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1 What is space?

- When we think of space we think about space where we live but in mathematics it more then that In mathamatics any generic abstract collection of elements are called space

2 Function Space

- Space of all possible function $F(x)$
- Idea is come from linear algebra
- It is like vector space with infinit dimenstions

2.1 Hilbert space

Space all possible wave functions

3 Vectors

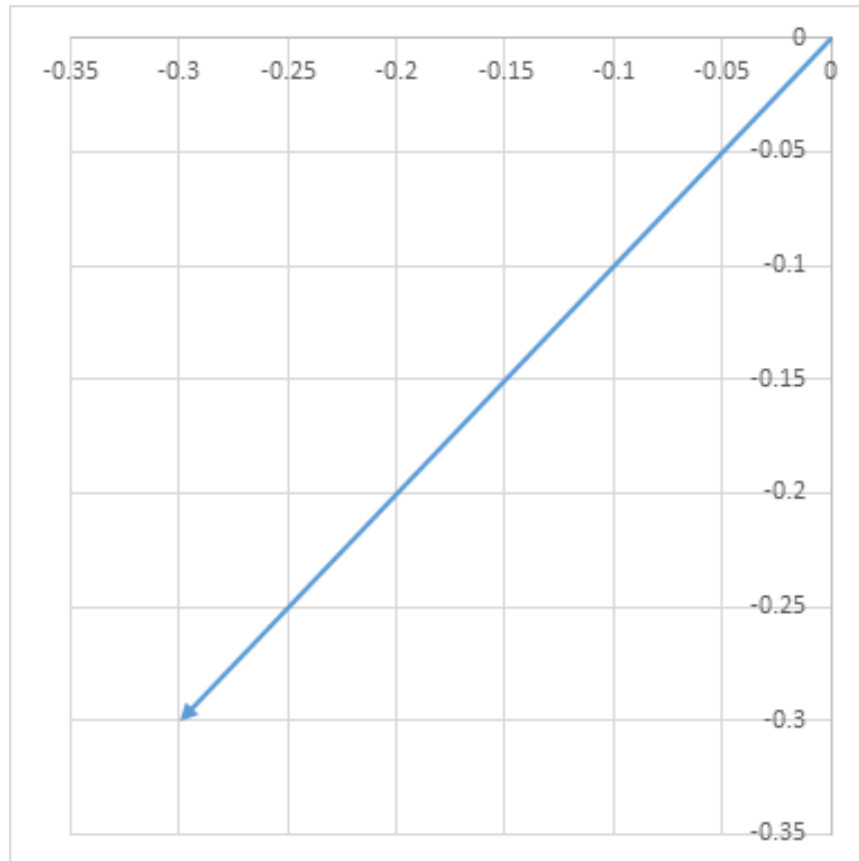
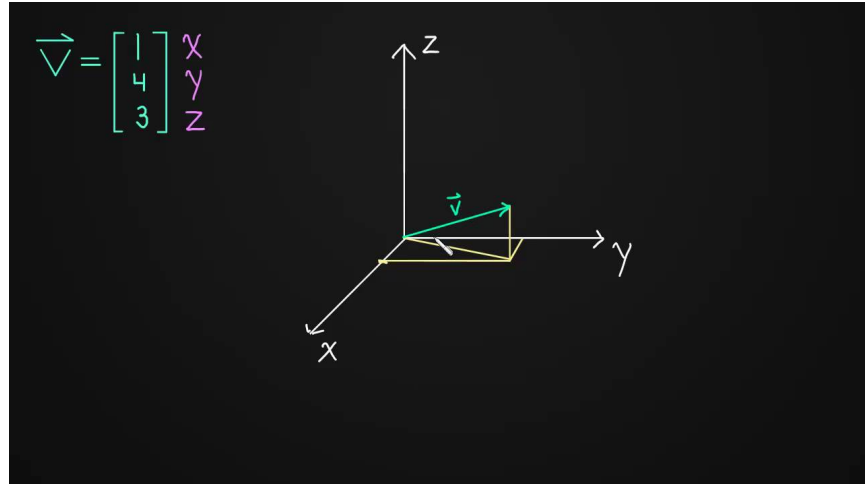


