# Summary of $\chi \mathrm{pod}$ / 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 un-averaged  $\epsilon_{\chi}$  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 reprocessing 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.
- Data including surface convection was identified and excluded in the analysis. The mixed layer depth was identified using a criteria of  $\sigma$ - $_{surface} = 0.04$ . This depth is shown in figures 1 and 2.

- 3 Results
- 3.1 Overview

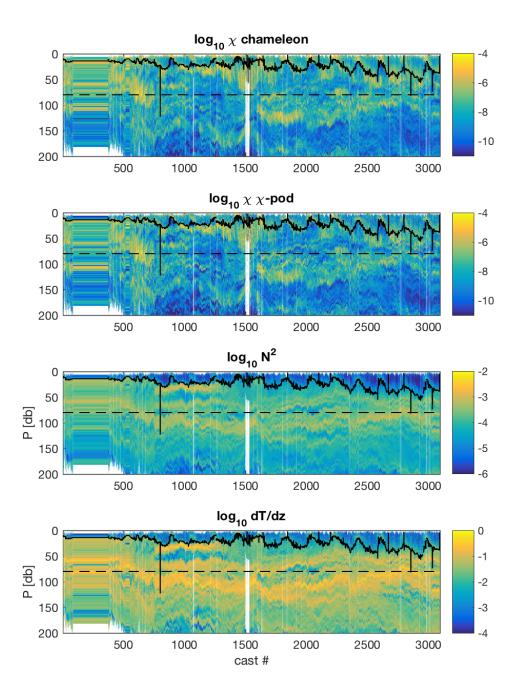


Figure 1: Comparison of  $\chi$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Black line shows shows convective regions excluded in further analysis.

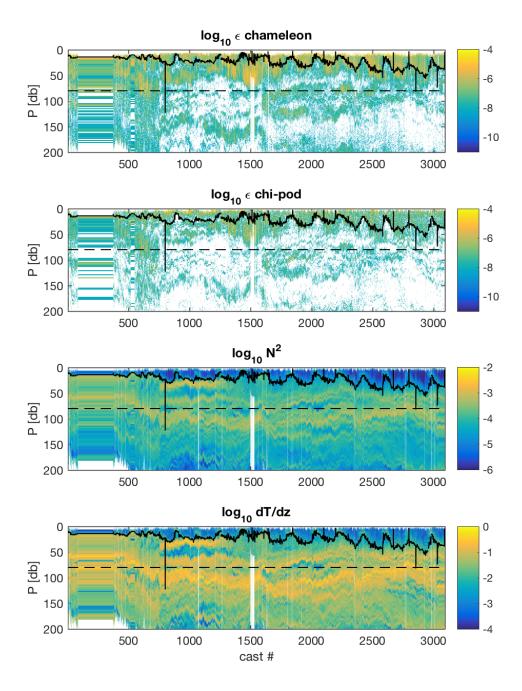
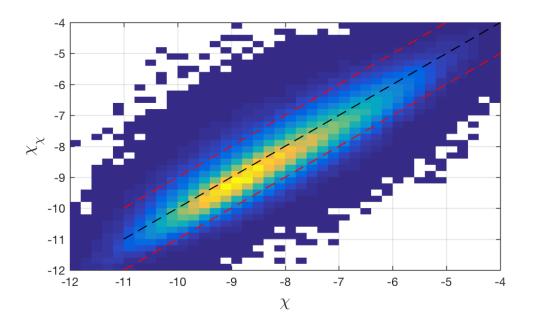


Figure 2: Comparison of  $\epsilon$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Values of  $\epsilon_{\chi}$  and  $\epsilon$  below chameleon noise floor  $(log_{10}[\epsilon] = -8.5)$  have been nan'd out. Black line shows shows convective regions excluded in further analysis.



