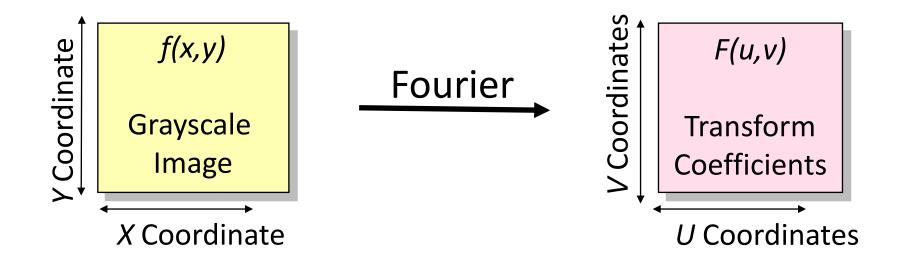
Fourier Transform of Pictures: Change of Basis in 2D

Image Domain

Frequency Domain Natural Basis, Real Numbers Fourier Basis, Complex Numbers



- The Image & Transform coefficients are arranged in a 2D array.
- Both image f and Fourier transform F are Cyclic / Periodic

2D Discrete Fourier

Fourier Transform

$$F(u,v) = \frac{1}{N} \sum_{y=0}^{N-1} \sum_{x=0}^{N-1} f(x,y) e^{\frac{-2\pi i(ux+vy)}{N}}$$

$$-\infty < \forall u, v > \infty$$

Inverse Fourier Transform

$$f(u,v) = \frac{1}{N} \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} F(u,v) e^{\frac{2\pi i(ux+vy)}{N}} -\infty < \forall x, y > \infty$$

$$-\infty < \forall x, y > \infty$$

$$F(0,0) = \frac{1}{N} \sum_{y=0}^{N-1} \sum_{x=0}^{N-1} f(x,y) e^{\frac{-2\pi i(0x+0y)}{N}} = \frac{1}{N} \sum_{y=0}^{N-1} \sum_{x=0}^{N-1} f(x,y) = N\bar{f}$$

2D Fourier Basis Functions

(For
$$-2 \le u, v \le 2$$
)

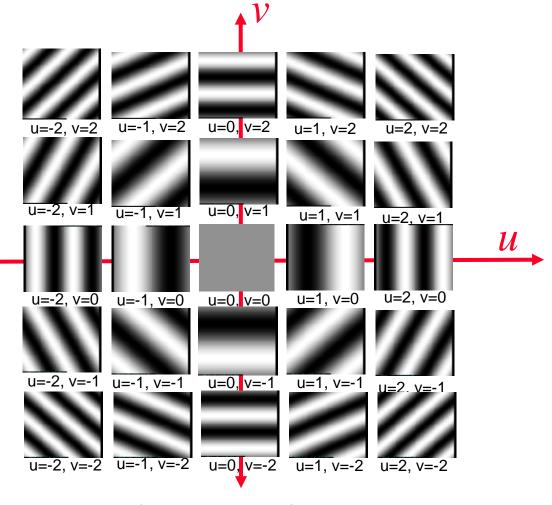
(Imaginary Part – How can we tell?)

$$e^{\frac{2\pi i(ux+vy)}{N}}$$

$$\sin(\frac{2\pi}{N}(ux+vy))$$

The original image is a weighted sum of all basis <u>functions</u>, where the basis function for frequency (u,v) is multiplied by the complex number F(u,v) specifying:

- 1. Amplitude
- 2. Phase (Shift)



- 25 basis <u>functions</u> for images 5×5.
- Each function is a 5×5 matrix.
- (0,0) at center of each image
- -1 is black, +1 is white, 0 is grey

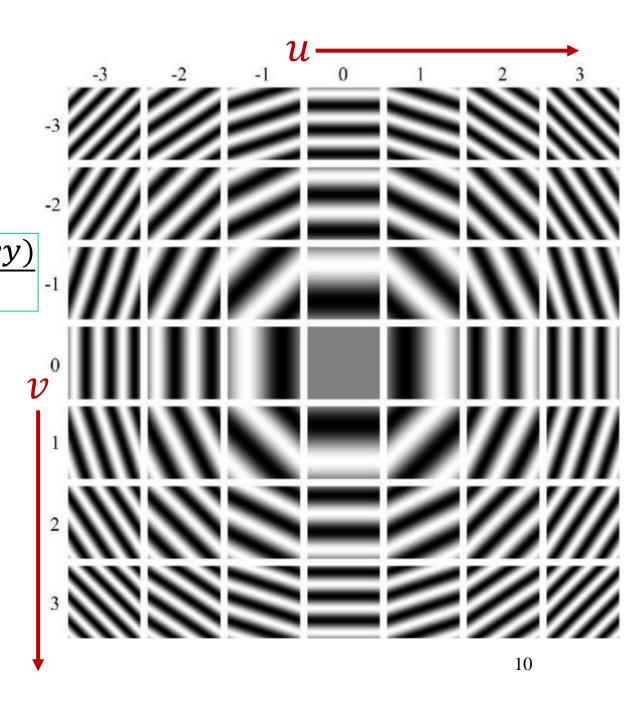
2D Fourier Basis Functions

(For $-3 \le u, v \le 3$) (Imaginary Part)

The original image is a weighted sum of all basis functions $\frac{2\pi i(ux+vy)}{2\pi i(ux+vy)}$

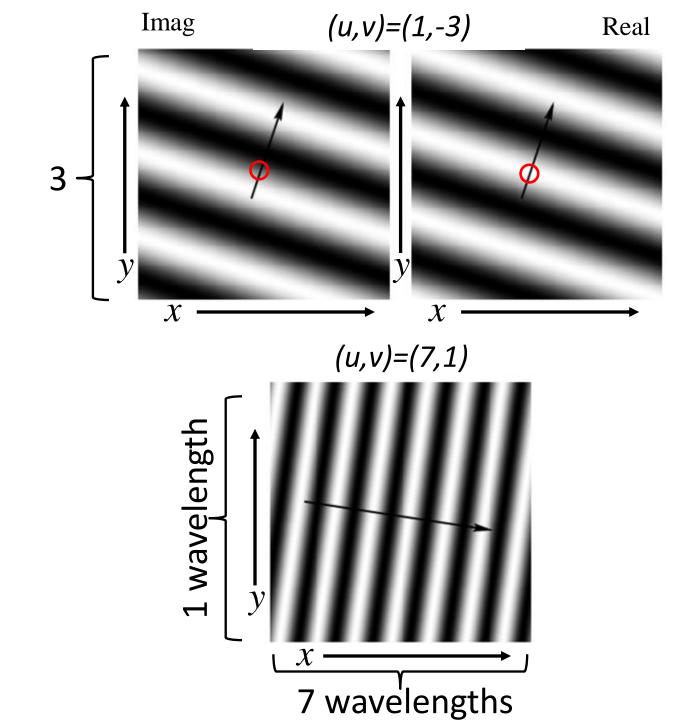
Showing
$$\sin(\frac{2\pi}{N}(ux+vy))$$

- 49 basis functions (matrices) for all images of size7×7.
- Each basis function is a 7×7 matrix.
- (0,0) at center of each image
- -1 is black, +1 is white, 0 is grey