Figure 1: This is Vector

- It has both direction and magnitude

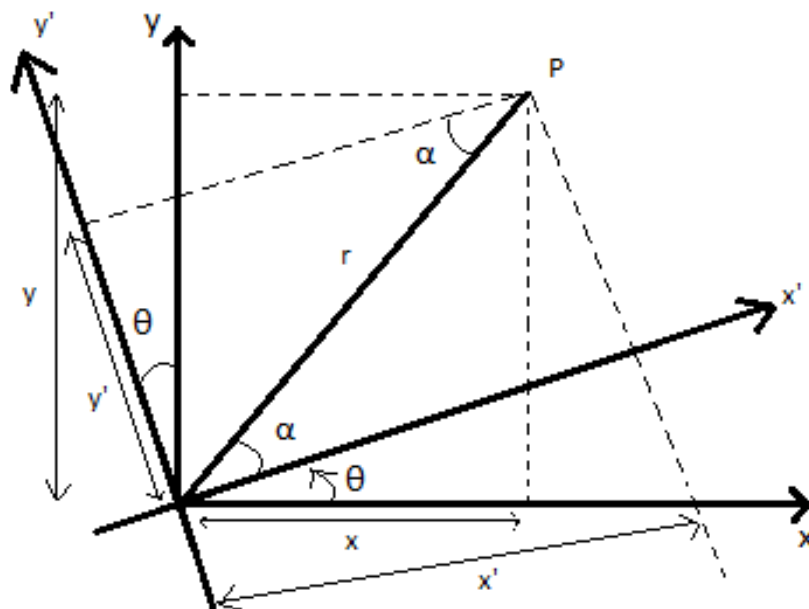
3.1 Vectors



3.2 Component representation

- $\vec{a} = x\hat{i} + y\hat{j} + z\hat{k}$
- Also i can write this way
- $\vec{a} = \frac{\vec{a} \cdot \hat{i}}{||\hat{i}||^2} + \frac{\vec{a} \cdot \hat{j}}{||\hat{j}||^2} + \frac{\vec{a} \cdot \hat{k}}{||\hat{k}||^2}$

4 coordinate transformation



4.1 coordinate transformation

- I can write p in terms of x and y
- $\vec{p} = \frac{\vec{p} \cdot \vec{x}}{||\vec{x}'||^2} + \frac{\vec{p} \cdot \vec{y}}{||\vec{y}'||^2}$
- I can also write in terms of x' and y'
- $\vec{p} = \frac{\vec{p} \cdot \vec{x}'}{||\vec{x}'||^2} + \frac{\vec{p} \cdot \vec{y}'}{||\vec{y}'||^2}$
- here dot product gives projection on to each axis and tells how much that vector is in point in x direction
- Also when dot product is zero the vectors are orthogonal to each other

5 Inner product of functions

- It is just like dot product between two vectors
- A function can be thought of as a vector of infinite dimension

- For defining innerproduct just discretize our functions f and g in some interval a and b
- We have $[f_1 \ f_2 \ \dots \ f_n]$ and $[g_1 \ g_2 \ \dots \ g_n]$ Now we can find innerproduct of this it just two vectors

5.1 Inner product continue

- Now the innerproduct is

•

$$\sum_{k=0}^{n-1} f_k g_k$$

- This as one problem when n increses this changes by huge amound so we need to normalize this by Δx

5.2 Inner product continue

- Now after normalization by Δx the equation become

•

$$\sum_{k=0}^{n-1} f_k g_k \Delta x$$

- This is Riemann approximation of integral

5.3 Innerproduct

- Now the equation become
- $\langle f(x), g(x) \rangle = \int_a^b f(x)g(x) \, dx$