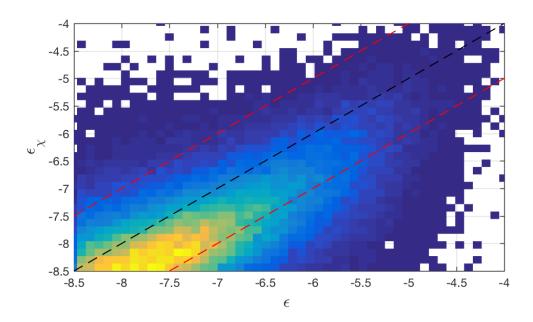


Figure 3: Comparison of  $\chi$  (top) and  $\epsilon$  (lower) from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Values of below chameleon noise floor ( $log_{10}[\epsilon] = -8.5$ ) have been nan'd out. Black line is 1:1, red lines are +/-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

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

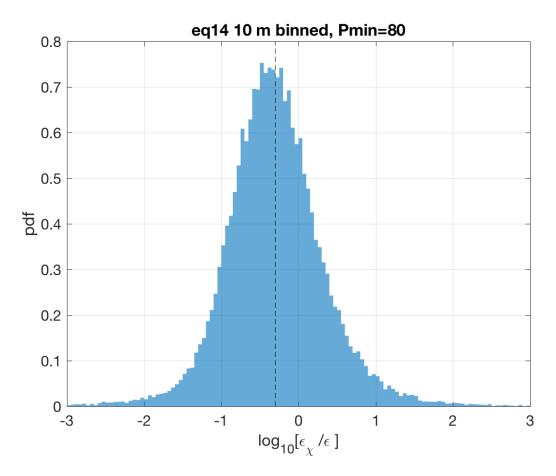


Figure 4: EQ14: Histogram of the ratio of  $\epsilon$  estimates from  $\chi$ pod method to the chameleon values, for  $\chi$ pod method applied to 1m binned profiles, and applied to just patches. Estimates for each profile were averaged in 10m depth bins. Vertical line shows mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$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

#### 3.3 Normalized eps vs chi plots

Assuming that

$$\gamma = \frac{N^2 \chi}{2\epsilon < T_z > 2} \tag{1}$$

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

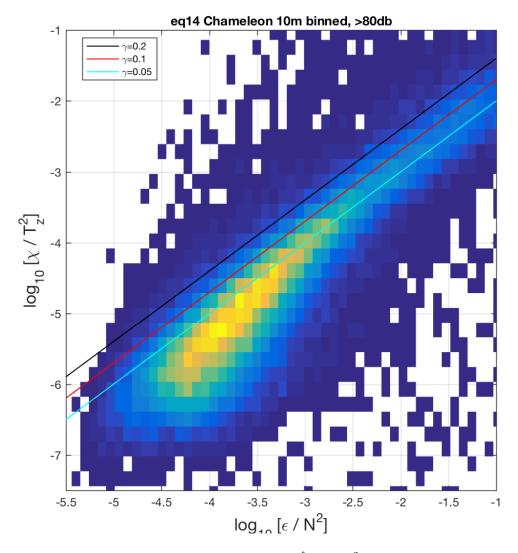


Figure 5: EQ14: 10m binned chameleon  $\epsilon/N\hat{2}$  vs  $\chi/t_z^2$ . Lines show different values of  $\gamma$ . Values of  $\epsilon$  below noise floor ( $log_{10}\epsilon < -8.5$ ) are discarded also.

#### 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\_diffNprof\_profavgs/zsm10m\_fmax7Hz\_respcorr0\_fc\_99hz\_gamma20\_nfft\_128\_screen\_chi\_1.

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.

