# Summary of $\chi \mathrm{pod}$ / Chameleon EQ08 Analysis

## Andy Pickering

## $\mathrm{June}\ 13,\ 2017$

# Contents

1	Ove	rview	2
2	Data	a and Processing	3
3 Resul		lts	
		Overview	
		$\epsilon_{\chi} \text{ vs } \epsilon$	
	3.3	Comparing individual estimates of $\epsilon$	8
	3.4	Dependence of bias on actual $\epsilon$	12
	3.5	Normalized eps vs chi plots	14
	3.6	Averaging multiple profiles	15
	3.7	Averaging over different-sized depth bins	18

#### 1 Overview

- This document is an attempt to provide an overview/summary of some analysis i've done with the EQ08 data. 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 'EQ08' experiment.
- γpod method is applied to thermistor data from Chameleon profiles in: ComputeChi\_Chameleon\_Eq08.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  $log_{10}[\epsilon] = -5$  was also applied.
- I re-processed the Chameleon data (run\_eq08\_AP.m) using a fmax of 15hz for the  $\chi$  calculations, based on looking at spectra and where they rolled off.
- 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. These depths were found w/ Identify\_ML\_eq08.m
- The figures in this document are made w/ Make\_Overview\_Plots\_eq08.m.

- 3 Results
- 3.1 Overview

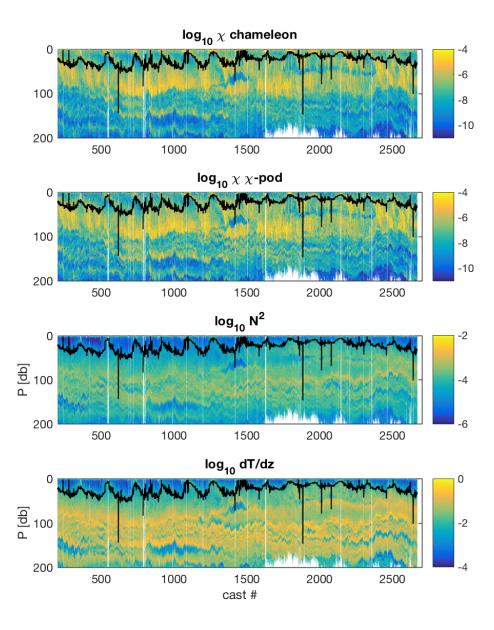


Figure 1: Comparison of  $\chi$  from chameleon method and chi-pod method, for eq08 chameleon profiles. Date from each profile were averaged in 2m bins. Black line shows shows convective regions excluded in further analysis.

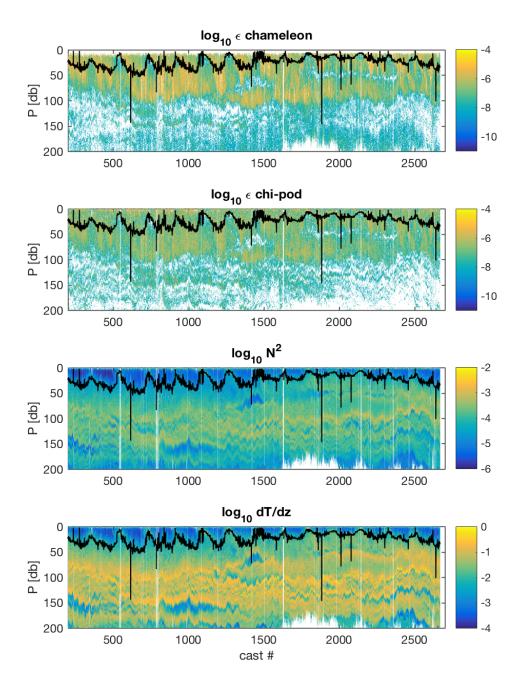


Figure 2: Comparison of  $\epsilon$  from chameleon method and chi-pod method, for eq08 chameleon profiles. Each profile was averaged in 2m bins. Values of  $\epsilon_{\chi}$  and  $\epsilon$  below chameleon noise floor ( $log_{10}[\epsilon] = -8.5$ ) have been nan'd out. Black line shows shows convective regions excluded in further analysis.

## 3.2 $\epsilon_{\chi}$ vs $\epsilon$

I first compared  $\epsilon_{\chi}$  vs  $\epsilon_{\chi}$  vs  $\epsilon$  for the Chameleon data, where  $\epsilon_{\chi} = N^2 \chi/2/\gamma/T_z^2$ , using  $\gamma = 0.2$ .  $\epsilon_{\chi}$  underestimates  $\epsilon$ , showing that that relationship (assumed in  $\chi$ pod processing does not hold here.

