# 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  $(\chi, \epsilon)$ .

#### 2 Data and Processing

- Data are from Chameleon profiles near the equator during the 'EQ14' experiment.
  On my laptop, they are located in the folder: /Cham\_Eq14\_Compare/Data/chameleon/processed/
- 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 normally-assumed 32Hz. These data are located at: /Cham\_Eq14\_Compare/Data/chameleon/processed\_AP\_7hz/
- The χpod method is applied to Chameleon profiles (thermistor only, not shear probe) from EQ14 in ComputeChi\_Chameleon\_Eq14.m
- 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 (Identify\_mixedlayer\_eq14.m) using a criteria of  $\sigma$ -surface = 0.04. This depth is shown in figures 1 and 2.
- Make\_Overview\_Plots.m Makes the figures in this document.

# 3 Results

### 3.1 Overview

• Both  $\chi_{\chi}$  and  $\epsilon_{\chi}$  appear to capture the depth and time structure of  $\chi$  and  $\epsilon$  well (Figures 1,2).

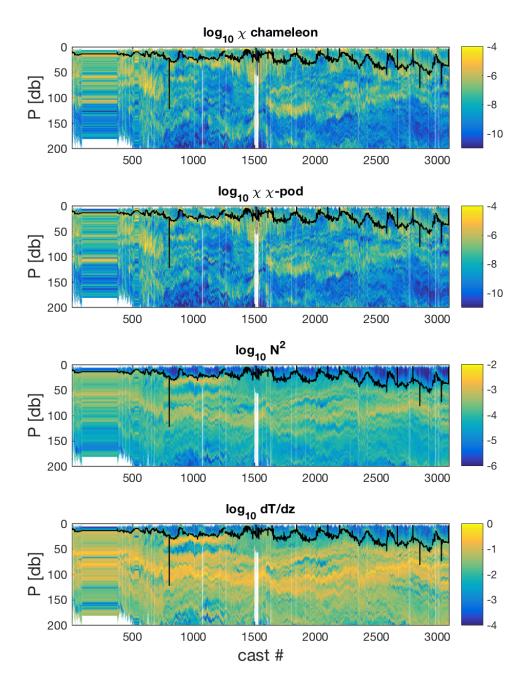


Figure 1: Comparison of  $\chi$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Black line shows shows convective regions excluded in further analysis.

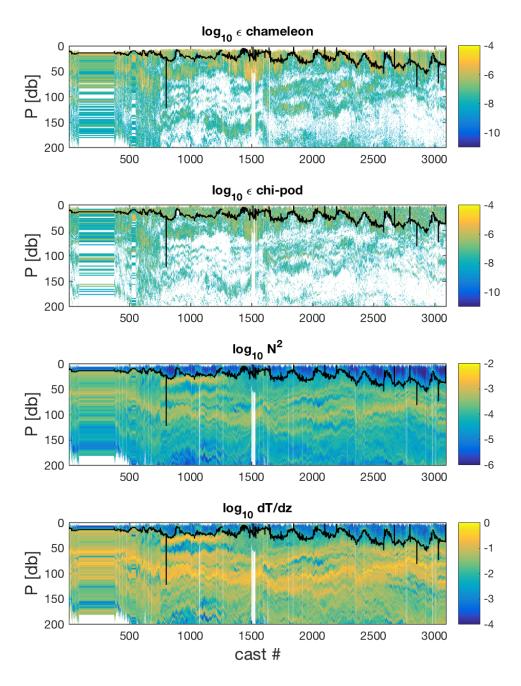
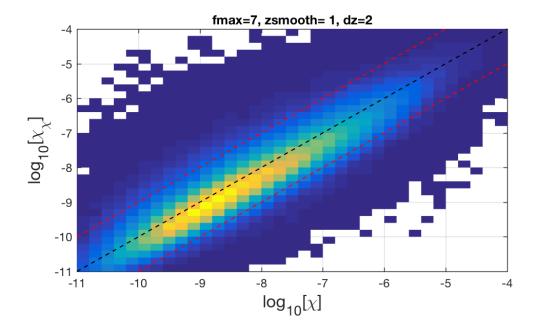


Figure 2: Comparison of  $\epsilon$  from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Values of  $\epsilon_{\chi}$  and  $\epsilon$  below chameleon noise floor  $(log_{10}[\epsilon] = -8.5)$  have been nan'd out. Black line shows shows convective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

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

- Both  $\chi_{\chi}$  and  $\epsilon_{\chi}$  are biased low (Figures 3,4); the  $\epsilon_{\chi}$  bias is larger (more negative).
- The bias in  $\chi$  is relatively constant with depth; the bias in  $\epsilon$  is more negative at shallower depths (Figure 5).



