```
%Matlab Practical 8
%Zach Viq
clearvars;
clear;
close all;
%q1.1
load swarm catalog NC.mat
%q1.2
index 123 = swarm index==123;
lats = swarm quake(index 123,7);
lons = swarm quake(index 123,8);
depth = swarm quake(index 123,9);
figure(1);
plot3(lons, lats, depth, 'k+'); set(gca, 'zdir', 'reverse'); title('Earthquake
Hypocenters');
km deglat = 111;
R = 6378;
avg lat = mean(lats);
km deglong = 2*pi*R earth*cos(avg lat*pi/180)/360;
a1 = 1/km deglong; %km/degree of longitude
a2 = 1/km deglat; %km/degree of latitude
a3 = 1; %km
daspect([a1 a2 a3]);
%q1.3
응 {
    The morphology of this swarm is mostly linear with depth.
응 }
%q1.4
lats km norm = (lats-mean(lats)) .* km deglong;
longs km norm = (lons-mean(lons)) .* km deglong;
dep norm = depth-mean(depth);
C = cov([lats km norm longs km norm dep norm]);
응 {
    0.1762, 0.1115, -0.4608
    0.1115, 0.1699, -0.3964
    -0.4608, -0.3964, 4.6093
응 }
%q1.5
eig(C)
    The eigenvalues tell us that one principal component is able to explain
\sim95\% of the variance in the dataset.
```

```
응 }
%q1.6
index 170 = swarm index == 170;
lats = swarm quake(index 170,7);
lons = swarm quake(index 170,8);
depth = swarm quake(index 170,9);
figure(2);
plot3(lons, lats, depth, 'k+'); set(gca, 'zdir', 'reverse'); title('Earthquake
Hypocenters');
daspect([a1 a2 a3]);
응 {
    The morphology of this swarm is mostly planar.
응 }
lats km norm = (lats-mean(lats)) .* km deglong;
longs km norm = (lons-mean(lons)) .* km deglong;
dep norm = depth-mean(depth);
C = cov([lats km norm longs km norm dep norm]);
응 {
    0.6677
              0.4847
                       -0.0150
    0.4847
             0.8800
                        0.2930
   -0.0150
              0.2930
                         2.2133
응 }
eig(C)
응 {
    The eigenvalues tell us that it now takes two dimensions to explain >95%
of the dataset, consistent with our observation that this swarm is planar.
응 }
%q2.1
load Data for Practical8.mat
%q2.2
figure (3); subplot (2,1,1); hold on;
plot(waveform(:,1), waveform(:,2), "DisplayName", "waveform");
plot(waveform1(:,1), waveform1(:,2), "DisplayName", "waveform1");
xlabel("Time (s)"); ylabel('Amplitude')
legend()
title('No-Noise');
hold off;
%q2.3
응 {
    There is about a 5 second delay between waveform and waveform1.
응 }
```

```
%q2.4
subplot(2,1,2);
[c,lags] = xcorr(waveform(:,2), waveform1(:,2), 'coef');
time per lag = 2*(waveform(end,1)-waveform(1,1))/length(lags);
lag time = lags.*time per lag;
plot(lag time,c); xlabel("Lag Time (s)"); ylabel("Correlation Coefficient");
%q2.5
응 {
    The maximum cross-correlation coefficient occurs at 5 seconds, which is
what we observed in the time-domain plot.
응 }
%q2.6
sig = 100;
noisy = waveform(:,2) + sig .* randn(length(waveform(:,2)),1);
noisy1 = waveform1(:,2) + sig .* randn(length(waveform1(:,2)),1);
figure(4); subplot(2,1,1); hold on;
plot(waveform(:,1),noisy,"DisplayName","waveform+noise");
plot(waveform1(:,1),noisy1,"DisplayName","waveform1+noise");
xlabel("Time (s)"); ylabel('Amplitude')
legend()
title('With Some Noise');
hold off;
subplot(2,1,2);
[c, lags] = xcorr(noisy, noisy1, 'coef');
plot(lag time,c); xlabel("Lag Time (s)"); ylabel("Correlation Coefficient");
    It is still clear in the time domain that the offset if 5 seconds, and
the cross-correlation confirms this.
응 }
%q2.7
sig = 4000;
noisy = waveform(:,2) + sig .* randn(length(waveform(:,2)),1);
noisy1 = waveform1(:,2) + sig .* randn(length(waveform1(:,2)),1);
figure (5); subplot (2,1,1); hold on;
plot(waveform(:,1),noisy,"DisplayName","waveform+noise");
plot(waveform1(:,1),noisy1,"DisplayName","waveform1+noise");
xlabel("Time (s)"); ylabel('Amplitude')
legend()
title('With Lots of Noise');
hold off;
subplot(2,1,2);
[c, lags] = xcorr(noisy, noisy1, 'coef');
plot(lag time,c); xlabel("Lag Time (s)"); ylabel("Correlation Coefficient");
응 {
    It is unclear in the time domain what the offset is, but the cross-
correlation still clearly shows the offset as 5 seconds.
응 }
%q2.8
```

응 {

In the presence of heavy noise, correlation-based meurements are preferable to picking by a human because it offers a more robust and repeatable method than what a human can produce using their eye. %}

ans =

0.0352

0.1171

4.6567

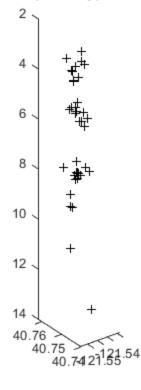
ans =

0.2576

1.2235

2.2799

## Earthquake Hypocenters



## Earthquake Hypocenters

