

# Summary of $\chi$ pod Chameleon EQ14 Analysis

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# 1 Overview

- This document is an attempt to provide an overview/summary of what i've found in my  $\chi$ pod analysis so far.
- The motivation/goal for all this work is to show if /how well the CTD- $\chi$ pod method works for estimating  $\chi, \epsilon, K_T$ , etc from fast temperature (thermistor) profiles. The idea is to deploy  $\chi$ pods on regular CTD casts on WOCE/CLIVAR cruises etc. to making mixing measurements.
- Before dealing with all the issues with the CTD deployments (depth loops, entrained water, rosette-induced turbulence etc.), I wanted to verify that the method itself worked w/out these complications.
- The Chameleon microstructure profiler has both thermistor and shear probes, so this seemed like an ideal way to test the method. I would apply the  $\chi$ pod method to the chameleon thermistor data only ( $\chi_\chi, \epsilon_\chi$ ), and compare to the 'true' results computed using the shear probes ( $\chi, \epsilon$ ).
- I found that the estimates of  $\chi$  agreed well, but  $\epsilon_\chi$  was biased low compared to  $\epsilon$  (Figure 1,2,3).
- The  $\chi$ pod method requires assuming a mixing efficiency, and uses the normal assumption that  $\gamma = 0.2$ . I computed gamma from the chameleon data (formula) and found that it was about an order of magnitude smaller than 0.2; hence the low epsilon estimates?
- The comparison of  $\epsilon_\chi$  to  $\epsilon$  seems to improve with increased averaging (of either multiple profiles or larger depth ranges).

## 2 Data and Processing

- Data are from Chameleon profiles near the equator during the ‘EQ14’ experiment.
- Sally shared w/ me Chameleon data that she and Jim processed. I ended up re-processing it using a smaller fmax (7Hz) because it looked like the thermistor spectra rolled off much lower than the assumed 32Hz.
- `ComputeChi_Chameleon_Eq14.m` : Applies  $\chi$ pod method to Chameleon profiles from EQ14.
- `Make_Overview_Plots.m` Makes almost all the figures in this document.
- The noise floor of Chamleon  $\epsilon$  was determined to be  $\log_{10}[\epsilon] = -8.5$ . Values below this threshold were discarded.  $\chi$ pod values below this threshold were also discarded, in order to make a valid comparison. An upper limit of  $\log_{10}[\epsilon] = -5$  (determined by Jim?) was also applied.

## **3 Results**

### **3.1 Overview**

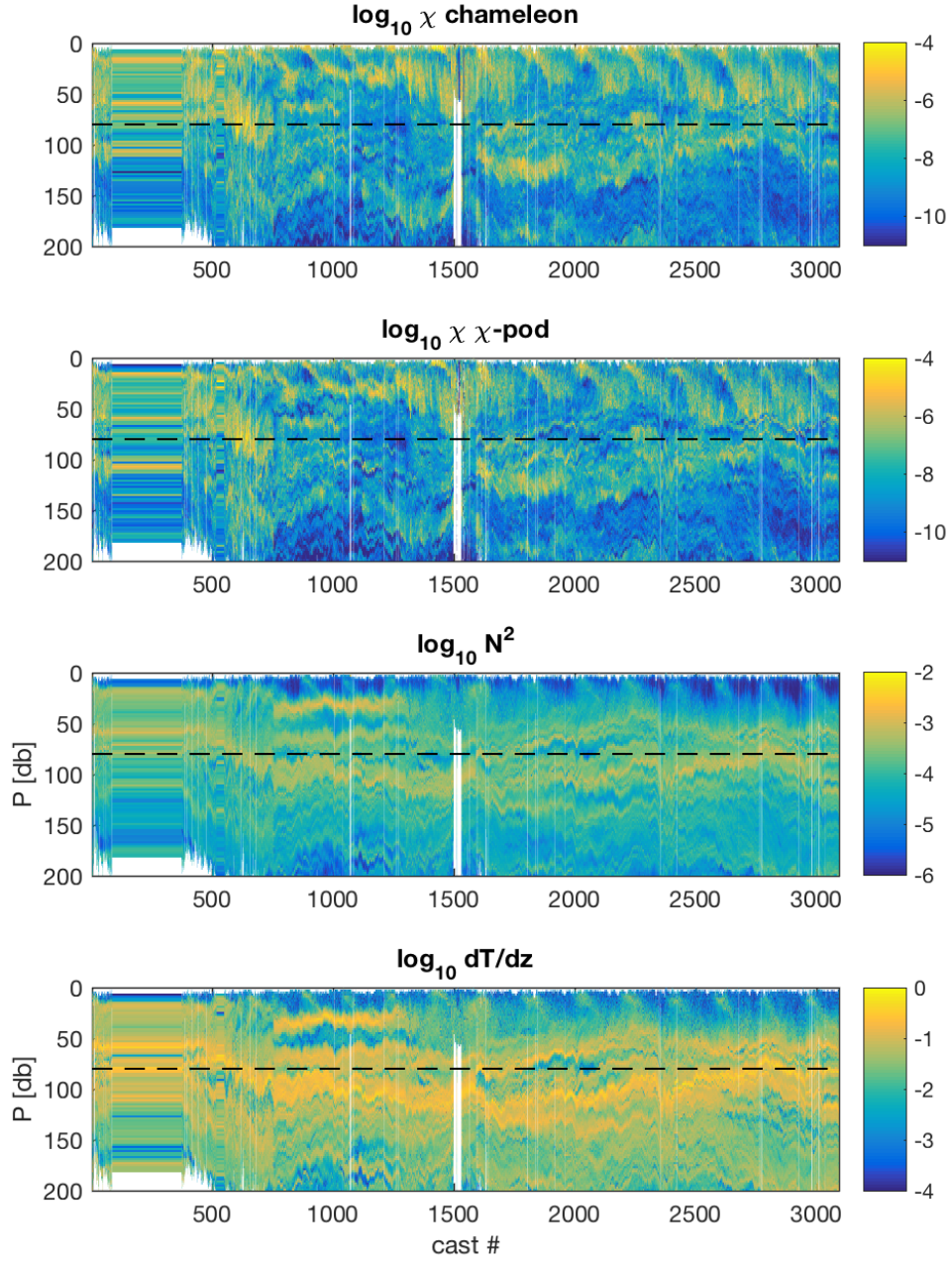


Figure 1: Comparison of  $\chi$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Horizontal line at 80m shows region excluded in further analysis because it contains near-surface convection.