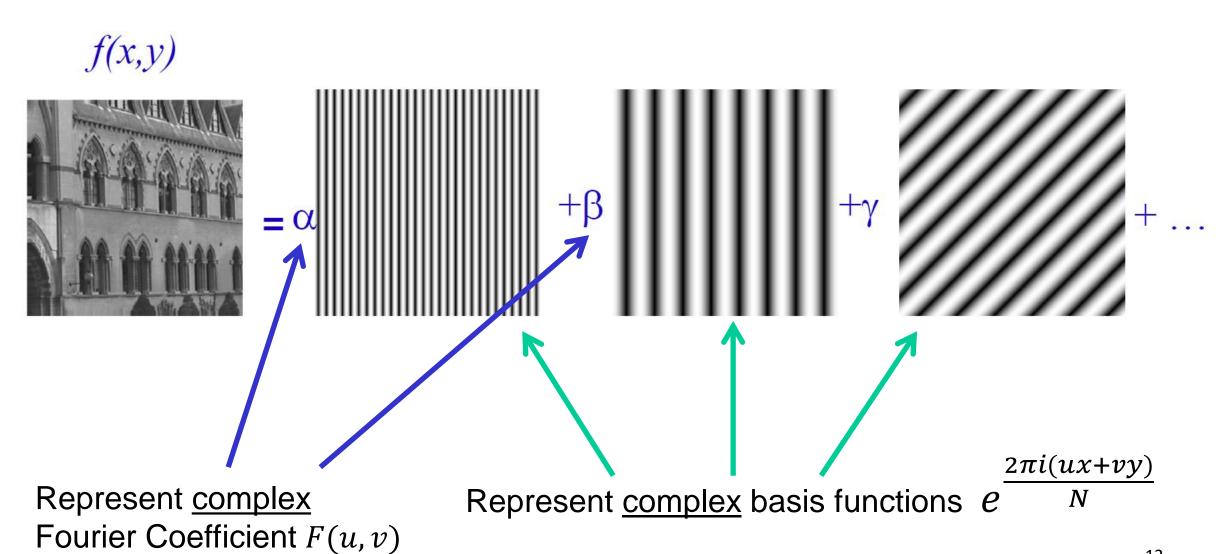


$$e^{\frac{2\pi i(ux+vy)}{N}}$$

Is (0,0)at center of image



Summary



Fourier Spectrum

Fourier Coefficient (complex number):

$$F(u) = R(u) + iI(u)$$

Fourier Spectrum (positive number)

$$|F(u)| = \sqrt{R^2(u) + I^2(u)}$$

Fourier Phase (Angle)

$$\theta(u) = \tan^{-1}(I(u)/R(u))$$

Fourier Coefficient (complex number):

$$F(u) = |F(u)| e^{i\theta}$$

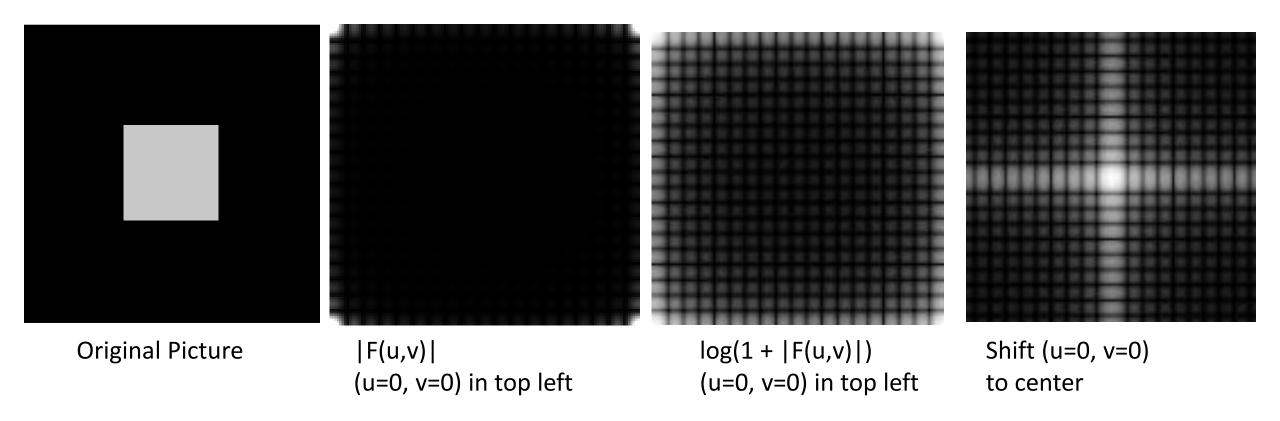
Display Fourier Spectrum as Picture

- 1. Compute $\log(|F(u)| + 1)$
- 2. Scale to full grey-level range (E.g. 0..255 or 0..1)
- 3. Move (u=0, v=0) to center of image (Shift by N/2)

Log Scale of [0..100]

Original F	100	4	2	1	0
Log (1+ F)	4.62	1.61	1.01	0.69	0
Scaled to 100	100	35	22	15	0

Display Fourier Spectrum (0 is black)



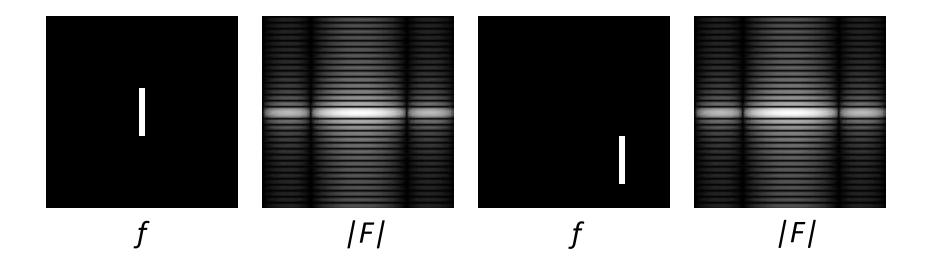
- Question: Why does |F(0,0)| always have the highest value in the Fourier spectrum of an image?
- Hint: The image intensities are always positive

