

# Digital Image Processing

(1)

## Lecture 7

### The 2D Discrete Fourier Transform

In Signals and Systems, DSS, we

Learned About The Fourier Transform

Continuous time

Fourier transforms

Discrete-time

Fourier transform

Discrete

Fourier Transform

$X[k]$  are samples of  $X(w)$  at evenly spaced intervals

No continuous stuff

→ 2D Discrete Fourier Transform: (DFT)

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

OUTPUT IMAGE  $(u \times v)$

Input Image  $M \times N$

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Rewrite it

$$F[u, v] = \sum_{x=0}^{N-1} \sum_{y=0}^{M-1} f[x, y] e^{-j\frac{2\pi ux}{N} - j\frac{2\pi vy}{M}}$$

INPUT IMAGE  $M \times N$

$u: 0 \dots N-1$

$v: 0 \dots M-1$

Inverse transform

$$f[x, y] = \frac{1}{MN} \sum_{u=0}^{N-1} \sum_{v=0}^{M-1} F[u, v] e^{j\frac{2\pi ux}{N} + j\frac{2\pi vy}{M}}$$

(IF  $M=N$ )

$$f[u, v] = \frac{1}{N} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f[x, y] e^{j\frac{2\pi ux}{N} + j\frac{2\pi vy}{N}}$$

$$f[u, v] = \frac{1}{N} \sum_{x=0}^{N-1} f[x, v] e^{j\frac{2\pi ux}{N}}$$

Another way of writing.

$$F[u, v] = \sum_{x=0}^{N-1} \left( \sum_{y=0}^{M-1} f[x, y] e^{-j\frac{2\pi vy}{M}} \right) e^{-j\frac{2\pi ux}{N}}$$

1D DFTs along columns

1D DFTs Along Rows

$\Rightarrow$  2D-DFT is SEPARABLE

(MATLAB: fft2)

$$\begin{matrix} & \sum_{x=0}^{N-1} & \sum_{y=0}^{M-1} \\ f[x, y] & \sum_{x=0}^{N-1} & \sum_{y=0}^{M-1} \\ & \sum_{y=0}^{M-1} & \end{matrix} = [r, c]$$

open top  
 $N \times M$

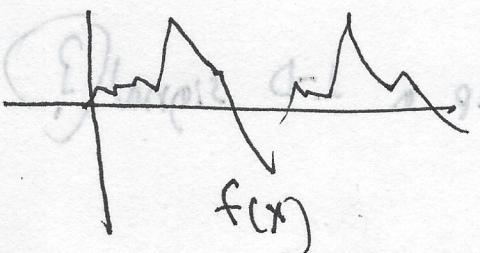
3D AM TUTORIAL  
 $C(N \times M)$

As with 1D DFT, 2D DFT is like a  
decomposition of an image into complex exponentials

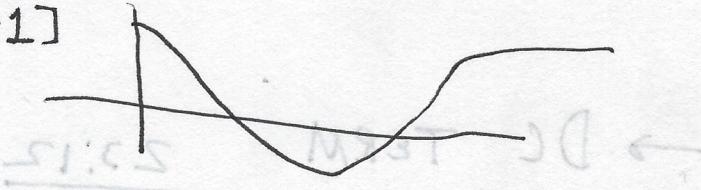
(sines + cosines)

$$-\frac{e^{j\frac{2\pi ux}{N}}}{c} = \cos\left(\frac{2\pi ux}{N}\right) - j\sin\left(\frac{2\pi ux}{N}\right)$$

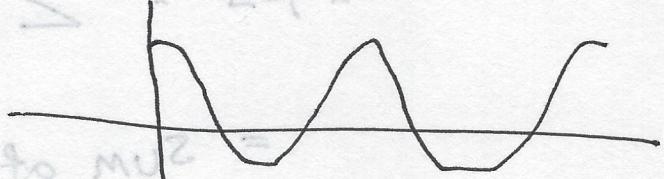
recall for 1-D signal



$$f(x) = F[0] + \sum F[k] \cos\left(\frac{2\pi kx}{N}\right) + F[-k] \sin\left(\frac{2\pi kx}{N}\right)$$