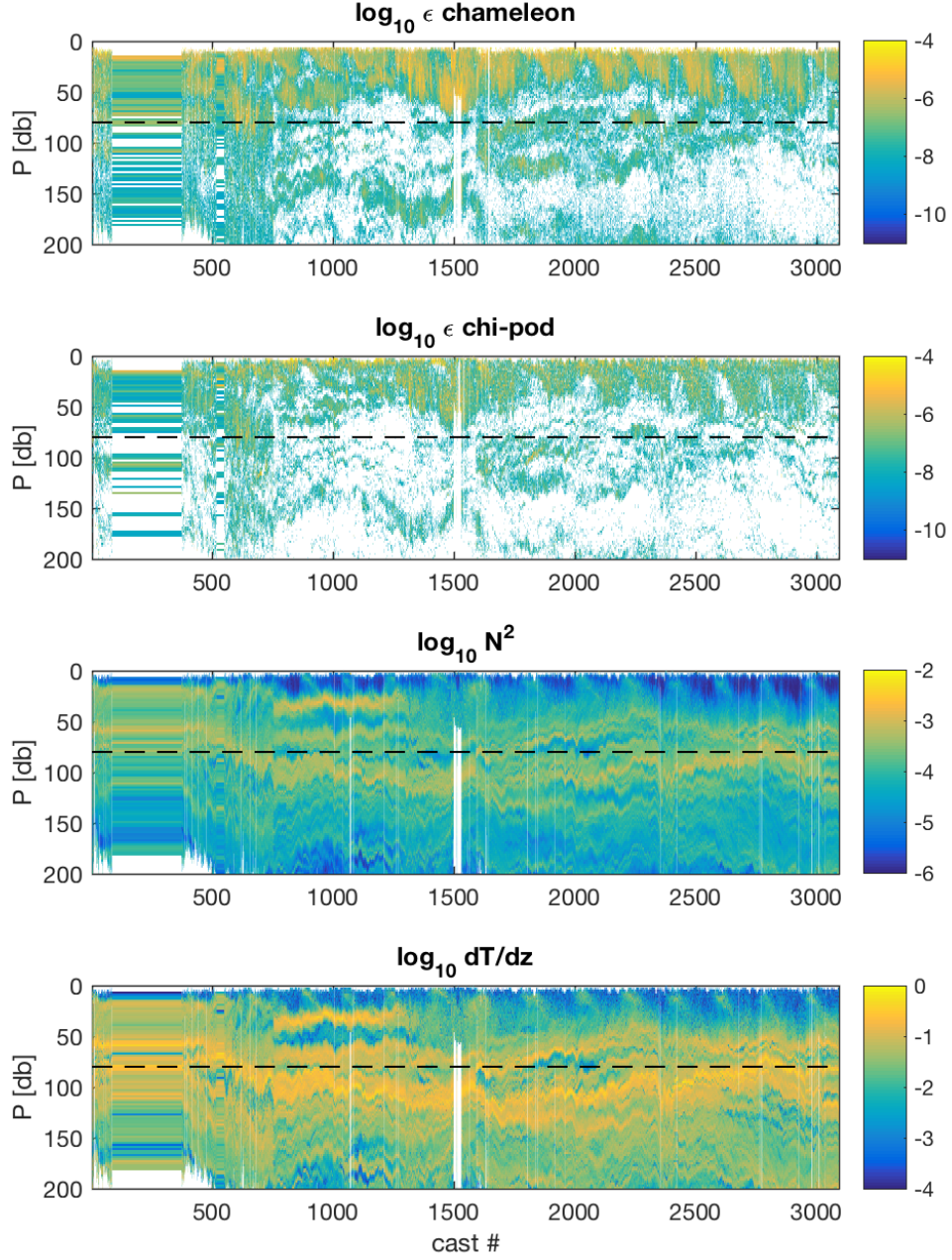


Figure 2: Comparison of  $\epsilon$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Values of  $\epsilon$  below chameleon noise floor ( $\log_{10}[\epsilon] = -8.5$ ) have been nan'd out. Horizontal line at 80m shows region excluded in further analysis because it contains near-surface convection.

