

Physics G4080/G6080 – Problem Set 3

Spring 2015

Due by 5:00 PM Monday March 30, 2015

1 Statistical Analysis of Mock Dataset

In this problem, you will analyze a mock data set that is available on the web site:

https://rbc.phys.columbia.edu/rdm/G4080_G6080_Spring_2015

There are 5 plain text files given, called v1, v2, v3, v4 and v5. (v1 is short for variable 1, etc.) Each text file contains 1,600,000 lines, which correspond to measurements of the given variable from a simulation. There are autocorrelations (correlations in time or measurement number) for each variable. There are also correlations between variables. In this problem, you need to analyze this data, using the methods discussed in class, to find average values, standard deviations of means, autocorrelations, etc.

In MATLAB, you can load these files in using the “Import Data” button on the main MATLAB window.

We will use M to represent the total number of values, *i.e.* $M = 1,600,000$. M is large enough so that it represents an effectively “infinite” sample size. Thus, the true data mean can be well approximated by averaging over all M values. Averages, standard deviations, etc. determined from the “infinite” sample will be denoted with a hat, *i.e.* \hat{v}_1 , etc.

We will use N to represent the number of measurements in a sample of the data. N corresponds to the amount of data you might actually collect in a simulation.

1. Determine the true means, \hat{v}_a , for v_1, v_2, \dots, v_5 from all M values.
2. Consider two values for our sample size: $N = 1,000$ and $N = 10,000$. There are M/N samples of this size in our M values. **Histogram the sample means for these values of N** and determine the true standard deviation of the means $\hat{\sigma}_{\bar{v}_a, N}$. You can determine $\hat{\sigma}_{\bar{v}_a, N}$ by assuming each of the M/N samples is independent. This is reasonable, provided the autocorrelations in the data are smaller than the values of N you use. Do your two values for $\hat{\sigma}_{\bar{v}_a, N}$ show the correct behavior with N ?

Need to make sure it's the “true” mean

3. You can now determine the true autocorrelation function for each variable, $\hat{C}_{v_a, n}$, which is given by

$$\hat{C}_{v_a, n} = \frac{1}{M - n} \sum_{i=1}^{M-n} (v_{a, i+n} - \hat{v}_a) (v_{a, i} - \hat{v}_a) \quad (1)$$

Here n goes from 0 to some maximum value n_{cut} with $n_{\text{cut}} \ll M$. Plot $\hat{C}_{v_a, n} / \hat{C}_{v_a, 0}$ versus n for $a = 1 \dots 5$.

4. Find the integrated autocorrelation times

$$\hat{\tau}_{\text{int},v_a} \equiv \frac{1}{2} \frac{1}{\hat{C}_{v_a,0}} \sum_{n=-n_{\text{cut}}}^{n_{\text{cut}}} \hat{C}_{v_a,n} \quad (2)$$

Estimate a value for n_{cut} from your plots. n_{cut} should be large enough that $\hat{C}_{v_a,n}/\hat{C}_{v_a,0}$ has gotten close enough to zero that the value of $\hat{\tau}_{\text{int},v_a}$ is not effected by modest changes in n_{cut} .

5. Calculate the true standard deviation of the data, *i.e.*

$$\hat{\sigma}_{v_a}^2 \equiv \frac{1}{M-1} \sum_{i=1}^M (v_{a,i} - \hat{v}_a)^2 \quad (3)$$

For a sample of size N , we should have

$$\hat{\sigma}_{\bar{v}_a,N} = \sqrt{\frac{2\hat{\tau}_{\text{int},v_a}}{N}} \hat{\sigma}_{v_a} \quad (4)$$

Does this relation hold for your analysis?

6. Calculate the true covariance matrix for the data, defined by

$$\hat{c}_{v_a,v_b} = \frac{1}{M} \sum_{i=1}^M (v_{a,i} - \hat{v}_a) (v_{b,i} - \hat{v}_b) \quad (5)$$

It is customary to define a normalized version of \hat{c}_{v_a,v_b} by

$$\hat{\rho}_{v_a,v_b} \equiv \frac{\hat{c}_{v_a,v_b}}{\hat{\sigma}_{v_a} \hat{\sigma}_{v_b}} \quad (6)$$

Since $\hat{c}_{v_a,v_a} = \hat{\sigma}_{v_a}^2$, $\hat{\rho}_{v_a,v_b}$ has ones on the diagonals and the off diagonals give a ready measure of the covariance between variables. Make a 3d surface plot of $\hat{\rho}_{v_a,v_b}$, using, for example, MATLAB's `surf()` function.