5.4 Inner product matlab

```
clear all, close all, clc
```

```
f = [0 0 .1 .2 .25 .2 .25 .3 .35 .43 .45 .5 .55 .5 .4 .425 .45 .425 .4 .35 .3 .25 .
```

```
g = [0 0 .025 .1 .2 .175 .2 .25 .25 .3 .32 .35 .375 .325 .3 .275 .275 .25 .225 .225 .
```

```
x = 0.1*(1:length(f));
```

```

xf = (.01:.01:x(end));
ff = interp1(x,f,xf,'cubic')

gf = interp1(x,g,xf,'cubic')

plot(xf(20:end-10),ff(20:end-10),'k','LineWidth',1.5)
hold on
plot(x(2:end-1),f(2:end-1),'bo','MarkerFace','b')
plot(xf(20:end-10),gf(20:end-10),'k','LineWidth',1.5)
plot(x(2:end-1),g(2:end-1),'ro','MarkerFace','r')

xlim([.1 2.7])
ylim([- .1 .6])
set(gca,'XTick',[.2:.1:2.6],'XTickLabels',{},'LineWidth',1.2)
set(gca,'YTick',[]);
box off

set(gcf,'Position',[100 100 550 250])

set(gcf,'PaperPositionMode','auto')
print('-depsc2', '-loose', '../figures/InnerProduct');

% %%
% xc = x;
% fc = f;
% n = length(x);
% hold on
% fapx = 0*ff;
% dx = xc(2)-xc(1);
% L = xc(end)-xc(1);
% L = 2.5
% A0 = (1/pi)*sum(fc.*ones(size(xc)))*dx*L;
% fapx = fapx + A0/2;
% for k=1:10
%     Ak = (1/pi)*sum(fc.*cos(2*pi*k*xc/L))*dx*L;
%     Bk = (1/pi)*sum(fc.*sin(2*pi*k*xc/L))*dx*L;
%
%     fapx = fapx + Ak*cos(2*k*pi*xf/L) + Bk*sin(2*k*pi*xf/L);

```

```
% end
%      plot(xf,fapx,'k')
```

6 Orthogonal Functions

- In vectors to check orthogonality we do dot product if dot product is zero then the vectors is orthogonal to each other
- $\vec{a} \cdot \vec{b} = |\vec{a}| |\vec{b}| \cos(\theta) = 0$
- mean $\theta = 90^\circ$
- In functions we can do the same thing

6.1 Orthogonal Functions continue

- In function space if f and g are orthogonal to each other then inner-product is zero
- $\int_a^b f(x)g(x) \, dx = 0$

6.2 Why Important

- In vectorspace we represents vectors in terms of orthogonal basis
- Same can do in Function Space Represent any function interms of orthogonal functions
- One example of this is Fourier Transform
- It reprasent f(x) interms of orthogonal sins and cosins

7 Fourier Series

- It is a cordinate transformation
- It is made for solving heat equation in 1800s
- It decompose the signal f into sins and cosins
- sins and cosins are form a orthogonal basis for function space

7.1 Fourier Series

- Any periodic signals can be represented in terms of sum of sines and cosines
-

$$f(x) = \frac{A_0}{2} + \sum_{k=1}^{\infty} (A_k \cos(kx) + B_k \sin(kx))$$

7.2 FS continue

- It can be thought as this
- $f(x) = \sum_{k=0}^{\infty} (\langle f(x), \cos(kx) \rangle \frac{\cos(kx)}{\|\cos(kx)\|^2} + \langle f(x), \sin(kx) \rangle \frac{\sin(kx)}{\|\sin(kx)\|^2})$

7.3 Fs

- $A_k = \frac{1}{\|\cos(kx)\|^2} \langle f(x), \cos(kx) \rangle$
- $B_k = \frac{1}{\|\sin(kx)\|^2} \langle f(x), \sin(kx) \rangle$
- $\|f(x)\|^2 = \langle f(x), f(x) \rangle$

7.4 Complex Fourier Series

- it uses complex exponential to represent signal
- Coefficient can be found exactly same as that of Fourier series
- project function into each complex exponential basis you get the coefficient c_k