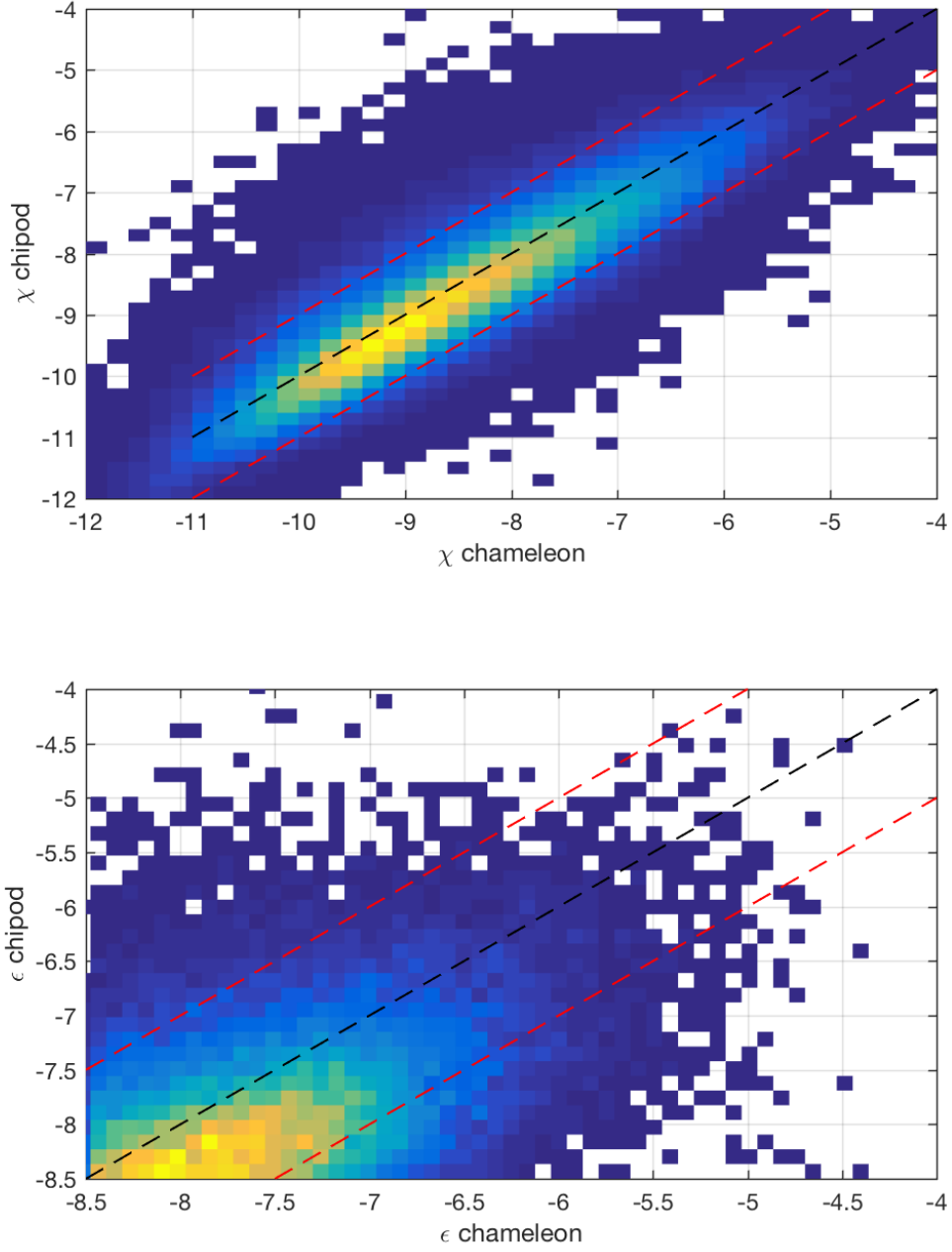


Figure 3: Comparison of  $\chi$   $\epsilon$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Values of  $\epsilon$  below chameleon noise floor ( $\log_{10}[\epsilon] = -8.5$ ) have been nan'd out. Black line is 1:1, red lines are  $\pm$  order of magnitude. Only data below 80m is used.

### 3.2 Comparing individual estimates of $\epsilon$

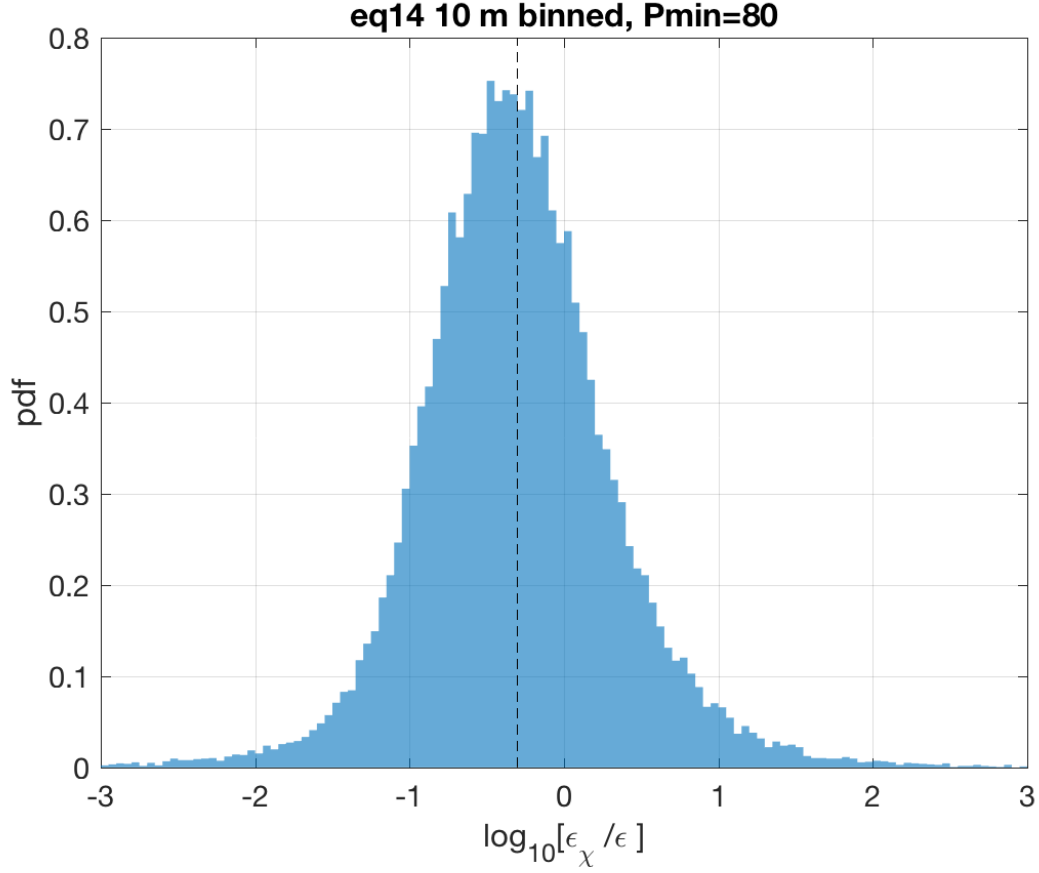


Figure 4: EQ14: Histogram of the ratio of  $\epsilon$  estimates from  $\chi$ pod method to the chameleon values, for  $\chi$ pod method applied to 1m binned profiles, and applied to just patches. Estimates for each profile were averaged in 10m depth bins. Vertical line shows mean of  $\log_{10}[\epsilon_{\chi}/\epsilon]$ .



### 3.3 Normalized eps vs chi plots

Assuming that

$$\gamma = \frac{N^2 \chi}{2\epsilon \langle T_z \rangle^2} \quad (1)$$

, plotting  $[\chi/t_z^2]$  vs  $[\epsilon/N^2]$  should follow a straight line with slope equal to  $2\gamma$ .

