




Misc stuff

MATLAB



lookfor command - to look for commands
based on "keyword"

(searches all m files in path, including your files, for the keyword).

what command - lists matlab related files
(returns structure with fields for m, mat, mex, mdl, classes, and packages
files).,





Help window (pull down menu).

Helpdesk (internet)

<http://www.mathworks.com/access/helpdesk/help/helpdesk.shtml>



workspace

```
>> help workspace
```

```
WORKSPACE Open Workspace browser to manage workspace
```

```
WORKSPACE Opens the Workspace browser with a view of the  
variables in the current Workspace. Displayed variables may  
be viewed, manipulated, saved, and cleared.
```


path

```
>> help path
```

PATH Get/set search path.

PATH, by itself, prettyprints MATLAB's current search path. The initial search path list is set by PATHDEF, and is perhaps individualized by STARTUP.

P = PATH returns a string containing the path in P. PATH(P) changes the path to P. PATH(PATH) refreshes MATLAB's view of the directories on the path, ensuring that any changes to non-toolbox directories are visible.

PATH(P1,P2) changes the path to the concatenation of the two path strings P1 and P2. Thus PATH(PATH,P) appends a new directory to the current path and PATH(P,PATH) prepends a new directory. If P is already on the path, then PATH(PATH,P) moves P to the end of the path, and similarly, PATH(P,PATH) moves P to the beginning of the path.

For example, the following statements add another directory to MATLAB's search path on various operating systems:

Unix: path(path, '/home/myfriend/goodstuff')

Windows: path(path, 'c:\tools\goodstuff')

format command

```
>> help format
```

```
. . .
```

```
FORMAT Set output format.
```

```
. . .
```

```
FORMAT does not affect how MATLAB computations are done.
```

To separate multiple commands on one line
use ";" for no output, and ',' for output

Cmd line editing

arrows: move cursor by character

ctrl arrows l and r: move cursor by word

ctrl a, e: move cursor to beginning, end line

ctrl u, d, h, k: clear line, delete char at
cursor, delete char before cursor, delete to
end of line.

Running an m file from the command line
(should not be interactive)

```
>> matlab < somefile.m
```

Don't show output on terminal (send to bit
bucket), run in background to not lock up
terminal

```
matlab -nosplash < eig_mov.m > /nl &
```


Global variables

When you define a variable at the matlab prompt, it is defined inside of matlab's "workspace."

Running a script does not affect this, since a script is just a collection of commands, and they're actually run from the same workspace.

If you define a variable in a script, it will stay defined in the workspace.

Global variables

Functions, on the other hand, do not share the same workspace.

A function won't know what a variable is unless

- the function gets the variable as an argument, or
- the variable is defined as a variable that is shared by the function and the matlab workspace, i.e. a global variable.

Global variables

To use a global variable, every place (function, script, or at the matlab prompt) that needs to share that variable must have a line near the top identifying it as a global variable, ie:

```
global phi;
```

Then when the variable is assigned a value in one of those places, it will have a value in all the other places that have the global statement.



Linear Algebra (a la Matlab) Review



MATLAB

```
>> a=[1 2 3;4 5 6;7 8 9]
```

```
a =
```

```
1 2 3
4 5 6
7 8 9
```

```
>> c=trace(a)
```

```
c =
15
```

```
>> a=[.96 -.28; .28 .96]
```

```
a =
```

```
0.9600 -0.2800
0.2800 0.9600
```

```
>> inv(a)
```

```
ans =
```

```
0.9600 0.2800
-0.2800 0.9600
```

```
>> a'*a
```

```
ans =
```

```
1 0
0 1
```

```
>>
```

trace of a Matrix is $\text{Tr}(A) =$

$$\sum_{i=1}^N a_{ii}$$

(the sum of the diagonal entries)

In matlab use `trace(A)`

a matrix **A** is **orthonormal** if

$$A^T = A^{-1}$$

and in this case

$$AA^T = I$$

Block matrices

```
>> b=[2 2;1 3]
```

```
b =
```

```
    2    2  
    1    3
```

```
c =
```

```
    0    2    3  
    5    4    7
```

```
>> d=[1 0]
```

```
d =
```

```
    1    0
```

```
>> e=[-1 6 0]
```

```
e =
```

```
   -1    6    0
```

```
>> a=[b c;d e]
```

```
a =
```

```
    2    2    0    2    3  
    1    3    5    4    7  
    1    0   -1    6    0
```


Linear Dependence

- A set of vectors is **linearly dependent** if one of the vectors can be expressed as a linear combination of the other vectors.

Example:

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$$

$$a \quad b \quad c$$

$$2a + 1b = c$$

Linear Independence

- A set of vectors is **linearly independent** if none of the vectors can be expressed as a linear combination of the other vectors.

Example:

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}$$

a

b

c

There is no simple, linear combination of a and b what will produce c.

Rank of a matrix

- The **rank** of a matrix is the number of linearly independent columns of the matrix.

Examples:

$$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{has rank 2}$$

$$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{has rank 3}$$

- Note: the rank of a matrix is also the number of linearly independent *rows* of the matrix.


```
>> a=[1 2; 3 4]
```

```
a =
```

```
    1    2  
    3    4
```

```
>> rank(a)
```

```
ans =
```

```
    2
```

```
>> rref(a)
```

```
ans =
```

```
    1    0  
    0    1
```

```
>> >> a=[1 2 3;4 5 6;7 8 9]
```

```
a =
```

```
    1    2    3  
    4    5    6  
    7    8    9
```

```
>> rank(a)
```

```
ans =
```

```
    2
```

```
>> rref(a)
```

```
ans =
```

```
    1    0   -1  
    0    1    2  
    0    0    0
```

There are a lot of matrix math functions

```
>> help matfun
```

```
Matrix functions - numerical linear algebra.
```

```
Matrix analysis.
```

```
norm      - Matrix or vector norm.  
normest   - Estimate the matrix 2-norm.  
rank      - Matrix rank.  
det       - Determinant.  
trace     - Sum of diagonal elements.  
null      - Null space.  
orth      - Orthogonalization.  
rref      - Reduced row echelon form.  
subspace  - Angle between two subspaces.
```

```
Linear equations.
```

```
/ and /    - Linear equation solution; use "help slash".  
linsolve   - Linear equation solution with extra control.  
inv        - Matrix inverse.  
rcond      - LAPACK reciprocal condition estimator  
cond       - Condition number with respect to inversion.  
condest    - 1-norm condition number estimate.  
normest1   - 1-norm estimate.
```

cholinc - Incomplete Cholesky factorization.
ldl - Block LDL' factorization.
lu - LU factorization.
luinc - Incomplete LU factorization.
qr - Orthogonal-triangular decomposition.
lsqnonneg - Linear least squares with nonnegativity constraints.
pinv - Pseudoinverse.
lscov - Least squares with known covariance.

Eigenvalues and singular values.

eig - Eigenvalues and eigenvectors.
svd - Singular value decomposition.
gsvd - Generalized singular value decomposition.
eigs - A few eigenvalues.
svds - A few singular values.
poly - Characteristic polynomial.
polyeig - Polynomial eigenvalue problem.
condeig - Condition number with respect to eigenvalues.
hess - Hessenberg form.
schur - Schur decomposition.
qz - QZ factorization for generalized eigenvalues.

ordschur - Reordering of eigenvalues in Schur decomposition.

ordqz - Reordering of eigenvalues in QZ factorization.

ordeig - Eigenvalues of quasitriangular matrices.

Matrix functions.

expm - Matrix exponential.

logm - Matrix logarithm.

sqrtn - Matrix square root.

funm - Evaluate general matrix function.

Factorization utilities

qrdelete - Delete a column or row from QR factorization.

qrinsert - Insert a column or row into QR factorization.

rsf2csf - Real block diagonal form to complex diagonal form.

cdf2rdf - Complex diagonal form to real block diagonal form.

balance - Diagonal scaling to improve eigenvalue accuracy.

planerot - Givens plane rotation.

cholupdate - rank 1 update to Cholesky factorization.

qrupdate - rank 1 update to QR factorization.

>>



Data Analysis


MATLAB



Flow Charts

2 tasks

Understanding How a Process Works



Communicating How a Process Works
(translating it into computer code,
communicating to the computer.)



A flow chart can therefore be used to:

Define and analyze processes;

Build a step-by-step picture of the process
for analysis, discussion, communication, and
coding;

and



Define, standardize or find areas for
improvement in a process.



