^{N_y-1} e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} n_y\right)}\\ &= \left(\frac{1}{N_x} \sum_{n_x=0}^{N_x-1} e^{-j\frac{2\pi k_x}{N_x} n_x}\right) \left(\frac{1}{N_y} \sum_{n_y=0}^{N_y-1} e^{-j\frac{2\pi k_y}{N_y} n_y}\right)\\ &= \delta[k_x] \delta[k_y] \end{split}$$

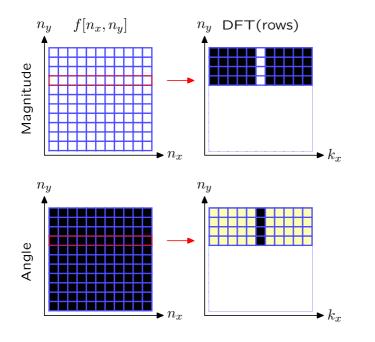
 $1 \stackrel{\text{DFT}}{\Longrightarrow} \delta[k_x]\delta[k_y]$

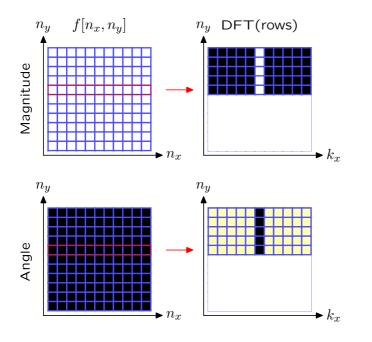


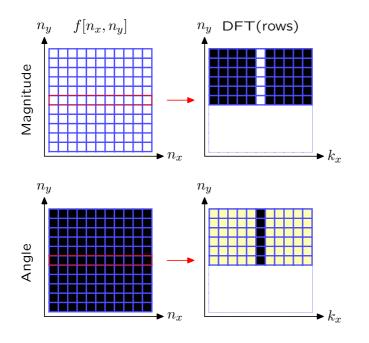


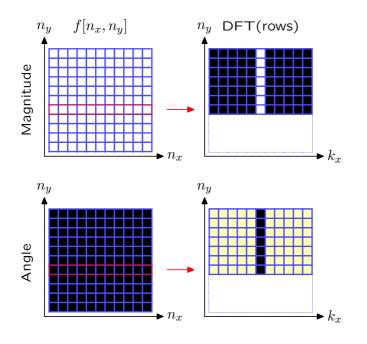


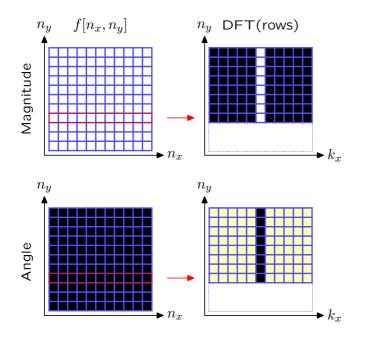


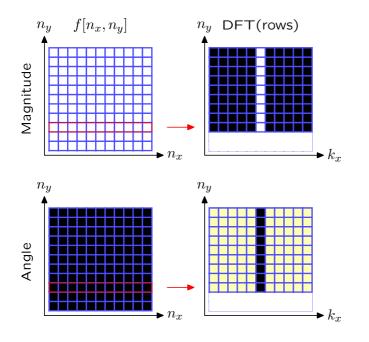


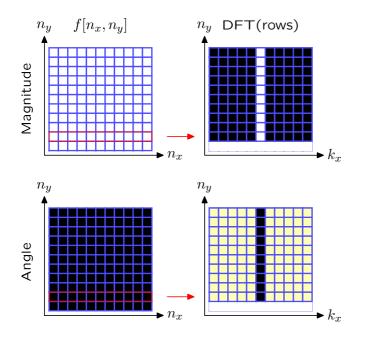


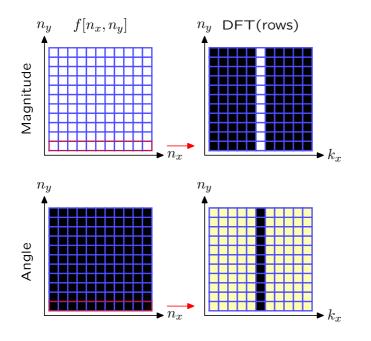


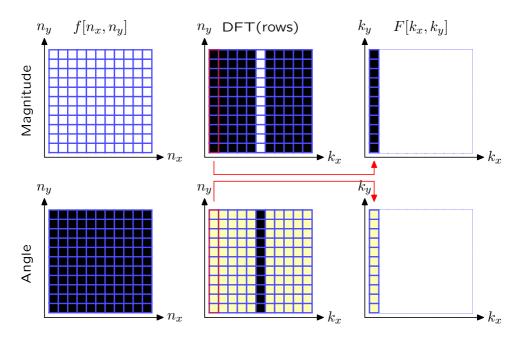


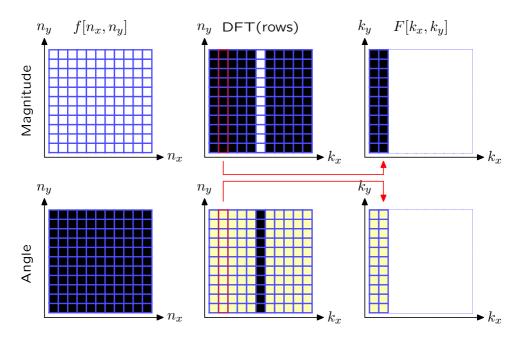


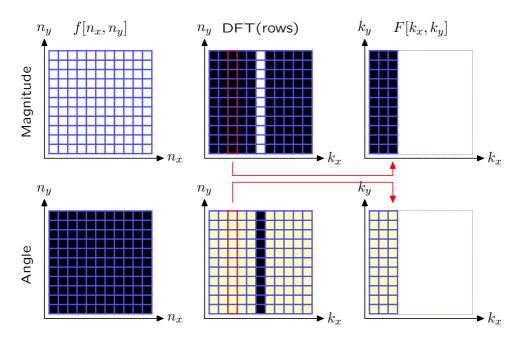


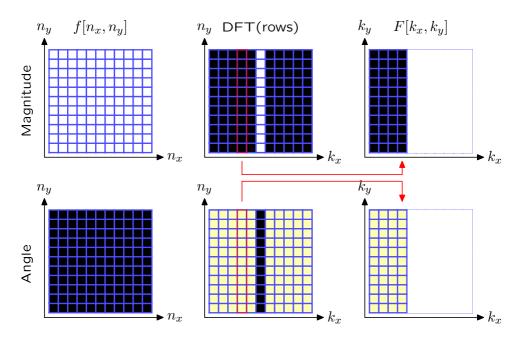


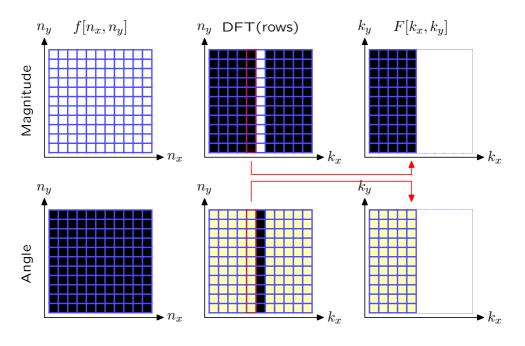


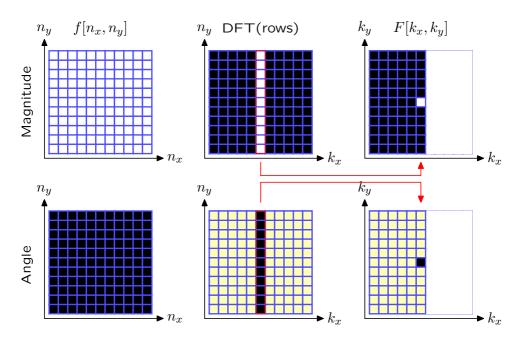


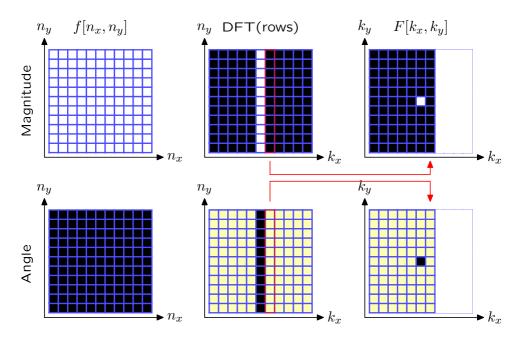


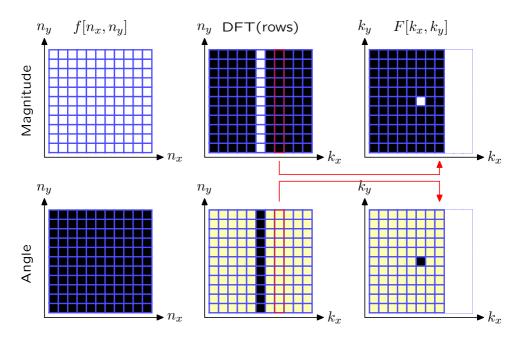


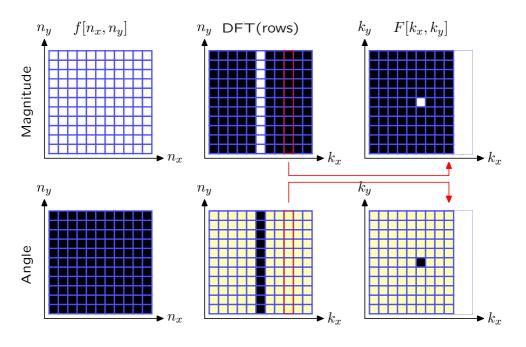


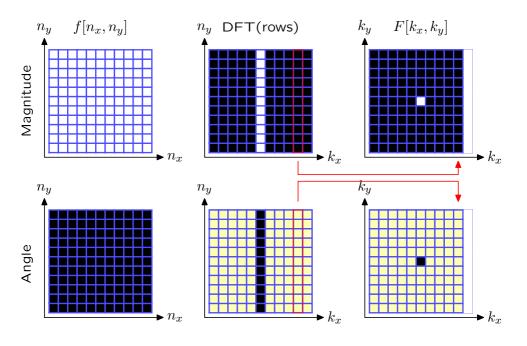