Translation

$$f(x-x_0, y-y_0) \Leftrightarrow F(u,v)e^{\frac{2\pi i(ux_0+vy_0)}{N}}$$

$$F(u-u_0, v-v_0) \Leftrightarrow f(x,y)e^{\frac{-2\pi i(u_0x+v_0y)}{N}}$$

Fourier Spectrum is invariant to image translation



Decomposition Equation

$$F(u,v) = \frac{1}{N} \sum_{x=0}^{N-1} \sum_{v=0}^{N-1} f(x,y) e^{\frac{-2\pi i(ux+vy)}{N}} e^{\frac{-2\pi i(ux+vy)}{N}} = \left(e^{\frac{-2\pi iux}{N}}\right) \left(e^{\frac{-2\pi ivy}{N}}\right)$$

$$F(u,v) = \frac{1}{N} \sum_{x=0}^{N-1} e^{\frac{-2\pi i u x}{N}} \sum_{y=0}^{N-1} f(x,y) e^{\frac{-2\pi i v y}{N}} \qquad F(x,v) = \sum_{y=0}^{N-1} f(x,y) e^{\frac{-2\pi i v y}{N}}$$

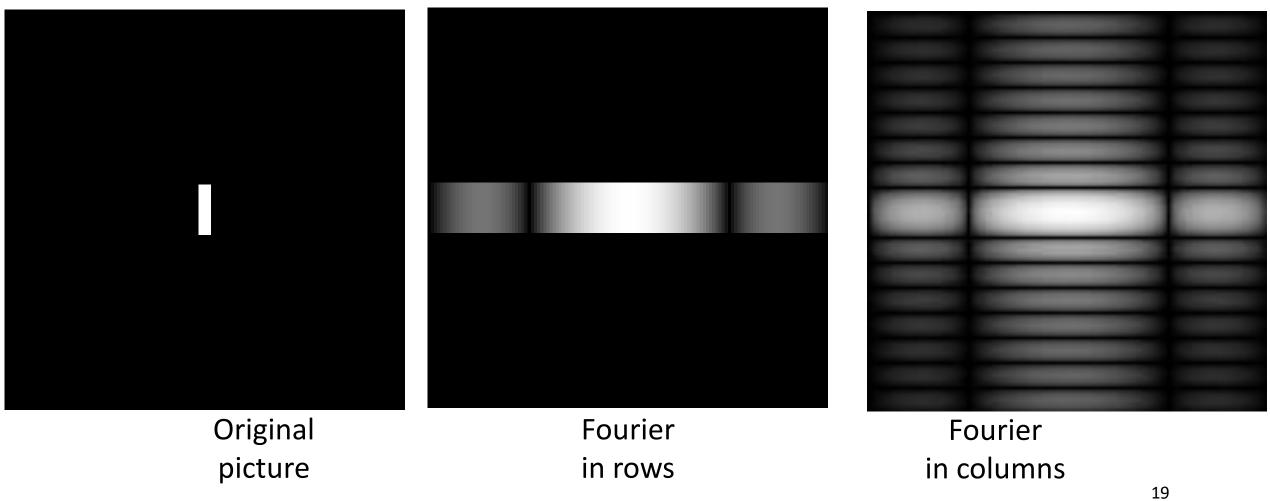
$$F(u,v) = \frac{1}{N} \sum_{n=0}^{N-1} e^{\frac{-2\pi i u x}{N}} F(x,v)$$

- Compute 1-D Fourier on each columnOn result:
- Compute 1-D Fourier on each row

Decomposition Conclusions

- 2-D Fourier Transform can be computed using 1-D Fourier
 - Compute 1-D Fourier on each column On result:
 - Compute 1-D Fourier on each row
 - (Multiply by N?)
- 1-D Fourier Transform is enough to compute Fourier of ANY dimension

Decomposition Example



Periodicity & Symmetry

$$F(u,v) = F(u + N, v) = F(u, v + N) =$$

= $F(u + N, v + N)$

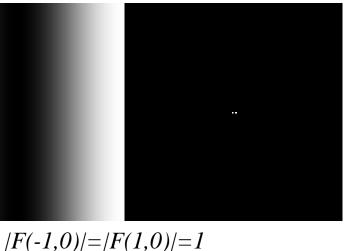
$$F(u,v) = F^*(-u,-v)$$

$$(a+bi)^* = (a-bi)$$

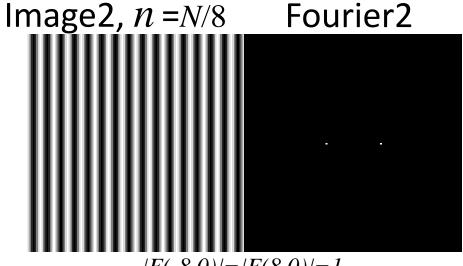
$$|F(u,v)| = |F(-u,-v)|$$

Fourier Spectrum of $Cos(2\pi nx/N)$ [N samples]

Image1, *n*=1 Fourier1



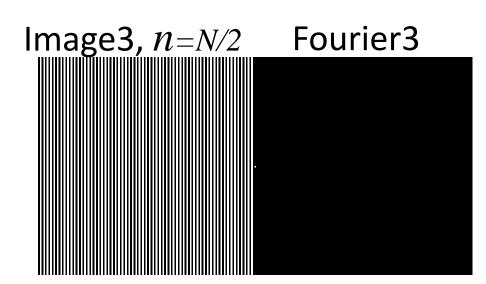
Why 2 points in Fourier Spectrum?



|F(-8,0)| = |F(8,0)| = 1

Note: Image of Cos has negative values, -1 is black, 1 is white

 $Cos(\pi x)$ 1, -1, 1, ...



