A modified peak-bagging technique for fitting low-$\ell$ solar p-modes.

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We introduce a modified version of a standard power spectrum ‘peak-bagging’ technique which is designed to gain some of the advantages that fitting the entire low-degree p-mode power spectrum simultaneously would bring, but without the problems involved in fitting a model incorporating many hundreds of parameters. Employing Monte-Carlo simulations we show that by using this modified fitting code it is possible to determine the true background level in the vicinity of the p-mode peaks. In addition to this we show how small biases in other mode parameters, which are related to inaccurate estimates of the true background, are also consequently removed.

1 Introduction

Determining the various parameter values of the resonant modes of oscillation of the Sun is an important process in the field of helioseismology. Parameters such as the mode frequencies, lifetimes and amplitudes can all be used to determine the conditions of the solar interior. Over the years the quality of helioseismic data has improved significantly due to the length of data sets increasing, signal-to-noise ratios being improved and more continuous observations being made, both from ground-based and space-borne missions. This has led to the estimated parameter values being constrained to increasingly greater precision.

As the precision of the parameter estimates increases so more subtle physical characteristics of the solar interior are being uncovered. Examples of this include the discovery of asymmetric peaks in the p-mode power spectrum (e.g., Duvall et al. 1993, Chaplin et al. 1999), which supported theoretical predictions (e.g., Gabriel 1992, 1995) and gave evidence to the belief that acoustic waves are generated within a well-localized region of the solar interior. Also, long sets of observations have allowed investigations to be carried out on the dependence of the mode parameters on solar activity (e.g., for asymmetry dependence see Chaplin et al. 2007).

Determining subtle effects such as these requires analysis techniques that return robust estimates of the mode parameters. If this is not the case and the returned parameters are biased, the inaccuracies can be confused with mode characteristics that have true physical significance.

Methods of determining solar mode parameters are often referred to as peak-bagging techniques. For low-degree (low-$\ell$) Sun-as-a-star observations, a common method of peak-bagging involves dividing the p-mode power spectrum into a series of ‘fitting windows’ centered on the $\ell = 0/2$ and $1/3$ pairs. The modes are then fitted, pair by pair, to determine how the mode parameters depend on both frequency (overtone number) and angular degree without the need to fit the entire spectrum simultaneously.

In this paper we use a Monte-Carlo type approach to estimate the extent of biases seen in the fitted parameters when using this ‘standard’ method of fitting, and introduce a new modified technique designed to reduce these problems.

2 Fitting Techniques

In this section the peak-bagging technique described in the introduction is explained in more detail. Reasons why the standard fitting method may return biased parameter estimates are discussed and the new modified fitting method, designed to limit these problems, is introduced.

We begin by summarizing briefly the main elements of the standard peak-bagging method. Within each fitting window the mode peaks are modeled using a modified Lorentzian equation (Nigam & Kosovichev 1998). This is fitted to the data using an appropriate maximum-likelihood estimator (Anderson et al. 1990). As we elaborate in Section 3 only simulated low-$\ell$ data has been used in this initial analysis. All modes within the frequency range $1500 \leq \nu \leq 4600$ $\mu$Hz and with angular degree $0 \leq \ell \leq 3$ were fitted. In Fletcher (2007) it was shown that if the weak $\ell = 4$ and $5$ modes are not accounted for, they can often impact on the fitted parameters of the stronger modes. This was also commented upon in Jiménez-Reyes et al. (2007) and investigated in Jiménez-Reyes et al. (2008). Therefore, in the regions of the spectrum where the modes are strongest, our fitting model was modified to also
include parameters for the $\ell = 4$ and 5 modes. This adaption was only made to the model when fitting the $\ell = 0/2$ pairs, as $\ell = 4$ and 5 modes never lie within the fitting windows of $\ell = 1/3$ pairs.

Within each fitting window the following parameters were varied iteratively until they converged on their best-fitting values:

1. A central frequency for each mode.
2. A parameter describing the symmetric rotational splitting pattern for each mode (not applicable at $\ell = 0$).
3. A linewidth for each mode, for which the logarithm was varied (the $m$ components in a mode were assumed to have the same widths).
4. A single maximum height – that of the outer, sectoral $m$ components – for each mode, for which the logarithm was varied (the relative $m$ component height ratios were assumed to take fixed values).
5. A single peak asymmetry parameter (the $m$ components of the modes in the pair were assumed to have the same asymmetry).
6. A flat, background offset for the fit, whose logarithm was varied.

It should be noted that although the simulated mode peaks were symmetric in frequency, a parameter characterising asymmetry was still used so as to be consistent with fits to real data.

For the most part the parameter values returned via this method are very accurate, especially the mode frequencies. However, there is a problem associated with restricting the fitting to only a small slice of the spectrum. For the model to be completely accurate there would have to be no power within the fitting window from modes lying outside it, which of course, is not true. The main problem associated with this is that the background parameter will be overestimated in order to account for the extra power from the wings of the outlying modes. The effect has previously been documented in the literature (Chaplin et al. 2003; Fletcher 2007) and is illustrated again in this paper in Section 4.