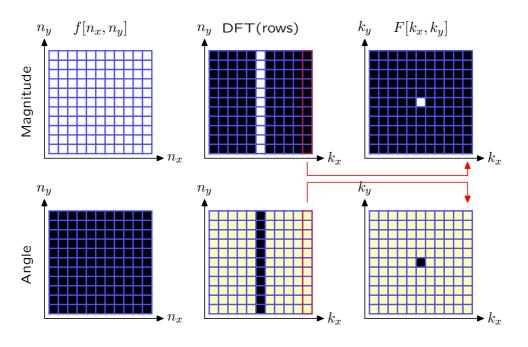


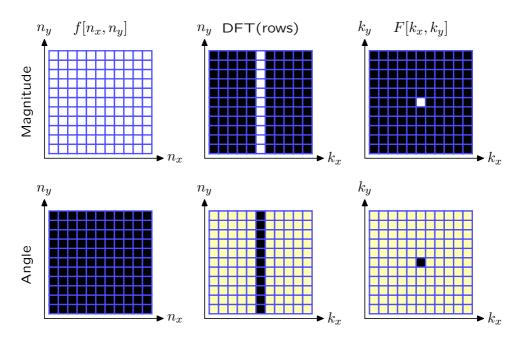


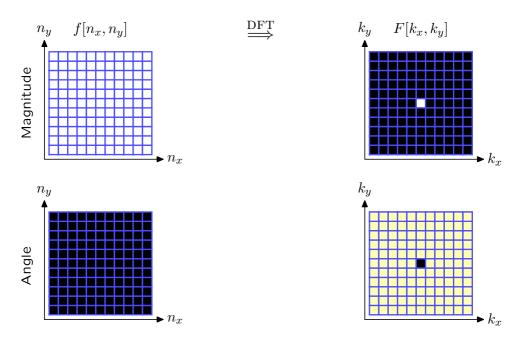






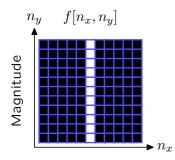


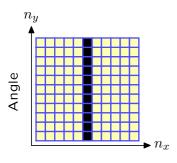


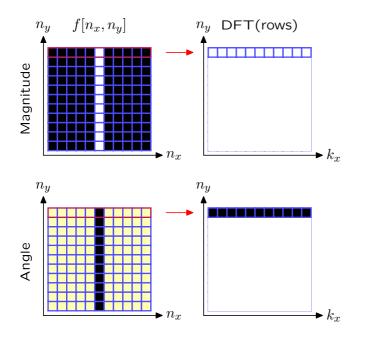


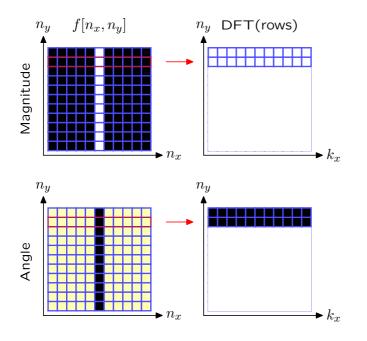
$$\begin{aligned} f_v[n_x, n_y] &= \delta[n_x] = \begin{cases} 1 & n_x = 0\\ 0 & \text{otherwise} \end{cases} \\ F_v[k_x, k_y] &= \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x-1} \sum_{n_y=0}^{N_y-1} \delta[n_x] e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} n_y\right)} \\ &= \frac{1}{N_x N_y} \sum_{n_x=0}^{0} \sum_{n_y=0}^{N_y-1} e^{-j\left(\frac{2\pi k_x}{N_x} 0 + \frac{2\pi k_y}{N_y} n_y\right)} = \frac{1}{N_x N_y} \sum_{n_y=0}^{N_y-1} e^{-j\frac{2\pi k_y}{N_y} n_y} \end{aligned}$$

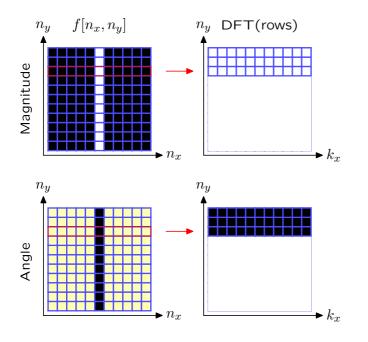
But
$$\sum_{ny=0}^{N_y-1} e^{-j\frac{2\pi ky}{N_y}ny} = \begin{cases} N_y & k_y = 0\\ 0 & \text{otherwise} \end{cases}$$
$$F_v[k_x, k_y] = \frac{1}{N_x N_y} N_y \delta[k_y] = \frac{1}{N_x} \delta[k_y]$$
$$\delta[n_x] \stackrel{\text{DFT}}{\Longrightarrow} & \frac{1}{N_x} \delta[k_y] \end{cases}$$