7.5 Representation

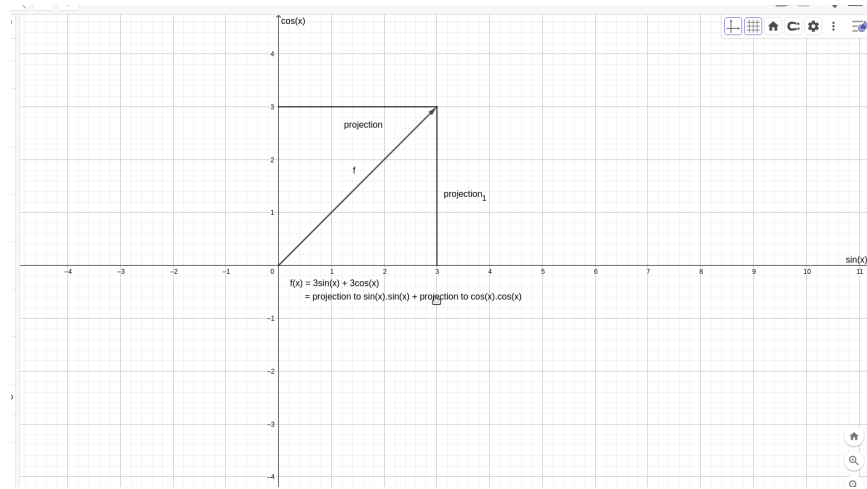
-

$$f(x) = \sum_{k=-\infty}^{\infty} C_k e^{j\omega_0 k t}$$

- $C_k = \frac{1}{2\pi} \langle f(x), e^{jk\omega_0 t} \rangle$

7.6 Example

- Assume we have a signal $f(x) = 3\sin(x) + 3\cos(x)$ then it will look like this



7.7 Matlab

```
clear all, close all, clc
```

```
kmax = 7;
```

```
dx = 0.001;
```

```
L = pi;
```

```
x = (-1+dx:dx:1)*L;
```

```
f = 0*x;
```

```
n = length(f);
```

```
nquart = floor(n/4);
```

```
nhalf = floor(n/2);
```

```
f(nquart:nhalf) = 4*(1:nquart+1)/n;
```

```
f(nhalf+1:3*nquart) = 1-4*(0:nquart-1)/n;
```

```
subplot(3,1,1)
```

```
plot(x,f,'-','Color',[0 0 0],'LineWidth',1.5)
```

```
ylim([-0.2 1.5])
```

```
xlim([-1.25*L 1.25*L])
```

```
set(gca,'LineWidth',1.2)
```

```

set(gca,'XTick',[-L 0 L],'XTickLabels',{});%{'-L','0','L','2L'})
set(gca,'YTick',[0 1],'YTickLabels',{});
box off

CC = colormap(jet(8));
% CCsparse = CC(5:5:end,:);
% CCsparse(end+1,:) = CCsparse(1,:);
CCsparse = CC(1:3:end,:);
%
subplot(3,1,2)
L = pi;
A0 = sum(f.*ones(size(x)))*dx;
plot(x,A0+0*f,'-','Color',CC(1,:)*.8,'LineWidth',1.2);
hold on
fFS = A0/2;
for k=1:kmax
    A(k) = sum(f.*cos(pi*k*x/L))*dx;
    B(k) = sum(f.*sin(pi*k*x/L))*dx;
    plot(x,A(k)*cos(k*pi*x/L),'-','Color',CC(k,:)*.8,'LineWidth',1.2);
%    plot(x,B(k)*sin(2*k*pi*x/L),'k-','LineWidth',1.2);
    fFS = fFS + A(k)*cos(k*pi*x/L) + 0*B(k)*sin(k*pi*x/L);
end
ylim([-0.7 0.7])
xlim([-1.25*L 1.25*L])
set(gca,'LineWidth',1.2)
set(gca,'XTick',[-L 0 L],'XTickLabels',{});%{'-L','0','L','2L'})
set(gca,'YTick',[-0.5 0 0.5],'YTickLabels',{});
box off
%
subplot(3,1,1)
hold on
plot(x,fFS,'-','Color',CC(7,:)*.8,'LineWidth',1.2)
l1=legend('','')
set(l1,'box','off');
l1.FontSize = 16;

subplot(3,1,3)
A0 = sum(f.*ones(size(x)))*dx;
plot(x,A0+0*f,'-','Color',CC(1,:), 'LineWidth',1.2);