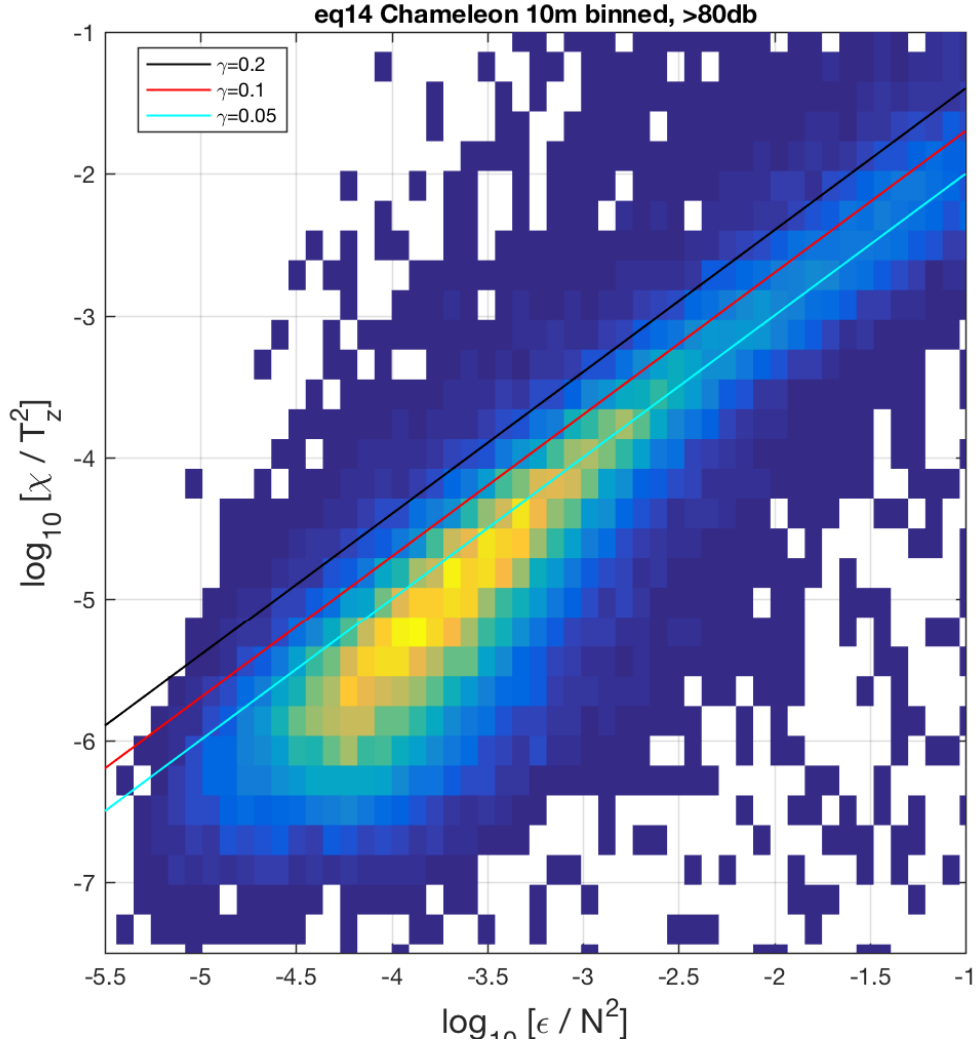


Figure 5: EQ14: 10m binned chameleon  $\epsilon/N^2$  vs  $\chi/t_z^2$  for \*below 80db\*. Lines show different values of  $\gamma$ . Values of  $\epsilon$  below noise floor ( $\log_{10}\epsilon < -8.5$ ) are discarded also.

### 3.4 Averaging many profiles of $\epsilon$

Figure 6 shows one example. A folder with many profiles is located at: [https://github.com/OceanMixingGroup/Analysis/tree/master/Andy\\_Pickering/eq14\\_patch\\_gamma/figures/chi\\_eps\\_profiles\\_40profavgs](https://github.com/OceanMixingGroup/Analysis/tree/master/Andy_Pickering/eq14_patch_gamma/figures/chi_eps_profiles_40profavgs).

I tried making plots of normalized chi vs eps, and scatterplots of chi-pod vs chameleon epsilon, for data averaged across different numbers of profiles.

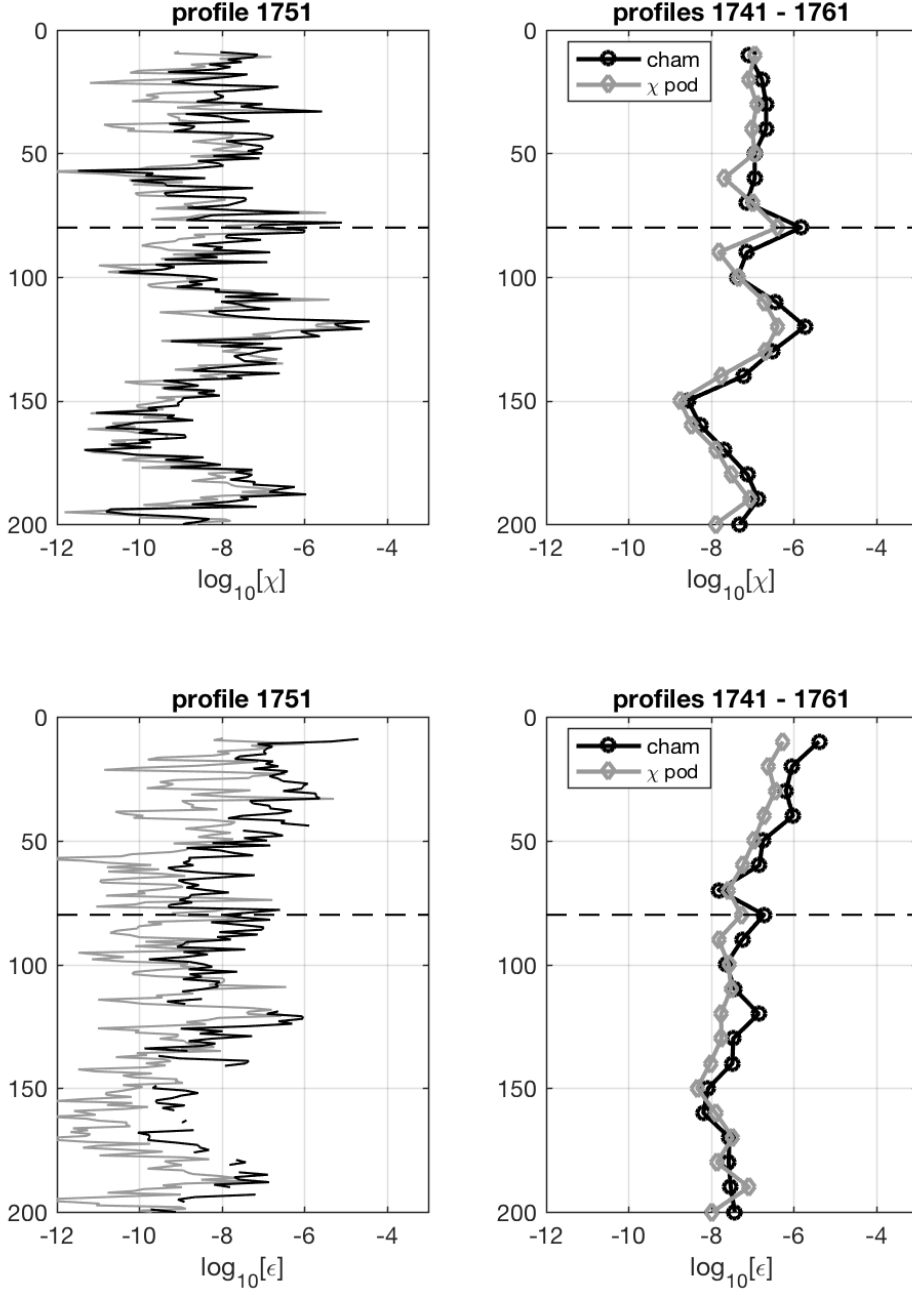


Figure 6: Example of averaging multiple profiles together. Left panels show a single profile from chamleon and chi-pod method. Right panels show average of  $\pm 40$  profiles, averaged in 10m depth bins.