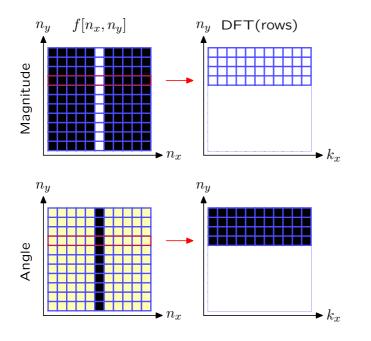


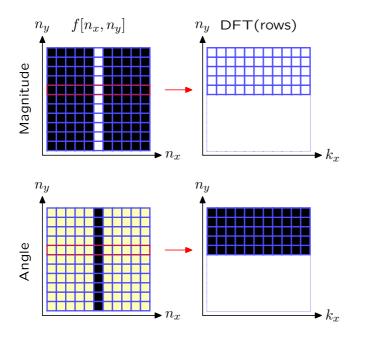


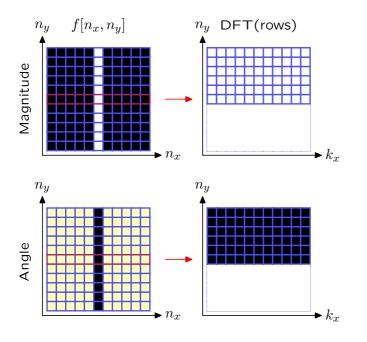


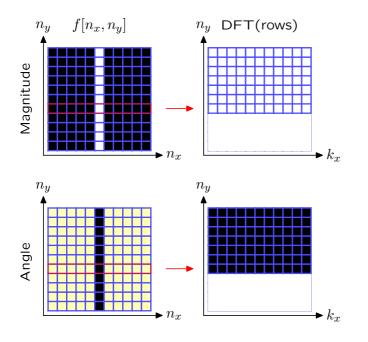


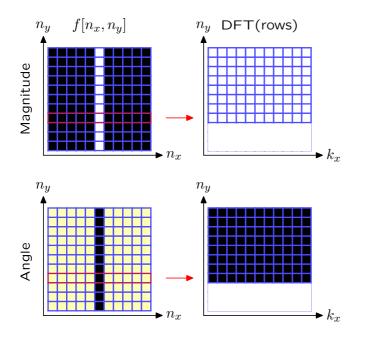


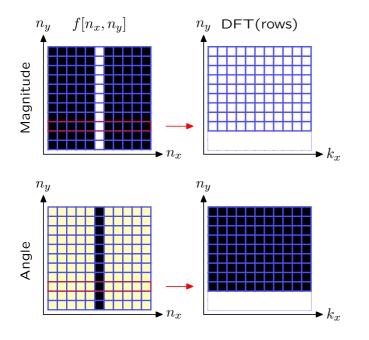


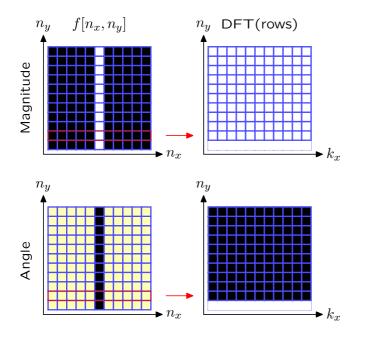


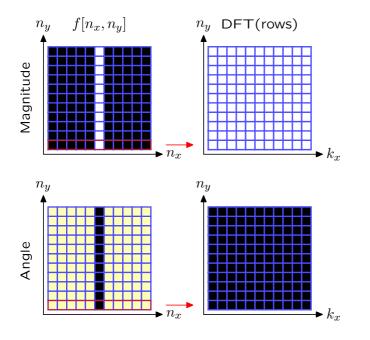


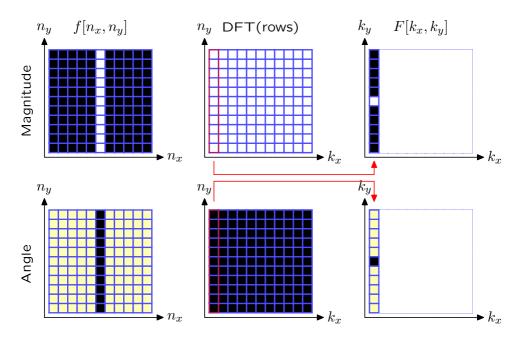


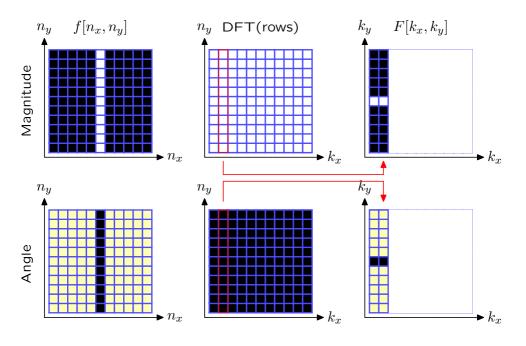


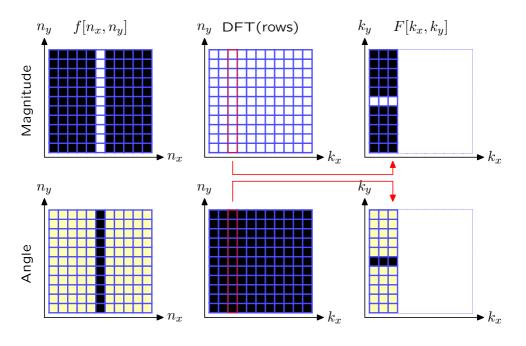


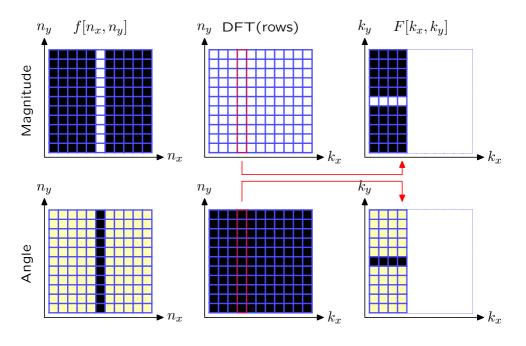


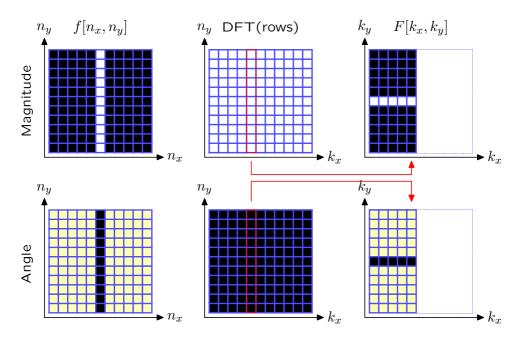


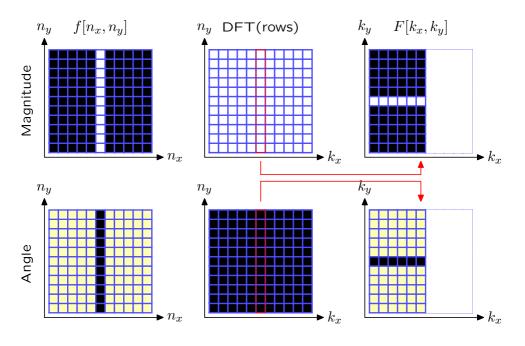


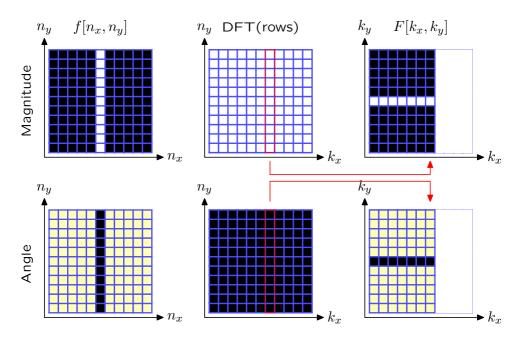


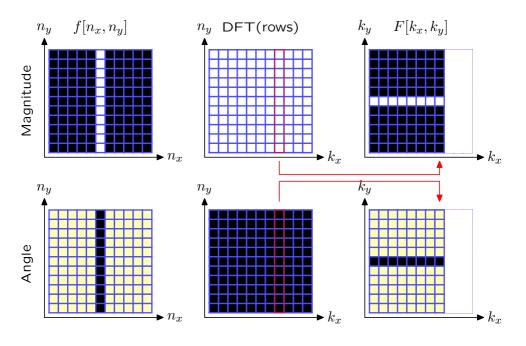


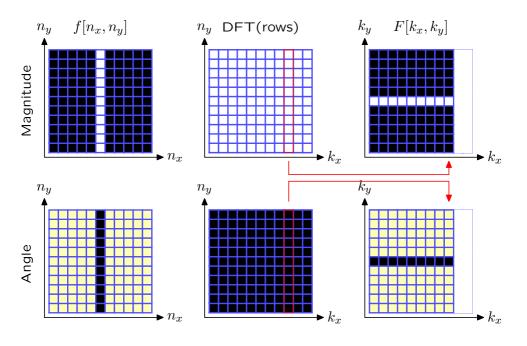


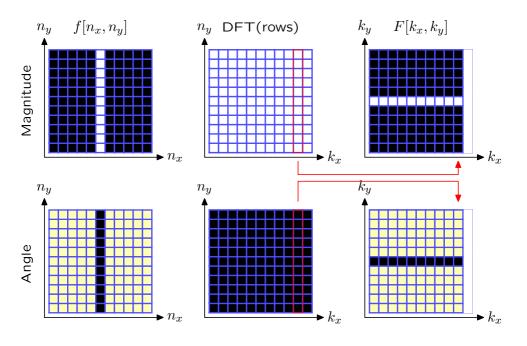


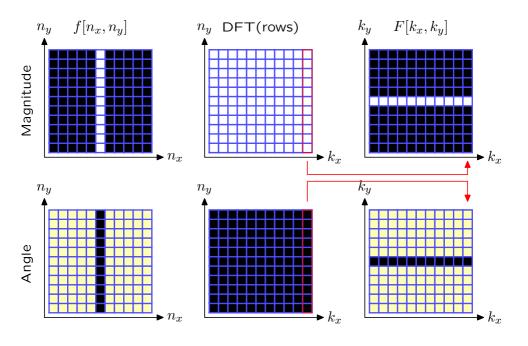


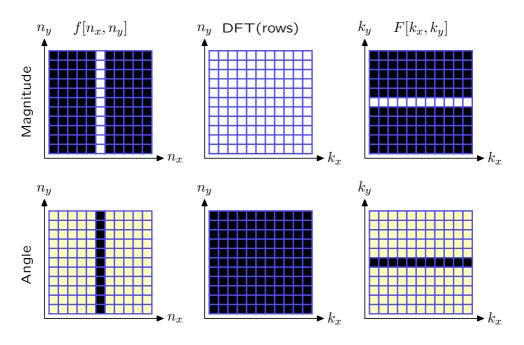






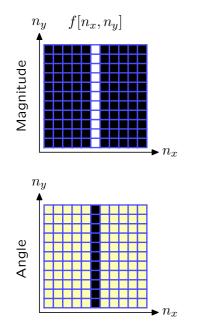


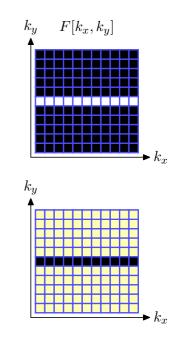




Example: Find the DFT of a vertical line.

 $\underline{\text{DFT}}$