```

```

hold on
fFS = A0/2;
for k=1:7
    Ak = sum(f.*cos(pi*k*x/L))*dx;
    Bk = sum(f.*sin(pi*k*x/L))*dx;
    plot(x,Ak*cos(k*pi*x/L),'-', 'Color',CC(k,:)*.8,'LineWidth',1.2);
%    plot(x,Bk*sin(2*k*pi*x/L),'k-', 'LineWidth',1.2);
    fFS = fFS + Ak*cos(k*pi*x/L) + 0*Bk*sin(k*pi*x/L);
end
ylim([-0.06 0.06])
xlim([-1.25*L 1.25*L])
set(gca,'LineWidth',1.2)
set(gca,'XTick',[-L 0 L],'XTickLabels',{'-L','0','L','2L'})
set(gca,'YTick',[-0.05 0 0.05],'YTickLabels',{'-0.05','0','0.05'});
box off

set(gcf,'Position',[100 100 550 400])
set(gcf,'PaperPositionMode','auto')
print('-depsc2', '-loose', '../figures/FourierTransformSines');

%% Plot amplitudes
clear ERR
clear A
fFS = A0/2;
A(1) = A0/2;
ERR(1) = norm(f-fFS);
kmax = 100;
for k=1:kmax
    A(k+1) = sum(f.*cos(2*pi*k*x/L))*dx*2/L;
    B(k+1) = sum(f.*sin(2*pi*k*x/L))*dx*2/L;
%    plot(x,B(k)*sin(2*k*pi*x/L),'k-', 'LineWidth',1.2);
    fFS = fFS + A(k+1)*cos(2*k*pi*x/L) + 0*B(k+1)*sin(2*k*pi*x/L);
    ERR(k+1) = norm(f-fFS)/norm(f);
end
thresh = median(ERR)*sqrt(kmax)*4/sqrt(3);
r = max(find(ERR>thresh));
r = 7;
subplot(2,1,1)
semilogy(0:1:kmax,A,'k','LineWidth',1.5)
hold on

```

```

semilogy(r,A(r+1),'bo','LineWidth',1.5)
xlim([0 kmax])
ylim([10^(-7) 1])
subplot(2,1,2)
semilogy(0:1:kmax,ERR,'k','LineWidth',1.5)
hold on
semilogy(r,ERR(r+1),'bo','LineWidth',1.5)
xlim([0 kmax])
ylim([3*10^(-4) 20])
set(gcf,'Position',[100 100 500 300])
set(gcf,'PaperPositionMode','auto')
% print('-depsc2', '-loose', '../figures/FourierTransformSinesERROR');

```

8 Fourier Transform

- Fourier series is for periodic signals
- If signal is not periodic then we can't use fourier series
- Fourier transform is limiting case of fourier series when $L \rightarrow \infty$

8.1 FT

•

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

•

$$F(\omega) = \int_{-\infty}^{\infty} f(x) e^{-j\omega x} dx$$

8.2 Work in progress

9 Discrete Fourier Transform

- In real life the data could be in measurements in some time
- We get time series instead of nice continuous function
- So the discrete fourier transform is discretized version of fourier transform

9.1 DFT

- In dft the integration become summation
- DFT
- $F(k) = \sum_{n=0}^{N-1} f_n e^{-2\pi j n \frac{k}{N}}$
- $k \in 0$ to $N-1$

9.2 Inverse DFT

- To come back to time series
- $f(n) = \sum_{k=0}^{N-1} F_k e^{2\pi j k \frac{n}{N}}$
- $n \in 0$ to $N-1$

9.3 DFT

- let $\omega_n = e^{j\frac{2\pi}{N}}$
- Then we can represent DFT in matrix form

9.4 Matrices form

$$\begin{pmatrix} F_0 \\ F_1 \\ \vdots \\ F_{N-1} \end{pmatrix} = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & \omega & \dots & \omega^{N-1} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{N-1} & \dots & \omega^{(N-1)^2} \end{bmatrix} \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_{N-1} \end{pmatrix}$$