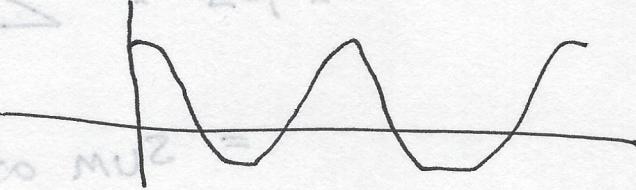
$$+ F[1], F[-1]$$



$$+ F[2], F[-2]$$

...  $\rightarrow$   $\sum F[k] \cos\left(\frac{2\pi kx}{N}\right)$

$$(x, x_0, \dots, x_N) \in \mathbb{C}^N$$



...  $\rightarrow$   $\sum F[k] \cos\left(\frac{2\pi kx}{N}\right)$

We are taking the original signal (Image) and writing it in terms of a different set of basis functions.

Value at each pixel versus weights on sinusoids

1H:ES do J term

④ useful for image compression (A representation of different basis)

MatLab: 14:57

We are seeing the basis function of the

Image

Are these figures some component of a real image, or are they parallel to the different

frequencies of the sinusoids that make up a 1-D signal?

thing up to make the image

→ DC TERM 23:12

What's

$$F[0,0] = \sum \sum f(x,y)$$

= sum of all pixel intensities

→ We have some really large number

Note: We want to show these (magnitude!) in the log scale to really see the other values.

Recall, white is a large number

MatLab 23:41

\* I should spend several hours working and doing these Matlab examples (5)

In the example, I see ~~that~~ that the middle of the image has very high frequencies.

Low frequency corner corresponds to the white noise

→ Due to the addition of the DFT

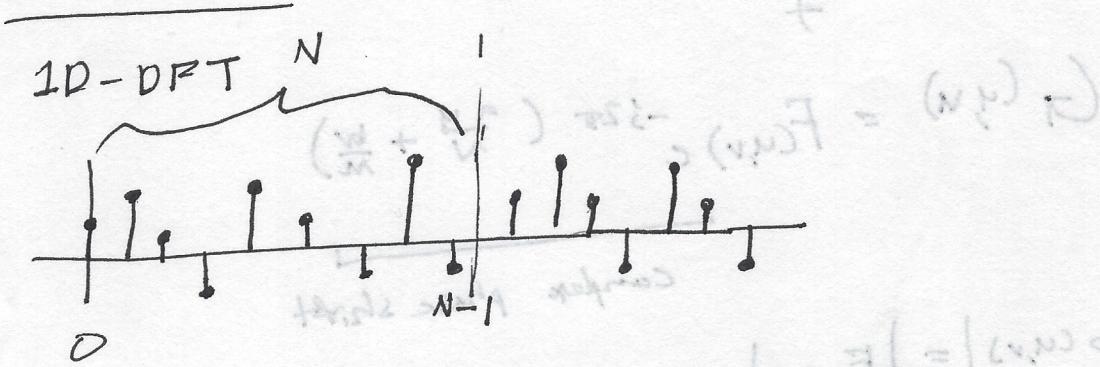
I suppose the sincs lessen the value of  $\text{Re}\{F[u,v]\}$

fft shift  $\approx \underline{25:18}$

↑ moves the DC value to the center

\* high frequency correspond edges image

\* low smoothly varying → low frequency



edges across image boundary

Issue



Is that image smoothly reconstructed? (corner is white)

or reorganized such that the edges are in pieces (center white)

6 Subways left 1st floor has obscure letters have more noise  
27:52 <sup>normal</sup>  
 shifting image we have tile images

