

Image Processing - two-dim. DSPSimple Property of two-dim. FFT:

IF Special case $y_{mn} = y_m * y_n$ (image factors into
fctn. of x times fctn. of y)

then transform is

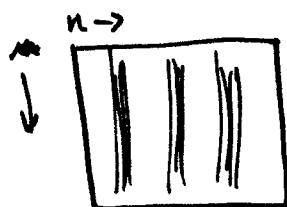
$$\begin{aligned}
 Y_{kl} &= \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} (y_m * y_n) e^{-j \frac{2\pi k m}{N}} e^{-j \frac{2\pi l n}{N}} \\
 &= \left(\sum_{m=0}^{N-1} y_m e^{-j \frac{2\pi k m}{N}} \right) * \left(\sum_{n=0}^{N-1} y_n e^{-j \frac{2\pi l n}{N}} \right) \\
 &= Y_k * Y_l \\
 &\quad \uparrow \quad \uparrow \\
 &\quad \text{transform} \quad \text{transform} \\
 &\quad \text{of } m\text{-fctn.} \quad \text{of } n\text{-fctn.}
 \end{aligned}$$

(then transform also factors)

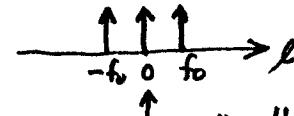
EXAMPLE:
("washboard")

Sinusoid in horizontal direction (n)

constant in vertical direction (m)

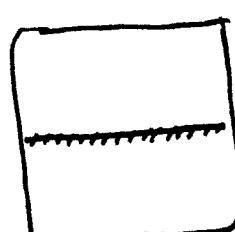
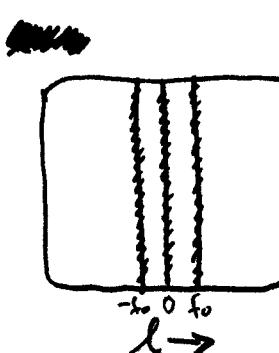


transform of sinusoid:

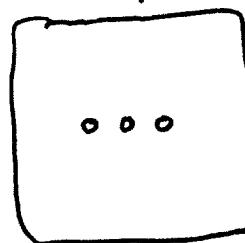


"DC"
Component -
because intensity
is always nonnegative

transform of constant:



=



\leftrightarrow
total
transform

```
#include "ppm.h"
#include <math.h>
```

washboard.c

5.3.2

```
void
main()
{
int x, y, width, height;
double tx, ty, temp;
#define pi (3.14159265358979)
#define f (8.)
#define window 0
Image *image1;

image1 = ImageCreate(256, 256);
ImageClear(image1, 255, 255, 255); /* all white */

width = ImageWidth(image1);
height = ImageHeight(image1);

for (y = 0; y < height; y++){
    for (x = 0; x < width; x++){
        if(1){
            temp = sin(2.*f*pi*((double)x/(width-1)));
            tx = 255.*temp*temp;
            if(window==1) tx *= (0.54 + 0.46*cos(2.*pi*x/(width-1)-pi));
            ImageSetPixel(image1, x, y, 0, tx );
            ImageSetPixel(image1, x, y, 1, tx );
            ImageSetPixel(image1, x, y, 2, tx ); }}}

ImageWrite(image1, "washboard.ppm");
}
```