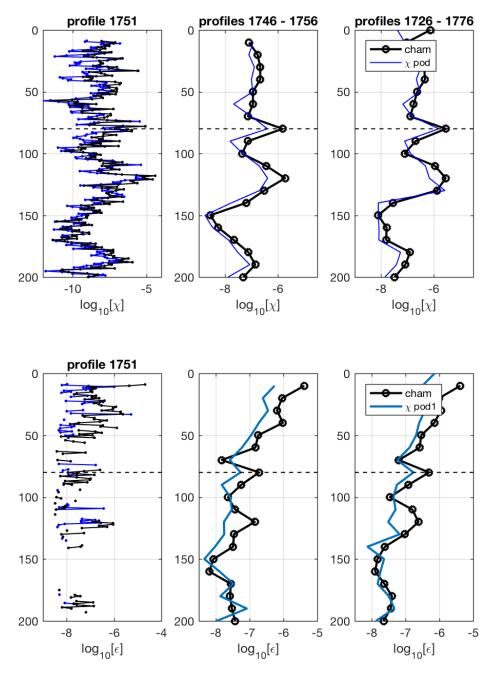


Figure 6: Example of averaging multiple profiles together. Left panels show a single profile from chamleeon and chi-pod method. Right panels show average of +/- 40 profiles, averaged in 10m depth bins.

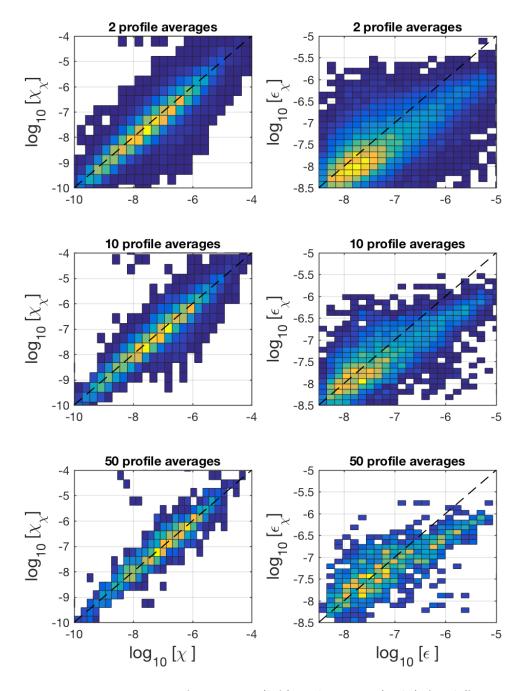
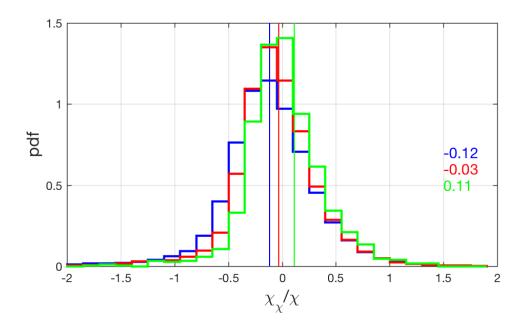


Figure 7: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) for different numbers of profiles averaged.



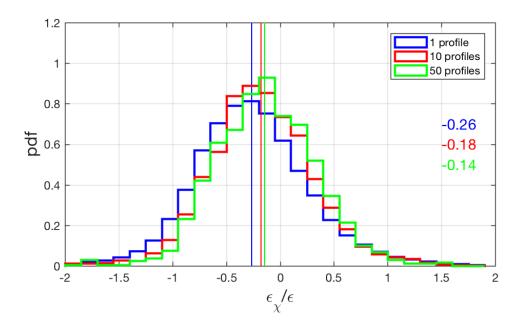


Figure 8: (log10) Ratio of  $\epsilon_{\chi}/\epsilon$  for different numbers of profiles averaged. Consecutive chunks of N profiles were averaged, and then (normalized) histogram of the ratios was plotted. Vertical lines and numbers to right are mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$  for each 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.