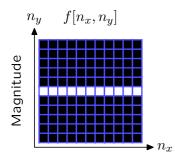


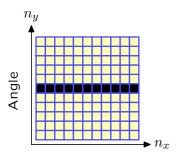
$$\begin{split} f_h[n_x, n_y] &= \delta[n_y] = \begin{cases} 1 & n_y = 0\\ 0 & \text{otherwise} \end{cases} \\ F_h[k_x, k_y] &= \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x-1} \sum_{n_y=0}^{N_y-1} \delta[n_y] \, e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} n_y\right)} \\ &= \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x-1} \sum_{n_y=0}^{0} e^{-j\left(\frac{2\pi k_x}{N_x} n_x + \frac{2\pi k_y}{N_y} 0\right)} = \frac{1}{N_x N_y} \sum_{n_x=0}^{N_x-1} e^{-j\frac{2\pi k_x}{N_x} n_x} \end{split}$$

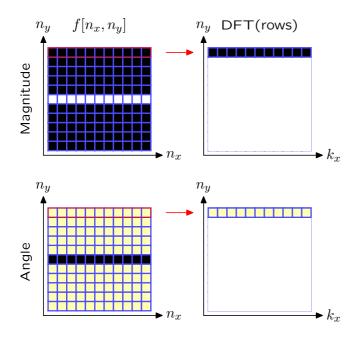
But
$$\sum_{n_x=0}^{N_x-1} e^{-j\frac{2\pi k_x}{N_x}n_x} = \begin{cases} N_x & k_x=0\\ 0 & \text{otherwise} \end{cases}$$

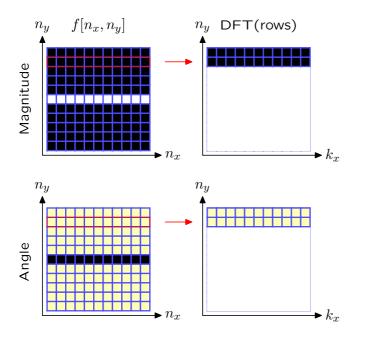
$$F_h[k_x,k_y] = \frac{1}{N_x N_y} N_x \delta[k_x] = \frac{1}{N_y} \delta[k_x]$$

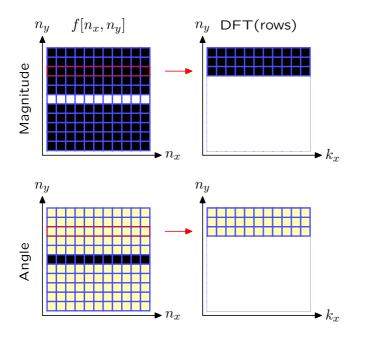
$$\delta[n_y] \stackrel{\text{DFT}}{\Longrightarrow} \frac{1}{N_y} \delta[k_x]$$