What will be F for $1+\cos(2\pi nx/N)$

|F(-N/2,0)| = |F(N/2,0)| = 1

Linearity (Φ is Transform Fourier)

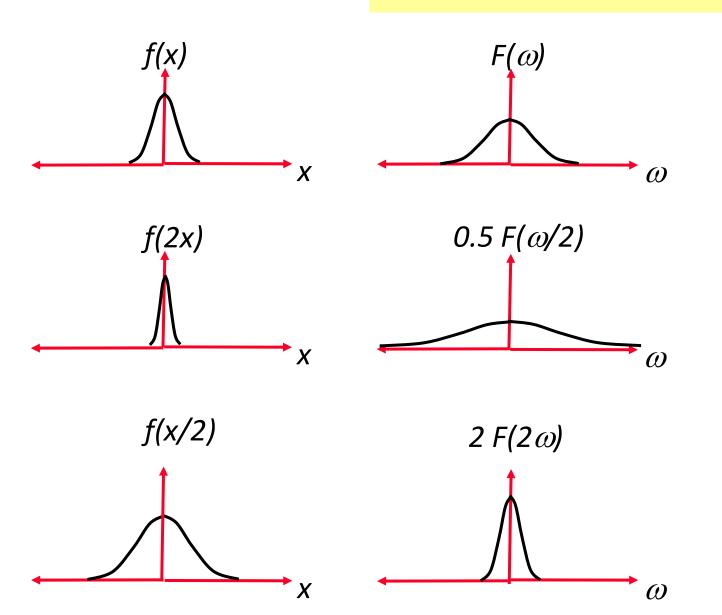
$$\Phi(f_1(x,y) + f_2(x,y)) = \Phi(f_1(x,y)) + \Phi(f_2(x,y))$$

$$\Phi(a \cdot f(x, y)) = a \cdot \Phi(f(x, y))$$

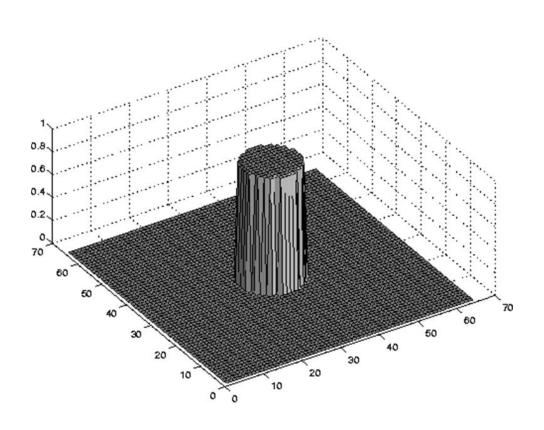
$$\Phi(f(ax,by)) = \frac{1}{|ab|} F(u/a, v/b)$$

Change Scale: Examples

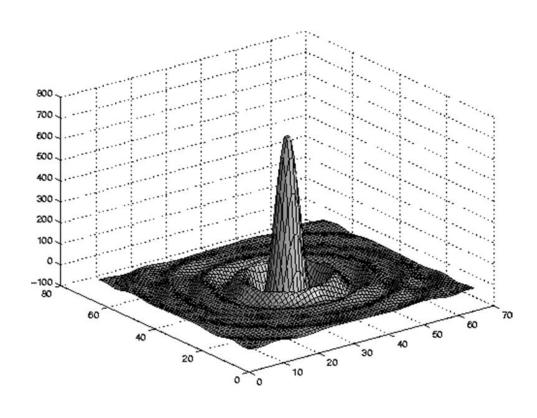
$$\Phi(f(ax, by)) = \frac{1}{|ab|} F(u/a, v/b)$$



2D Fourier



Original Function



spectrogram

Discrete Derivatives Using Fourier

Derivatives are defined over continuous functions. How can we compute derivative of a sequence of numbers? One Possibility: Use Fourier Transform

$$f(x) = \sum_{u} F(u) e^{\frac{2\pi i u x}{N}}$$

Derivative – Increase high frequencies

$$f'(x) = \left(\sum_{u} F(u) e^{\frac{2\pi i u x}{N}}\right)' = \sum_{u} F(u) \left(e^{\frac{2\pi i u x}{N}}\right)' =$$

$$= \frac{2\pi i}{N} \sum_{u} u F(u) e^{\frac{2\pi i u x}{N}}$$

$$F(u) \rightarrow uF(u)$$

2D Derivatives using Fourier

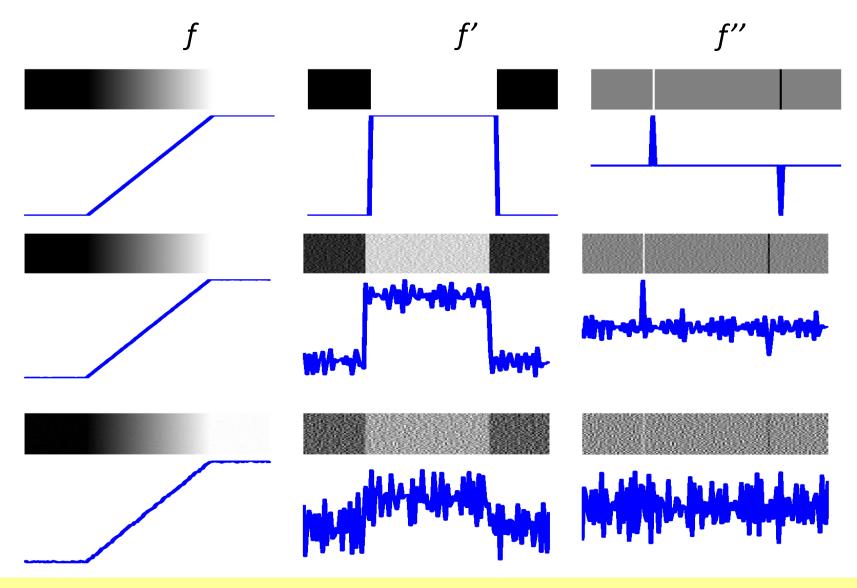
- To compute the x derivative of f(x,y):
 - 1. Compute the Fourier Transform F(u,v)
 - 2. Multiply Fourier coefficient F(u,v) by $\frac{2\pi i}{N}u$
 - 3. Compute the Inverse Fourier Transform
- To compute the y derivative of f(x,y):
 - 1. Compute the Fourier Transform F(u,v)
 - 2. Multiply Fourier coefficient F(u,v) by $\frac{2\pi i}{N}v$
 - 3. Compute the Inverse Fourier Transform

Derivative as a Fourier Filter

$$f'(x) = \frac{2\pi i}{N} \sum_{u} u F(u) e^{\frac{2\pi i u x}{N}}$$

- Multiply Fourier Coefficient F(u) with $\frac{2\pi i}{N}u$
- Amplifies higher frequencies (and Noise)
 - Since Noise has more high frequency than normal images
 - Derivatives amplify noise
- Cancels DC (F(0)), the image average, becomes zero)

Effect of Noise on Derivatives

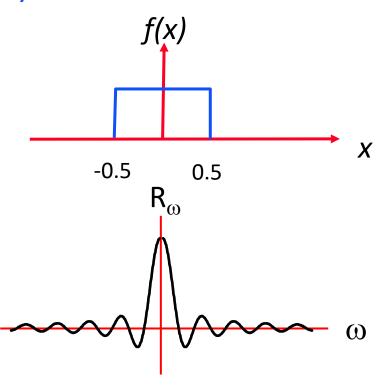


Note: This slide neglects the cyclic effect, and derivatives here do not sum to 0.