Also

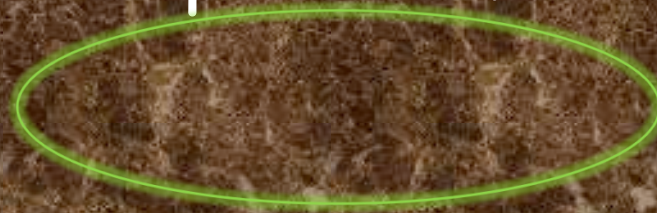
by conveying the information or processes in
a step-by-step flow, you can then
concentrate more intently on each individual
step,



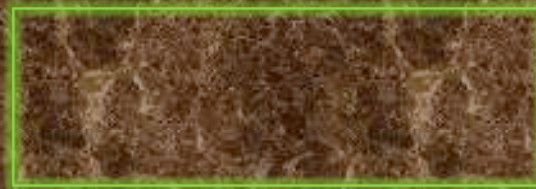
without feeling overwhelmed by the bigger
picture.

Most flow charts are made up of three main types of symbol:

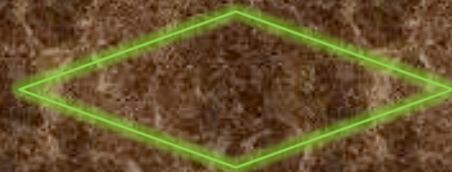
Elongated circles, signify start or end of a process;




Rectangles, show instructions or actions;




Diamonds, show decisions to be made



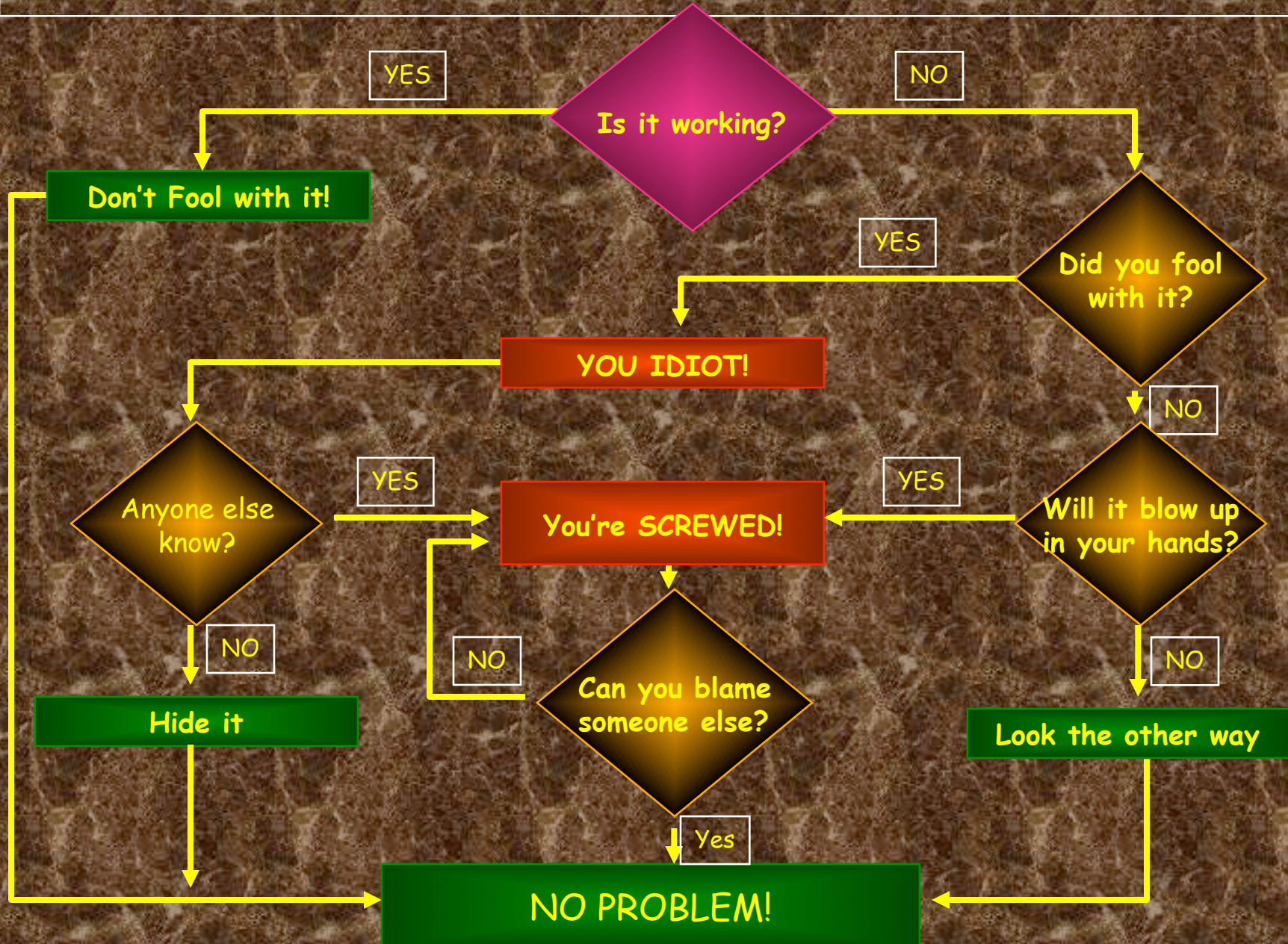


Within each symbol, write down what the symbol represents. This could be the start or finish of the process, the action to be taken, or the decision to be made.



Symbols are connected one to the other by arrows, showing the flow of the process.

Worlds most famous Flowchart: General Flowchart For Problem Resolution -



Coding and Flow Charts

Today's presentation will focus on understanding Chuck's matlab script for polarization analysis using 3 component recordings of body and/or surface waves

Chucks example codes:

<http://www.ceri.memphis.edu/people/langston/matlab/programming.html>



GOAL: solve for polarization using 3
component seismic data

Starting data: 3 component single station
SAC formatted data


Result: Identify the azimuth(s) of the
primary wave(s) recorded in the data

How to we get from A to B?



Principal Component Analysis


Principal component analysis (PCA) is a vector space transform often used to reduce multidimensional data sets to lower dimensions for analysis.





Principal Component Analysis

PCA involves the calculation of the eigenvalue decomposition of a data covariance matrix or singular value decomposition of a data matrix, usually after mean centering the data for each attribute.



Its operation can be thought of as revealing the internal structure of the data in a way which best explains the variance in the data.

What we are looking for:

New set of axis (basis) that maximizes the correlation of $HT(Z)$ with R , and minimizes the correlations between both $HT(Z)$ and R with T .

We are not using the full power of PCA, since we already have some model for the result of the analysis

(and have therefore preprocessed the data by taking the Hilbert transform of the z component).

What is the idea?


Seismic waves are polarized

P wave longitudinal (V and R)

S wave transverse with SH and SV
polarizations (T, V and R)


Rayleigh waves (V and R)


Love waves (T).



If we take a short time period we can think of each component as a vector of n terms.


If we take the dot product of each vector with itself and with the other two components we can find the "angle" between them.





We can also make these dot products by making a $3 \times n$ array using each seismogram as a row.

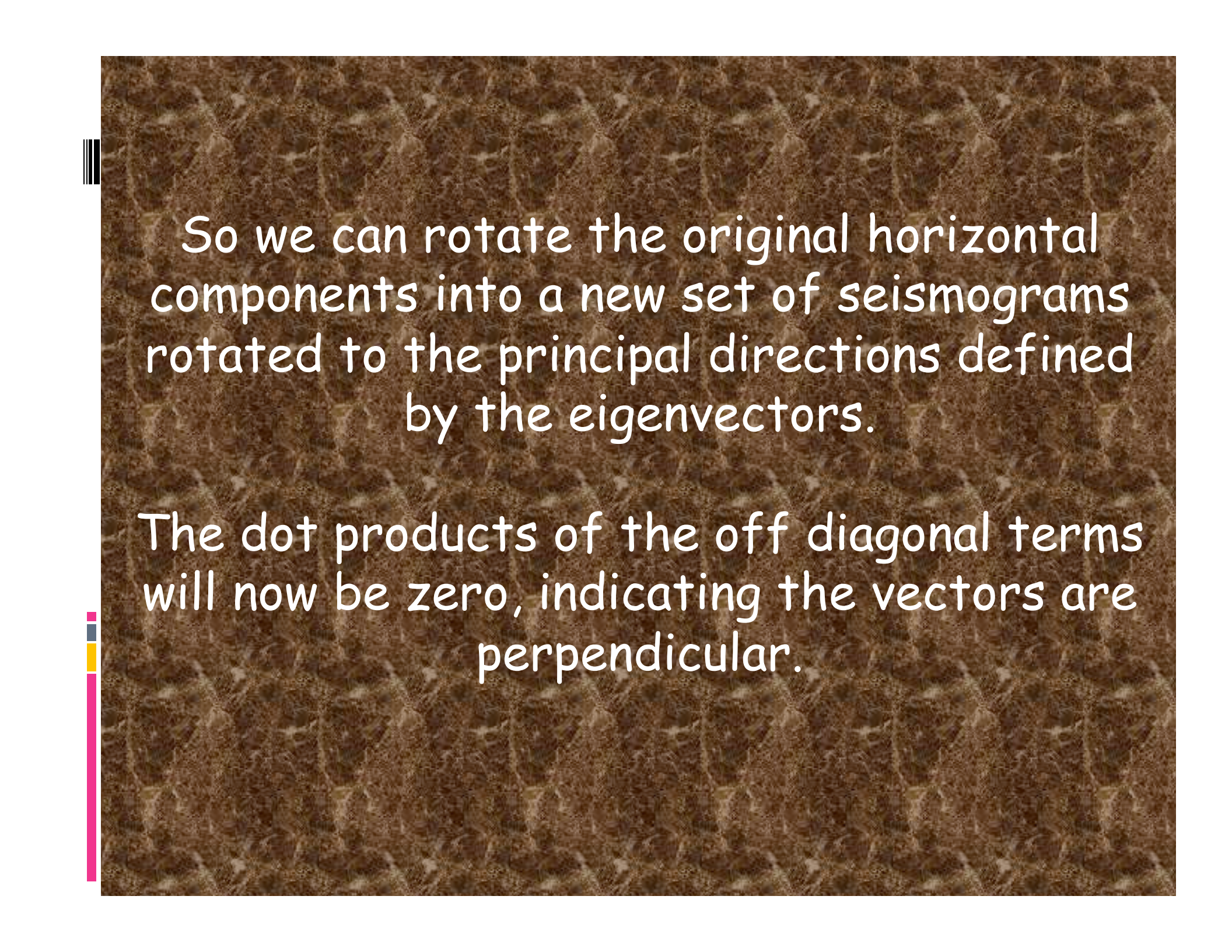
Multiplying this array with its transpose results in a 3×3 matrix with the various dot products in the elements of the matrix.