9.5 Beauty of matrices

- DFT matrix

•

$$\begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & \omega & \dots & \omega^{N-1} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{N-1} & \dots & \omega^{(N-1)^2} \end{bmatrix}$$

9.6 Matlab code for DFT matrix

```
clear all, close all, clc
n = 256;
w = exp(-i*2*pi/n);

% Slow
for i=1:n
    for j=1:n
        DFT(i,j) = w^((i-1)*(j-1));
    end
end

% Fast
[I,J] = meshgrid(1:n,1:n);
DFT = w.^((I-1).*(J-1));
imagesc(real(DFT))
```

9.7 Matlab Gibbs phenomena

```
clear all, close all, clc

kmax = 7;

dx = 0.001;
L = pi;
x = (-1+dx:dx:1)*L;
f = 0*x;
n = length(f);
nquart = floor(n/4);
nhalf = floor(n/2);

f(nquart:nhalf) = 4*(1:nquart+1)/n;
f(nhalf+1:3*nquart) = 1-4*(0:nquart-1)/n;
subplot(3,1,1)
plot(x,f,'-', 'Color',[0 0 0], 'LineWidth',1.5)
ylim([-0.2 1.5])
xlim([-1.25*L 1.25*L])
set(gca, 'LineWidth',1.2)
```

```

set(gca,'XTick',[-L 0 L],'XTickLabels',{});%{'-L','0','L','2L'})
set(gca,'YTick',[0 1],'YTickLabels',{});
box off

CC = colormap(jet(8));
% CCsparse = CC(5:5:end,:);
% CCsparse(end+1,:) = CCsparse(1,:);
CCsparse = CC(1:3:end,:);
%
subplot(3,1,2)
L = pi;
A0 = sum(f.*ones(size(x)))*dx;
plot(x,A0+0*f,'-','Color',CC(1,:)*.8,'LineWidth',1.2);
hold on
fFS = A0/2;
for k=1:kmax
    A(k) = sum(f.*cos(pi*k*x/L))*dx;
    B(k) = sum(f.*sin(pi*k*x/L))*dx;
    plot(x,A(k)*cos(k*pi*x/L),'-','Color',CC(k,:)*.8,'LineWidth',1.2);
%    plot(x,B(k)*sin(2*k*pi*x/L),'k-','LineWidth',1.2);
    fFS = fFS + A(k)*cos(k*pi*x/L) + 0*B(k)*sin(k*pi*x/L);
end
ylim([-0.7 0.7])
xlim([-1.25*L 1.25*L])
set(gca,'LineWidth',1.2)
set(gca,'XTick',[-L 0 L],'XTickLabels',{});%{'-L','0','L','2L'})
set(gca,'YTick',[-0.5 0 0.5],'YTickLabels',{});
box off
%
subplot(3,1,1)
hold on
plot(x,fFS,'-','Color',CC(7,:)*.8,'LineWidth',1.2)
l1=legend('','')
set(l1,'box','off');
l1.FontSize = 16;

subplot(3,1,3)
A0 = sum(f.*ones(size(x)))*dx;
plot(x,A0+0*f,'-','Color',CC(1,:), 'LineWidth',1.2);