Sinusoid in x-direction

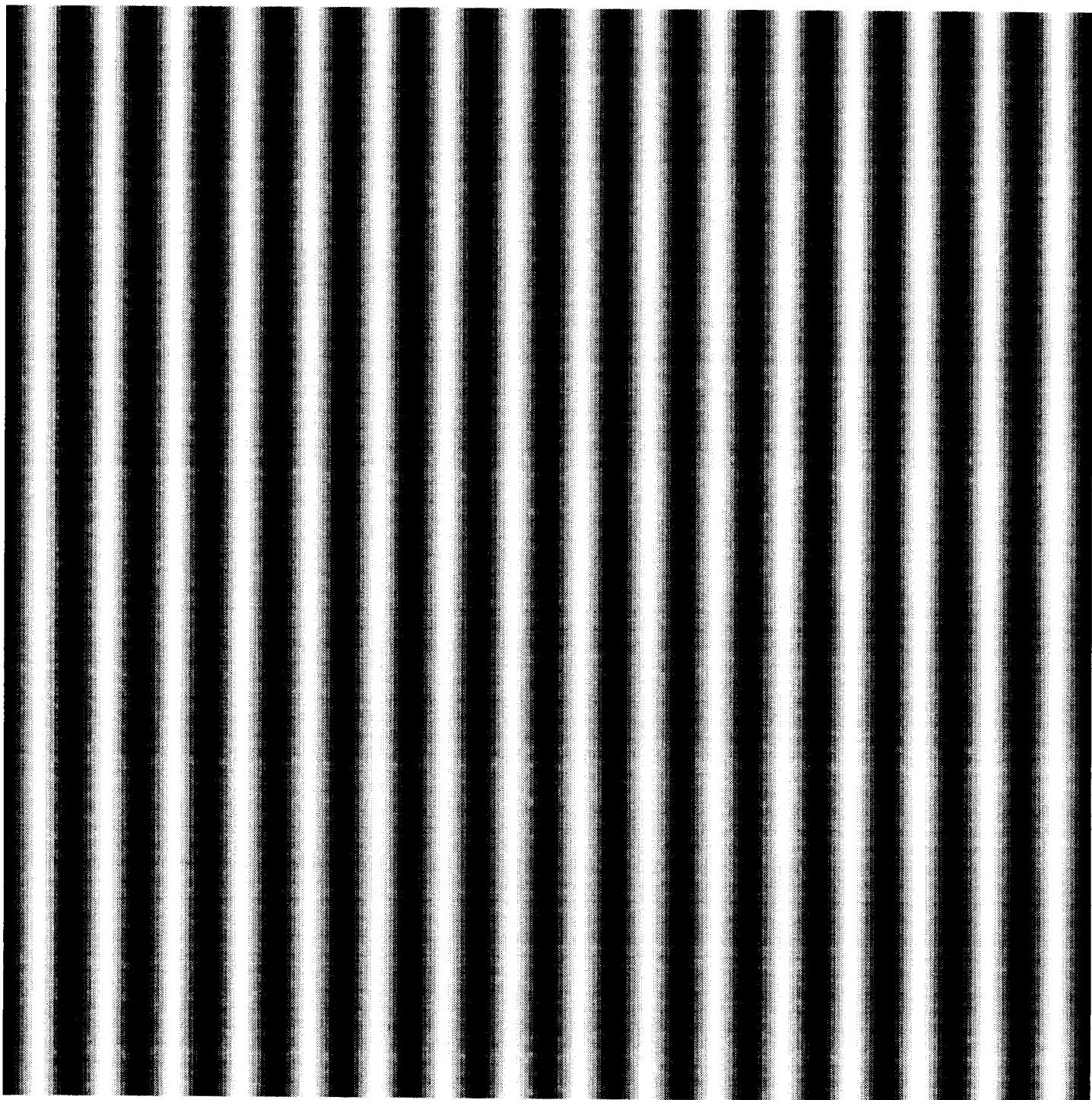
$$\sin^2 \theta = \frac{1}{2}(1 - \cos 2\theta)$$

0-freq.