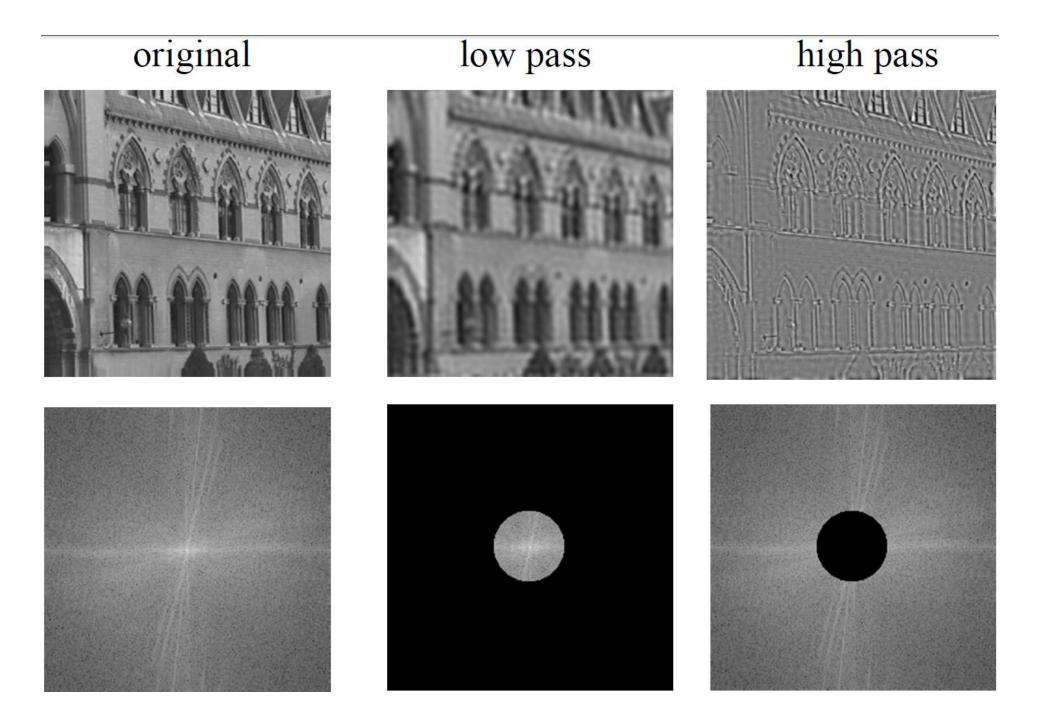
Fourier of Special Functions

The Window (Box) Function (Rect):

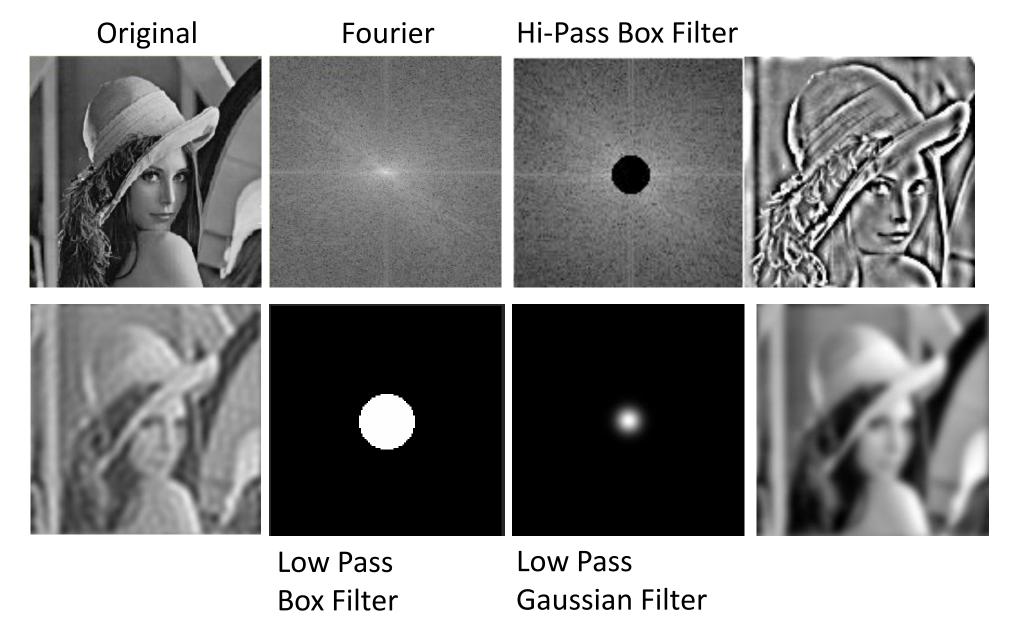
$$rect(x) = \begin{cases} 1 & \text{if}|x| < \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$



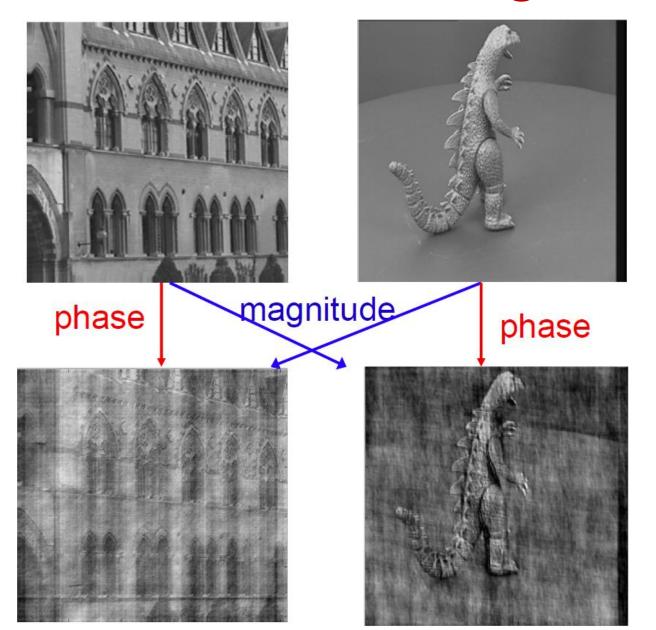
$$F(\omega) = \frac{\sin(\pi\omega)}{\pi\omega} = \operatorname{sinc}(\pi\omega)$$