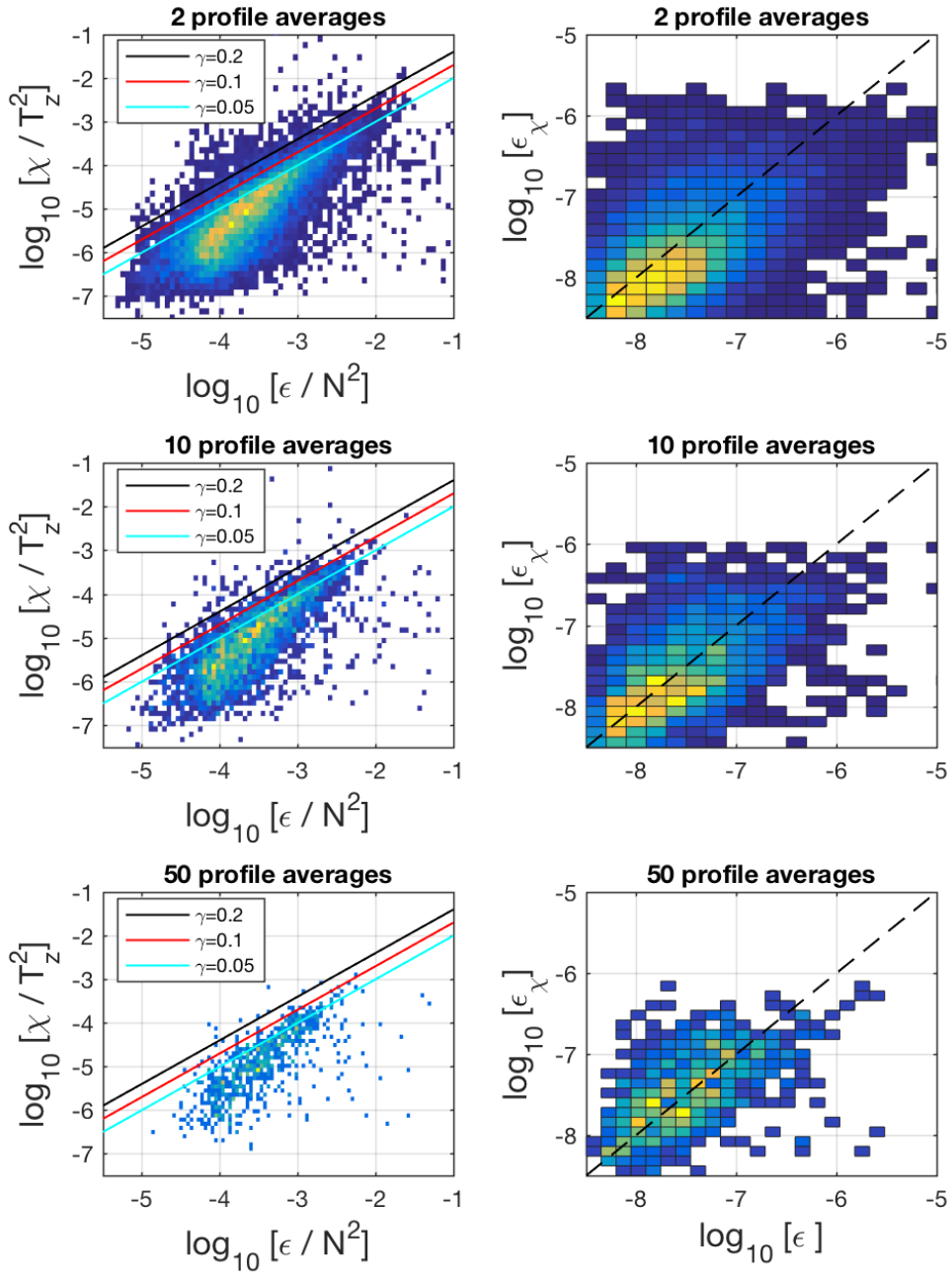


Figure 7:

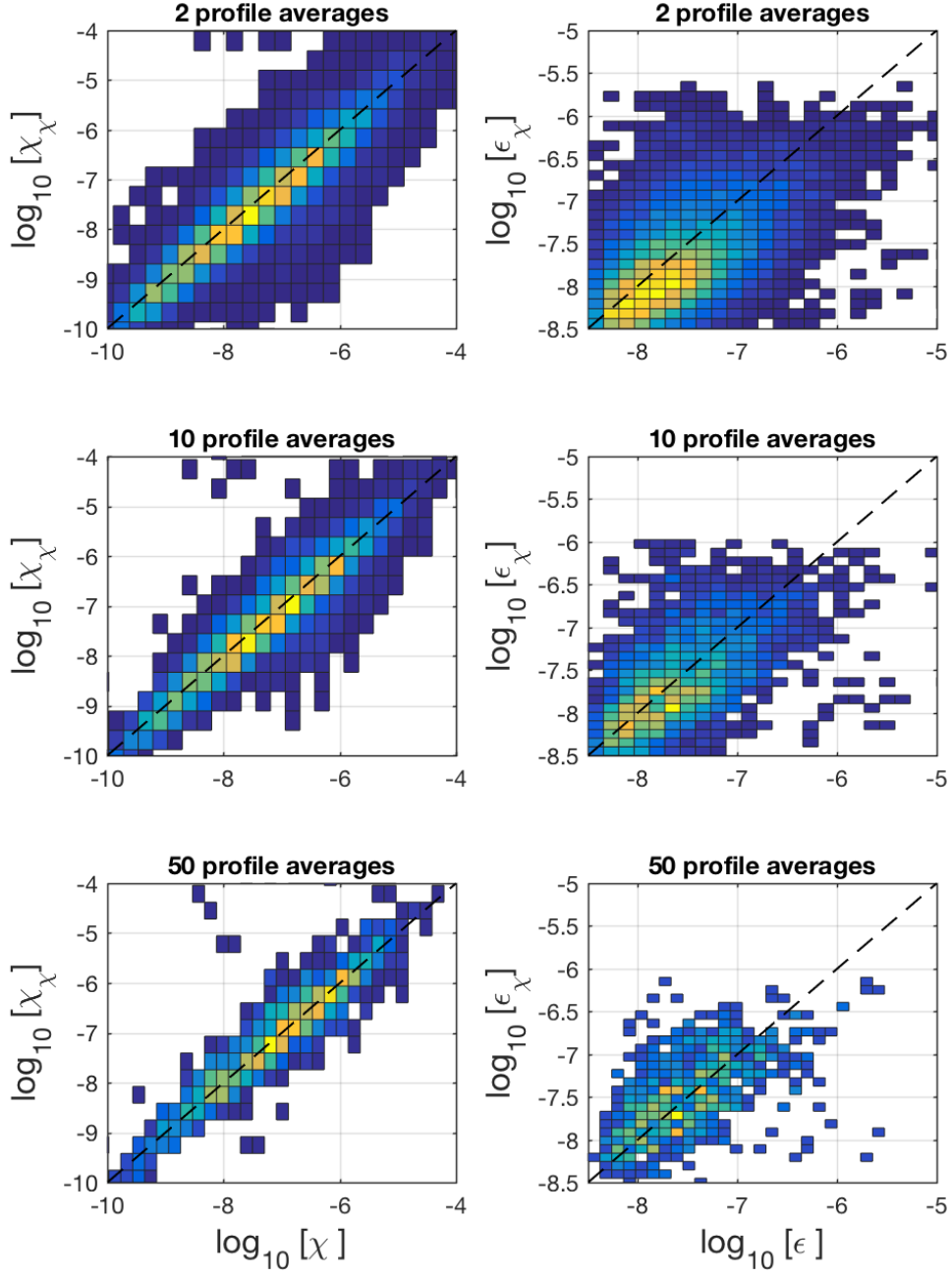


Figure 8: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) for different numbers of profiles averaged.

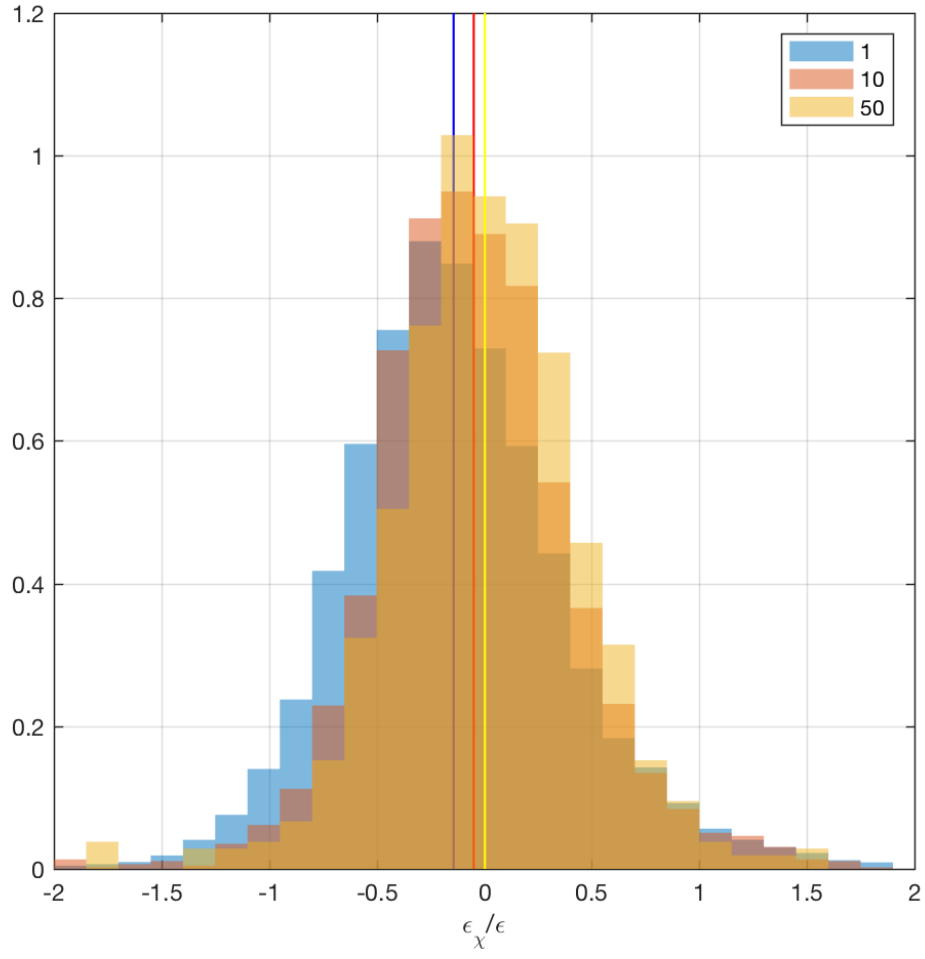


Figure 9:  $(\log_{10})$  Ratio of  $\epsilon_{\chi}/\epsilon$  for different numbers of profiles averaged. Consecutive chunks of  $N$  profiles were averaged, and then (normalized) histogram of the ratios was plotted. Vertical lines are mean of  $\log_{10}[\epsilon_{\chi}/\epsilon]$  for each distribution.

### **3.5 Effects of averaging in different-sized depth bins**

I also looked at the effects of averaging each profile in different sized depth bins instead of averaging profiles.