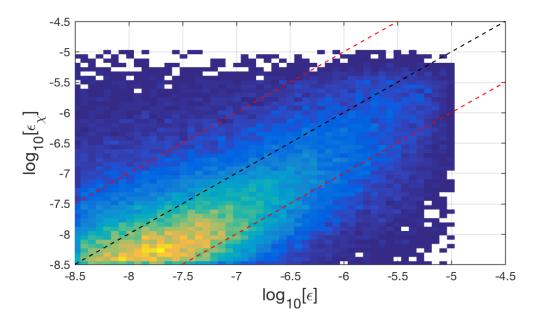


Figure 3: Comparison of  $\chi$  (top) and  $\epsilon$  (lower) from chameleon method and chi-pod method, for EQ14 chameleon profiles. Each profile was averaged in 2m bins. Values of below chameleon noise floor ( $log_{10}[\epsilon] = -8.5$ ) have been nan'd out. Black line is 1:1, red lines are +/-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

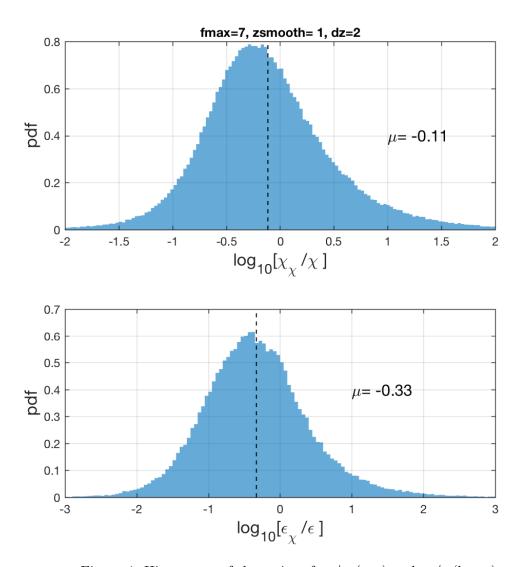
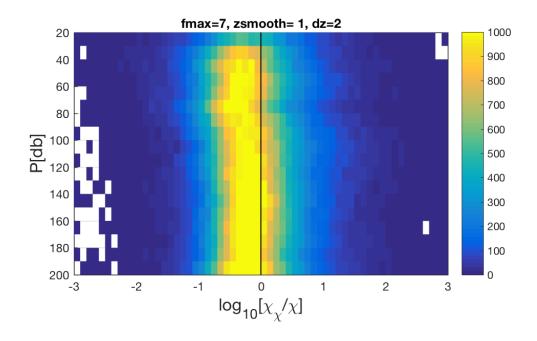


Figure 4: Histograms of the ratios of  $\chi_\epsilon/\chi$  (top) and  $\epsilon_\chi/\epsilon$  (lower) .



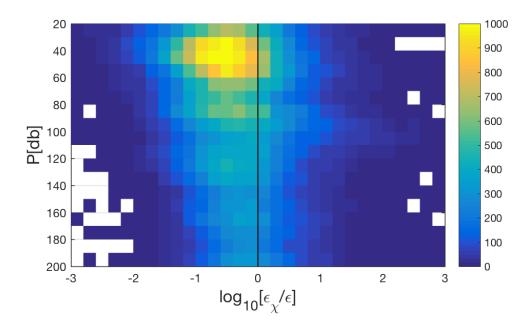
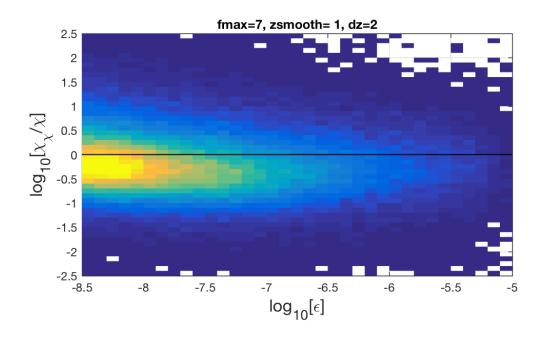


Figure 5: 2D histograms of ratios  $\chi_{\chi}$  and  $\epsilon_{\chi}$  ratios vs depth.