Compare Filters



Effects of Phase and Magnitude

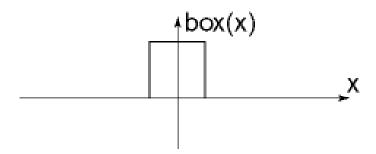


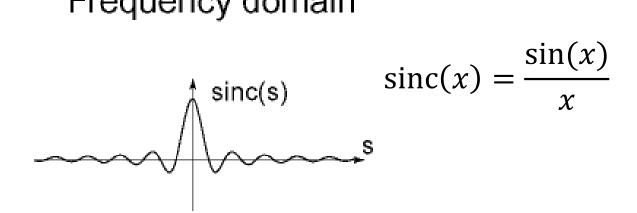
Fourier Transform Pairs

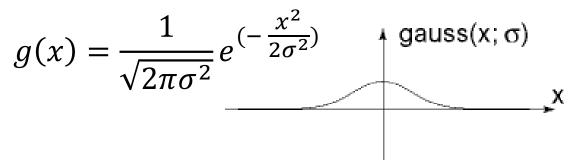
$$F(s) = \int_{-\infty}^{\infty} f(x)e^{-i2\pi sx}dx$$

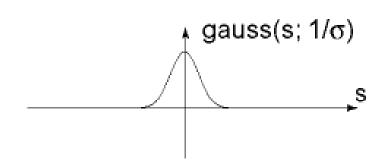
Spatial domain

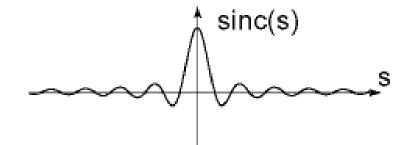
Frequency domain

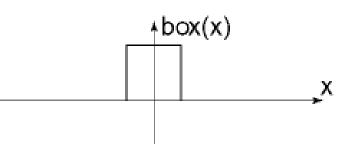










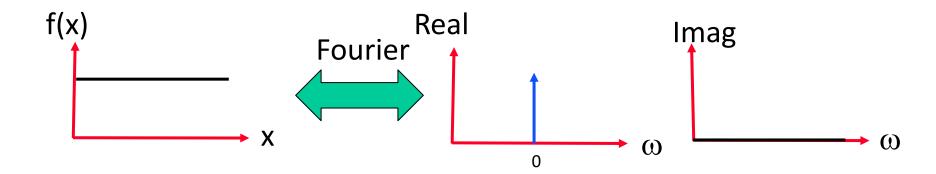


Fourier of Special Functions

The Constant Function:

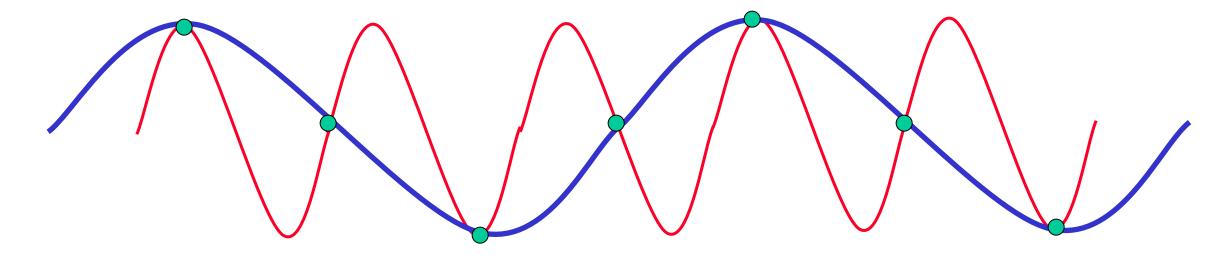
$$f(x) = 1$$

$$F(u) = \frac{1}{N} \sum_{x=0}^{N-1} 1 e^{\frac{-2\pi i ux}{N}}$$



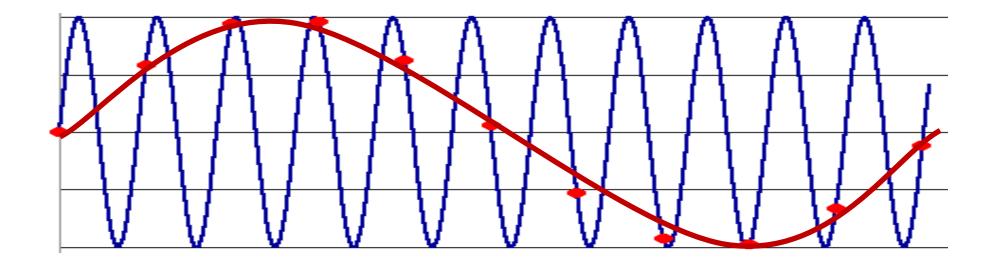
Aliasing

- Sampling can result in aliasing.
- Example: Sampling at 1.5π



• To avoid aliasing, distance between sampling points should be less than ½ of wavelength (Niquest...) 37

Aliasing



Sampling The most important slide in course!

- Blur before you sample (Low-pass filter: reduce the highest frequencies)
- Sampling without low-pass results in aliasing.

- How NOT to shrink an image:
 - sample every other pixel

• Blur before you sample!