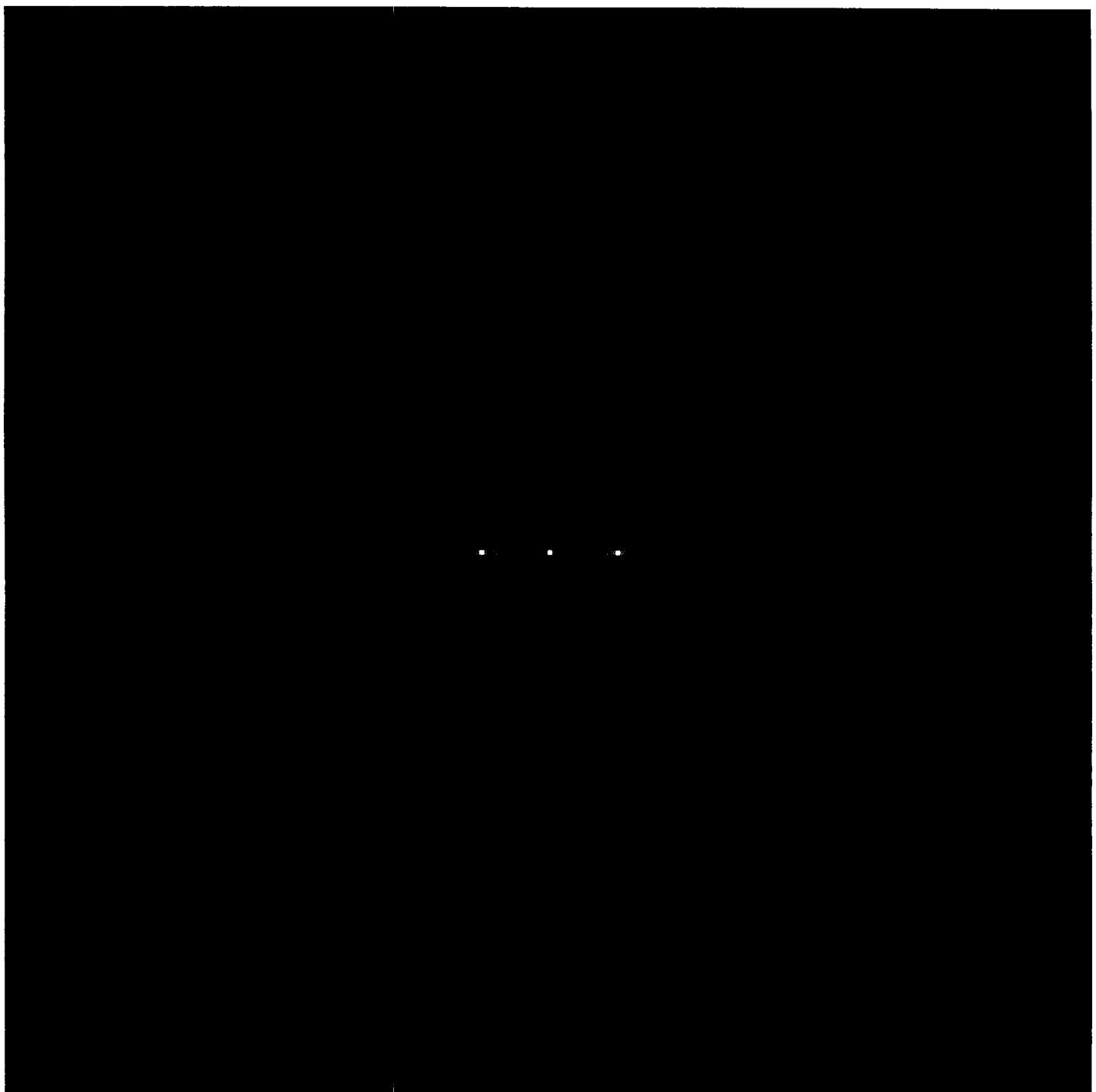
washboard image

5.3.3

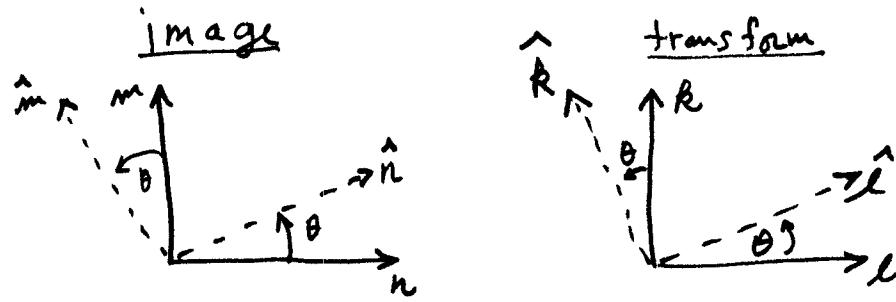


5.2.4

5.2.4



Another Impotent Property of two-dim. FFT:



rotate image by $\Theta \Rightarrow$ rotate transform by Θ

Proof

$$\text{transform } Y_{kl} = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} y_{mn} e^{-j \frac{2\pi}{N} (mk + nl)}$$

this is dot product between two vectors

$$\begin{bmatrix} n \\ m \end{bmatrix} \cdot \begin{bmatrix} l \\ k \end{bmatrix} = \text{length}\left(\begin{bmatrix} n \\ m \end{bmatrix}\right) \cdot \text{length}\left(\begin{bmatrix} l \\ k \end{bmatrix}\right) \cdot \cos \gamma_{\text{between}}$$

- invariant with respect to coordinate system rotation

$$\therefore \hat{m} \hat{k} + \hat{n} \hat{l} = mk + nl \quad (\hat{m}, \hat{n}; \hat{k}, \hat{l} \text{ are rotated})$$

$$\therefore Y_{kl} = \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} y_{mn} e^{-j \frac{2\pi}{N} (\hat{m} \hat{k} + \hat{n} \hat{l})}$$

↑ rotated transform ↑ rotated image

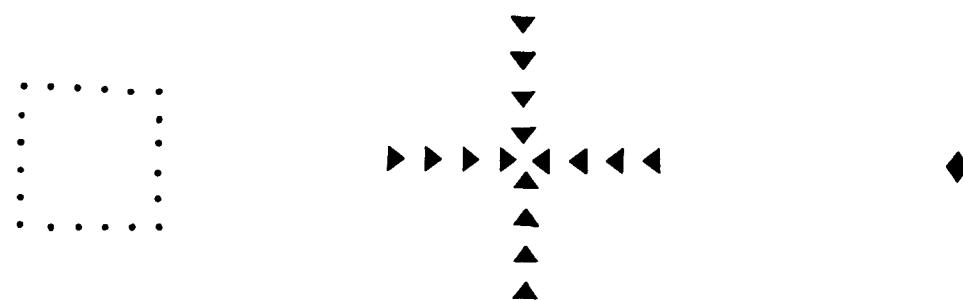
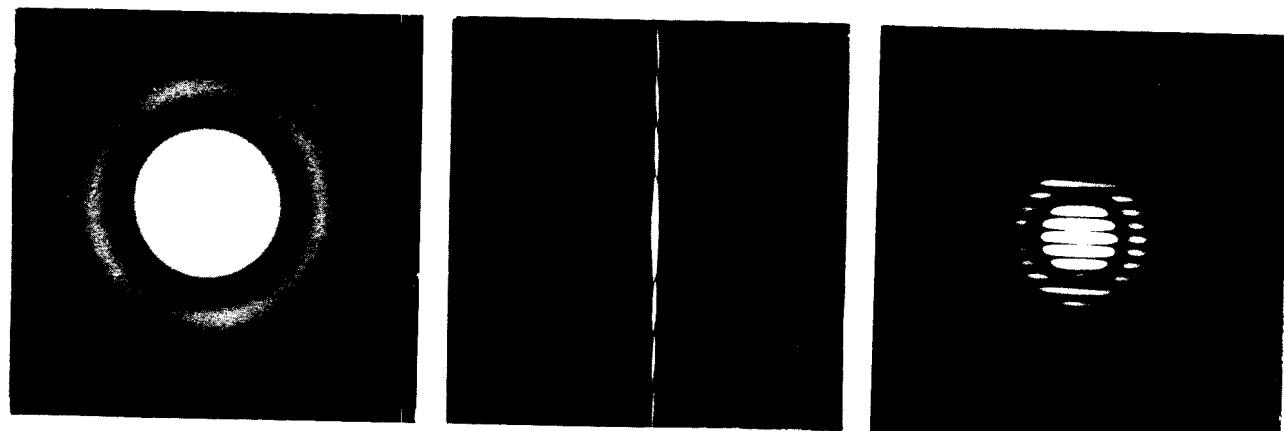
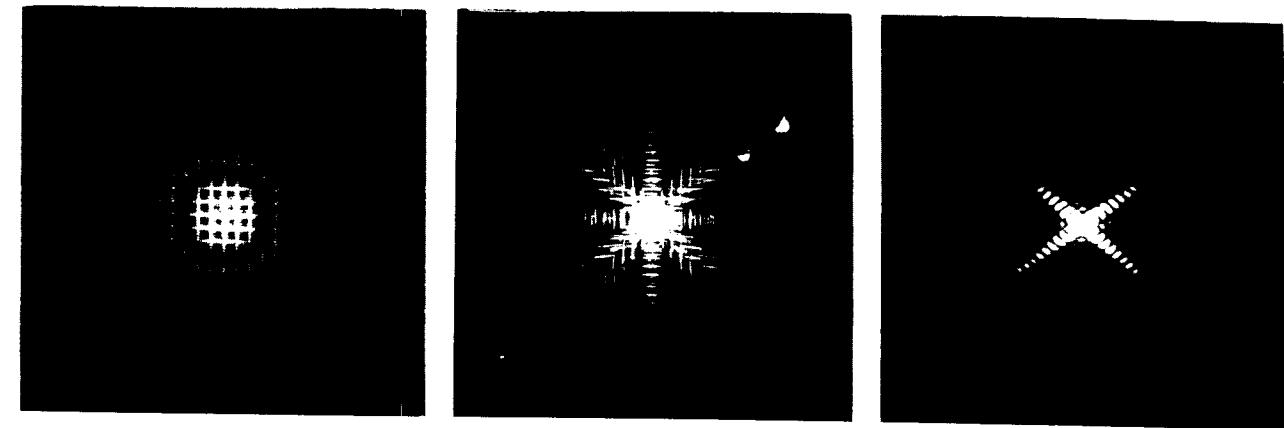
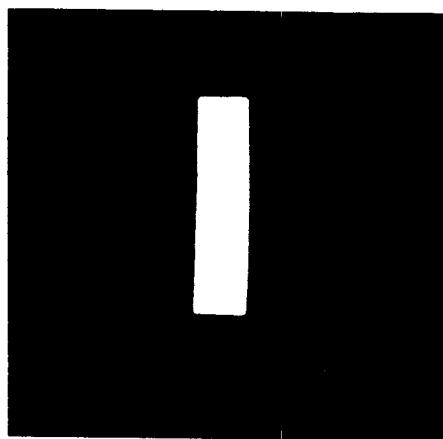
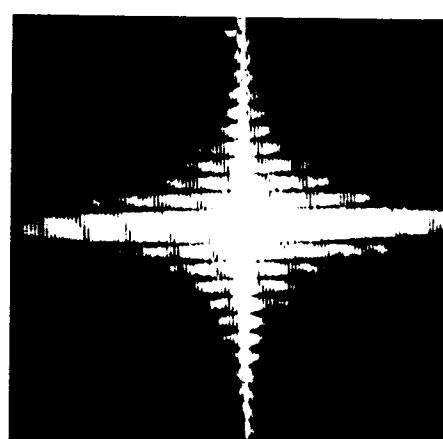


Figure 3.3 Some two-dimensional functions and their Fourier spectra.

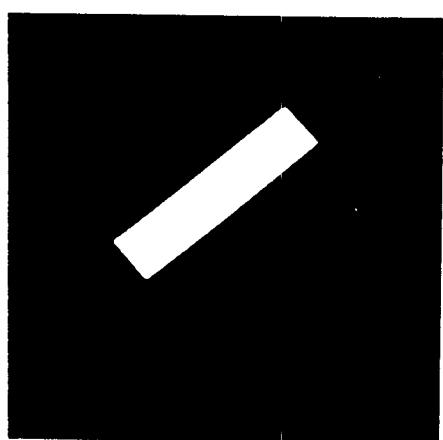
Figure 3.3 (Continued.)