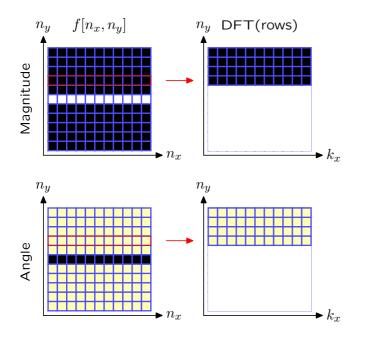


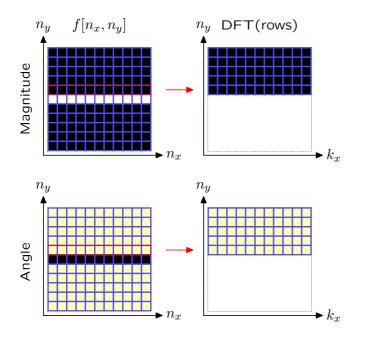


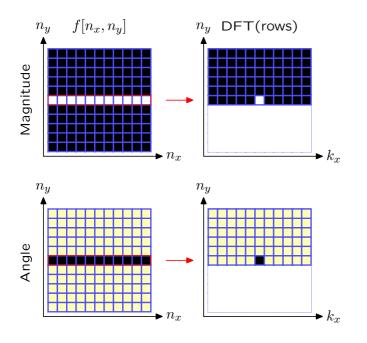


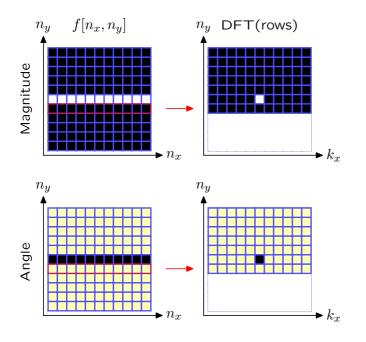


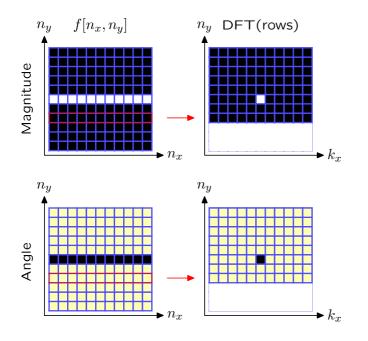


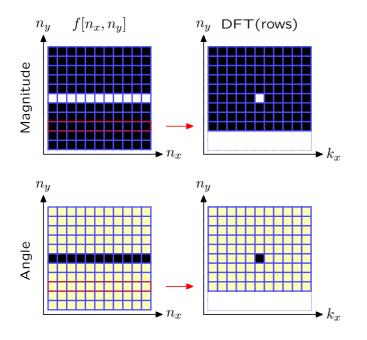


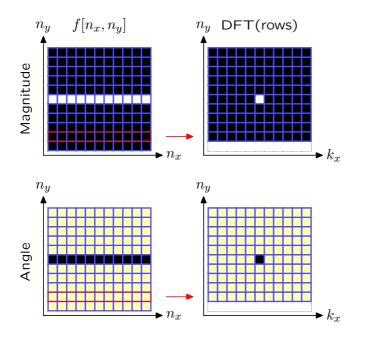


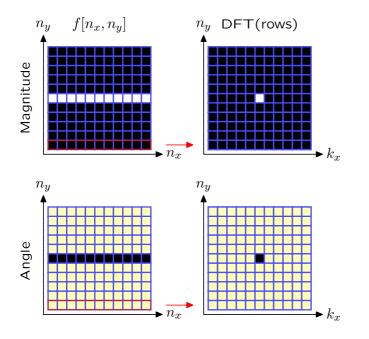


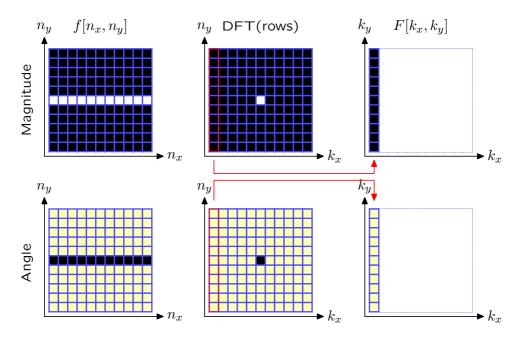


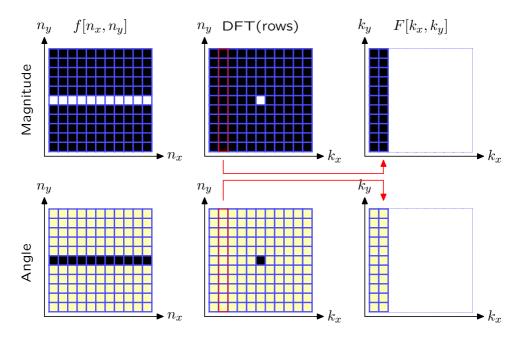


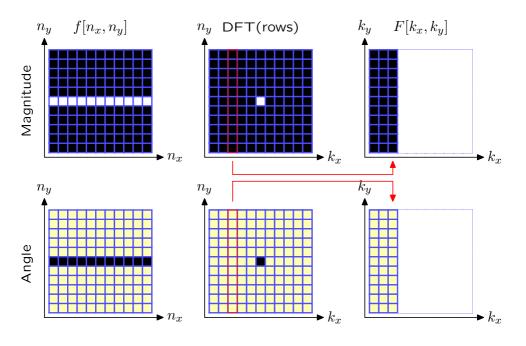


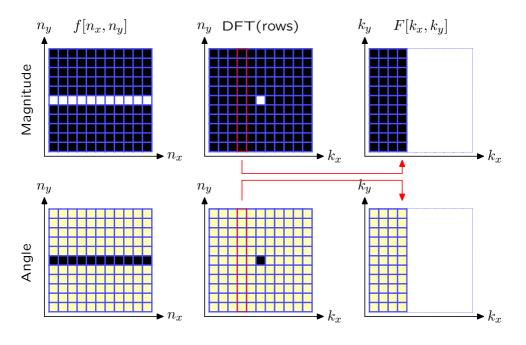


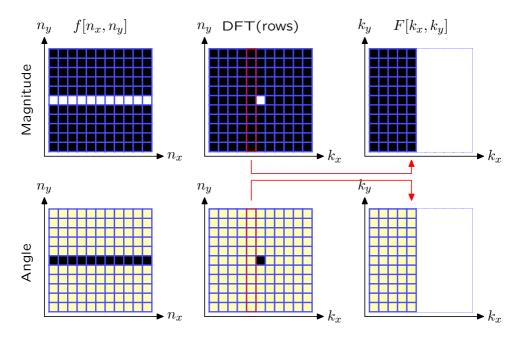


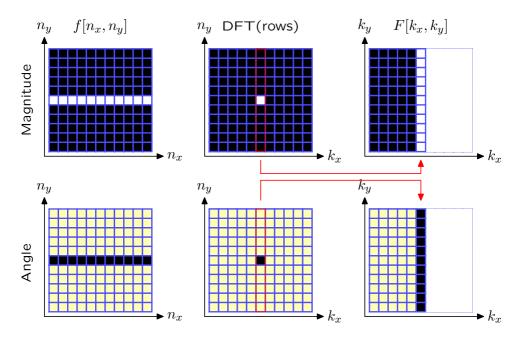


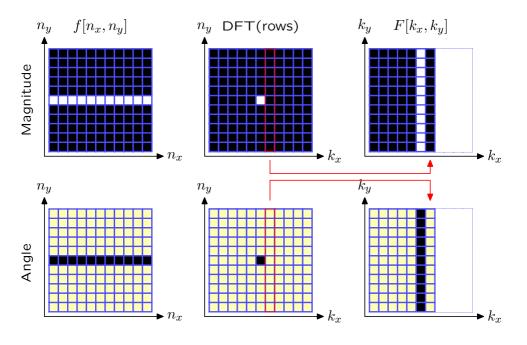


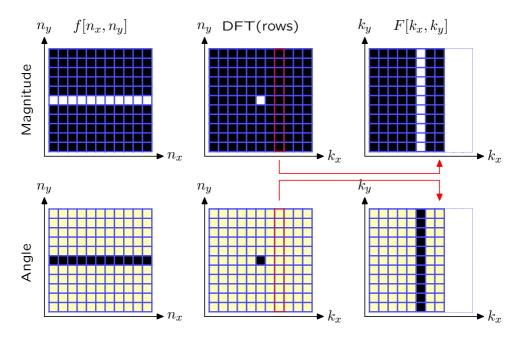


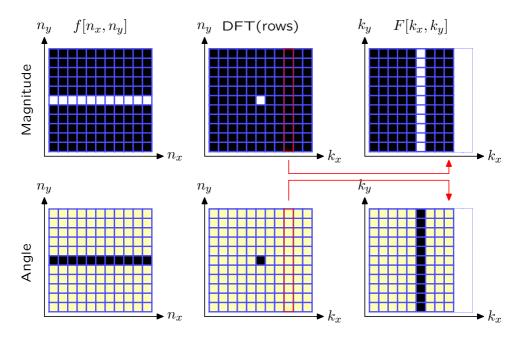


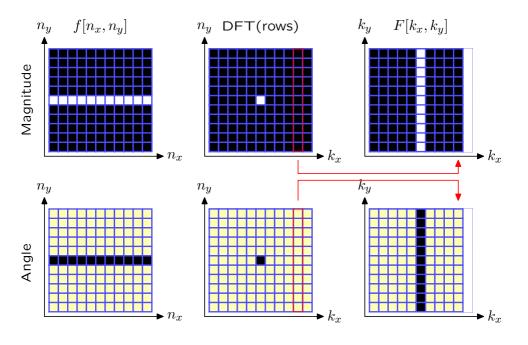


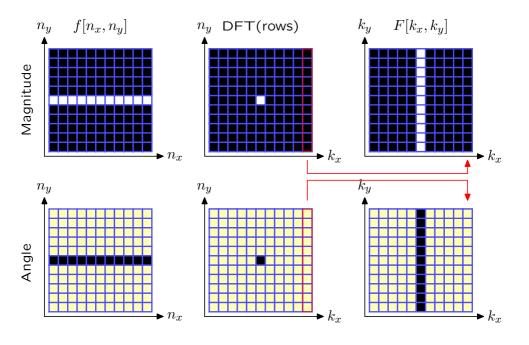


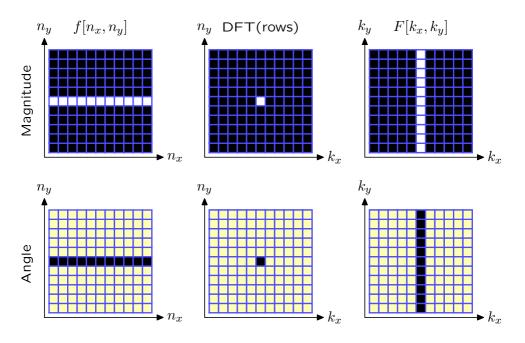






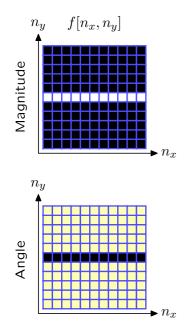


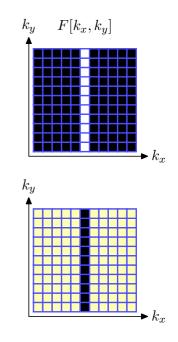




Example: Find the DFT of a horizontal line.

 $\underline{\text{DFT}}$



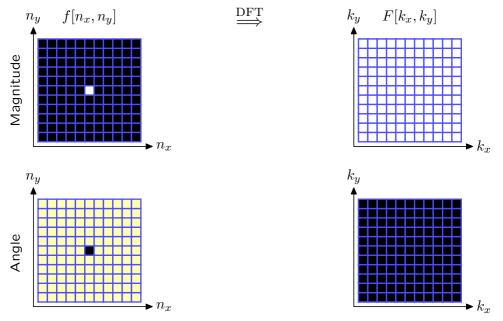


Translating (Shifting) an Image

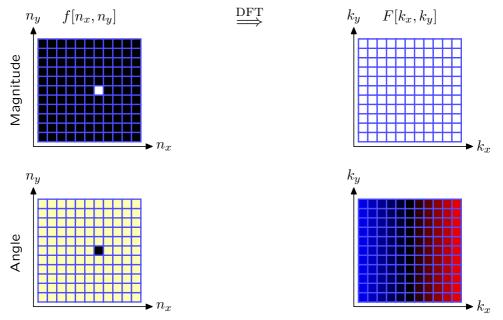