Now find the eigenvectors and eigenvalues of this matrix.

From the eigenvectors we can make a rotation matrix that will rotate our matrix to a diagonal matrix.

The off diagonal elements are now all zero and from the geometric interpretation of the dot product this means that the two vectors used to make that dot product are perpendicular.

The background is a brown, marbled texture. On the left side, there are vertical bars: a black and white striped bar near the top, and a bar with blue, yellow, and pink segments further down.

So we can rotate the original horizontal components into a new set of seismograms rotated to the principal directions defined by the eigenvectors.

The dot products of the off diagonal terms will now be zero, indicating the vectors are perpendicular.

GOAL: Solve for polarization

- load SAC data
- remove mean
- plot waveforms
- filter waveforms to highlight waves of interest
- plot filtered waveforms

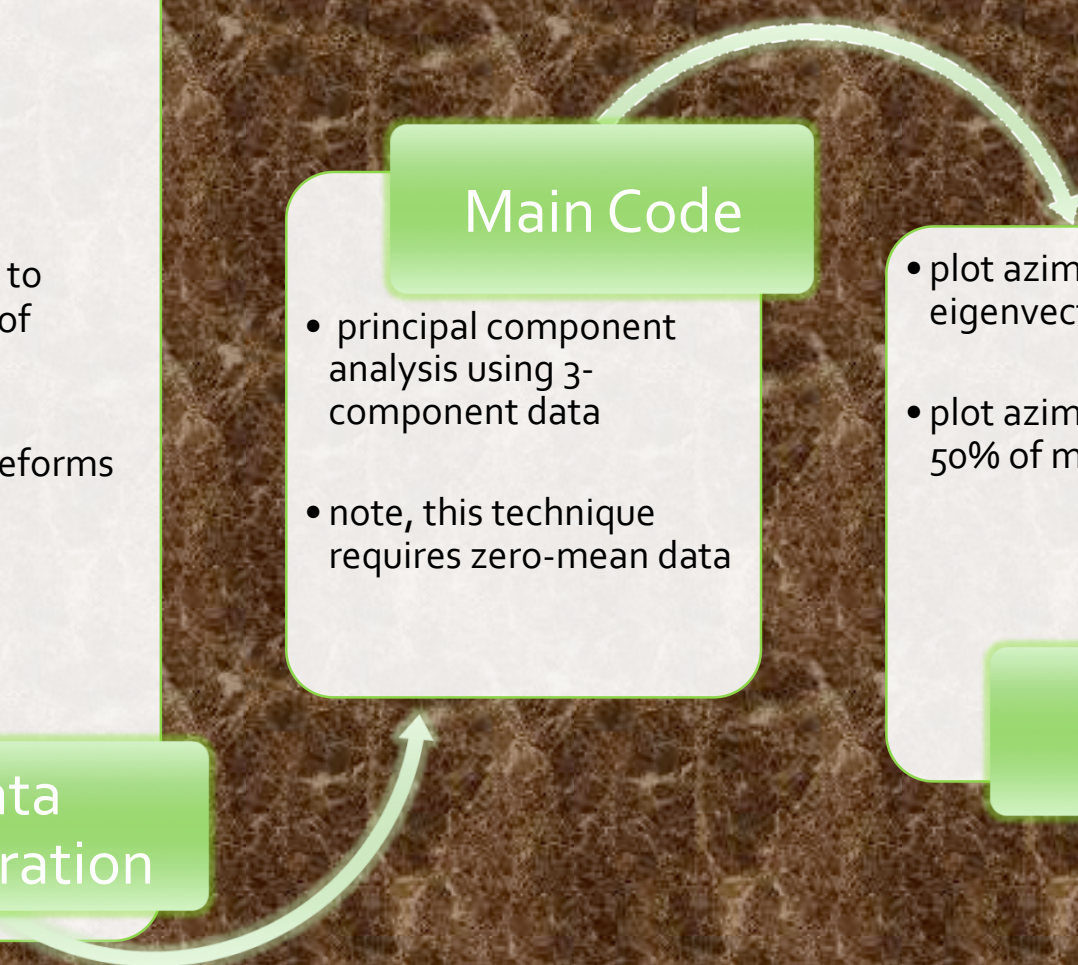
Data
Preparation

Main Code

- principal component analysis using 3-component data
- note, this technique requires zero-mean data

- plot azimuths of eigenvectors
- plot azimuths exceeding 50% of maximum value

Results



Step 1: Data Preparation

Load SAC data

3 files (Z,E,N).... suggests using a subroutine

Provide station name

Remove the mean

Plot waveforms vs time

Filter waveforms to highlight waves of interest

Design a filter; allow flexible input of corner info

Work on 3 files...suggests a subroutine

Plot filtered waveforms

Create function 'polarize'

```
function polarize(station,delt,ttot,twin,hilb,flp,fhi)
%
% function polarize(station,delt,ttot,twin,hilb,flp,fhi)
%
% Program to read in 3 component waveform data
% Create the covariance matrix for a moving time window
% Find the principal components and infer polarization
%
% input:
%     station = station name for sacfile prefix
%     delt = sampling interval
%     ttot = total number of seconds to analyze in traces
%     twin = time window length, each time shift will be 1/2 of the
%           window length
%     hilb = 0, no hilbert transform of vertical component
%           = 1, hilbert transform
%     flp = low passband corner frequency of a 2nd order butterworth
%           filter used to filter the data, if 0, then no filtering
%     fhi = hi passband corner frequency of the filter
```


Loading SAC data

Many matlab scripts exist to read in sac data.

I modified one so it is

- not sensitive to byte order
- returns the data, plus: npts, delta, and begin point of the SAC file
- data is a column vector

Read the data


```
[e,npts,delt,date,hour,minu,seco,fname]=get_sac_fn(' ../  
2007.308.20.37.16.3856.IU.SBA.00.BHE.R.SAC');  
  
[n,npts,delt,date,hour,minu,seco,fname]=get_sac_fn(' ../  
2007.308.20.37.16.3856.IU.SBA.00.BHN.R.SAC');  
  
[z,npts,delt,date,hour,minu,seco,fname]=get_sac_fn(' ../  
2007.308.20.37.16.3856.IU.SBA.00.BHZ.R.SAC');
```

Removing the data mean


We need to remove the mean of the data for principal component analysis (PCA).

We also need to transpose the column vector data into row vectors.