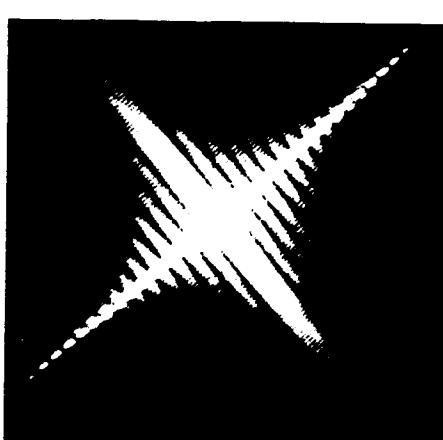
(a)



(b)

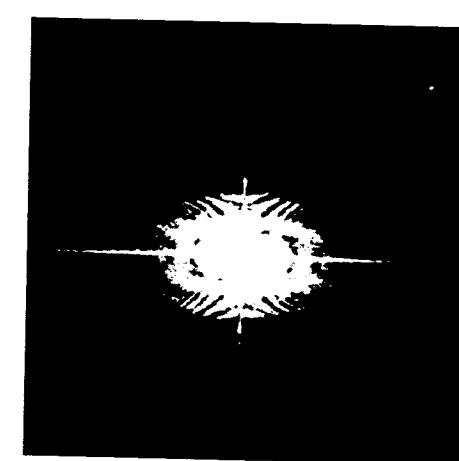
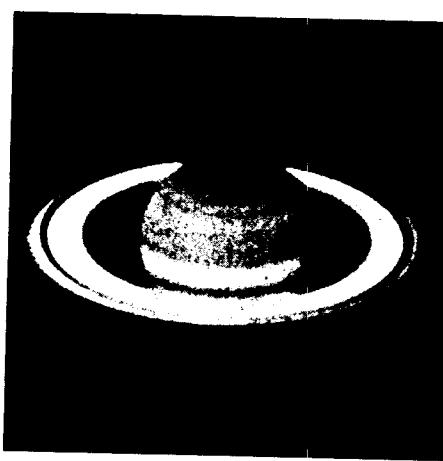


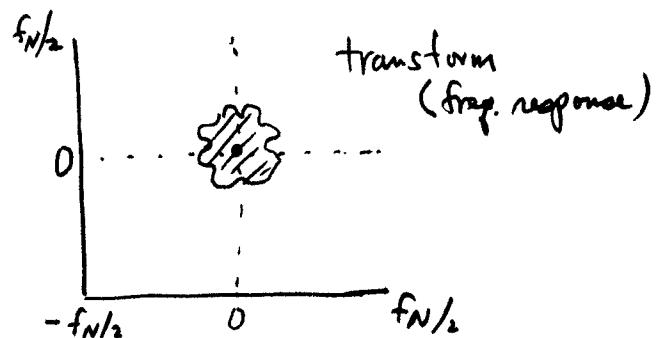
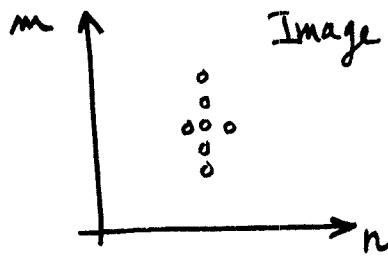
(c)



(d)

ILLUSTRATING ROTATION



third Property of two-dim. FFT:BALK to filtering

template, molecule:

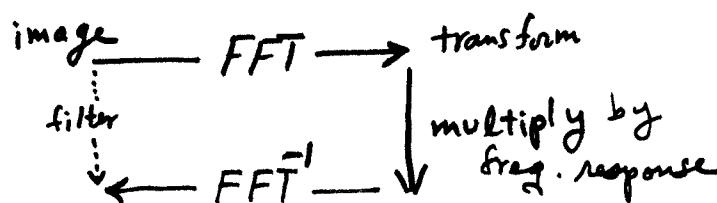
replace each point
by weighted average
of neighbours

called "convolution"
or "filtering"