Image Aliasing Example

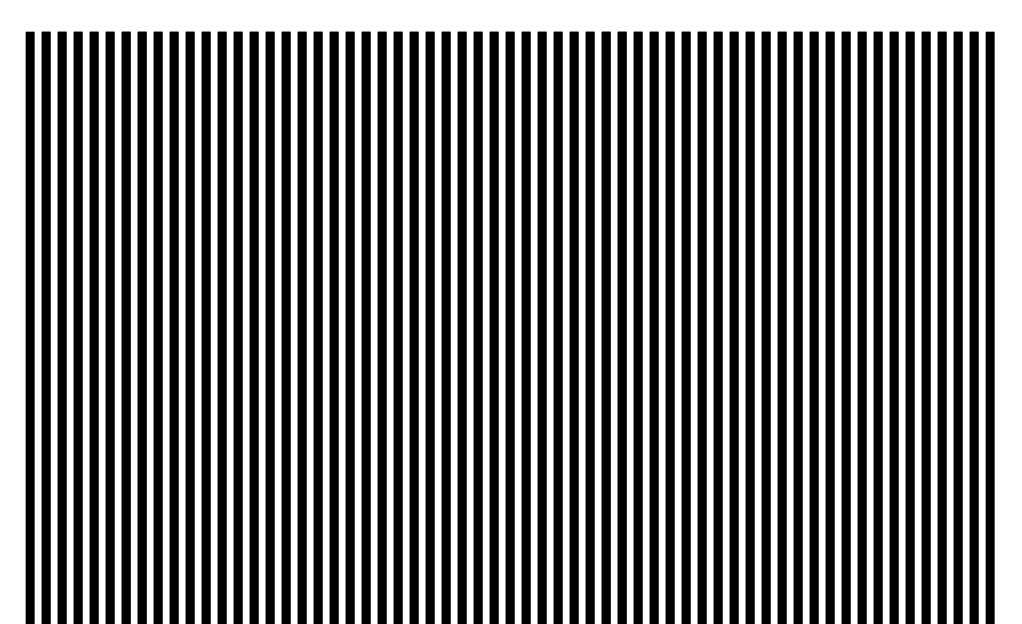


Image Aliasing Example – 1 sample/cycle

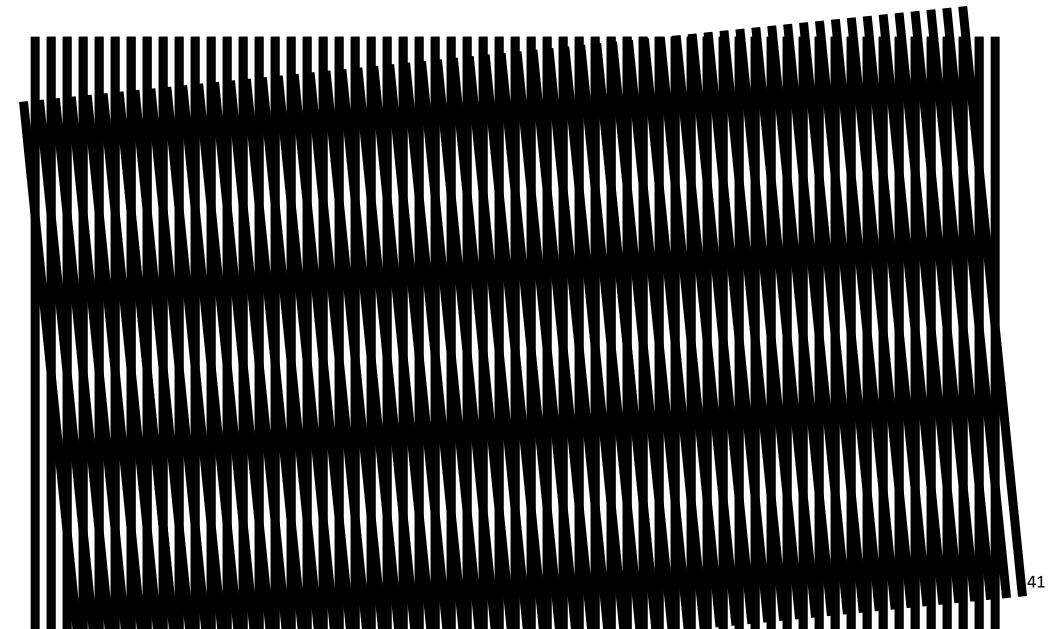


Image Aliasing Example – 3 samples/cycle

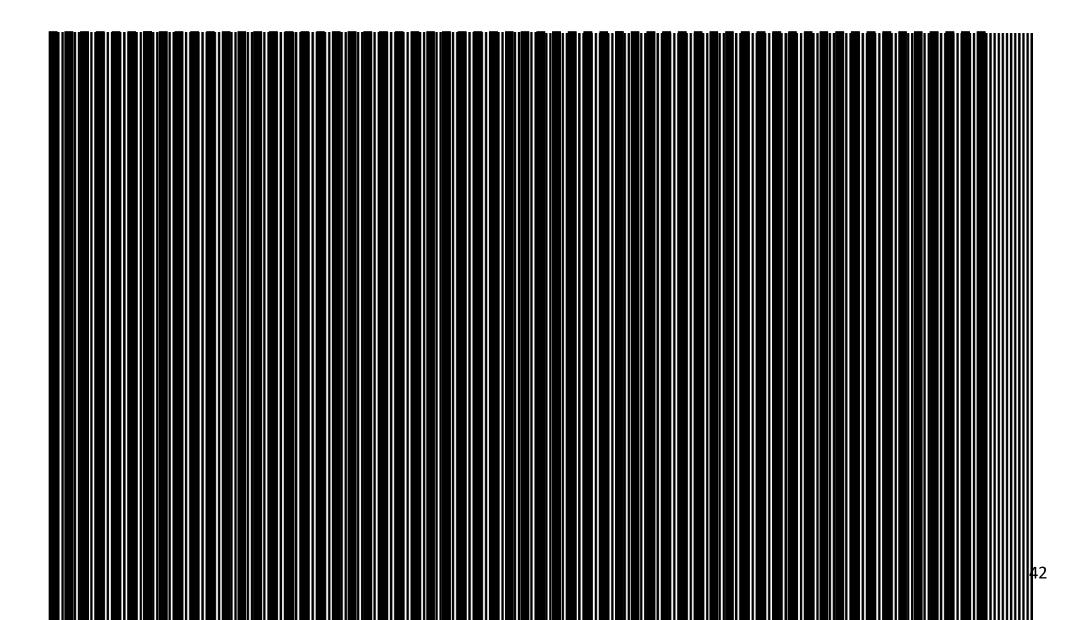
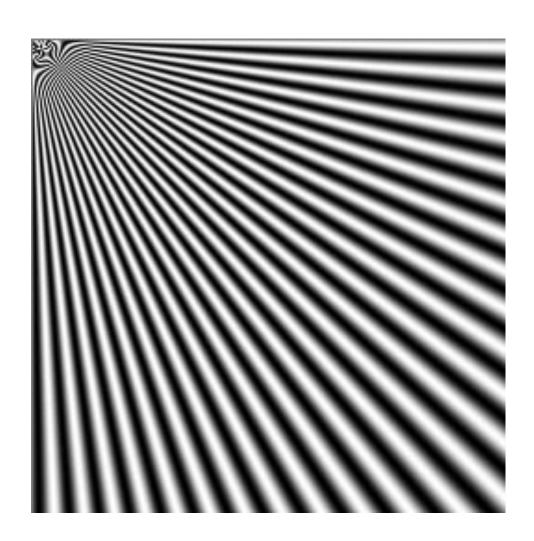


Image Aliasing Example



- Assume a column is 1 pixel wide. B/W cycle length is 2. Good sampling must be at distances <1.
- Sampling every second column will give either a solid black or a solid white.
- Blur before sample will give a solid gray regardless of shift.