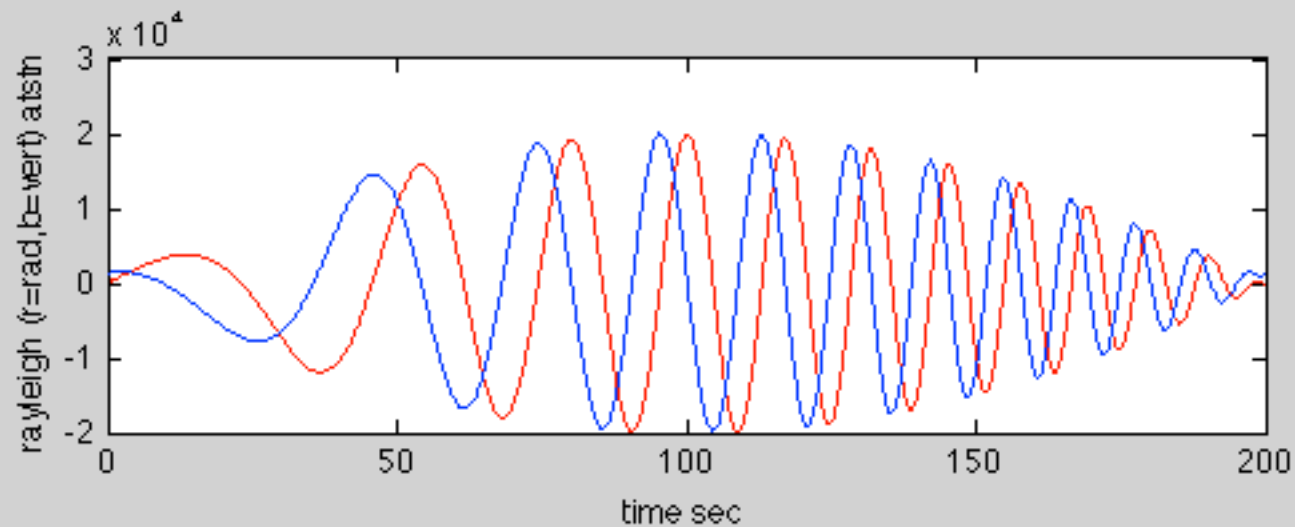
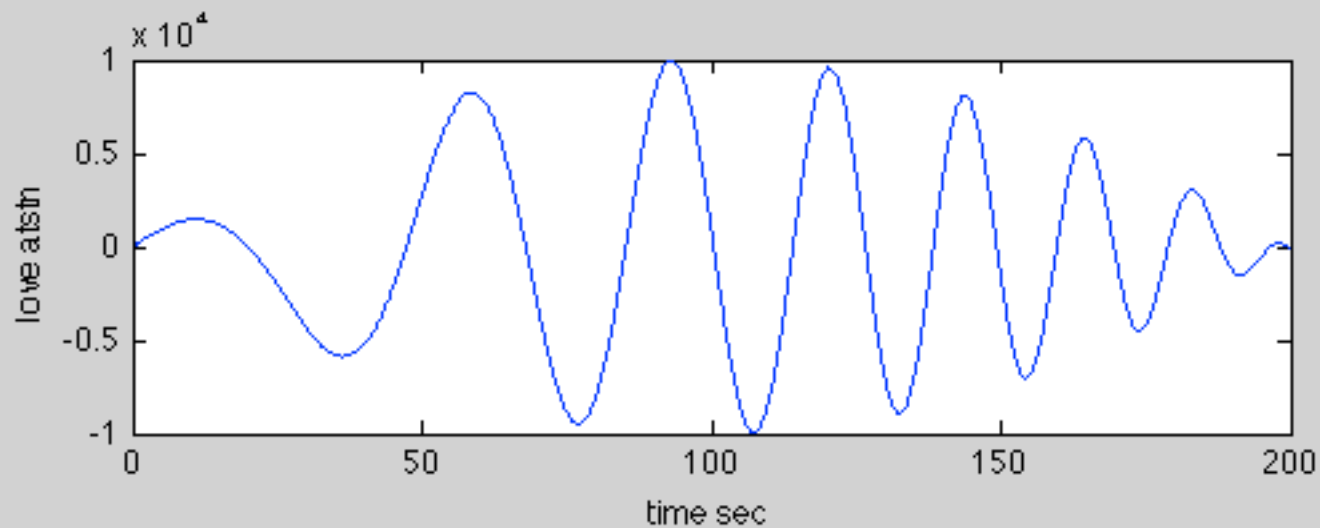
```
e=dmean(e');           % remove the mean from each
n=dmean(n');           % and transpose the data
z=dmean(z');
```

```
subroutine: dmean
function [a]=dmean(b)
%
% [a]=dmean(b)
%
%      Remove the mean from a row vector
m=mean(b);
a=b-m;
return;
```

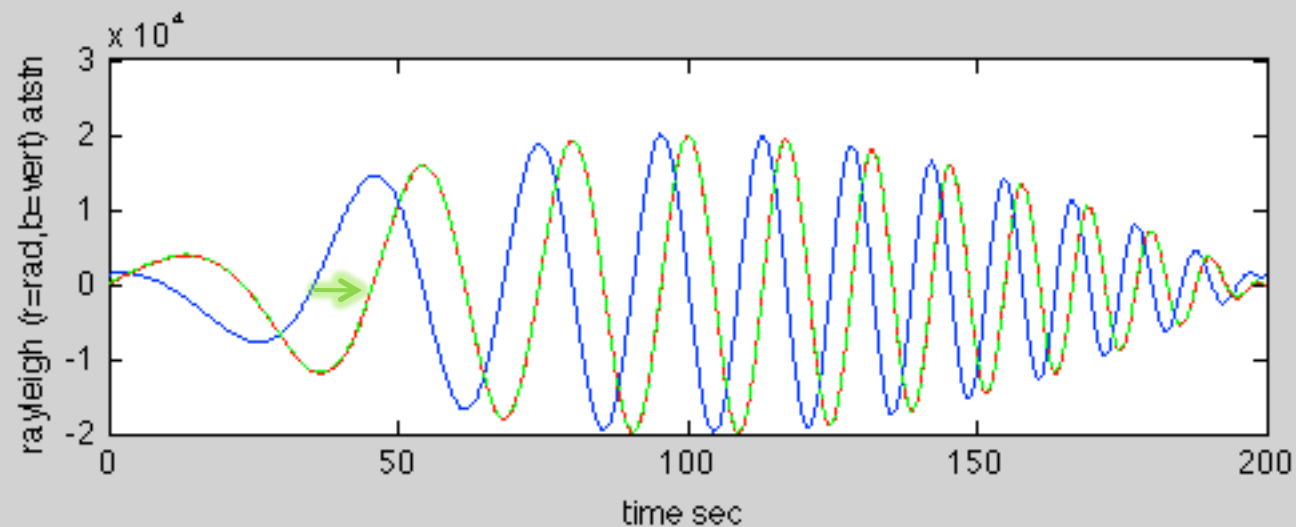
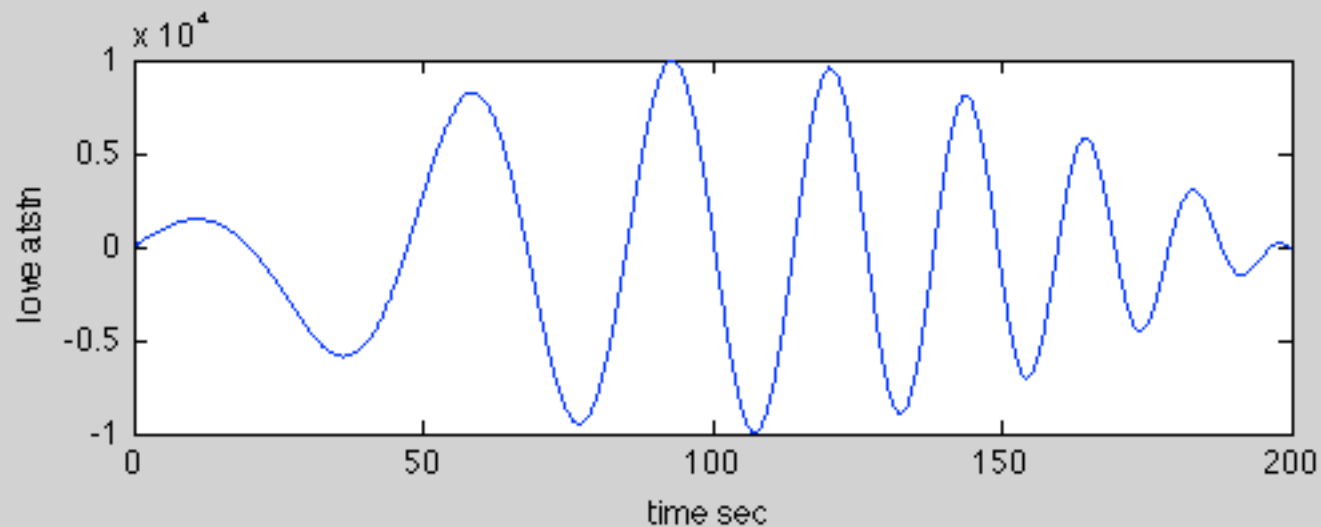


Make Love and Rayleigh waves (Z, R and T)