→ edges in horizontal, vertical

Volcano Image 29:13  
 gray scale the image or you get odd colors

38:06

rotates set of colors 3D with camera

Fourier

Transform Properties

Shift

$$\text{if } g(x, y) = f(x-a, y-b)$$

$$G(u, v) = F(u, v) e^{-j2\pi \left( \frac{ay}{N} + \frac{bx}{M} \right)}$$

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

Scale / Flip:

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

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

$$g(x, y) = f(ax, by)$$

$$G(u, v) = \frac{1}{|ab|} F\left(\frac{u}{a}, \frac{v}{b}\right)$$

if  $a \text{ or } b = -1$

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flip in spatial domain  $\Leftrightarrow$  flip in frequency domain

(check HIR fact out)

Rotate:

if  $g(x,y) = f(x,y)$  Rotated CCW by  $\theta^\circ$

then  $G(u,v) = F(u,v)$  Rotated CCW by  $\theta^\circ$

(we are rotating all of the contribution of the cosines)

~~Homework~~ problem is to prove this property.

Convolution:

$$h(x,y) = f(x,y) \star g(x,y)$$

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

$\star$  = circular convolution

Note: In Image Processing, the filters we use are very small compared to the image we are trying to process.

1Kx1K vs 7x7 or 3x3

so Matlab may be doing these filters in the spatial domain versus the frequency domain  $\Rightarrow$  (easy computations)

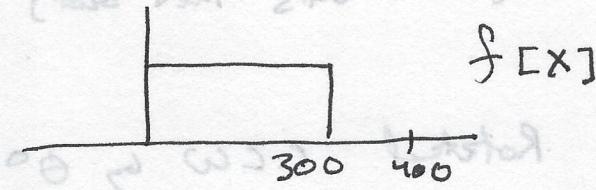
⑧

recall: circular convolution

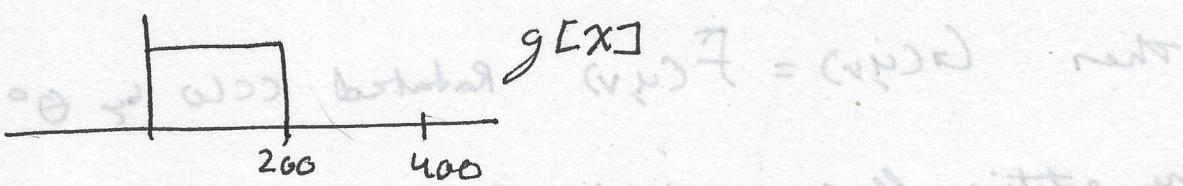
$f \circledast g = f \otimes g$

→ regular convolution

(two test 2D plots)

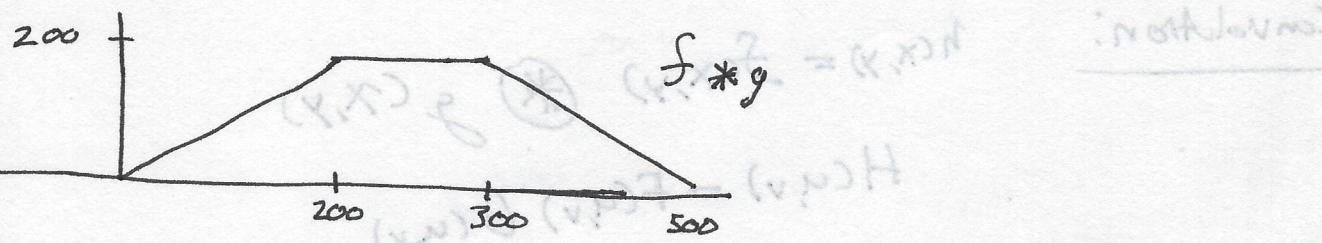
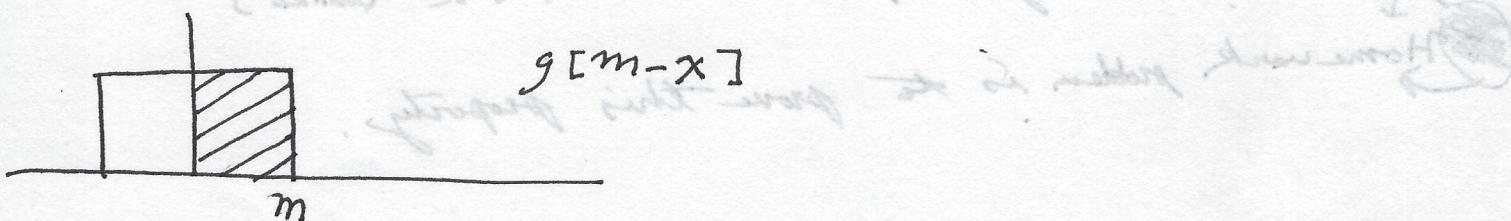