Effect of image translation (shifting) on its Fourier transform. Assume that $f_0[n_x, n_y] \stackrel{\text{DFT}}{\Longrightarrow} F_0[k_x, k_y].$ Find the 2D DFT of $f_1[n_x, n_y] = f_0[n_x - n_{x0}, n_y - n_{y0}]$ $F_1[k_x, k_y] = \sum \sum f_1[n_x, n_y] e^{-j\frac{2\pi k_x}{N_x}n_x} e^{-j\frac{2\pi k_y}{N_y}n_y}$ kr ku $=\sum \sum f_0[n_x - n_{x0}, n_y - n_{y0}]e^{-j\frac{2\pi kx}{N_x}n_x}e^{-j\frac{2\pi ky}{N_y}n_y}$ $k_T k_H$ Let $l_x = n_x - n_{x0}$ and $l_y = n_y - n_{y0}$. Then $F_1[k_x, k_y] = \sum \sum f_0[l_x, l_y] e^{-j\frac{2\pi k_x}{N_x}(l_x + n_{x0})} e^{-j\frac{2\pi k_y}{N_y}(l_y + n_{y0})}$ lr lu $= e^{-j\frac{2\pi k_x}{N_x}n_{x0}}e^{-j\frac{2\pi k_y}{N_y}n_{y0}}F_0[k_x, k_y]$

Translating an image adds linear (in k_x , k_y) **phase** to its transform.

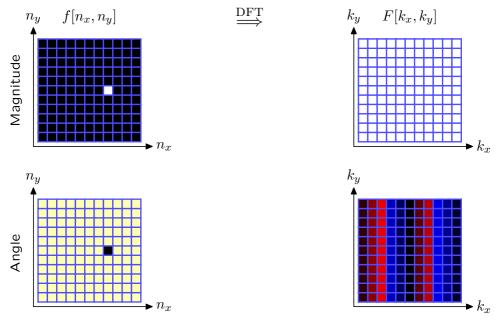
Example: Find the DFT of a shifted 2D unit sample.



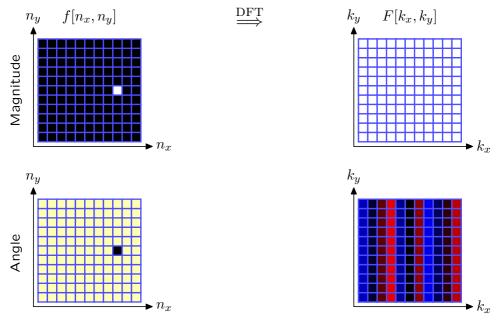
Example: Find the DFT of a shifted 2D unit sample.



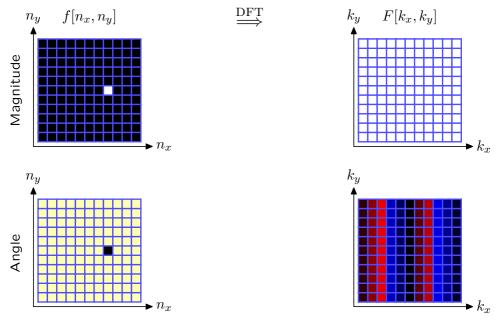
Example: Find the DFT of a shifted 2D unit sample.



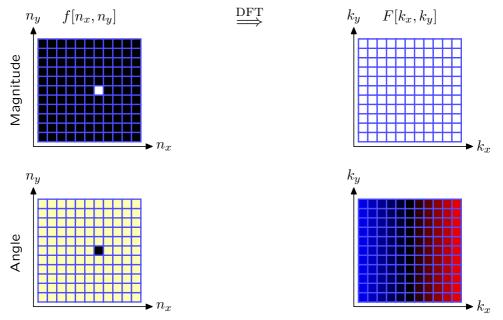
Example: Find the DFT of a shifted 2D unit sample.