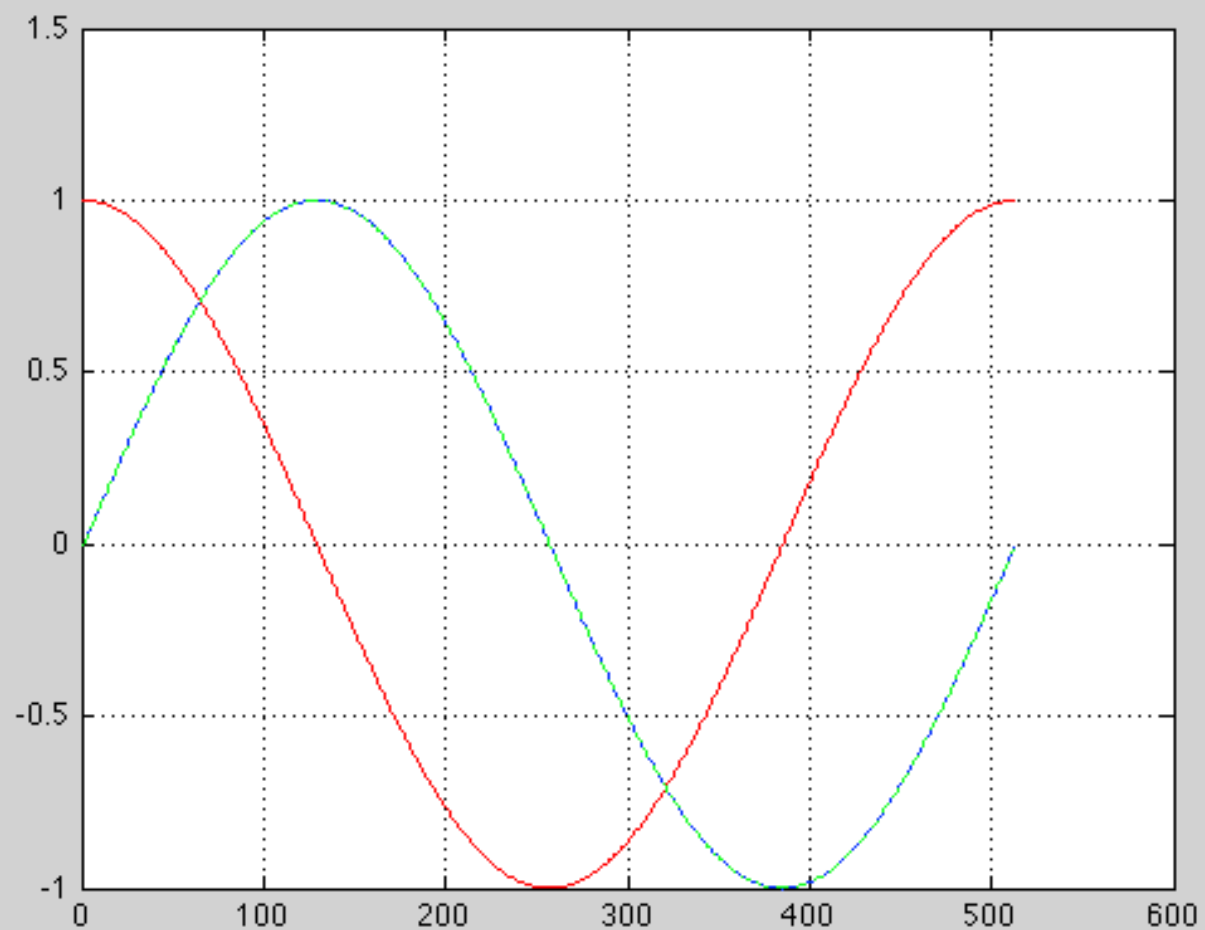


Rayleigh R and Z related by Hilbert X-form

(90° phase shift, blue trace is Hilbert Transformed to green trace, then overlays red trace.).



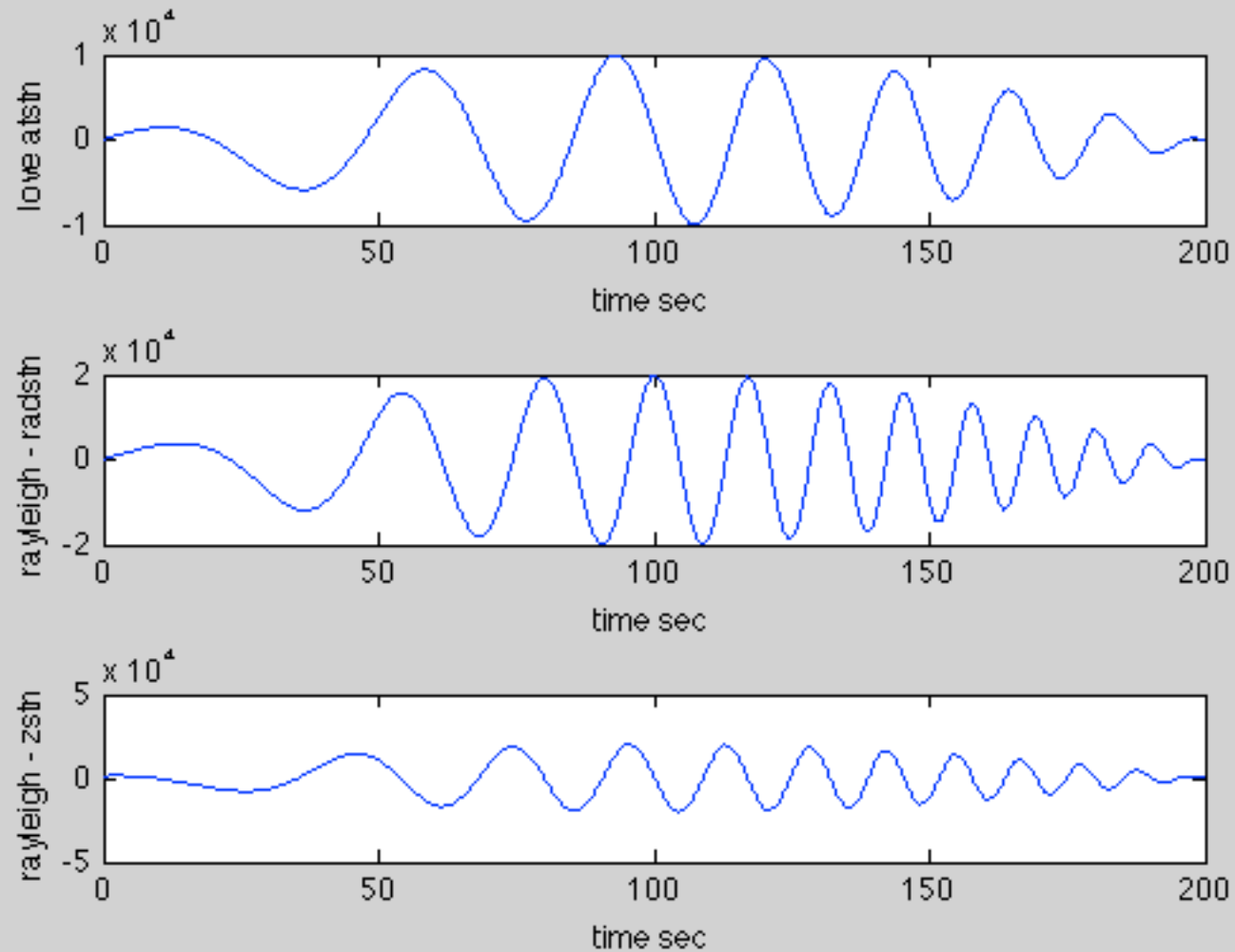

```
n=512;  
a=sin(2*pi*[0:(n-1)]/n);  
b=hilbert(a);  
clf  
plot(a)  
hold  
plot(-imag(b),'r')  
plot(real(b),'g--')  
grid
```