$$(f, g)^\text{F} = (f \circledast g)^\text{F}$$



(convolution with no rotations w.r.t. the pointer in  $m$ )

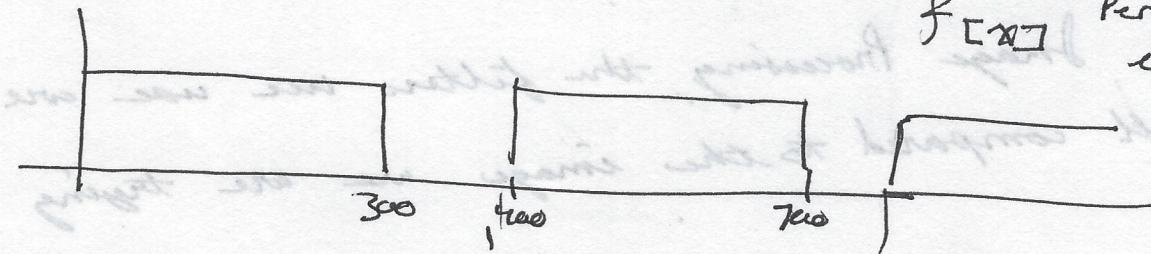
$g[m-x]$



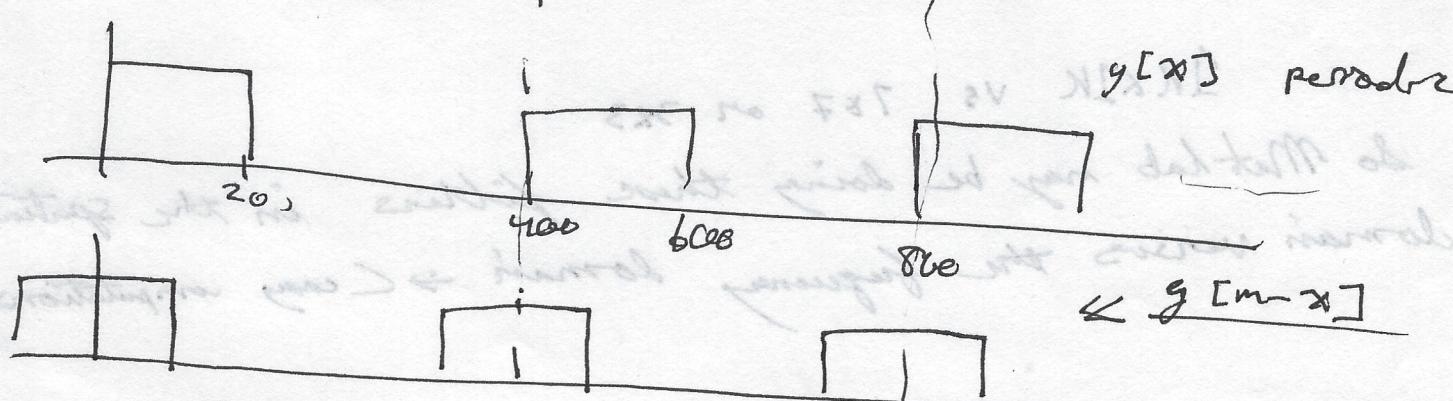
Circular convolution

