ENEE 324

Spring 2011

Homework #14

The goal of this homework is to load some real data, estimate several parameters, and then use bootstrap to estimate the validity of those parameters.

If you do turn this homework in, please answer the questions, record your observations when asked, supply the Matlab figures and results, and attach your Matlab code.

- Download the Matlab data file containing the data from the Homework page of the course website, or directly from < http://terpconnect.umd.edu/~jzsimon/enee324/hw/megExData.mat >, and load it into Matlab. It contains neural response data from a single (magnetoencephalography) channel while the subject repeatedly listened to a sound of duration 2 second.
 - a) Once you have loaded the data, use whos or the Matlab workspace to see what the new variables are that you have loaded. There should be 6 total. The most important are NS, data, fs, and time. NS is the number of trials that the stimulus was presented and response recorded. fs is the sampling frequency. time is a vector of time steps in ms. FYI, the stimulus turns on at t = 0 ms, which is not the earliest moment of time. data, the most important variable, contains the time-series data: the recorded magnetic field at each times step, for each of NS trials, with values ~ 10^{-13} Tesla.
 - b) Plot the first 4 trials, e.g. plot(time, data(:,m)) for 4 different m. Do they look similar? They should look more different than similar.
 - c) Below, you will need the indices of time that correspond to t = 0 ms and t = 400 ms, since we will only search for maxima and minima in that interval. Use
 t0index = find(diff(sign(time)));
 t400index = find(diff(sign(time-400)));

What is time(t0index)?—it should be near 0. What is time(t400index)? Why does this work?

- d) Calculate the following "plug-in" (i.e. potentially re-usable in bootstrap) estimates:
 - i) the mean response over all trials. Call it meandata. Hint: This should be as long as time. Plot it; it should have a coherent response for the first few hundred ms after t = 0 ms (the "onset response") and after t = 2000 ms ("offset response").
 - ii) the minimum (most negative) value of the mean response (call it minval) and the time at which it occurs (call it mintime), between 0 and 400 ms. You must use an automated method, not "by eye", because we will repeat this measurement hundreds of times in bootstrap. I use [minval,minindex] = min(meandata(t0index:t400index));

mintime = time(minindex+t0index-1)

(Why is it necessary to include "+t0index"? Why "-1"?) Do the results look agree with your eye's estimates? For this channel, it should be between 100 and 300 ms (e.g. "near" 200 ms).

- iii) Repeat (ii) for the maximum response between 0 and 400 ms, called maxval and maxtime. Do the results look agree with your eye's estimates? For this channel, it should be near 300 ms.
- e) Plotting with context:
 - i) Plot data and meandata overlapping in the same figure. Plot data first, in magenta, so that it doesn't overwhelm the plot, then plot meandata in black. You can do this in one line with plot(time, data,'m',time, meandata,'k')

Clearly, the individual signals have a lot of noise, in addition to signal (seen only in the mean).

ii) Plot meandata and put crosshairs where the minimum and maximum values occur. I use:

```
plot(time, meandata,'k')
line([mintime mintime], ylim,'color','r')
line([time(1) time(end)], [minval minval],'color','r')
line([maxtime maxtime], ylim,'color','b')
line([time(1) time(end)], [maxval maxval],'color','b')
Alternate a few times between axis tight, to get the big picture, and
axis([0 600 -2.5e-13 2.5e-13]), to see where the action is.
```

2) Prepare the bootstrap. We'll use B = 500 bootstrap samplings:

B = 500;

Now perform the bootstrap calculations. In a for loop, e.g. with for b = 1:B and end surrounding the calculations, recalculate the same parameters as in (1), but with the bootstrap samplings. Save the bootstrap values of mintime, minval, maxtime and maxval in arrays called mintimeB, minvalB, maxtimeB and maxvalB. Recommended techniques:

- a) Put the bootstrap resampled index into a variable called indexTmp. This variable should be of length NS and contain random integers between 1 and NS.
- b) To keep the time information correct and only shuffle the trial numbers, replace data with data(:,indexTmp). For example, if in (1) you had an equation of the form meandata = mean(...data...), replace it with meandataTmp = mean(...data(:,indexTmp)...) The temporary variable meandataTmp is explained next.
- c) (This is not crucial, but it keeps things tidy so there will be less chance for mistakes and fewer index gymnastics.) Put all calculations in temporary variables, as in (a), and as in [minvalTmp,minindexTmp] = min(meandataTmp(t0index:t400index));

```
mintimeTmp = time(minindexTmp+t0index-1);
```

and the after all temporary variables are created, only then put them into the bootstrap vectors:

```
...
mintimeB(b) = mintimeTmp;
minvalB(b) = minvalTmp;
maxtimeB(b) = maxtimeTmp;
maxvalB(b) = maxvalTmp;
```

- 3) Analyze the statistics of minvalB, maxvalB, mintimeB and maxtimeB.
 - a) What is the mean of minvalB? How does it compare to minval? What is the standard deviation of minvalB? Is the bias large enough to that you should use the mean of minvalB instead of minval? (A standard rule of thumb is to check whether |bias/σ| > 0.25). Use hist() to see the distribution of values of minvalB. Is it relatively smooth (in this case it should be)?
 - b) Repeat (a) but for mintimeB. In this case the distribution may *not* be smooth. You may need to fiddle with the number of bins used in hist(), since 1) you may have outliers, and 2) you want the width of the bins to be roughly two or three times the smallest time step (so they get filled more nicely). (What is the smallest time step?)
 - c) Repeat (a) but for maxvalB.
 - d) Repeat (b) for maxtimeB. Same precautions apply as in (b).
- 4) Create and examine the bootstrap mean time series meandataB. Since all of the previous estimates (e.g. mintime, minval) are single numbers, the bootstrap collection of them is a 1 x B vector. meandata, though, is a time series, so its bootstrap collection is a length(time) x B matrix, and it the boostrap loop you can fill it up like this: meandataB(:,b) = meandataTmp;
 - a) Plot the first 4 bootstrap mean time series, e.g. plot(time, meandataB (:,b)) for different b. Do they look similar? They should look more similar than different (unlike 1b).
 - b) Compute the mean and standard deviation of the bootstrap mean time series (across bootstrap samplings):

```
meandataBmean = mean(meandataB,2);
meandataBstd = std(meandataB,1,2);
```

 c) Plotting with context: Plot meandata and meandataB overlapping in the same figure. Plot meandataB first, in yellow, so that it doesn't overwhelm the plot, then plot meandata in black. Plot meandata + 2*meandataBstd and meandata - 2*meandataBstd, to act as error bars. You can do this in one line with:

plot(time, meandataB, 'y', time, meandata, 'k', time, ... meandata-2*meandataBstd, 'b', time, meandata+2*meandataBstd, 'b') The bootstrap versions of the response have much less noise than the original version (in data) plotted above in (1ei). Also the error bars (in blue) follow the yellow bootstrap cloud well. You may wish to restrict the axes to the important time region so the graph looks less busy: axis([0 600 -2.5e-13 2.5e-13])

d) Add estimates of the minimum and maximum values, with their own error bars, both in amplitude and time, to the same graph:

```
line([mintimeBmean mintimeBmean], minvalBmean+2*[-minvalBstd minvalBstd],'color','r')
line(mintimeBmean+2*[-mintimeBstd mintimeBstd], [minvalBmean minvalBmean],'color','r')
line([maxtimeBmean maxtimeBmean], maxvalBmean+2*[-maxvalBstd maxvalBstd],'color','r')
line(maxtimeBmean+2*[-maxtimeBstd maxtimeBstd], [maxvalBmean maxvalBmean],'color','r')
```