3.3 Dependence of bias on actual  $\epsilon$ 



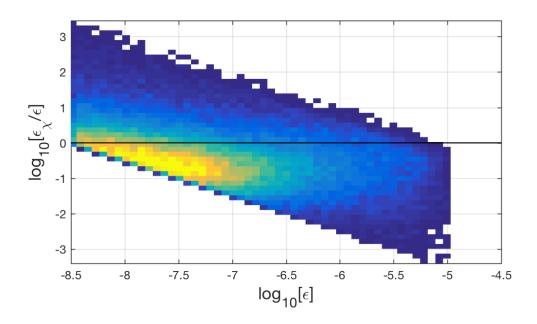


Figure 6: 2D histograms of ratios  $\chi_{\chi}/\chi$  and  $\epsilon_{\chi}/\epsilon$  ratios vs  $\epsilon$ .

#### 3.4 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  $2\gamma$ . Chameleon data from EQ14 tend to fall near  $\gamma=0.05$  (Figure 6).

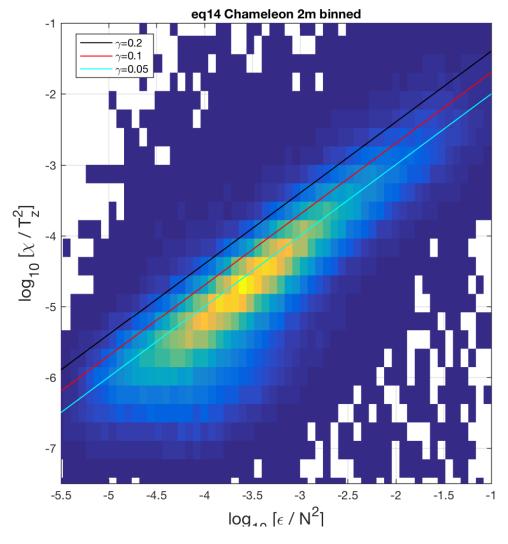


Figure 7: EQ14: 2m binned chameleon  $\epsilon/N\hat{2}$  vs  $\chi/t_z^2$ . Lines show different values of  $\gamma$ . Values of  $\epsilon$  below noise floor ( $log_{10}\epsilon < -8.5$ ) are discarded also.

### 3.5 Averaging multiple profiles of $\epsilon$

- Averaging over multiple profiles reduces the bias in both  $\chi$  and  $\epsilon$  (Figures 7,8).
- Averaging 10 profiles together seems to give the smallest bias.

Figure 9 shows one example.

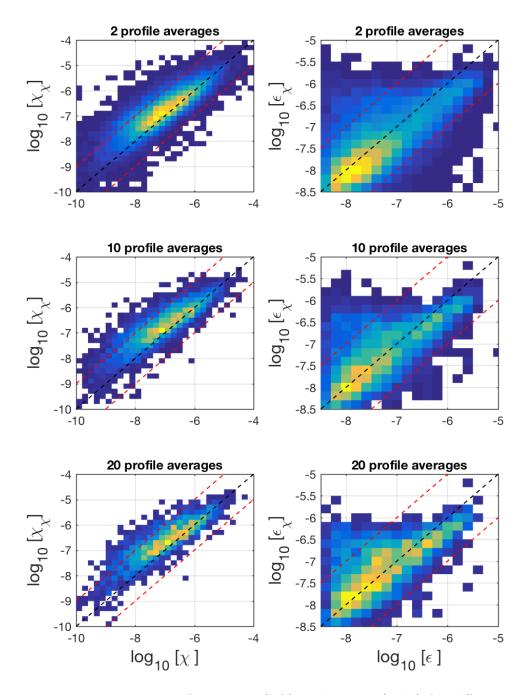


Figure 8: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) for different numbers of profiles averaged.

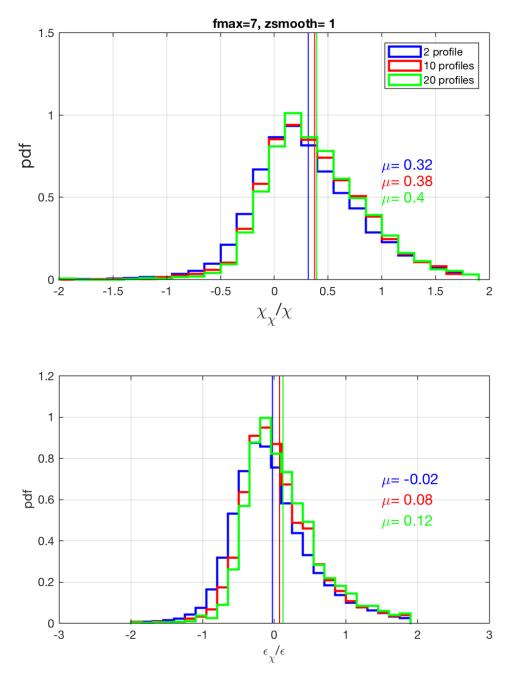


Figure 9: (log10) Ratio of  $\epsilon_{\chi}/\epsilon$  for different numbers of profiles averaged. Consecutive chunks of N profiles were averaged, and then (normalized) histogram of the ratios was plotted. Vertical lines and numbers to right are mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$  for each distribution.