no boundaries, natural =

$f[x]$  Periodically extended



$g[x]$  periodic

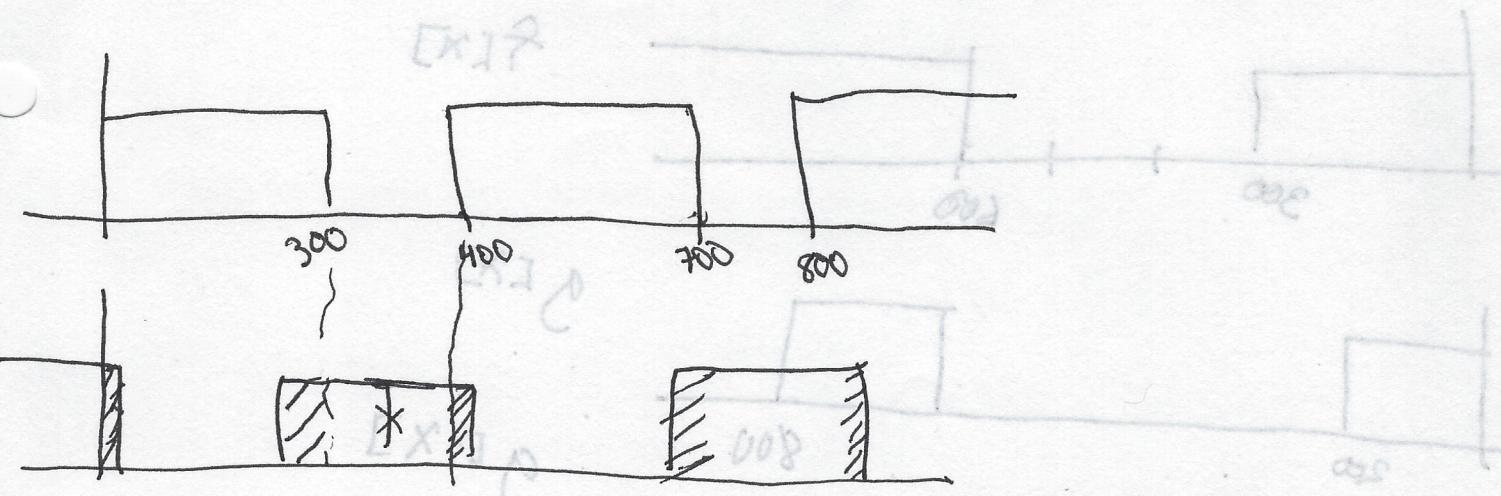


$\leftarrow g[m-x]$

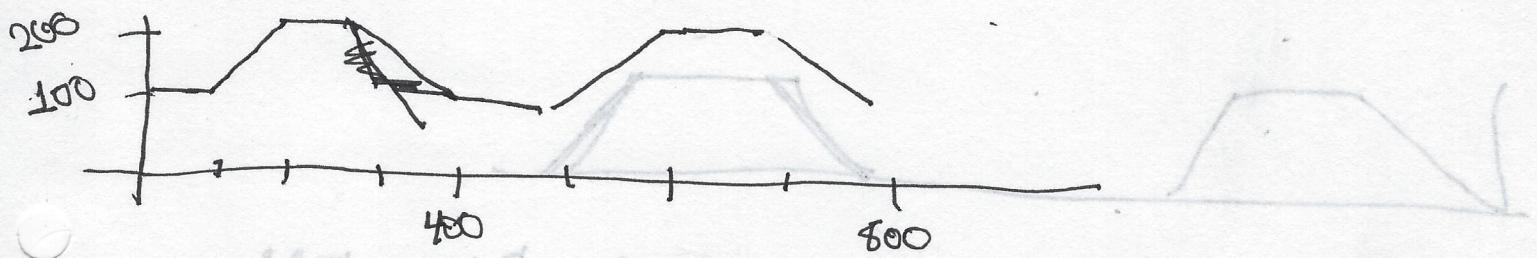
if  $g[n-x]$  is further back

008 008

(9)