7. Now, pick two groups of data from the full universe of data. One should have $N = 1,000$ and the other should have $N = 10,000$. These two groups represent results one might get from simulations. We want to see how well these groups reproduced the true statistical results for these data. Estimate the autocorrelation function $C_{v_a,n}$ from these two groups and the integrated autocorrelation time. Use these to determine the standard deviation of the mean $\sigma_{\bar{v}_a,N}$. Compare this with the results from the universe of data. Also compare the normalized covariance matrix ρ_{v_a,v_b} from these small samples with the universe of data.

2 Jackknife Analysis

In this problem, we will see how the jackknife method can be used to find errors on functions on the mean values of data, *i.e.* on $f(\bar{v}_a)$. We can use the universe of data to calculate the errors and then calculate the same error from a sample of size $N = 1,000$ or $N = 10,000$.

We will consider three functions in what follows

$$f_1(\bar{v}_a) = \bar{v}_1/\bar{v}_2 \quad (7)$$

$$f_2(\bar{v}_a) = \exp(\bar{v}_3 - \bar{v}_4) \quad (8)$$

$$f_3(\bar{v}_a) = \left(\frac{\bar{v}_1}{\bar{v}_2} + \frac{\bar{v}_3}{\bar{v}_4} \right) \log \bar{v}_5 \quad (9)$$

1. Break the M measurements up in to groups of size N , calculate \bar{v}_a for each group and then calculate $f_i(\bar{v}_a)$ for each group. Calculate these functions of the data means for all M/N groups and find the standard deviation for $f_i(\bar{v}_a)$, $\hat{\sigma}_{f_i,N}$.
2. Calculate $\hat{\sigma}_{f_i,N}$ from naive propagation of errors, *i.e.* using $\hat{\sigma}_{\bar{v}_a,N}$ and neglecting correlations between the v_i . Compare with your results from part 1.
3. We now want to estimate $\sigma_{\bar{v}_a,N}$ and $\sigma_{f_i,N}$ using the jackknife method from a single sample of size N . First we must deal with the autocorrelations in the data, and you have an idea of the integrated autocorrelation time from the first problem. We proceed as follows here. Average your N data values into bins of size b . This will produce N/b data values. Then use the jackknife method to estimate \bar{v}_a and $\sigma_{\bar{v}_a,N}$ from these N/b data values. The jackknife method resums these N/b values as done in class, *i.e.*

$$v'_{a,k} = \frac{1}{N/b - 1} \sum_{i=1, i \neq k}^{N/b} v_{a,i} \quad (10)$$

From these jackknife values, you can determine \bar{v}_a and $\sigma_{\bar{v}_a,N}$. Do this for a few different values of b comparable to the integrated autocorrelation time to check that your results do not depend strongly on b .

4. Now calculate $f_i(v'_{a,k})$ for each of the N/b jackknife blocks. You can then determine $\sigma_{f_i,N}$ from

$$\sigma_{f_i,N}^2 = \frac{N/b - 1}{N/b} \sum_{k=1}^{N/b} (f_i(v'_{a,k}) - f_i(\bar{v}_a))^2 \quad (11)$$

Again, do this for a few values of b that are comparable to the integrated autocorrelation time. How does $\sigma_{f_i,N}$ compare with $\hat{\sigma}_{f_i,N}$ from part 1?

Fun. Find τ for each N bucket, both sizes, then find σ_{τ} .

3 Integrated Autocorrelation Time Estimator - G6080 only

In class, we discussed the need to apply a cutoff when calculating τ_{int} , as in Equation (2). We can explore this numerically, given our universe of data. In particular, in this question you can investigate the error on τ_{int} as n_{cut} is increased.

Again choosing two values for N (1,000 and 10,000), estimate τ_{int} for each of the M/N samples of size N in the universe of data, as a function of n_{cut} . Then find the standard deviation $\sigma_{\tau, N}$ of $\tau_{\text{int}}(n_{\text{cut}}$ by using all M/N samples. Does the standard deviation with $n_{\text{cut}} \sim N$ decrease as N increases?

4 Argon Molecular Dynamics

We can now apply these statistical ideas to the results of your argon MD simulation. Run as long a simulation as is practical and make measurements of the temperature, potential energy and the time average of the virial, which is given by

$$\sum_i \sum_{j>i} r_{ij} \frac{\partial V_{ij}}{\partial r_{ij}} \quad (12)$$

every MD time step. You should be able to run a few thousand steps, after thermalization. Measure the autocorrelation times for the temperature, potential energy and virial. Also measure the covariance matrix for these 3 quantities. Use your estimate of the autocorrelation times, along with binning and the jackknife method to give an error on the pressure from your simulation.

Find σ_{Pressure} using binning/jackknife and τ_{Pressure}