```



```

hold on
fFS = A0/2;
for k=1:7
    Ak = sum(f.*cos(pi*k*x/L))*dx;
    Bk = sum(f.*sin(pi*k*x/L))*dx;
    plot(x,Ak*cos(k*pi*x/L),'-', 'Color',CC(k,:)*.8,'LineWidth',1.2);
%    plot(x,Bk*sin(2*k*pi*x/L),'k-', 'LineWidth',1.2);
    fFS = fFS + Ak*cos(k*pi*x/L) + 0*Bk*sin(k*pi*x/L);
end
ylim([-0.06 0.06])
xlim([-1.25*L 1.25*L])
set(gca,'LineWidth',1.2)
set(gca,'XTick',[-L 0 L],'XTickLabels',{'-L','0','L','2L'})
set(gca,'YTick',[-0.05 0 0.05],'YTickLabels',{'-0.05','0','0.05'});
box off

set(gcf,'Position',[100 100 550 400])
set(gcf,'PaperPositionMode','auto')
print('-depsc2', '-loose', '../figures/FourierTransformSines');

%% Plot amplitudes
clear ERR
clear A
fFS = A0/2;
A(1) = A0/2;
ERR(1) = norm(f-fFS);
kmax = 100;
for k=1:kmax
    A(k+1) = sum(f.*cos(2*pi*k*x/L))*dx*2/L;
    B(k+1) = sum(f.*sin(2*pi*k*x/L))*dx*2/L;
%    plot(x,B(k)*sin(2*k*pi*x/L),'k-', 'LineWidth',1.2);
    fFS = fFS + A(k+1)*cos(2*k*pi*x/L) + 0*B(k+1)*sin(2*k*pi*x/L);
    ERR(k+1) = norm(f-fFS)/norm(f);
end
thresh = median(ERR)*sqrt(kmax)*4/sqrt(3);
r = max(find(ERR>thresh));
r = 7;
subplot(2,1,1)
semilogy(0:1:kmax,A,'k','LineWidth',1.5)
hold on

```

```

semilogy(r,A(r+1),'bo','LineWidth',1.5)
xlim([0 kmax])
ylim([10^(-7) 1])
subplot(2,1,2)
semilogy(0:1:kmax,ERR,'k','LineWidth',1.5)
hold on
semilogy(r,ERR(r+1),'bo','LineWidth',1.5)
xlim([0 kmax])
ylim([3*10^(-4) 20])
set(gcf,'Position',[100 100 500 300])
set(gcf,'PaperPositionMode','auto')
% print('-depsc2', '-loose', '../figures/FourierTransformSinesERROR');

```

9.8 Work in progres

10 FFT

- FFT is an algorithm to compute DFT fast and efficiently
- It uses symmetry in DFT
- To compute DFT Without FFT it require $O(n^2)$ but FFT require only $O(n \log(n))$

10.1

11 Gabor Transform

11.1 Limitations of Fourier transform

- FT is good for representing smooth signal when there is sudden jump or discontinuity then it is not capture very well Gibbs phenomena
- FT is good for stationary signal
- Stationary means frequency of signal not change with time
- When we compute Fourier Transform we loss all of time information so we can't say when this frequency occurred

- non stationary signals example is audio signal which frequency changes with time

11.2 Gabor transform

- it solve the problem of FT
- Gabor Transform allow us to compute spectrogram a time frequency plot
- Also called windowed FT
- We take a window function multiply with the signal and translate the signal to get gabor transform

11.3 Gabor transform

- pull out both time and frequency content
- instead of computing FT of entire signal we divide into several sections and compute FT of each section
- Mathamatically we can write

•

$$G(f(t)) = \int_{-\infty}^{\infty} f(\tau) e^{-i\omega\tau} g(\tau - t) d\tau$$

- g is the window function it can be gaussian or rectangular
- We can't know what frequency exist at what time instead but we can know what frequency band exist at what time

11.4 picture

- gabor grid
- ./gab.gif

11.5 Problems of gabor transform

- Uncertainty principle
- It tells about when when you narrow the window you get better time resalution but you get poor frequency resalution
- when you stretch the window you get better frequency inforation but poor time information
- uncertainty principle tell us
- $\Delta t \Delta f \geq \frac{1}{4\pi}$

11.6 matlab code for spectrogram

```
clear all, close all, clc

n = 128;
L = 30;
dx = L/(n);
x = -L/2:dx:L/2-dx;
f = cos(x).*exp(-x.^2/25);           % Function
df = -(sin(x).*exp(-x.^2/25) + (2/25)*x.*f); % Derivative

%% Approximate derivative using finite Difference...
for kappa=1:length(df)-1
    dfFD(kappa) = (f(kappa+1)-f(kappa))/dx;
end
dfFD(end+1) = dfFD(end);

%% Derivative using FFT (spectral derivative)
fhat = fft(f);
kappa = (2*pi/L)*[-n/2:n/2-1];
kappa = fftshift(kappa); % Re-order fft frequencies
dfhat = i*kappa.*fhat;
dfFFT = real(ifft(dfhat));

%% Plotting commands
plot(x,df,'k','LineWidth',1.5), hold on
plot(x,dfFD,'b--','LineWidth',1.2)
plot(x,dfFFT,'r--','LineWidth',1.2)
```

```
legend('True Derivative','Finite Diff.','FFT Derivative')
```