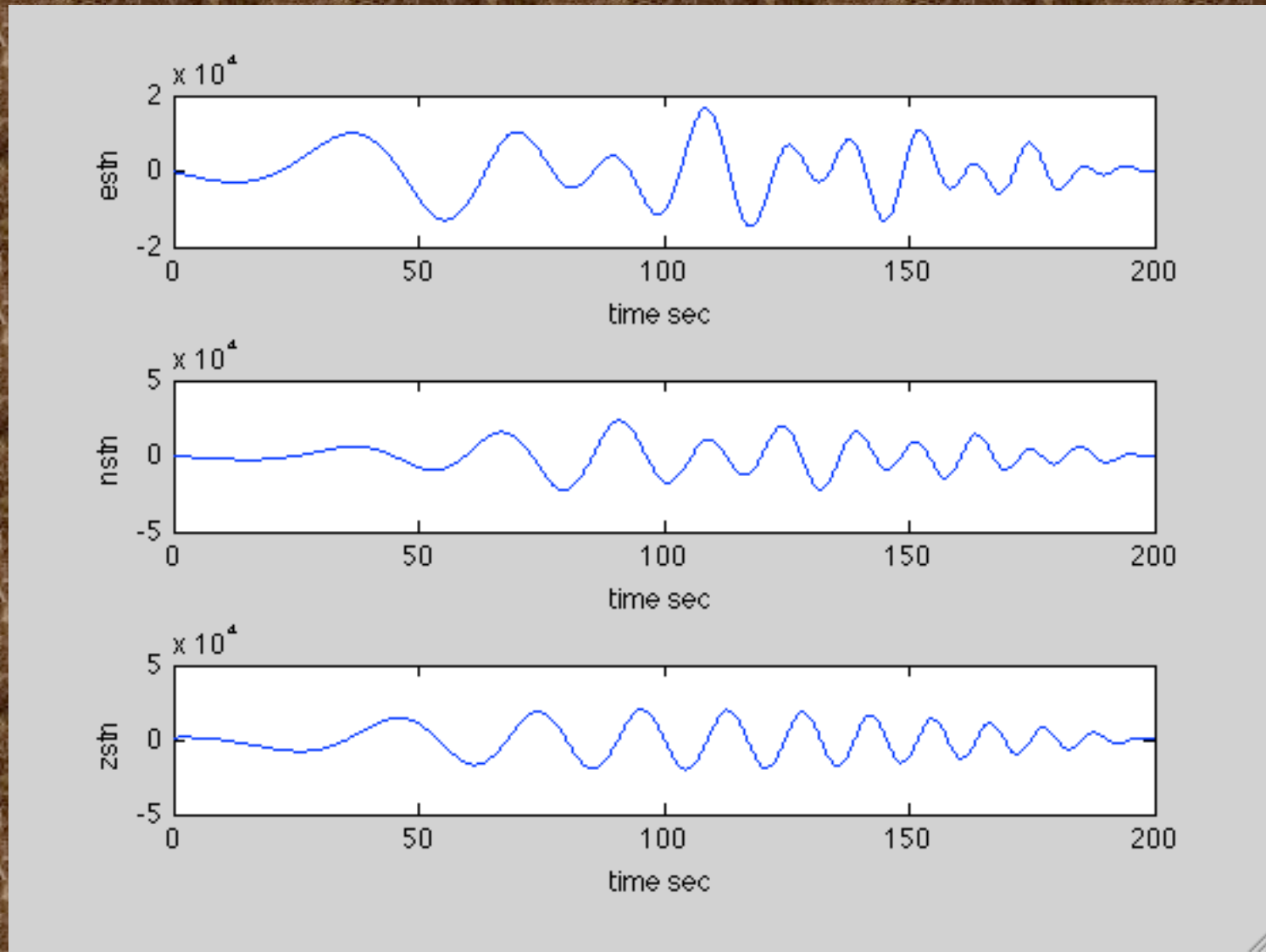
Hilbert Transform

```
if hilb ==1;           % hilbert transform the vertical component
    zh=hilbert(z);      % to make Rayleigh wave in phase on vert
                        % and horz
    z=-imag(zh);        % if present (z constructed from HT, so
                        % used +imag to make overlay for last
                        % figure)
else;
end;
```

Make Love and Rayleigh waves (Z, R and T)



Rotate horizontals into seismograms @ 30°.



Plot the data

```
% plot the raw data
```

```
f1=figure('name','DATA SEISMOGRAMS');
```

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

```
plot(t,e);
```

```
xlabel('time sec');
```

```
ylabel(strcat('EW Comp at ',station));
```

```
subplot(3,1,2);
```

```
plot(t,n);
```

```
xlabel('time sec');
```

```
ylabel(strcat('NS Comp at ',station));
```

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

```
plot(t,z);
```

```
xlabel('time sec');
```

```
ylabel(strcat('Z comp at ',station));
```

Filtering

Filtering is a two step process in Matlab

Design the filter
Apply the filter

There is a filter design GUI you can use to design the perfect filter called fdatool

Or you can design filters using pre-built filter types (Butterworth, Bessel, etc.)


```

function [d]=bandpass(c,flp,fhi,npts,delt)
%
% [d]=bandpass(c,flp)
%
% bandpass a time series with a 2nd order butterworth filter
%
% c = input time series
% flp = lowpass corner frequency of filter
% fhi = highpass corner frequency
% npts = samples in data
% delt = sampling interval of data
%
n=2;           % 2nd order butterworth filter
fnq=1/(2*delt); % Nyquist frequency
Wn=[flp/fnq fhi/fnq]; % non-dimensionalize the corner
frequencies
[b,a]=butter(n,Wn); % butterworth bandpass non-
dimensional frequency
d=filtfilt(b,a,c); % apply the filter: use zero
phase filter (p=2)
return;

```

Filter & plot the filtered data

```
% filter the data
```

```
%
```

```
if flp > 0;
```

```
    e1=bandpass(e,flp,fhi,npts,delt);
```

```
    n1=bandpass(n,flp,fhi,npts,delt);
```

```
    z1=bandpass(z,flp,fhi,npts,delt);
```

```
    e=e1;
```

```
    n=n1;
```

```
    z=z1;
```

```
%
```

```
% plot the filtered data
```

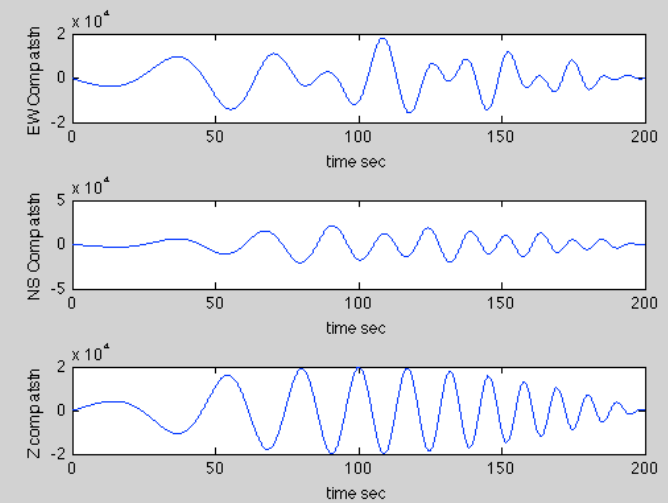
```
f2=figure('name','FILTERED SEISMOGRAMS');
```

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

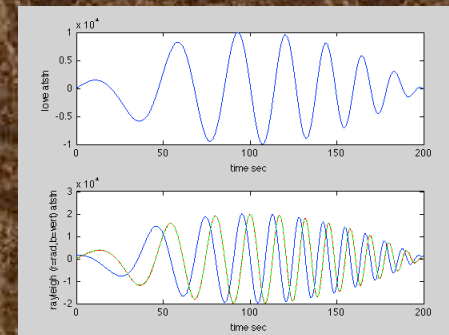
```
..... removed for clarity
```

```
else;
```

```
end;
```



*The vertical channel has also had a Hilbert transform applied so that the Rayleigh wave is in phase on the NS and Z components



GOAL: Solve for polarization

- Recognize that incoming seismic phases should represent the principal components, or the strongest signal, on the 3 component data
- The principal components, in turn, are equal to the eigenvectors of the covariance matrix of the 3 component matrix. This can be derived using PCA techniques
- Eigenvectors/values represent a spatial transformation which maximizes covariance between the 3 components, and they contain information on the azimuth from which the primary signal is derived
- **Since multiple phases may be present, we would prefer to look at short time windows of the 3 component data, or in other words, perform PCA on a running window through the continuous waveforms**