Aliasing Example



Aliasing Example

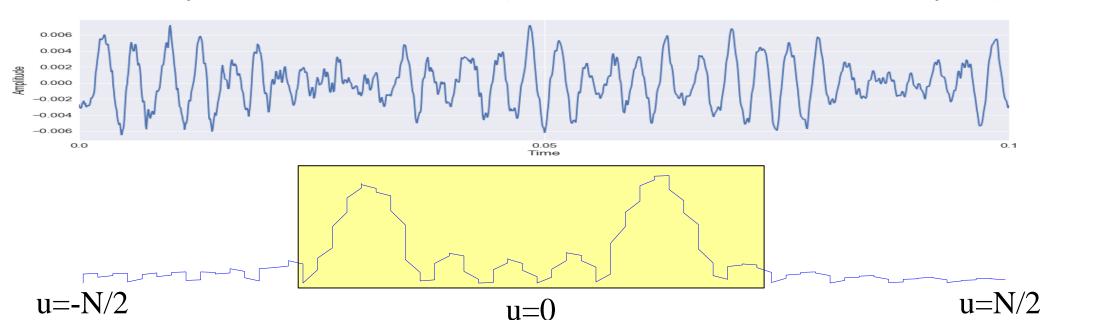


Signal Resizing (Example: *N samples* → *N/2 samples*)

Use Fourier

- $F(u) = \frac{1}{N} \sum_{x=0}^{N-1} f(x) e^{\frac{-2\pi i u x}{N}}$
- 1. Compute Fourier (N samples to N coefficients)
- 2. Bring u=0 to center (FFT Shift)

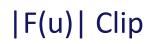
- $f(x) = \sum_{u=-N/4}^{N/4-1} F(u) e^{\frac{2\pi i u x}{N/2}}$
- 3. Crop Fourier from N to N/2 coefficients remove N/2 high frequencies
- 4. Compute Inverse Fourier (N/2 coefficients to N/2 Samples)

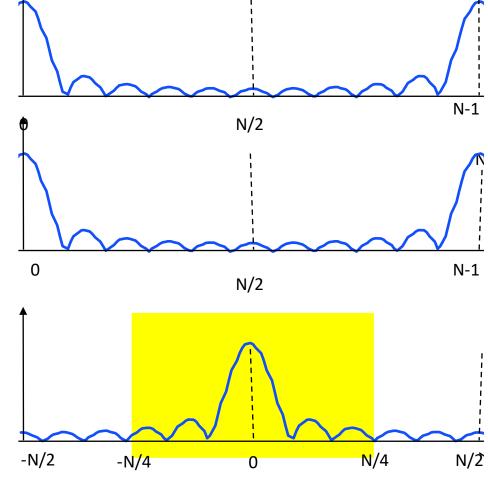


Resizing (1D)

FFT of Original Signal







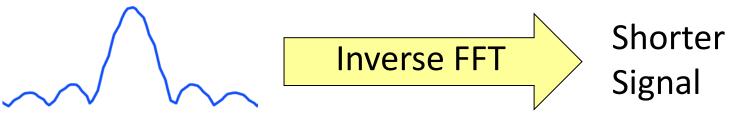
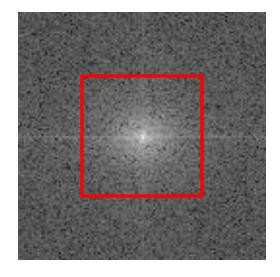


Image Reduction (E.g. $N \times N \rightarrow N/2 \times N/2$)

- 1. **Blur** and Subsample every 2nd pixel in every 2nd row
- 2. Use Fourier (Can resize to any new size)
 - 1. Compute Fourier $(N \times N)$, Shift (0,0) to center
 - 2. Crop Fourier (e.g. $N/2 \times N/2$): Ideal low pass
 - 3. Compute Inverse Fourier





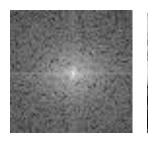
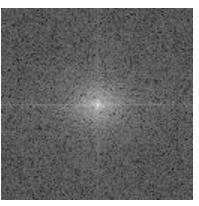


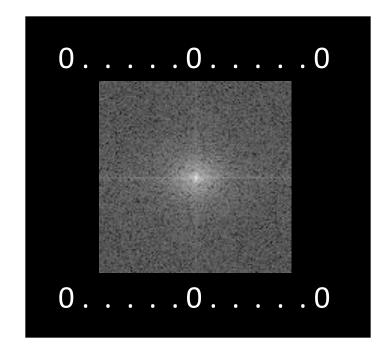


Image Expansion (E.g. $N/2 \times N/2 \rightarrow N \times N$)

- 1. Compute Fourier (*N×N*)
- 2. Pad Fourier with zeros (E.g. $2N \times 2N$)
- 3. Compute Inverse Fourier $(2N \times 2N)$

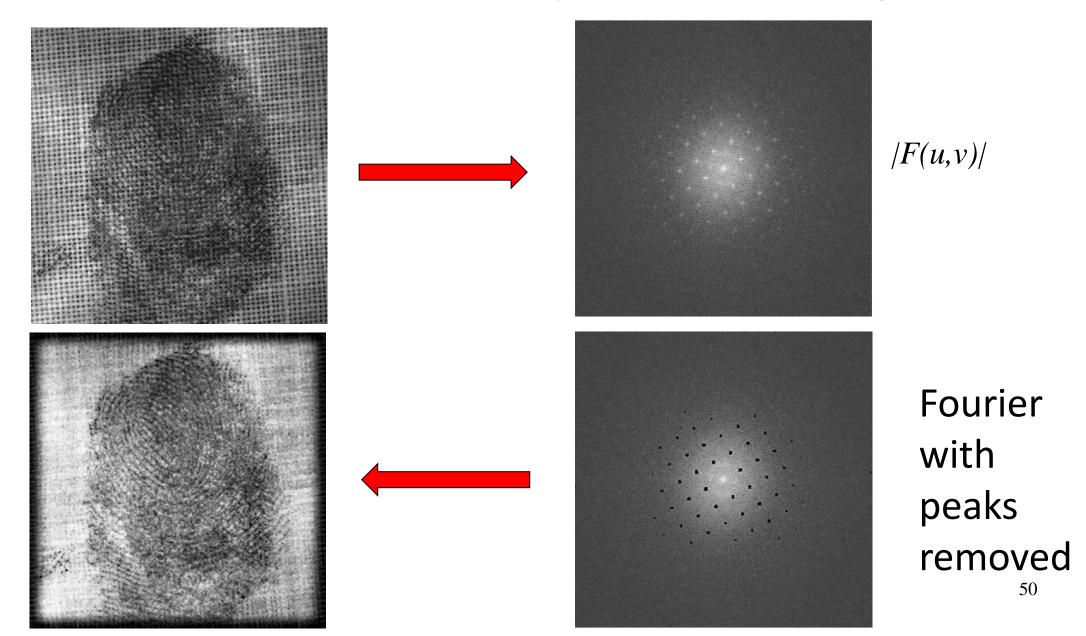








Enhancement – Remove periodic background



Enhancement – Remove periodic noise