Do our Periodic convolutions w/

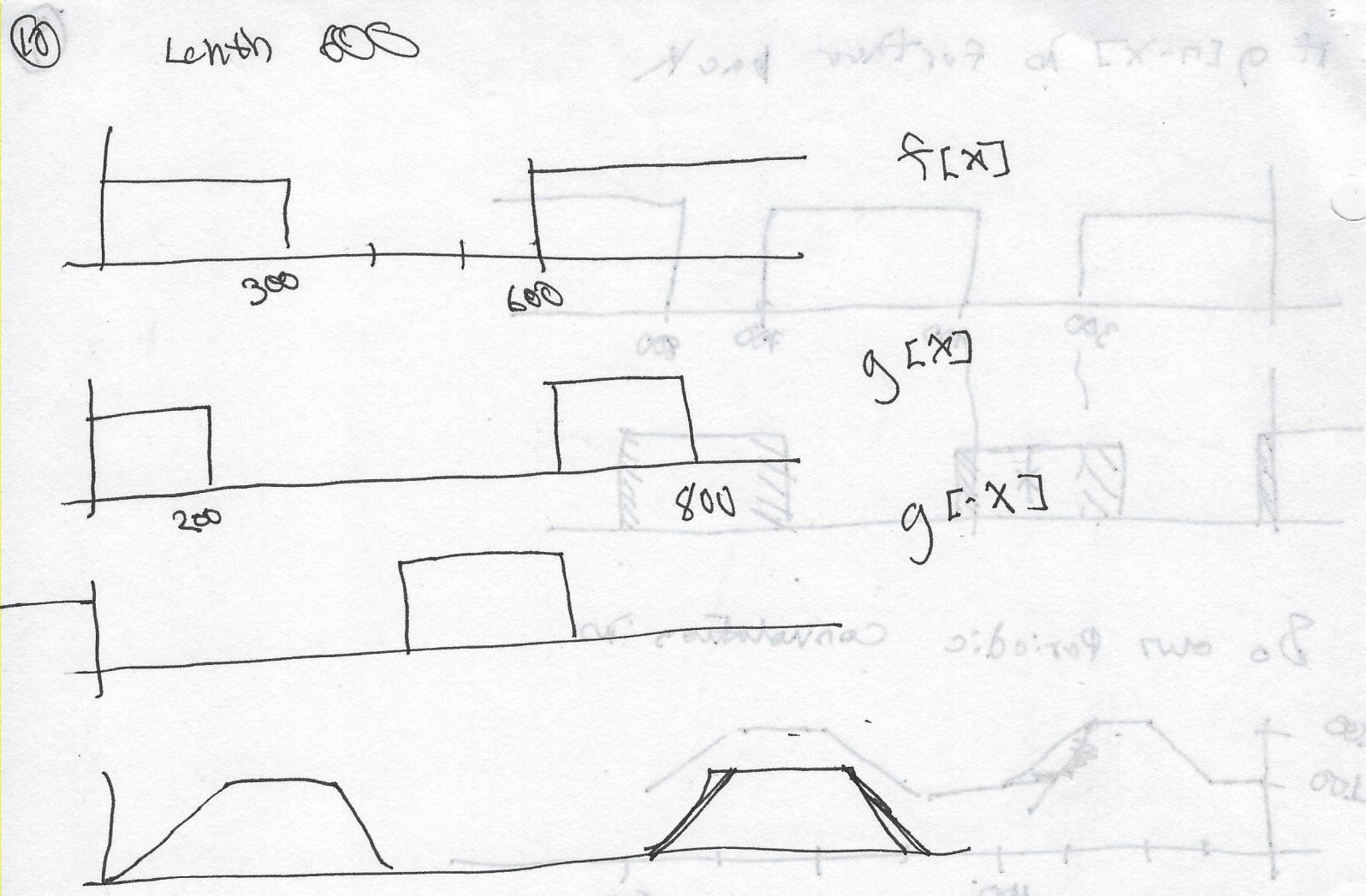


So  $g[n]$  is a periodic signal. It's like  $\sum_{k=0}^{\infty} g[n-kM]$

Problem May be: This may not be what you want

Since copies may introduce into regions we don't expect.

