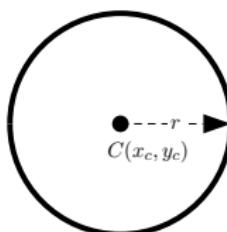


# CM2202: Scientific Computing and Multimedia Applications

## Lab Class Week 8

School of Computer Science & Informatics

# Circles



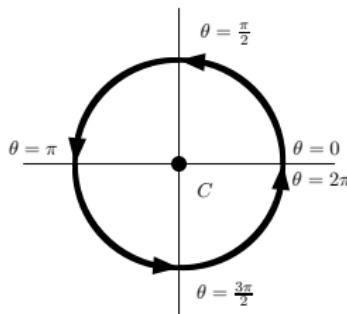
The implicit equation of a circle is the standard formula:

$$(x - x_c)^2 + (y - y_c)^2 - r^2 = 0$$

where the centre of the circle is  $C(x_c, y_c)$  and  $r$  is the radius of the circle.

This form is commonly used for whole circles.

# Circle (parametric form)



The parametric equation of a circle is given by:

$$\begin{aligned}x &= x_c + r \cos(\theta) \\y &= y_c + r \sin(\theta)\end{aligned}$$

- Parameterisation in terms of angle subtended at the circle centre,  $C$ .

# MATLAB Circle code

To create  $n$  points,  $p$ , equally space on circle of **centre** and radius, **r**:

Implicit form, [circle.imp.points.2d](#):

```
for i = 1 : n
    theta = ( 2.0 * pi * ( i - 1 ) ) / n;
    p(1,i) = center(1) + r * cos ( theta );
    p(2,i) = center(2) + r * sin ( theta );
end
```

Parametric form, is similar.

# Fourier Transform in MATLAB

## fft() and fft2()

MATLAB provides functions for 1D and 2D **Discrete Fourier Transforms (DFT)**:

- fft(X)** is the 1D discrete Fourier transform (DFT) of **vector** X. For **matrices**, the FFT operation is applied to **each column** — **NOT** a 2D DFT transform.
- fft2(X)** returns the 2D Fourier transform of matrix X. If X is a vector, the result will have the same orientation.
- fftn(X)** returns the N-D discrete Fourier transform of the **N-D array X**.

**Inverse DFT** **ifft()**, **ifft2()**, **ifftn()** perform the **inverse** DFT.

See appropriate MATLAB **help/doc** pages for **full details**.

Plenty of examples to Follow.

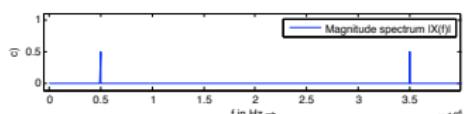
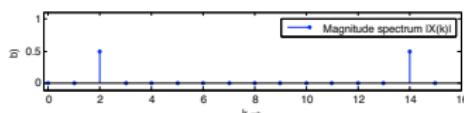
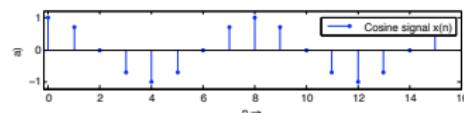
See also: **MATLAB Docs Image Processing → User's Guide → Transforms → Fourier Transform**

# Visualising the Fourier Transform

## Visualising the Fourier Transform

Having computed a DFT it might be useful to visualise its result:

- It's useful to visualise the Fourier Transform
- Standard tools
- Easily plotted in MATLAB



# The Magnitude Spectrum of Fourier Transform

Recall that the Fourier Transform of our **real** audio/image data is always **complex**

- **Phasors:** This is how we encode the **phase** of the underlying signal's **Fourier Components**.

How can we visualise a complex data array?

Back to Complex Numbers:

Magnitude spectrum **Compute the absolute value of the complex data:**

$$|F(k)| = \sqrt{F_R^2(k) + F_I^2(k)} \text{ for } k = 0, 1, \dots, N - 1$$

where  $F_R(k)$  is the **real** part and  $F_I(k)$  is the **imaginary** part of the  $N$  sampled Fourier Transform,  $F(k)$ .

**Recall MATLAB:** `Sp = abs(fft(X,N))/N;`  
**(Normalised form)**

# The Phase Spectrum of Fourier Transform

## The Phase Spectrum

### Phase Spectrum

The Fourier Transform also represent phase, the **phase spectrum** is given by:

$$\varphi = \arctan \frac{F_I(k)}{F_R(k)} \text{ for } k = 0, 1, \dots, N - 1$$

Recall MATLAB: `phi = angle(fft(X,N))`

# Relating a Sample Point to a Frequency Point

When **plotting graphs** of *Fourier Spectra* and doing other DFT processing we may wish to **plot** the x-axis in **Hz (Frequency)** rather than **sample point** number  $k = 0, 1, \dots, N - 1$

There is a **simple relation** between the two:

- The sample points go in steps  $k = 0, 1, \dots, N - 1$
- For a given sample point  $k$  the frequency relating to this is given by:

$$f_k = k \frac{f_s}{N}$$

where  $f_s$  is the *sampling frequency* and  $N$  the **number** of samples.