multiply by transform of
template



filtering \Leftrightarrow multiplication

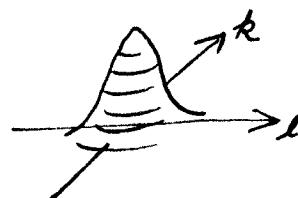
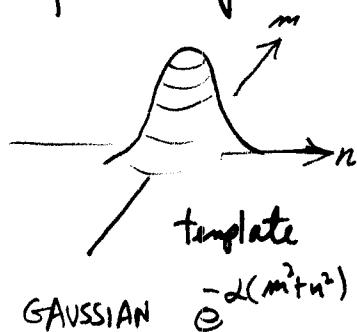


Guideline:

concentrated
in
Space domain \Leftrightarrow broad
in
freq. domain

& Vice-versa

"blur" in photoshop

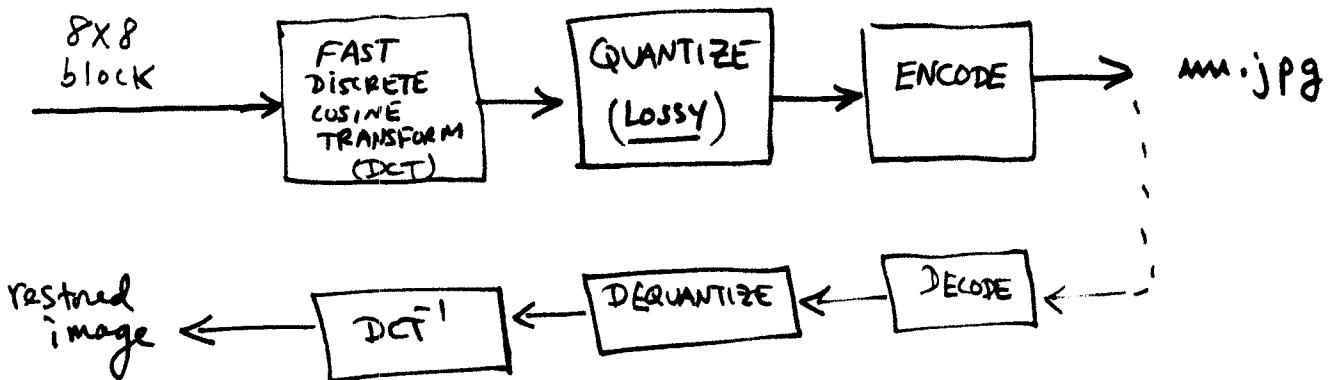


transform of template
GAUSSIAN $e^{-\beta(k^2+l^2)}$

GAUSSIAN is its own transform

JPEG BASELINE IMAGE CODING ALGORITHM [Kou 95]

Use 8×8 blocks



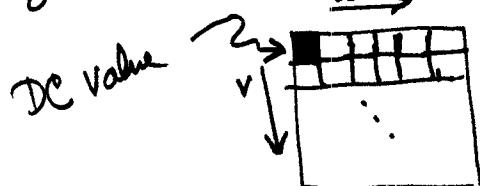
$$\text{DCT: } S_{vu} = \frac{1}{4} C_u C_v \sum_{x=0}^7 \sum_{y=0}^7 s_{yx} \cos \frac{(2x+1)u\pi}{16} \cos \frac{(2y+1)v\pi}{16}$$

$$\text{IDCT: } s_{yx} = \frac{1}{4} \sum_{u=0}^7 \sum_{v=0}^7 C_u C_v S_{vu} \cos \frac{(2x+1)u\pi}{16} \cos \frac{(2y+1)v\pi}{16}$$

$$\text{where } C_m = \begin{cases} 1/\sqrt{2} & m=0 \\ 1 & \text{else} \end{cases}$$

like FFT - has fast recursive algorithms
that have been hacked to death

yields a block of size 8×8 holding the transform S_{vu}



Each transform element is quantized by

$$\hat{S}_{vu} = \text{round} \left[\frac{S_{vu}}{Q_{vu}} \right] \quad (\text{integer})$$