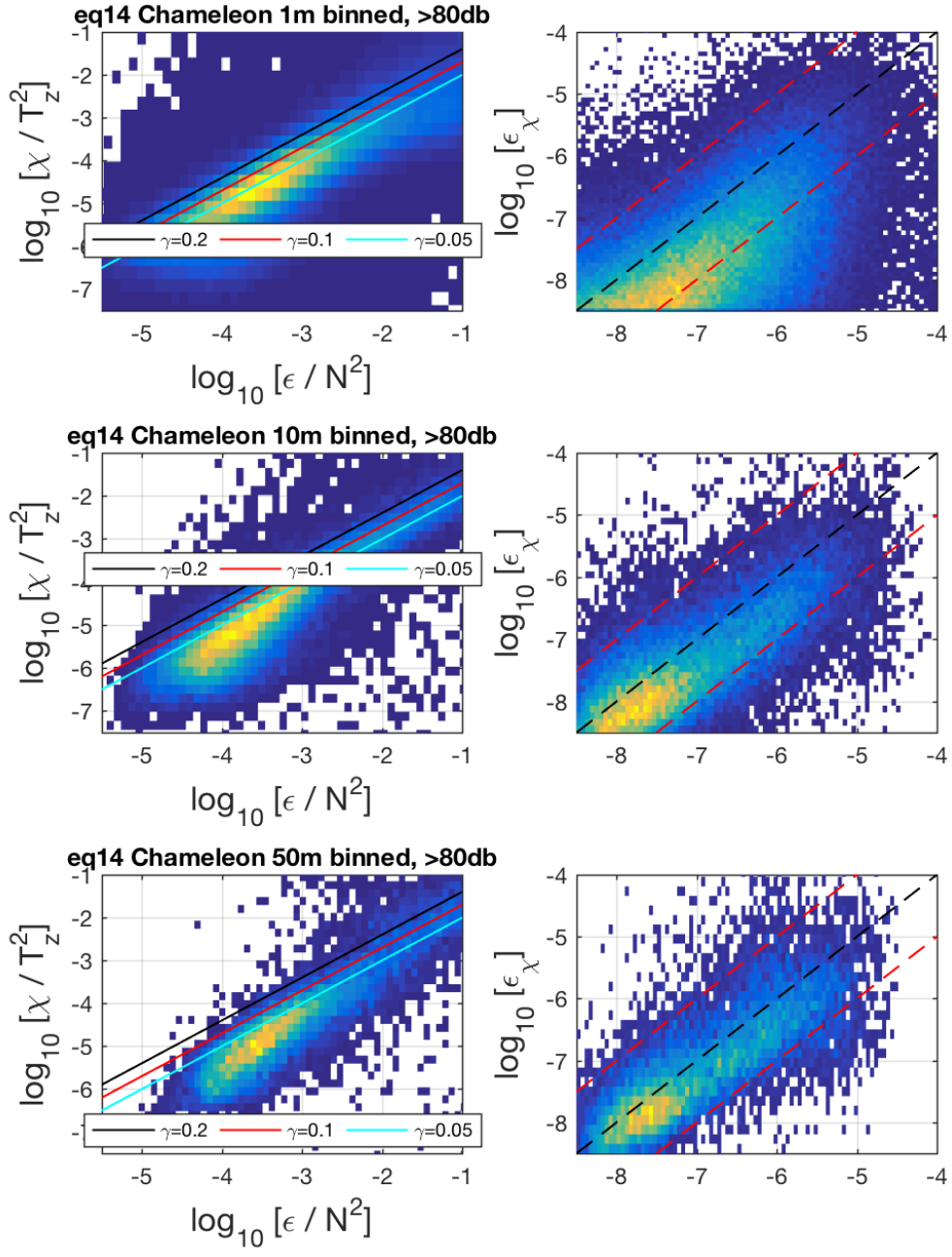


Figure 10: Normalized plots of  $\chi$  vs  $\epsilon$  for different amounts of vertical averaging.



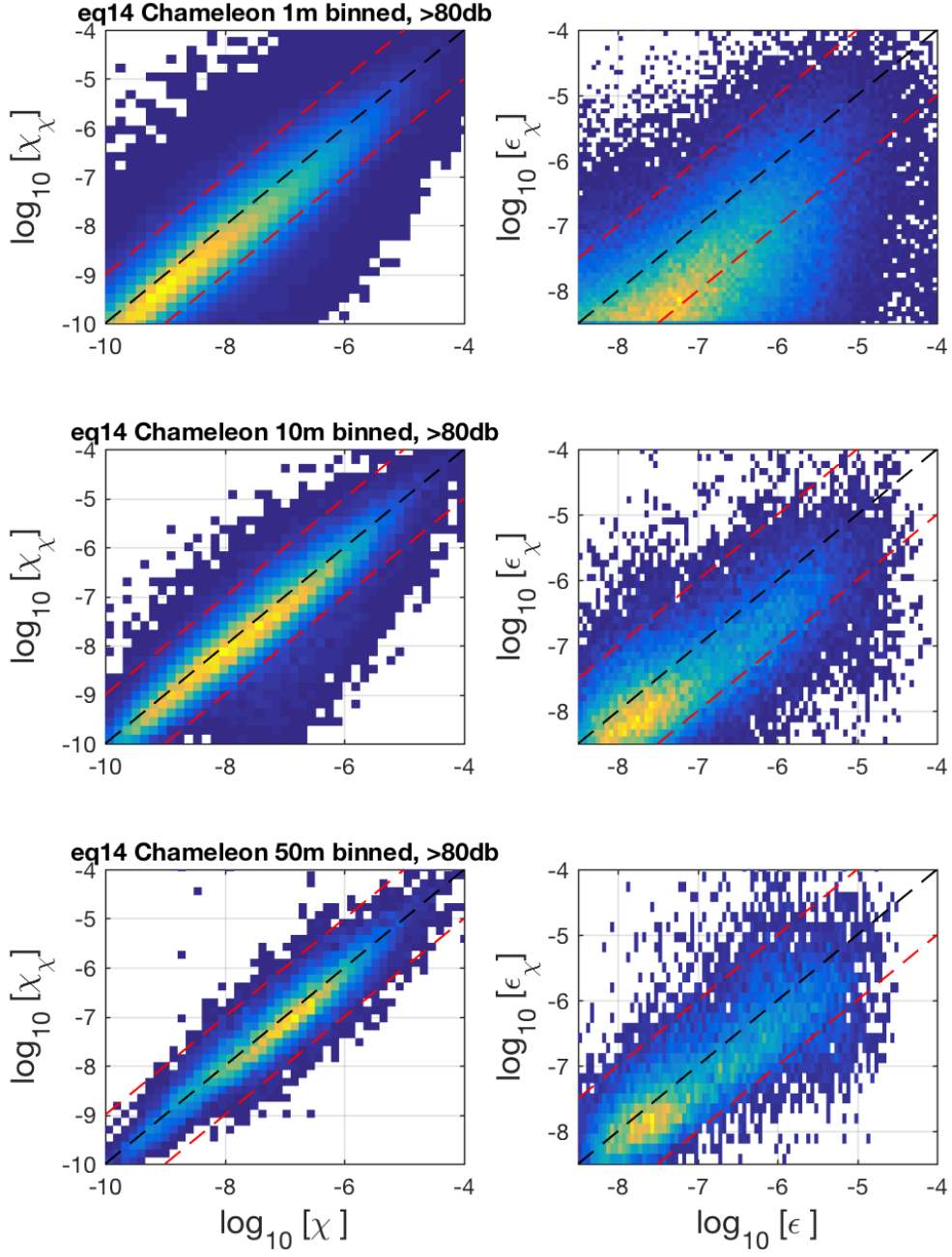


Figure 11: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) averaged over different size depth bins