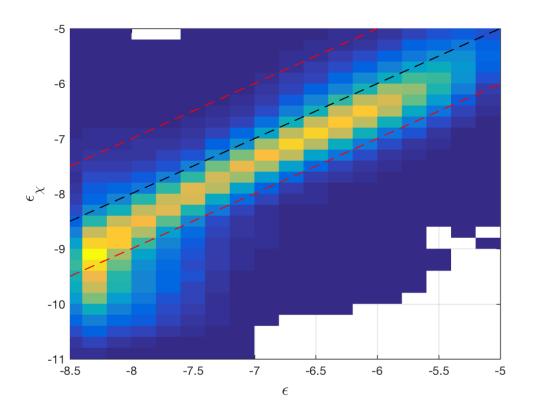
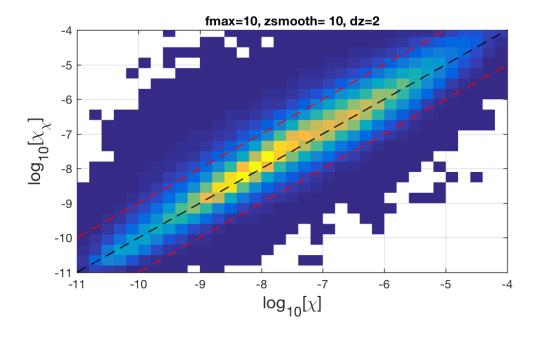


Figure 3: 2D histogram of  $\epsilon_\chi$  vs  $\epsilon$  for Chameleon data.

## 3.3 Comparing individual estimates of $\epsilon$

- The  $\chi$ pod method tends to slightly over-estimate  $\chi$ , and underestimate  $\epsilon$  (Figures 4,5).  $\epsilon$  seems to be more under-estimated at higher values of  $\epsilon$ .
- The bias in tends to be larger (more negative) at shallower depths (Figure 6).



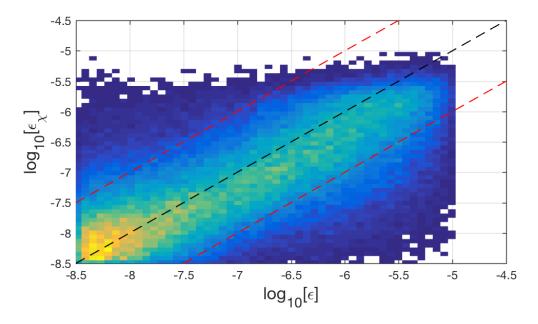


Figure 4: Comparison of  $\chi$  (top) and  $\epsilon$  (lower) from chameleon method and chi-pod method, for eq08 chameleon profiles. Each profile was averaged in 2m bins. Values of below chameleon noise floor ( $log_{10}[\epsilon] = -8.5$ ) have been nan'd out. Black line is 1:1, red lines are +/- order of magnitude.

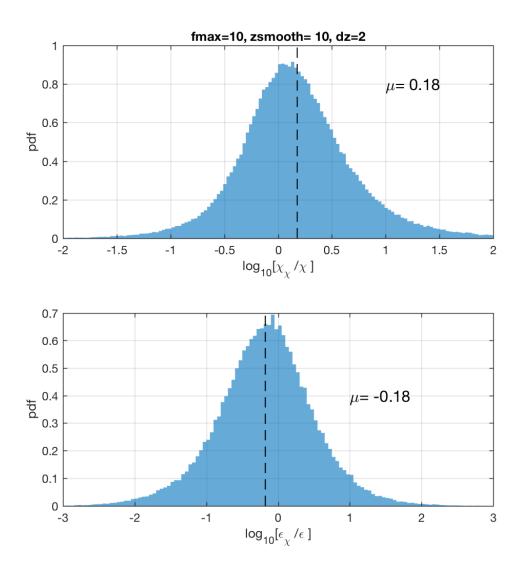
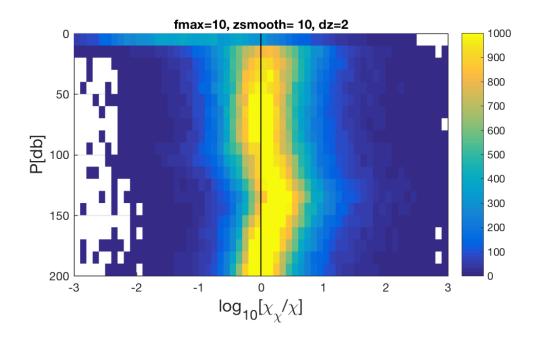


Figure 5: eq08: Histogram of the ratio of  $\epsilon$  estimates from  $\chi$ pod method to the chameleon values. Estimates for each profile were averaged in 10m depth bins. Vertical line shows mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$ .