$Q = \begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{bmatrix}$ is 8×8 fixed matrix provided as a plug-in

small $Q \Rightarrow$ many levels possible, accurate quantization

large $Q \Rightarrow$ few levels possible, rough quantization

to (approximately) restore, $\hat{\mathbf{x}}_{\text{rec}} \approx Q_{\text{rec}} \hat{\mathbf{x}}_{\text{rec}}$



not perfectly restored

\rightarrow "lossy" compression

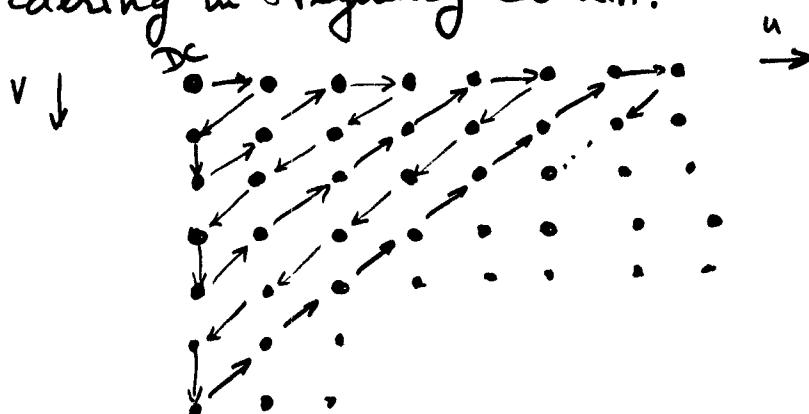
this is first source of compression -

most information is contained in low-frequency

components, so we make Q ~~large~~^{small} for low frequencies,
large for high frequencies

the DC component $[0,0]$ element is especially important,
and is treated separately, using ~~the~~ change from previous
block. (Second source of compression)

to encode Non-DC components, use zig-zag
ordering in frequency domain:



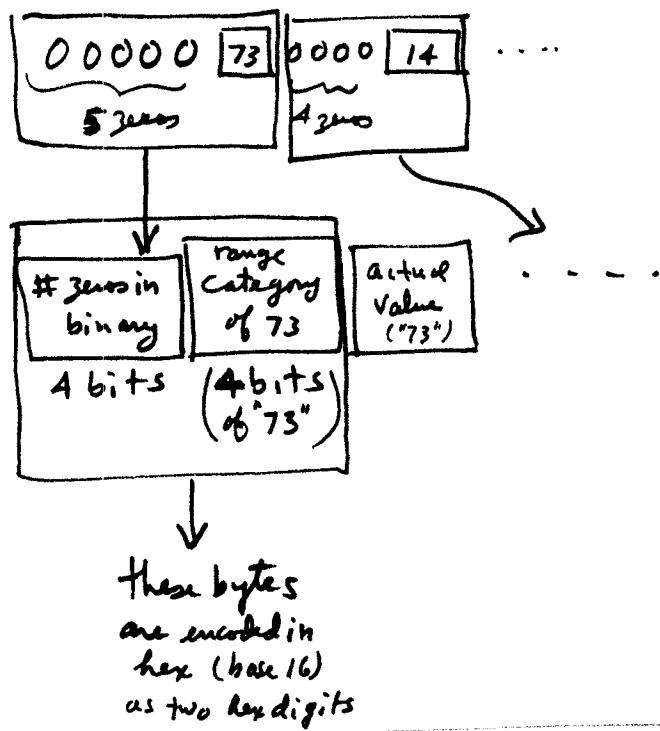
this puts low-frequency information first.

We then use run-length coding, looking for runs of consecutive zeros.

Special code values:

$\text{EOB} = "00000000"$ (byte) = rest of block is zero
$\text{ZRL} = "11110000"$ (byte) = run of 16 zeros

Runs of zeros of length less than 16 are encoded as



the fact that the zig-zag ordering may result in many zeros at the end, & in general produces runs of zeros, is a third source of compression

BTW, JPEG = Joint Photographic Exerts Group

many elaborations - JPEG baseline is simplest & most common