For fitting purposes the background is usually assumed to be flat in frequency across the extent of the fitting window. In the case of the true background this is a fairly safe assumption as the sizes of the fitting windows tend to be quite small, and over the region of the spectrum where $p$ modes are observed, the background is thought to be a relatively weak function of frequency. However, the presence of the wings from the surrounding modes means the effective background (i.e., background plus wings) will have a stronger dependence on frequency. Specifically, the effective background is observed to rise towards the extreme ends of the fitting window as these areas are nearer, in frequency, to the surrounding peaks. The combined effect of incorrectly fitting the true background and using a model that does not correctly account for the shape of the effective background, means other parameters may be impacted upon during the fitting process to minimise these inaccuracies.

This effect is shown in Fig. 1, where a simple single Lorentzian mode peak has been fitted. Either side of this mode, outside of the fitting window, there are two large peaks the effect of whose wings can be seen within the fitting window. (This is not a realistic scenario as the outlying peaks have been substantially increased in height in order to exaggerate bias in the fits). When a model is used which does not account directly for the wings of outlying modes, the fitted background is increased to compensate for the extra power in order to minimise the likelihood function. However, because the additional power from the outlying modes is not flat across the fitting window the likelihood function can be further minimised by changes in the linewidth and height of the mode. In this case the background is reduced and the height increased. This can be seen more clearly in the right-hand plot of Fig. 1 where the backgrounds have been removed allowing the linewidth and height of the fitted peak to be more easily compared with those of the true peak. We shall see in section 4 that similar biases are seen when fitting more complicated, more realistic simulated data.

An obvious way of fixing these problems is to fit the entire power spectrum simultaneously, in which case the wings of all the modes will be included in the model and the true background will be fitted. Unfortunately, doing this requires a model with many hundreds of parameters. As such attempts to fit entire power spectra would involve large computing times and will be susceptible to premature convergence. Even so, this technique has been employed relatively successfully when the time series used were of fairly short duration (i.e., a few hundred days) (e.g., Roca Cortés et al. 1998; Jiménez et al. 2002).

There are at least three possible methods that retain the benefits of full-spectrum fitting without the need to use a complex model. One method is to fit the full spectrum, but to describe the parameters as smooth functions of frequency thus reducing the number of parameters to be fitted (e.g., Jefferies & Vorontsov 2004). A second method is to fit the entire spectrum using a multi-step iterative process whereby groups of parameters (such as frequencies, linewidths etc.) are fitted separately. This method has been pioneered by Gelly (2002). Finally, a third approach is to employ an amalgamation of full-spectrum fitting and the standard ‘pair-by-pair’ fitting. In this scenario one would initially use parameter fits from the standard technique to create a first-guess model for the entire spectrum. This model can then be used to fit just a small slice of the spectrum in the same manner as the pair-by-pair fitting, allowing only the parameters associated with the target pair to vary, leaving the remaining model parameters associated with all the other modes fixed. The advantage of this approach is that, even though peaks of surrounding modes are not present in the fitting range, the effects of their wings will still be modeled.

All of these techniques are worthy of further study, however, in this paper it is the final of these three approaches that we investigate. A simple algorithm for this modified fitting technique is:
Fig. 1  Plot showing fits to a single mode. Large peaks situated outside the fitting window are not accounted for by the model. In the left hand plot the dashed line shows the spectrum to be fitted whereas the solid line gives the fitted result. The dotted line shows the true peak characteristics without the effects of the outlying modes. In the right hand plot the solid line shows the fitted peak minus the fitted background whereas the dashed line shows the true peak minus the true background. This plot highlights more clearly the bias in the linewidth and height.

1. Create a model for the full spectrum using the parameters determined from the standard fitting code.  
2. Perform a fit across a single fitting window using the full spectrum model, but only allow parameters associated with the peaks centered in the window to vary.  
3. Repeat the process for all fitting windows.

It should be noted that when forming the full spectrum model from the parameter estimates determined using the traditional fitting method, the asymmetry term is reset to zero. This is because the simplified formalisation of the Nigam & Kosovichev (1998) expression used for our fitting model is only valid within the vicinity of the peaks. The asymmetry is allowed to vary again during the second round of fitting.