eps\_boot\_prof\_cnum\_1576.png

Figure 10: Example of averaging multiple profiles together. Left panels show a single profile from chamleeon and chi-pod method. Right panel shows bootstrap average of 5 profiles, averaged in 10m depth bins, with 95% confidence intervals. Data in mixed layer and shallower than 20m have been excluded.

## 3.6 Averaging over different-sized depth bins

• Averaging over large depth bins reduces the bias in both  $\chi$  and  $\epsilon$  (Figures 10,11).

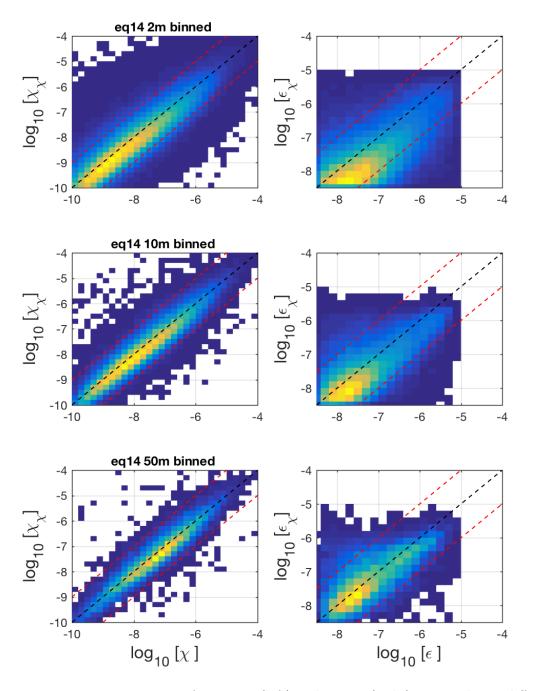


Figure 11: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) averaged over different size depth bins

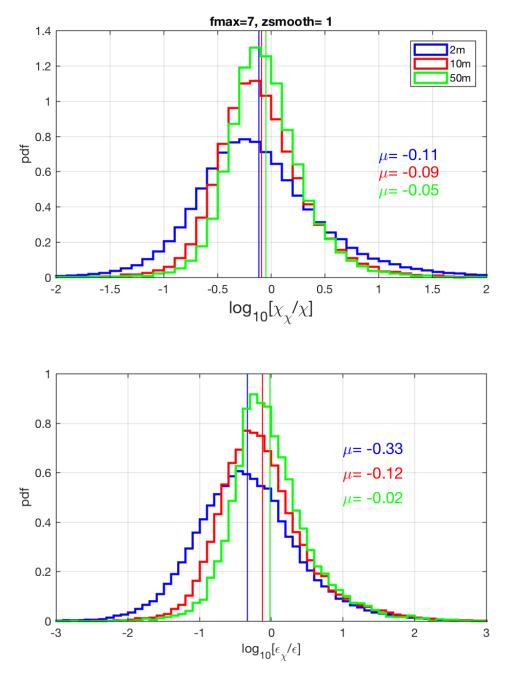


Figure 12: Histogram of log10 of ratio  $\epsilon_{\chi}/\epsilon$  for different amounts of vertical averaging. Vertical lines are mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$  for each distribution.

#### 3.7 $\gamma$ 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 12) but still slightly less than 0.2.

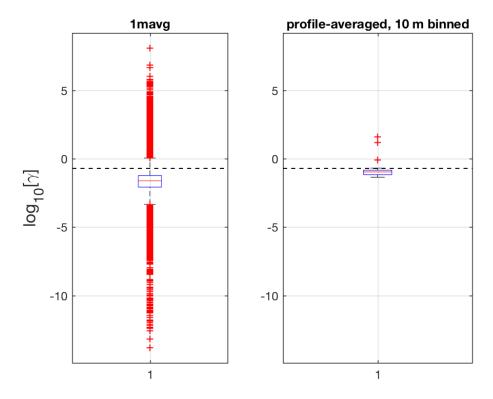


Figure 13: Boxplots of  $log_{10}[\gamma]$  for a set of profiles from EQ14. Left is for all 1m avg data. Right is for data from all profiles averaged in 10m bins. Horizontal dashed line indicates  $\gamma = 0.2$ .