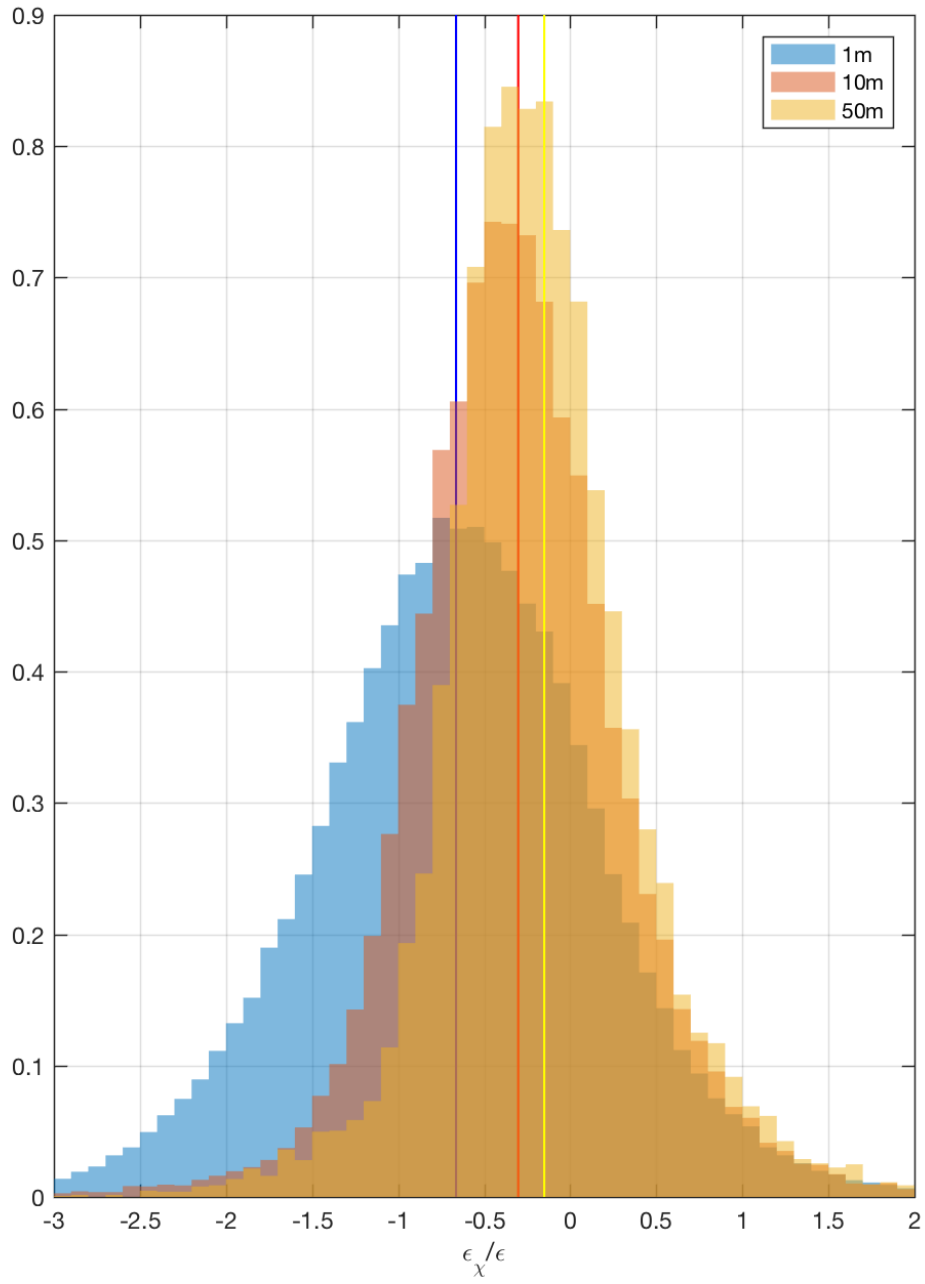


Figure 12: Histogram of  $\log_{10}$  of ratio  $\epsilon_\chi/\epsilon$  for different amounts of vertical averaging. Vertical lines are mean of  $\log_{10}[\epsilon_\chi/\epsilon]$  for each distribution.

### 3.6 $\gamma$ computed from averaged quantities

If we compute gamma from time-averaged  $N^2, T_z, \chi, \epsilon$  do we get  $\gamma = 0.2$  (or a different gamma)? Estimates from the averaged data are larger (Figures ??,13) but still slightly less than 0.2 .

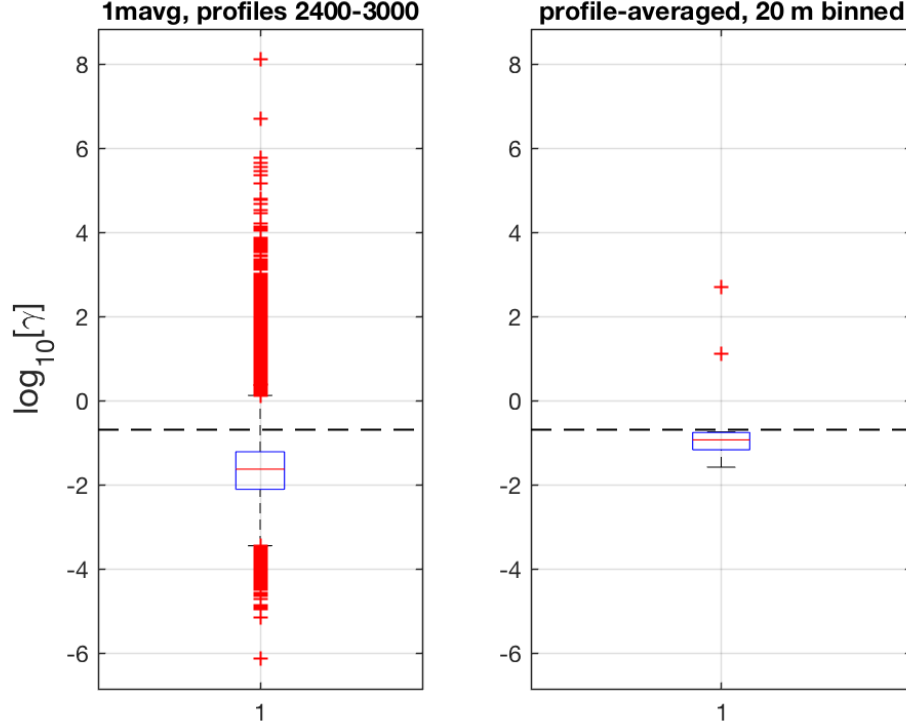


Figure 13: Boxplots of  $\log_{10}[\gamma]$  for a set of profiles from EQ14. Left is for all 1m avg data. Right is for data from all profiles averaged in 10m bins. Horizontal dashed line indicates  $\gamma = 0.2$ .

## 4 Summary

- Individual (and 10m binned)  $\chi$ pod estimates of  $\epsilon_\chi$  are biased low compared to Chameleon  $\epsilon$ .
- This appears to be because  $\gamma$  computed from the Chameleon data is lower than the assumed 0.2
- $\gamma$  computed from averaged (across profiles)  $N^2$ ,  $T_z$ ,  $\chi$ , and  $\epsilon$  is closer to 0.2
- Averaging many  $\epsilon$  profiles reduces the bias (if we use same noise floor for  $\epsilon$  as Chameleon).
- Increased depth-averaging also reduces bias.

Questions:

- Would be good to have 'standard' code to compute  $\chi$  from thermistor data etc.? Thermistor response/noise level varies a lot though, would need a standard way to determine correction.