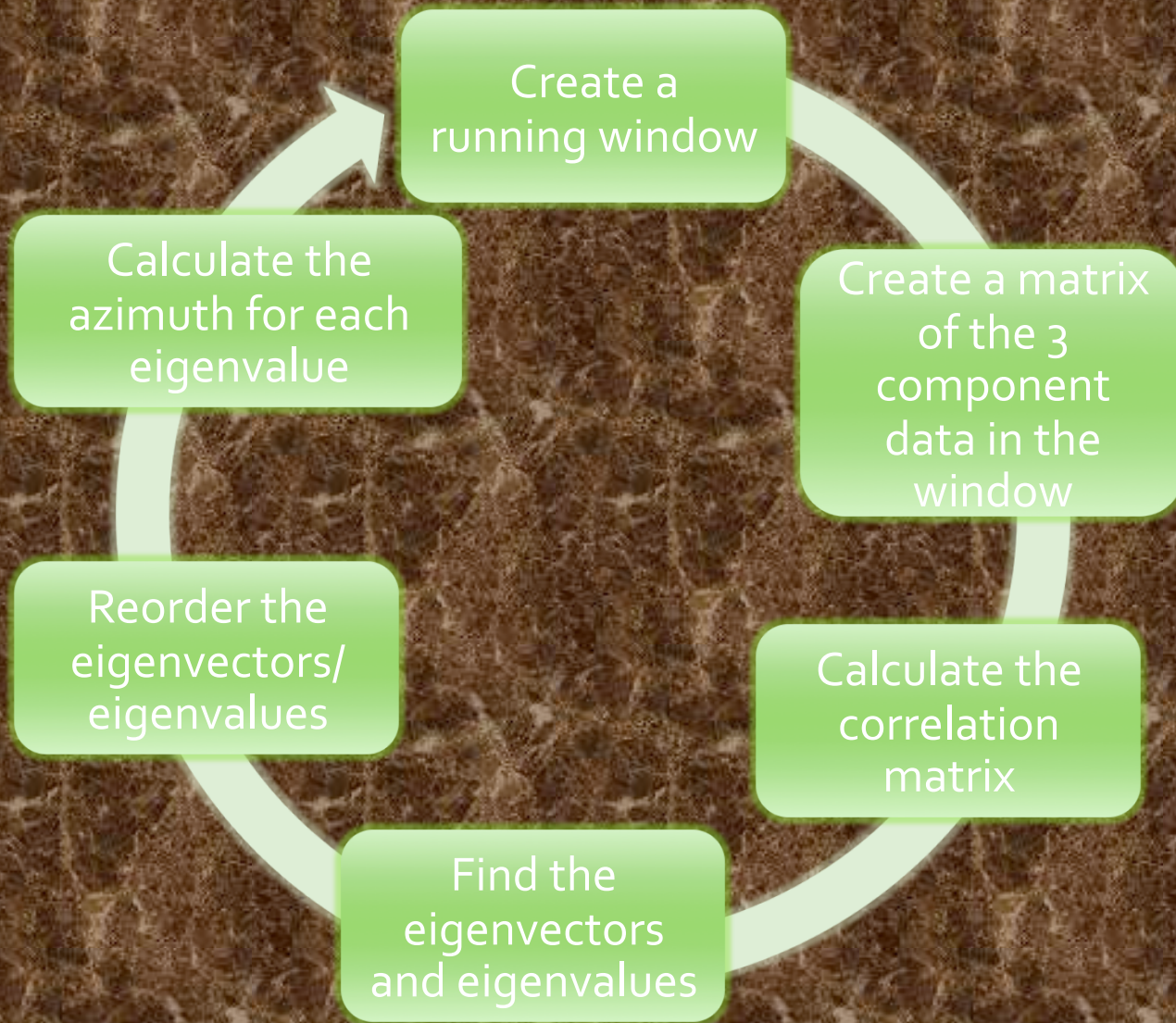
Results

- plot azimuths of eigenvectors
- plot azimuths exceeding 50% of maximum value

Main Code



Step 2: Main Code



Moving window using loops

```
% Moving window loop
```

```
%
```

```
npts1=fix(ttot/delt) + 1; % total number of samples to  
analyze
```

```
nwin=fix(twin/delt) + 1; % number of samples in a time  
window
```

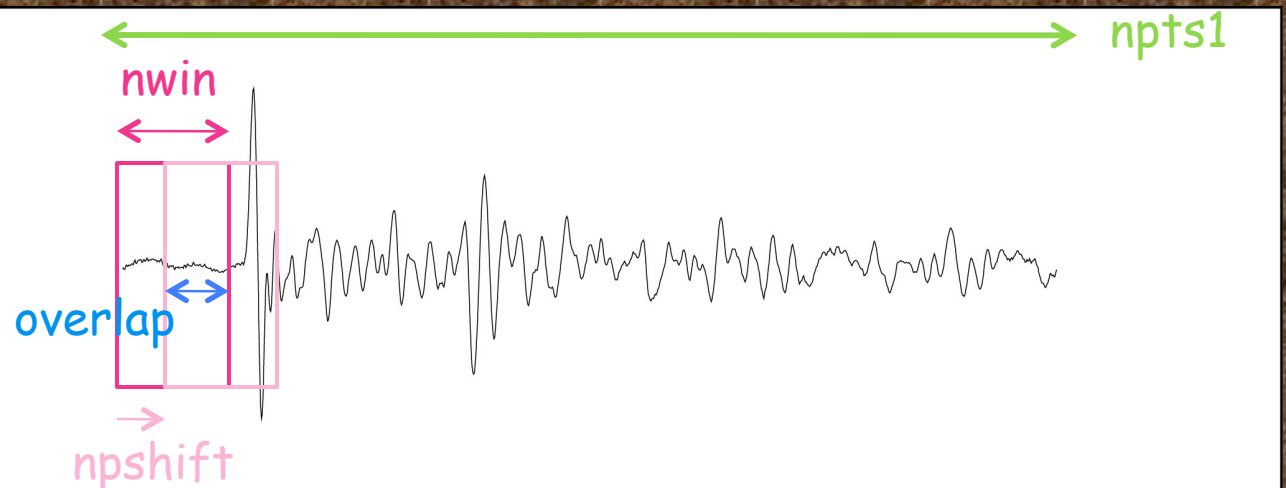
```
npshift=fix(twin/(2*delt))+1; % number of samples to shift  
over
```

```
kfin=fix((npts1-nwin)/(npshift+1))+1; % number of time windows  
considered
```

```
mxde1=0.;
```

```
mxde2=0.;
```

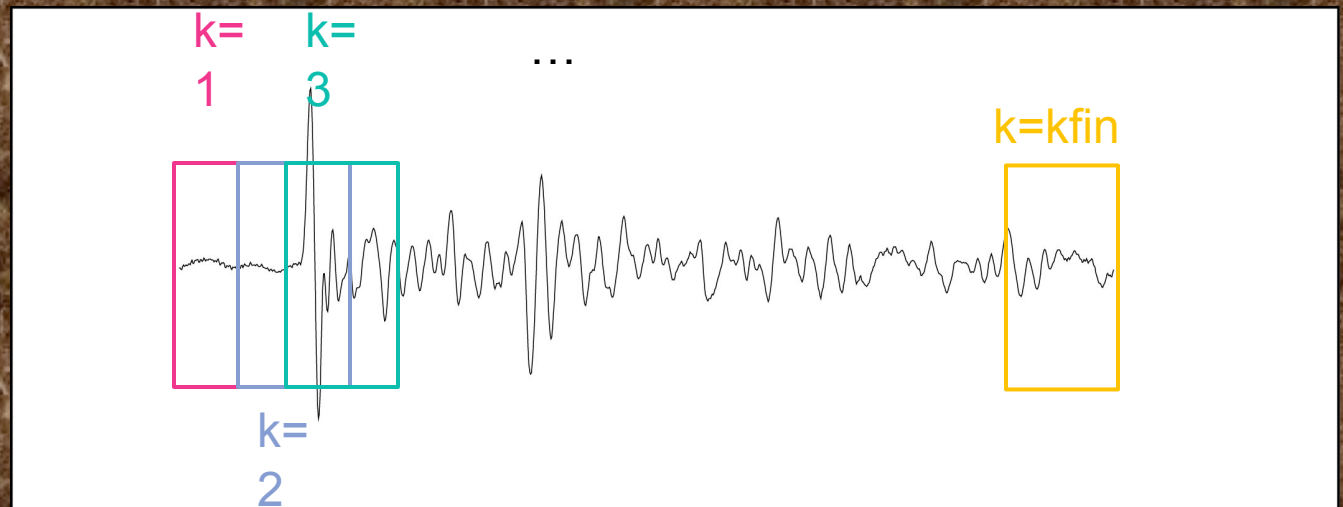
```
mxde3=0.;
```



```

for k=1:kfin;
    nwinst=(k-1)*(npshift-1)+1; % start of time window
    nwinfn=nwinst+nwin-1;      % end of time window
    ..... missing code to be supplied later
    t2(k)=delt*(nwinst-1);     % assign time for this window to
the window start
end;

```



Eigenvalues/Eigenvectors

When a Matrix multiplies a vector in general the direction and magnitude of the vector will change.

BUT there are special vectors where only the magnitude changes (on multiplication by the Matrix). These are called **eigenvectors** The value by which the length changes is the associated **eigenvalue**

We say that x is an eigenvector of A iff

$$Ax = \lambda x$$

In other words, x is an eigenvector if when you multiply it by A it returns a multiple of itself. λ is called the associated eigenvalue.