Solution: zero padding signal so that copies don't unexpectedly overlap  
→ extend period



zero-padding  $T_0$  at least Dimension

$$(M_1 + M_2 - 1) \times (N_1 + N_2 - 1)$$

- (Dimension of Images)

For circular convolution  
between  $M_1 \times N_1$   
 $M_2 \times N_2$  IMAGES

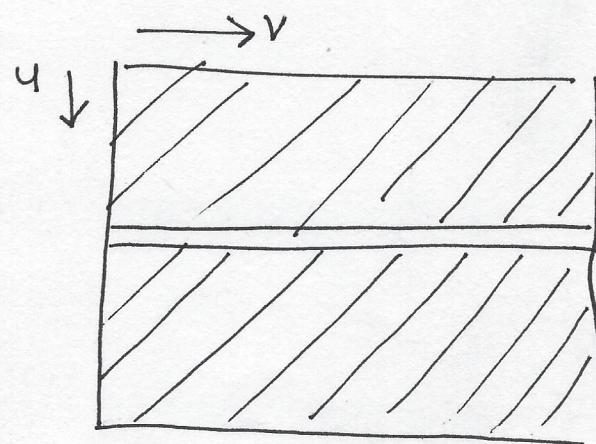
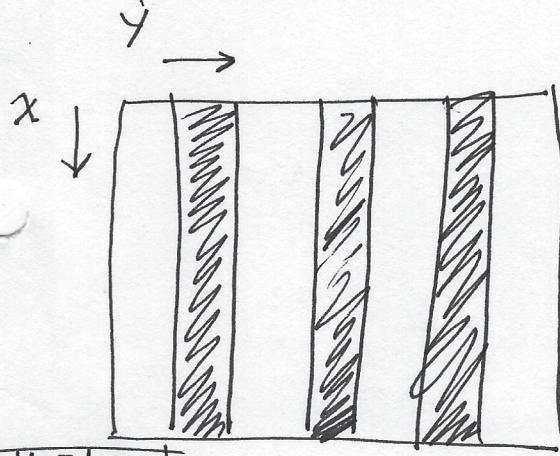
IS sufficient

Good News: filter/impulse In Matlab Does  
all this for you

MatLab crops down to get the image the user wants

NOTE: If an image has strong image edges at  $0^\circ$ , we see a strong contribution in the 2D-DFT at  $0+90^\circ \rightarrow$  why is that?

→ MatLab 1:00:35



$$\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} [255 \ 0 \ 255 \ 0 \ 255 \dots] \xrightarrow{\text{DFT}} \text{DC}$$

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} [0001000001000\dots]$$

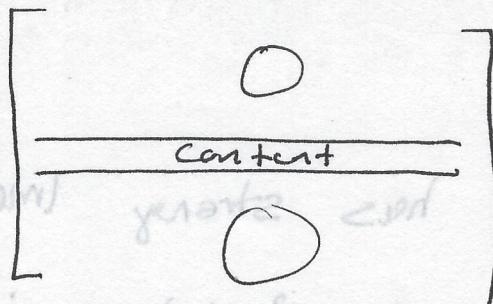
nothing in the middle  
→ DC, high frequency is away from DC.