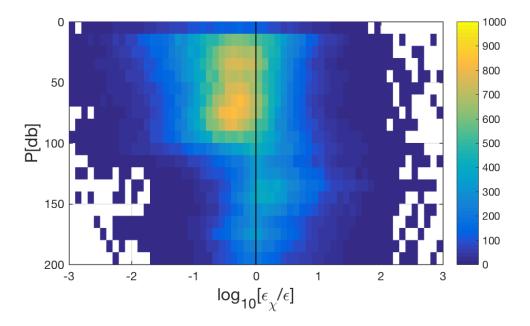
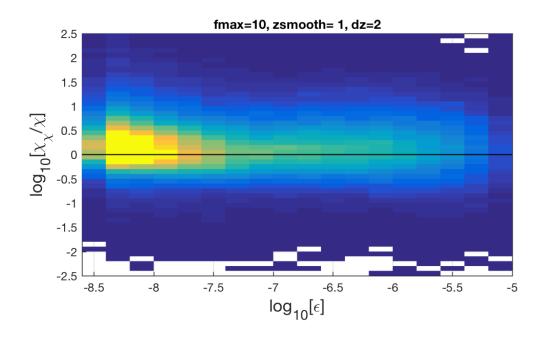


Figure 6: 2D histograms of ratios  $\chi_{\chi}/\chi$  and  $\epsilon_{\chi}/\epsilon$  ratios vs depth.

3.4 Dependence of bias on actual  $\epsilon$ 



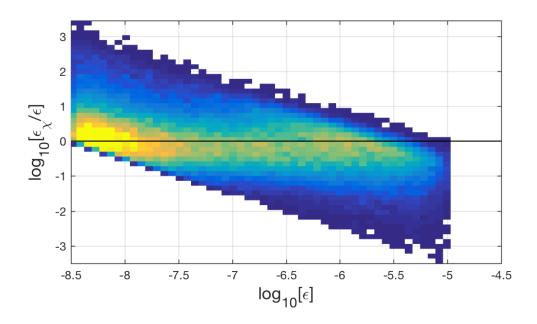


Figure 7: 2D histograms of ratios  $\chi_{\chi}/\chi$  and  $\epsilon_{\chi}/\epsilon$  ratios vs  $\epsilon$ .

#### 3.5 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  $2\gamma$ . The Chameleon data from EQ08 tend to fall near  $\gamma=0.1$  or slightly lower (Figure 8).

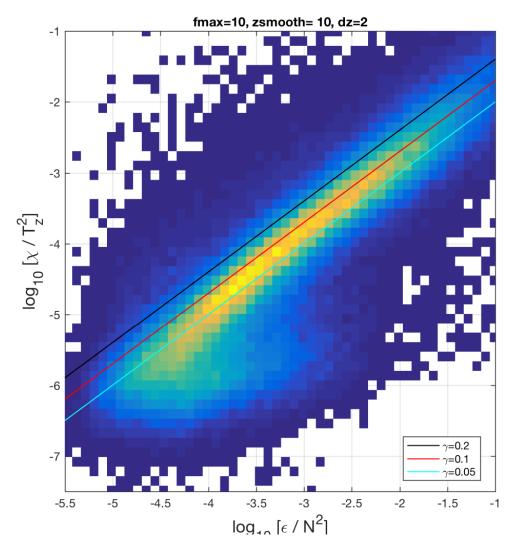


Figure 8: eq08: 10m binned chameleon  $\epsilon/N\hat{2}$  vs  $\chi/t_z^2$ . Lines show different values of  $\gamma$ . Values of  $\epsilon$  below noise floor  $(log_{10}\epsilon < -8.5)$  are discarded also.

3.6 Averaging multiple profiles

•

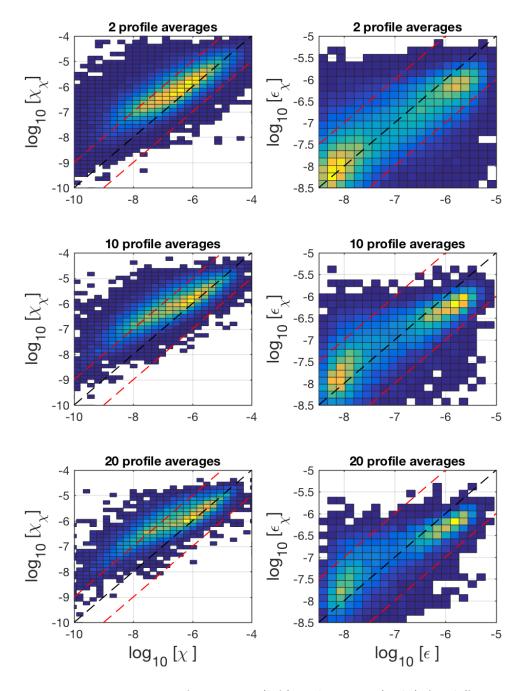
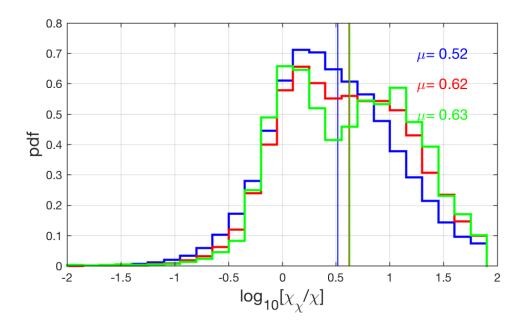