- Thus we have **equidistant frequency steps** of  $\frac{f_s}{N}$  ranging from 0 Hz to  $\frac{N-1}{N} f_s$  Hz

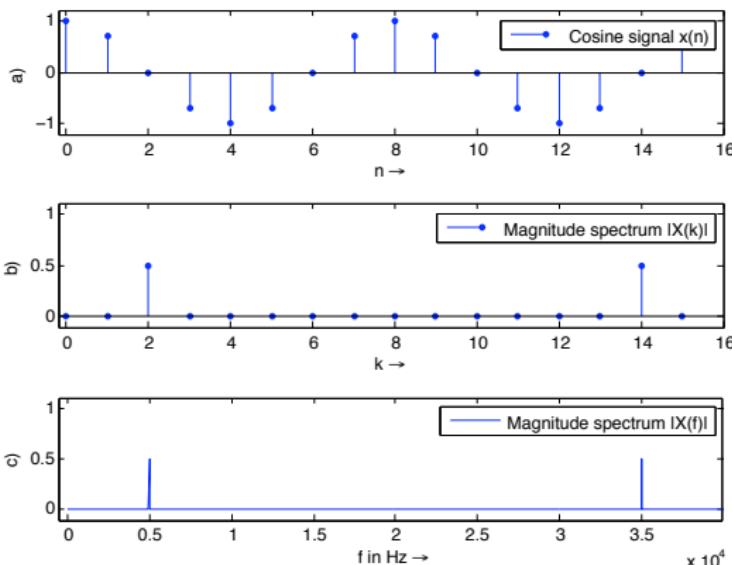
# MATLAB Fourier Frequency Spectra Example

## fourierspectraeg.m

```
N=16;  
x=cos(2*pi*2*(0:1:N-1)/N)';  
  
figure(1)  
subplot(3,1,1);  
stem(0:N-1,x,'.');//  
axis([-0.2 N -1.2 1.2]);  
legend('Cosine signal x(n)');//  
ylabel('a');//  
xlabel('n \rightarrow');//  
  
X=abs(fft(x,N))/N;  
subplot(3,1,2);stem(0:N-1,X,'.');//  
axis([-0.2 N -0.1 1.1]);  
legend('Magnitude spectrum |X(k)|');//  
ylabel('b');//  
xlabel('k \rightarrow');//  
  
N=1024;  
x=cos(2*pi*(2*1024/16)*(0:1:N-1)/N)';  
  
FS=40000;  
f=((0:N-1)/N)*FS;  
X =abs(fft(x,N))/N;  
subplot(3,1,3);plot(f,X);  
axis([-0.2*44100/16 max(f) -0.1 1.1]);  
legend('Magnitude spectrum |X(f)|');//  
ylabel('c');//  
xlabel('f in Hz \rightarrow')  
  
figure(2)  
subplot(3,1,1);  
plot(f,20*log10(X./(0.5)));  
axis([-0.2*44100/16 max(f) ...  
-45 20]);  
legend('Magnitude spectrum |X(f)| ...  
in dB');//  
ylabel('|X(f)| in dB \rightarrow');//  
xlabel('f in Hz \rightarrow')
```

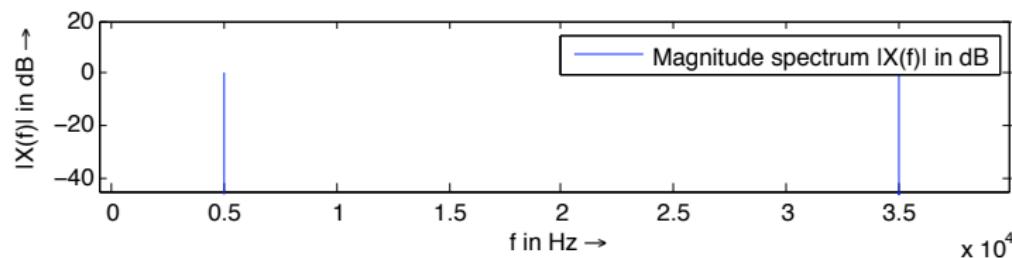
# MATLAB Fourier Frequency Spectra Example Output

fourierspectraeg.m produces the following:



# Magnitude Spectrum in dB

**Note:** It is common to plot both spectra magnitude (also frequency ranges not show here) on a dB/log scale:  
(Last Plot in [fourierspectraeg.m](#))



# Time-Frequency Representation: Spectrogram

## Spectrogram

It is often **useful** to look at the **frequency distribution** over a **short-time**:

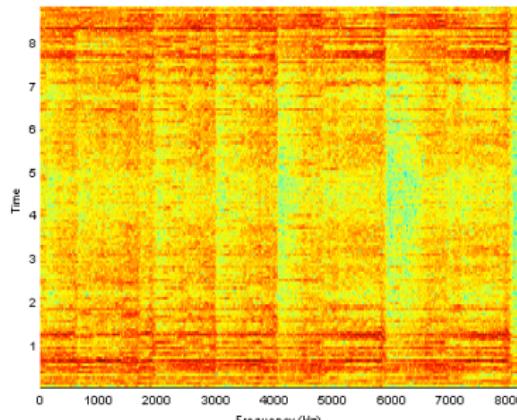
- Split signal into **N** segments
- Do a **windowed Fourier Transform** — **Short-Time Fourier Transform (STFT)**
  - Window needed to reduce **leakage** effect of doing a shorter sample SFFT.
  - Apply a **Blackman**, **Hamming** or **Hanning** Window
- MATLAB function does the job: **Spectrogram** — see **help spectrogram**
- See also MATLAB's **specgramdemo**

# MATLAB spectrogram Example

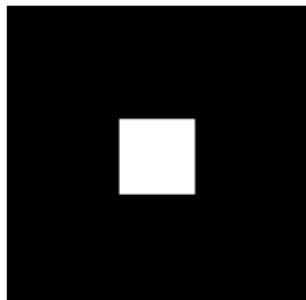
## spectrogrameg.m

```
load ('handel')
[N M] = size(y);
figure(1)
spectrogram(fft(y,N),512,20,1024,Fs);
```

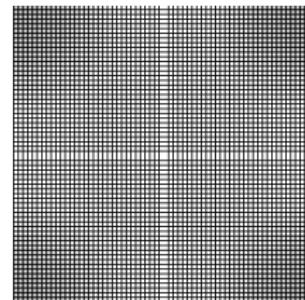
Produces the following:



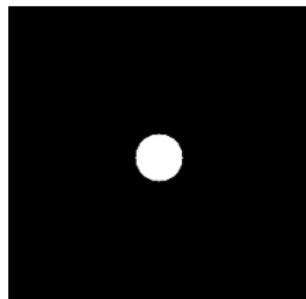
# Ideal Low Pass Filter Example 1



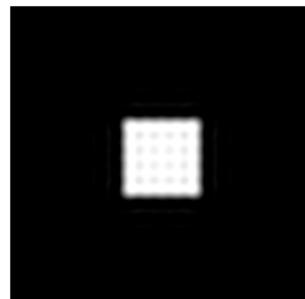
(a) Input Image



(b) Image Spectra



(c) Ideal Low Pass Filter



(d) Filtered Image

# Ideal Low-Pass Filter Example 1 MATLAB Code

## low pass.m:

```
% Compute Ideal Low Pass Filter
u0 = 20; % set cut off frequency

% Create a white box on a
% black background image
M = 256; N = 256;
image = zeros(M,N)
box = ones(64,64);
%box at centre
image(97:160,97:160) = box;

% Show Image
figure(1);
imshow(image);

% compute fft and display its spectra
F=fft2(double(image));
figure(2);
imshow(abs(fftshift(F)));

% display
figure(3);
imshow(fftshift(H));

% Apply filter and do inverse FFT
G=H.*F;
g=real(ifft2(double(G)));

% Show Result
figure(4);
imshow(g);
```

# Butterworth Low-Pass Filter Example Code

## butterworth.m:

```
% Load Image and Compute FFT as
% in Ideal Low Pass Filter Example 1
%%%%%
% Compute Butterworth Low Pass Filter
u0 = 20; % set cut off frequency

u=0:(M-1);
v=0:(N-1);
idx=find(u>M/2);
u(idx)=u(idx)-M;
idy=find(v>N/2);
v(idy)=v(idy)-N;
[V,U]=meshgrid(v,u);

for i = 1: M
    for j = 1:N
        %Apply a 2nd order Butterworth
        UVw = double((U(i,j)*U(i,j) + V(i,j)*V(i,j))/(u0*u0));
        H(i,j) = 1/(1 + UVw*UVw);
    end
end
% Display Filter and Filtered Image as before
```

# MATLAB Convolution Reverb (1)

Let's develop a fast convolution routine: [fconv.m](#)

```
function [y]=fconv(x, h)
% FCONV Fast Convolution
% [y] = FCONV(x, h) convolves x and h,
%       and normalizes the output to +-1.
%       x = input vector
%       h = input vector
%
Ly=length(x)+length(h)-1; %
Ly2=pow2(nextpow2(Ly));    % Find smallest power of 2 that is > Ly
X=fft(x, Ly2);            % Fast Fourier transform
H=fft(h, Ly2);            % Fast Fourier transform
Y=X.*H;                   % DO CONVOLUTION
y=real(ifft(Y, Ly2));     % Inverse fast Fourier transform
y=y(1:1:Ly);              % Take just the first N elements
y=y/max(abs(y));          % Normalize the output
```

**See also:** MATLAB built in function [conv\(\)](#)

# MATLAB Convolution Reverb (2)

## reverb\_convolution\_eg.m

```
% reverb_convolution_eg.m
% Script to call implement Convolution Reverb

% read the sample waveform
filename='..//acoustic.wav';
[x,Fs,bits] = wavread(filename);

% read the impulse response waveform
filename='impulse_room.wav';
[imp,Fsimp,bitsimp] = wavread(filename);

% Do convolution with FFT
y = fconv(x,imp);

% write output
wavwrite(y,Fs,bits,'out_IReverb.wav');
```