(DC term in the middle)

the DFT is separable, we can take the DFT of both vectors

(n)

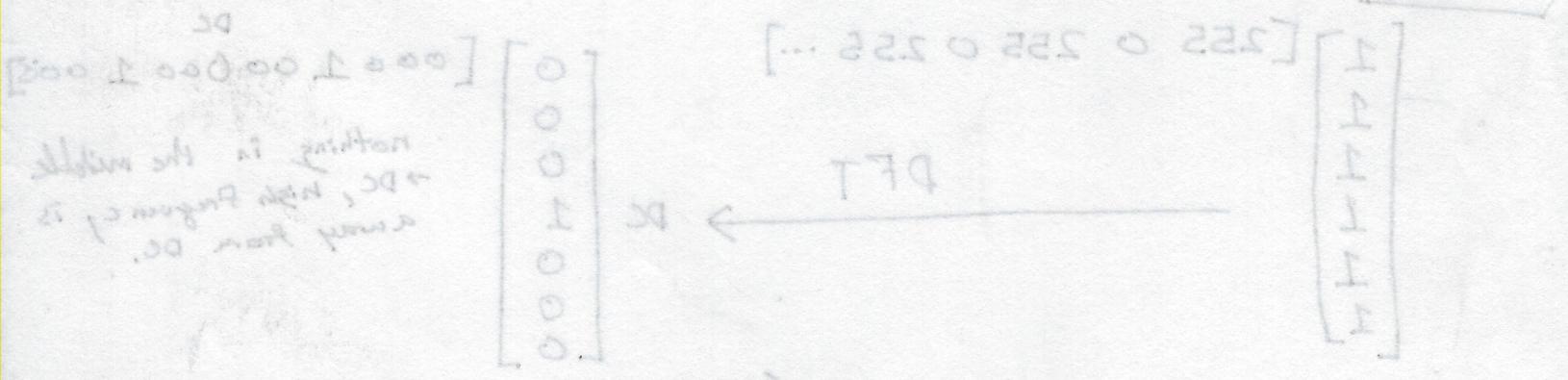
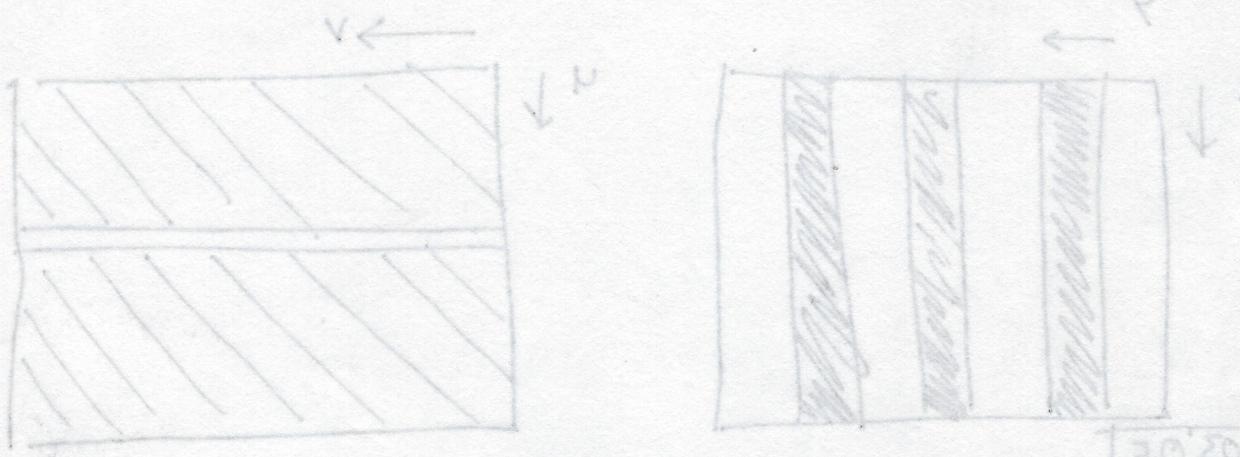
So as a result of most errors last fall  
from now on



the error generates and upon no fit bottom

The left in most fitting points we will do to  
stuff is pdw.  $\leftarrow$  DFT to TFD

DFT bottom



(Matters not DC)

TFD with what was we, difference in TFD at  
another time so