Figure 9: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) for different numbers of profiles averaged.



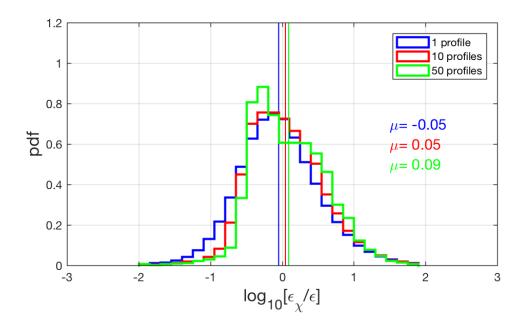


Figure 10: (log10) Ratio of  $\epsilon_{\chi}/\epsilon$  for different numbers of profiles averaged. Consecutive chunks of N profiles were averaged, and then (normalized) histogram of the ratios was plotted. Vertical lines and numbers to right are mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$  for each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

#### 3.7 Averaging over different-sized depth bins

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

• The bias in  $\epsilon$  is decreased with averaging over larger depth intervals, although the bias in  $\chi$  increases slightly (Figure 12).

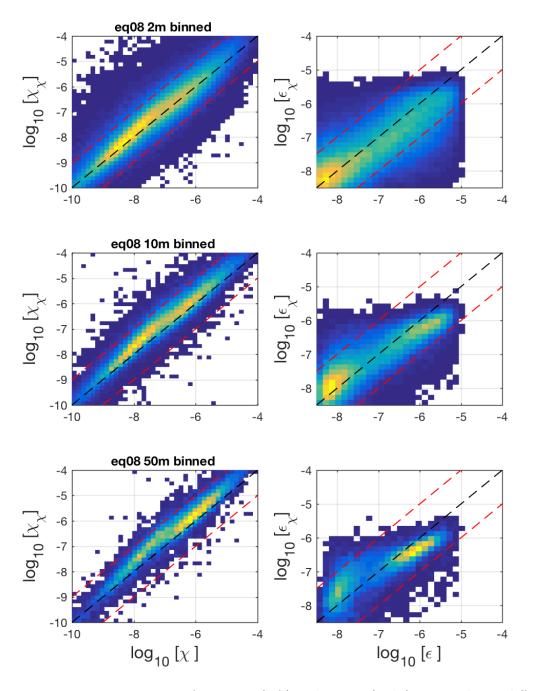


Figure 11: 2D Histograms of  $\chi_{chi}$  vs  $\chi$  (left) and  $\epsilon_{\chi}$  vs  $\epsilon$  (right) averaged over different size depth bins

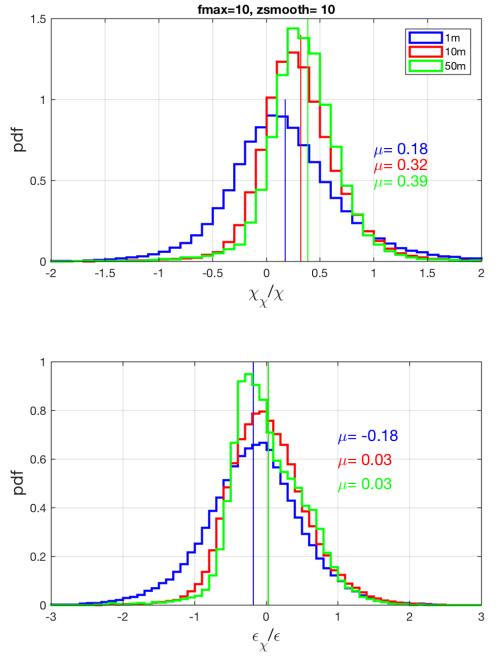


Figure 12: Histogram of log10 of ratio  $\epsilon_{\chi}/\epsilon$  for different amounts of vertical averaging. Vertical lines are mean of  $log_{10}[\epsilon_{\chi}/\epsilon]$  for each distribution.