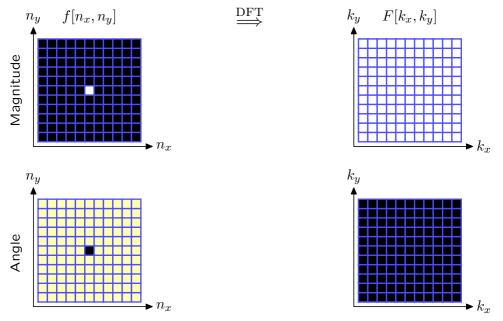
Example: Find the DFT of a shifted 2D unit sample.



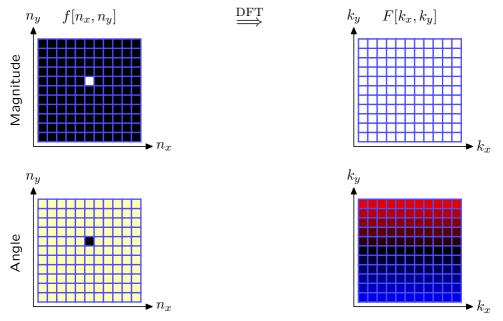
Example: Find the DFT of a shifted 2D unit sample.



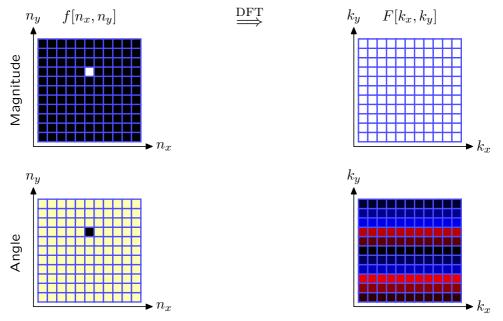
Example: Find the DFT of a shifted 2D unit sample.



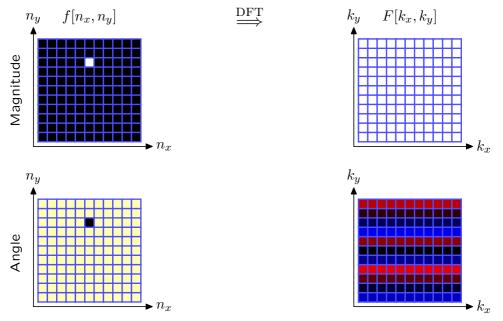
Example: Find the DFT of a shifted 2D unit sample.



Example: Find the DFT of a shifted 2D unit sample.



Example: Find the DFT of a shifted 2D unit sample.



Using Python

Calculating DFTs is most efficient in NumPy (Numerical Python).

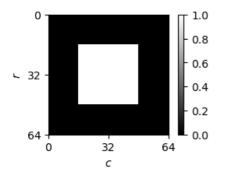
- NumPy arrays are homogeneous: their elements are of the same type
- Numpy operators (+, -, abs, .real, .imag) combine elements to create new arrays. e.g., (f+g) [n] is f[n]+g[n].
- 2D Numpy arrays can be indexed by tuples: e.g., f[r,c] = f[r][c].
- 2D Numpy arrays support negative indices: e.g., f[-1] = f[len(f)-1]
- 2D indices address row then column.

NumPy indexing is consistent with **linear algebra** (row first then column with rows increasing downward and columns increasing to the right). But it differs from **physical mathematics** (x then y with x increasing to the right and y increasing upward). You may do calculations either way, but row, column is often less confusing.

Numpy Example

Make a white square on a black background.

```
import numpy
from lib6003.image import show_image
f = numpy.zeros((64,64))
for r in range(16,48):
        for c in range(16,48):
            f[r,c] = 1
show_image(f,zero_loc='topleft')
```



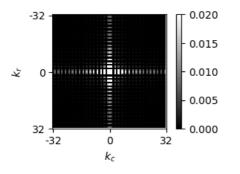
Numpy Example

Find the 2D DFT of the square.

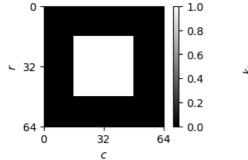
import numpy
from lib6003.image import show_image
from lib6003.fft import fft2

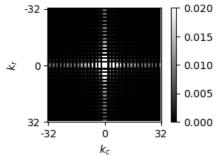
F = fft2(f)

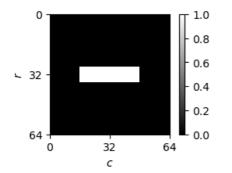
show_image(numpy.abs(F),zero_loc='center',vmin=0,vmax=0.02)

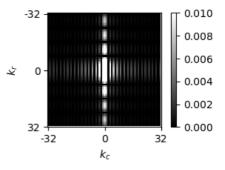


Big and Small



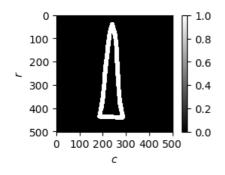






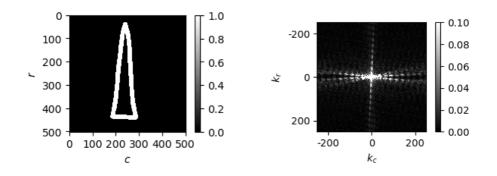
Triangle

What are the dominant features of the magnitude of the DFT of a triangle?



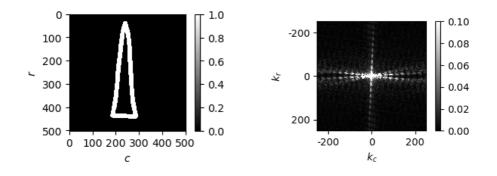
Triangle

What are the dominant features of the magnitude of the DFT of a triangle?



Triangle

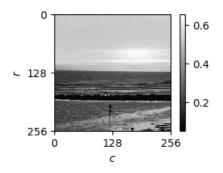
What are the dominant features of the magnitude of the DFT of a triangle?



The DFT has three nearly linear features, one for each edge of the triangle. Lines in the frequency domain are perpendicular to those in space domain.

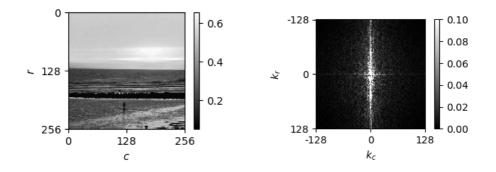
Ocean

What are the dominant features of the DFT magnitude of an ocean view?



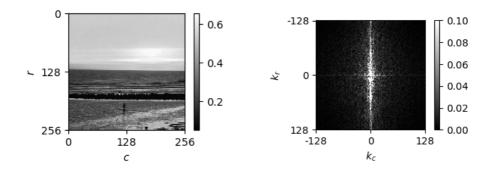
Ocean

What are the dominant features of the DFT magnitude of an ocean view?



Ocean

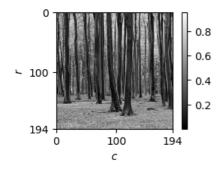
What are the dominant features of the DFT magnitude of an ocean view?



The horizontal features in the ocean view show up as a strong vertical line in the DFT.

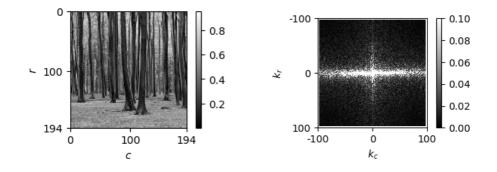
Trees

What are the dominant features of the DFT magnitude of these trees?



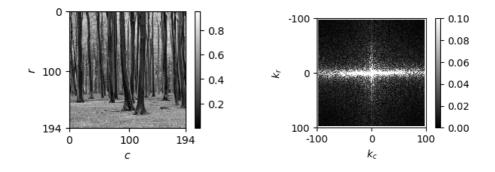
Trees

What are the dominant features of the DFT magnitude of these trees?