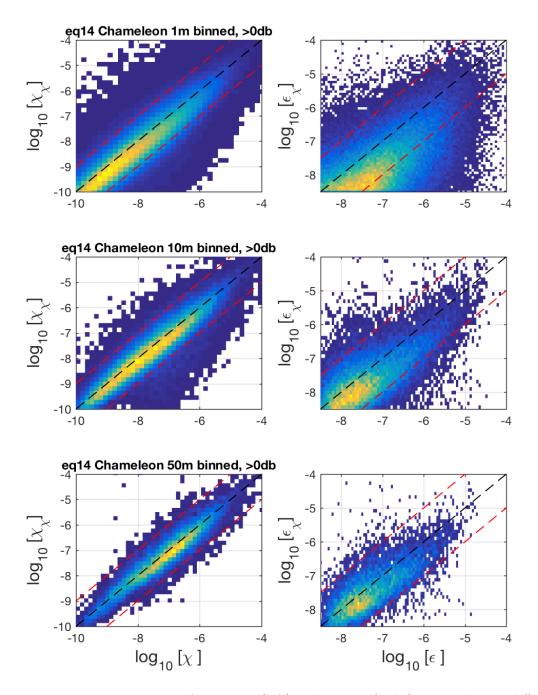
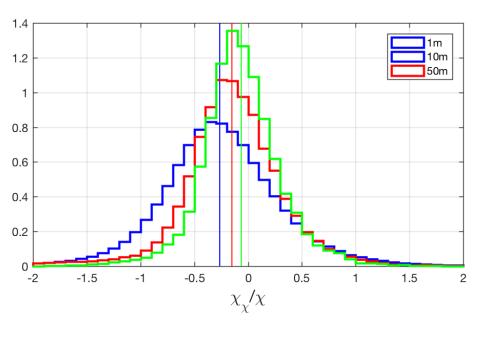


Figure 9: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) averaged over different size depth bins



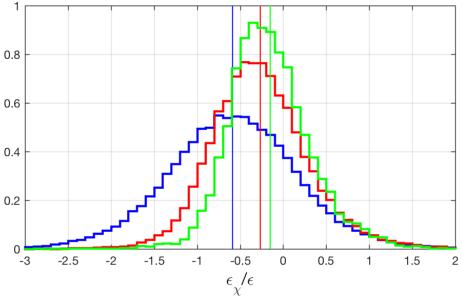


Figure 10: Histogram of log10 of ratio  $\epsilon_{\chi}/\epsilon$  for different amounts of vertical averaging. Vertical lines are mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$  for each distribution.

#### 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 (Figure 11) but still slightly less than 0.2.

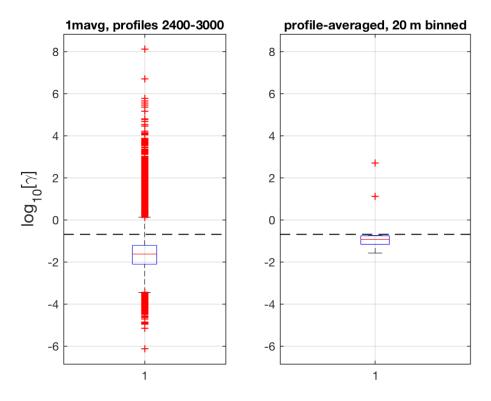


Figure 11: Boxplots of  $log_{10}[\gamma]$  for a set of profiles from EQ14. Left is for all 1m avg data. Right is for data from all profiles averaged in 10m bins. Horizontal dashed line indicates  $\gamma = 0.2$ .

## 4 Summary

- Inidivudal (and 10m binned)  $\chi$ pod estimates of  $\epsilon_{\chi}$  are biased low compared to Chameleon  $\epsilon$ .
- $\bullet$  This appears to be because  $\gamma$  computed from the Chameleon data is lower than the assumed 0.2
- $\gamma$  computed from averaged data (across profiles)  $N^2,\,T_z,\,\chi,$  and  $\epsilon$  is closer to 0.2
- Averaging many  $\epsilon$  profiles or averaging depth reduces the bias