In Matlab use $[V,D] = \text{eig}(A)$ to get a matrix V whose columns are the eigenvectors of A and a **diagonal** matrix D whose entries on the diagonal are the corresponding eigenvalues.

```
>> A
```

```
A =
```

```
1    2
3    4
```

```
>> [V,D] = eig(A)
```

```
V =
```

```
-0.8246   -0.4160
 0.5658   -0.9094
```

```
D =
```

```
-0.3723    0
         0    5.3723
```

Missing code from inside our loop

```
a=csigm(e,n,z,nwinst,nwinfn);    % signal matrix
c=a'*a;                          % covariance matrix
[v1,d1]=eig(c);                  % eigenvalue/eigenvectors
[v,d]=order(v1,d1);              % put eigenvalues & eigenvectors
                                in ascending order

% azimuth for each of the 3 eigenvalues
ang1(k)=atan2(v(1,1),v(2,1)) * 180/pi;
ang2(k)=atan2(v(1,2),v(2,2)) * 180/pi;
ang3(k)=atan2(v(1,3),v(2,3)) * 180/pi;

% incidence angle of the 3 eigenvalues
vang1(k)=acos(abs(v(3,1)))* 180/pi; %angle from the vertical
vang2(k)=acos(abs(v(3,2)))* 180/pi;
vang3(k)=acos(abs(v(3,3)))* 180/pi;
```


Still in loop

```
de1(k)=d(1);
```

```
de2(k)=d(2);
```

```
de3(k)=d(3);
```

```
mxde1=max(mxde1,de1(k)); % find the maximum values
```

```
mxde2=max(mxde2,de2(k));
```

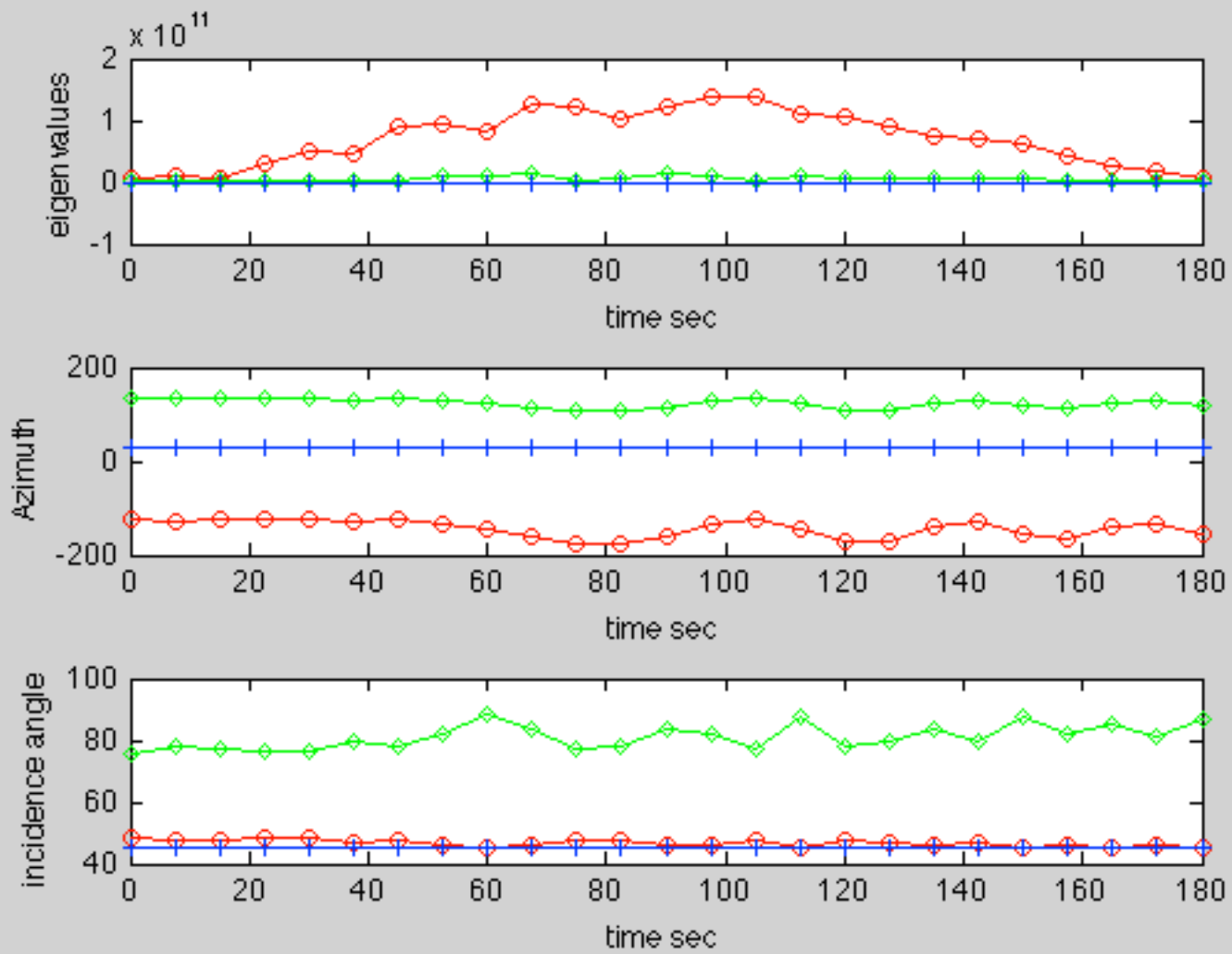
```
mxde3=max(mxde3,de3(k));
```

Outside of Loop again

```
f3=figure('name','Eigenvalues and Inferred Azimuth');  
subplot(3,1,1);  
plot(t2,de1,'-or',t2,de2,'-dg',t2,de3,'-+b');  
xlabel('time sec');  
ylabel('eigenvalues');
```

```
subplot(3,1,2);  
plot(t2,ang1,'-or',t2,ang2,'-dg',t2,ang3,'-+b');  
xlabel('time sec');  
ylabel('Azimuth');
```

```
subplot(3,1,3);  
plot(t2,vang1,'-or',t2,vang2,'-dg',t2,vang3,'-+b');  
xlabel('time sec');  
ylabel('incidence angle');
```



GOAL: Solve for polarization

Results

- plot azimuths of eigenvectors
- plot azimuths exceeding 50% of maximum value

Main Code



Rose Diagrams

```
% Rose plots
f4=figure('name','Azimuth Distribution');
subplot(2,3,1);
title('Azimuth - Largest Eigenvalue');
rose(ang1*pi/180,100);

subplot(2,3,2);
title('Azimuth - Intermediate Eigenvalue');
rose(ang2*pi/180,100);

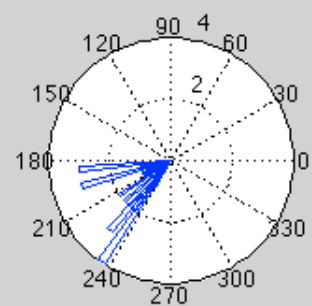
subplot(2,3,3);
title('Azimuth - Smallest Eigenvalue');
rose(ang3*pi/180,100);
```

```

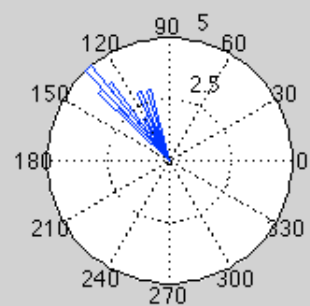
nskip=1;
if nskip == 1;
    else;
neig1=0;
neig2=0;
neig3=0;
for k=1:kfin;
    if de1(k) >= 0.5*mxde1;
        neig1=neig1+1;
        angm1(neig1)=ang1(k);
    else;
    end;
    if de2(k) >= 0.5*mxde2;
        neig2=neig2+1;
        angm2(neig2)=ang2(k);
    else;
    end;
    if de3(k) >= 0.5*mxde3;
        neig3=neig3+1;
        angm3(neig3)=ang3(k);
    else;
    end;
end;
subplot(2,3,4);
title('Azimuth - Largest Eigenvalue,50%
      Threshold');
rose(angm1*pi/180,100);
subplot(2,3,5);
title('Azimuth - Intermediate Eigenvalue,
      50% Threshold');
rose(angm2*pi/180,100);
subplot(2,3,6);
title('Azimuth - Smallest Eigenvalue,50%
      Threshold');
rose(angm3*pi/180,100);
end;

```

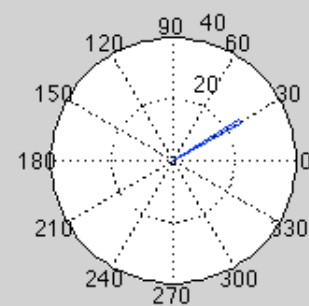

Azimuth - Largest Eigenvalue



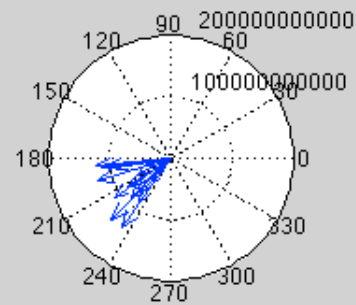
Azimuth - Intermediate Eigenvalue



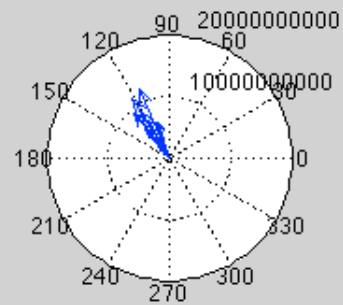
Azimuth - Smallest Eigenvalue



Azimuth - Largest Eigenvalue



Azimuth - Intermediate Eigenvalue



Azimuth - Smallest Eigenvalue