Trees

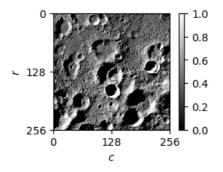
What are the dominant features of the DFT magnitude of these trees?



Now there is a strong horizontal line in the DFT.

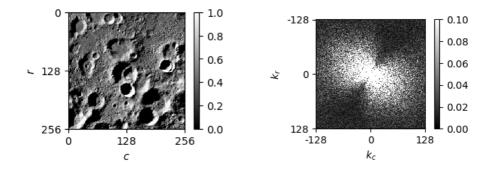
Moon

What are the dominant features of the DFT magnitude of the moon?



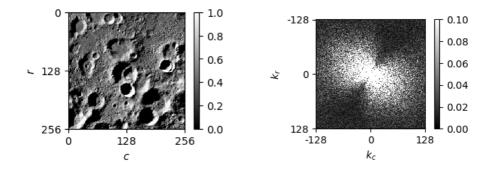
Moon

What are the dominant features of the DFT magnitude of the moon?



Moon

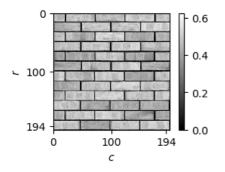
What are the dominant features of the DFT magnitude of the moon?



Large distribution of frequencies. Concentration along r = c axis due to illumination from upper left.

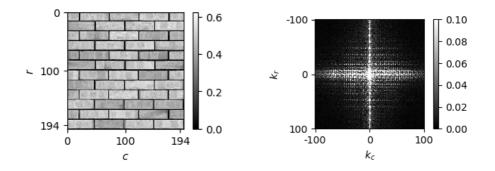
Bricks

What are the dominant features of the DFT magnitude of this brick wall?



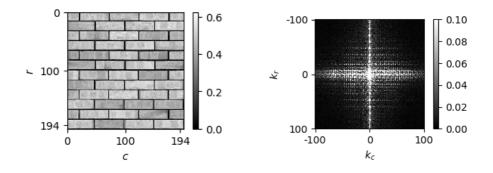
Bricks

What are the dominant features of the DFT magnitude of this brick wall?



Bricks

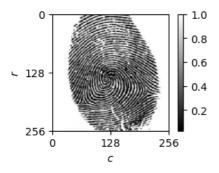
What are the dominant features of the DFT magnitude of this brick wall?



Strong horizontal "layer" in space \rightarrow strong vertical line in frequency. Strong vertical features but broken up periodically \rightarrow periodicity in frequency.

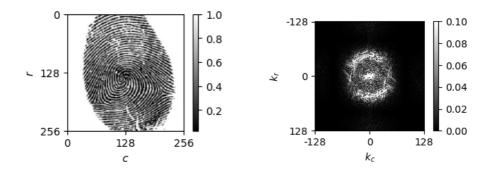
Fingerprint

What are the dominant features of the DFT magnitude of this fingerprint?



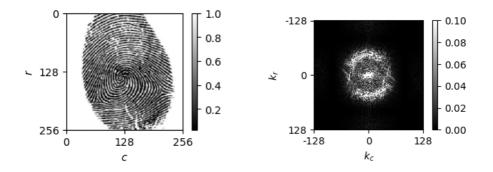
Fingerprint

What are the dominant features of the DFT magnitude of this fingerprint?



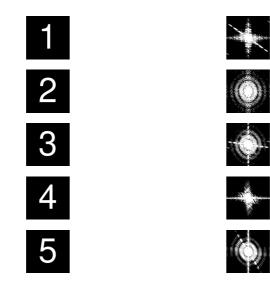
Fingerprint

What are the dominant features of the DFT magnitude of this fingerprint?

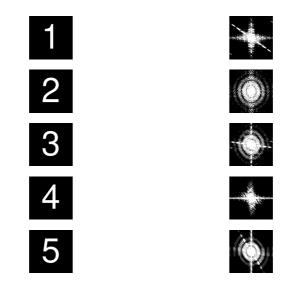


About 40 fingerprint ridges $\rightarrow k \approx 40$.

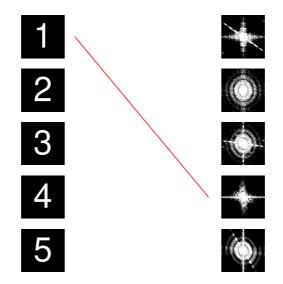
Which panel on right shows the mag of the DFT of each digit on the left?



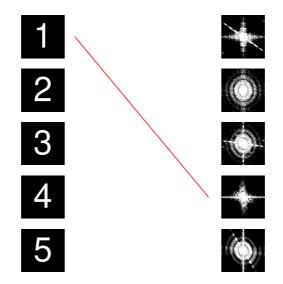
Which panel on right shows the mag of the DFT of the digit 1?



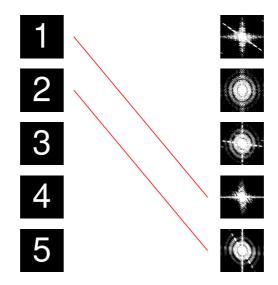
Which panel on right shows the mag of the DFT of the digit 1?



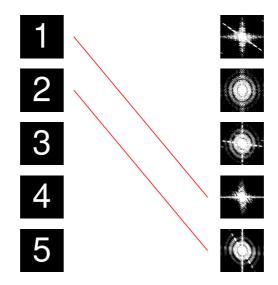
Which panel on right shows the mag of the DFT of the digit 2?



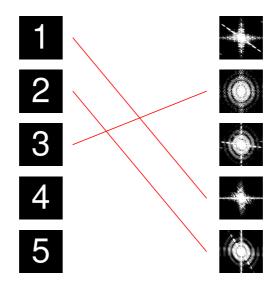
Which panel on right shows the mag of the DFT of the digit 2?



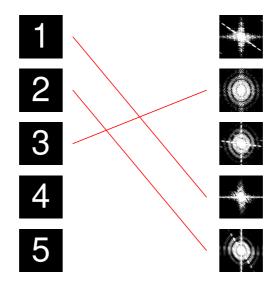
Which panel on right shows the mag of the DFT of the digit 3?



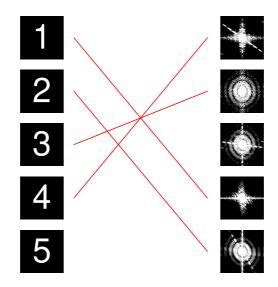
Which panel on right shows the mag of the DFT of the digit 3?



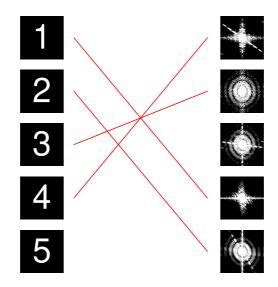
Which panel on right shows the mag of the DFT of the digit 4?



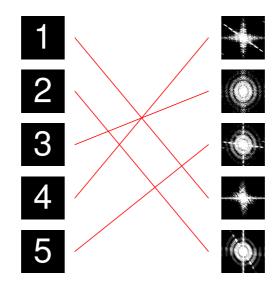
Which panel on right shows the mag of the DFT of the digit 4?



Which panel on right shows the mag of the DFT of the digit 5?



Which panel on right shows the mag of the DFT of the digit 5?



Summary

Introduced 2D signal processing.

• generally simple extensions of 1D ideas

Introduced 2D Fourier representations.

- Fourier kernel comprises the sum of an x part and a y part
- basis functions are complex exponentials

Properties of 2D DFT

- transform all of the rows then transform all of the columns
- transform all of the columns then transform all of the rows

Question of the Day

Sketch the magnitude of the 2D Fourier Transform of a checkmark.