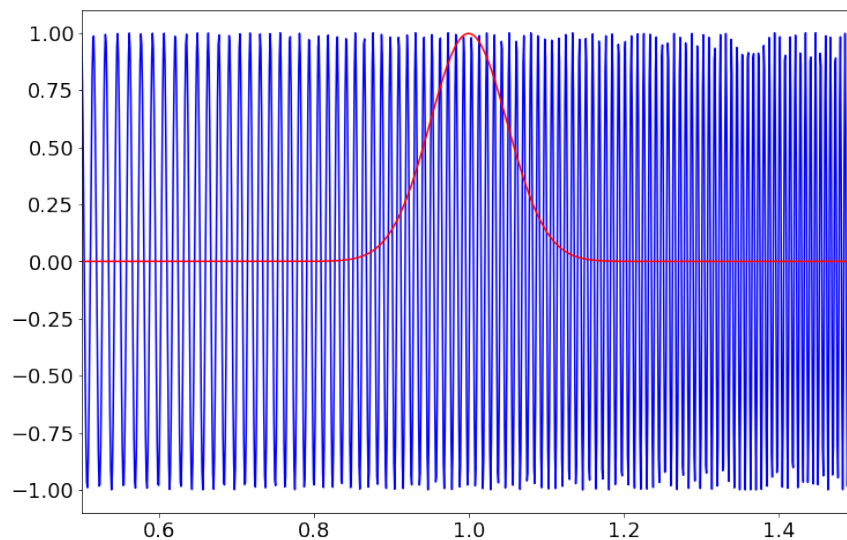
11.7 beethoven code matlab

```
clear all, close all, clc
```

```
% If you download mp3read, you can use this code
% also, need to download mp3read from
% http://www.mathworks.com/matlabcentral/fileexchange/13852-mp3read-and-mp3write
% [Y,FS,NBITS,OPTS] = mp3read(' ../../DATA/beethoven.mp3'); % add in your own song
% T = 40; % 40 seconds
% y=Y(1:T*FS); % First 40 seconds
load ../../DATA/beethoven_40sec.mat
%% Spectrogram
spectrogram(y,5000,400,24000,24000,'yaxis');

%% SPECTROGRAM
% uncomment remaining code and download stft code by M.Sc. Eng. Hristo Zhivomirov
% wlen = 5000;
% h=400; % Overlap is wlen - h
% % perform time-frequency analysis and resynthesis of the original signal
% [S, f, t_stft] = stft(y, wlen, h, FS/4, FS); % y axis range goes up to 4000 HZ
% imagesc(log10(abs(S)));
% load CC.mat
% colormap(ones(size(CC))-(CC))
%
% axis xy, hold on
% XTicks = [1 300 600 900 1200 1500 1800 2100];
% XTickLabels = {'0','5','10','15','20','25','30','35'};
% YTicks = [0 1000 2000 3000];
% YTickLabels = {'0','4000','8000','12000'};
% set(gca,'XTick',XTicks,'XTickLabels',XTickLabels);
% set(gca,'YTick',YTicks,'YTickLabels',YTickLabels);
%
% % plot a frequency
% freq = @(n)((2^(1/12))^(n-49))*440;
% freq(40) % frequency of 40th key = C
```

11.8 Idea



12 Wavelet Transform

- supercharged Fourier transform
- Generalize Fourier transform
- Represent signals in terms of other orthogonal functions
-

12.1 Wavelet

- Wavelets are new basis functions also act as window function
- Wavelets are some wave like oscillating functions in limited durations
- There are so many wavelets available

12.2 Haar Wavelet