As in the first set of fits all modes within the frequency range $1500 \leq \nu \leq 4600 \mu Hz$ and angular degree $0 \leq \ell \leq 3$ (along with detectable $\ell = 4$ and 5 modes) were fitted. However only modes with frequencies up to $4000 \mu Hz$ were analysed as modes with higher frequencies will begin to experience contamination from outlying modes that were not fitted using the pair-by-pair technique. The same parameters that were varied in the standard fitting techniques were also varied in the modified version. However, in order to reduce the complexity of the code it was decided to increase the size of the fitting window to include both $\ell =0/2$ and $\ell = 1/3$ pairs (and consequently the weaker $\ell = 4$ and 5 modes as well). Because of the extended size of the fitting window, it was also decided to allow the background to vary as a function of one over frequency (which is assumed to be a reasonable assumption for how the background varies with frequency in real solar data). This is more important at low frequencies where the background varies more rapidly with frequency. It should be noted that this frequency dependence was only applied across the extent of each fitting window and there were no constraints forcing a fitting window at higher frequencies to have a smaller background than was fitted for a window at lower frequencies. However, for the most part the fitted backgrounds were found to ‘match-up’ from one window to the next.

3 Data used

A large set of simulated data was created enabling Monte Carlo simulations to be performed, in order to show whether, and if so where, the modified fitting code improved upon the standard code. The data were generated in the time domain using the first-generation solarFLAG simulator (Chaplin et al. 2006). A database of mode frequency, power, and linewidth estimates, based on analysis of spectra made from observations by the Birmingham Solar Oscillations Network (BiSON), was used in order to fix the various characteristics of each oscillator component. Visibility levels for the barely-detectable $\ell = 4$ and 5 modes were fixed at levels calculated by Christensen-Dalsgaard (1989). At the extreme ends of the spectrum – where there are no reliable, fitted estimates for the parameters – appropriate extrapolations of the known parameter values were made. In all cases the rotationally induced splitting between adjacent $m$ components was set at 0.4$\mu Hz$, again in order to match fitted estimates from observational data. In all a sequence of 60 ‘long’ 3456-day time series was generated, along with 240 ‘short’ 796-day time series.

4 Monte-Carlo simulations

In Fig. 2 the average fitted parameters determined from the Monte Carlo testing are shown. We concentrate initially on the background parameter as this is the parameter for which fits will most obviously be improved by accounting for the wings of modes that lie outside the fitting region. In Fig. 2(a) the average fitted backgrounds are plotted as a function of frequency for both the standard and modified fitting methods. For the standard technique the back-
grounds within each $\ell = 0/2$ and 1/3 fitting window are plotted, whereas for the modified technique the background levels at the frequency of each $\ell = 0$ mode are shown. (Recall that modes are fitted two pairs at a time by the modified technique and that the background is allowed to vary as one over frequency across the fitting window.)

The standard fitting code shows considerable overestimates of the background, as these estimates also incorporate the wings of the surrounding modes lying outside the fitting windows. There are also inconsistencies between the fits for different sets of mode pairs, with the estimated backgrounds in the window centered on $\ell = 0/2$ pairs being greater on average than those for $\ell = 1/3$. This problem can also be attributed to not allowing for the wings of the surrounding modes, and is a result of two distinct effects. The first is that the overall power in an $\ell = 1/3$ pair is slightly greater than that in an $\ell = 0/2$ pair (assuming similar mode frequencies). The second is dependent on the fact that, in the direction of lower to higher frequencies, $\ell = 1/3$ mode pairs are closer to the next 0/2 pair than $\ell = 0/2$ pairs are to the next 1/3 pair. This is important because modes at higher frequencies tend to have larger linewidths and (up until around 3100 $\mu$Hz) greater heights.

In contrast to the standard fitting, Fig. 2(a) shows that the modified code returns estimates of the background that closely match the input values, at least up to frequencies of around 3000 $\mu$Hz. The small overestimate of the background returned above this frequency is not entirely understood, although it does appear from the plots that the effect is dependent on the length of the time series.

It has previously been documented that many of the fitted mode parameters are correlated with one another (e.g., Fletcher2007). This suggests that the effect of not accurately fitting the true background, when using the standard fitting technique, may also impact on the estimates of the other parameters. This was highlighted in Section 2 in Fig. 1 where the height and widths were shown to be biased due to the presence of unmodeled outlying peaks.

In Fig. 2(b) average differences between the fitted and input frequencies are shown. The plot shows that for the most part the frequencies returned by the standard fitting technique are both robust and accurate and so there is little to improve upon when using the modified fitting code. The frequencies show the smallest correlation with the other parameters and therefore one might expect them to be particularly robust when fitting the data using the standard method.

In Fig. 2(c) the average differences in the natural logarithms of the fitted and input linewidths are shown. This time there is a small but clear systematic bias in the estimates returned by the standard fitting method. As the magnitude of the bias is quite small the differences in the natural logarithms closely approximate the fractional bias in the linear width values, (i.e., a difference in the natural logarithm of -0.05 indicates the fitted linear values of the widths underestimate the input value by about 5 percent). For the 3456-day power spectra, this bias is clearly reduced when using the modified fitting code, with the largest improvements coming at higher frequencies and for the higher-degree ($\ell = 2$ and 3) modes.

However, the results of the fits to the 796-day power spectra are not quite as clear cut. Over a large part of the frequency range the standard fitting code still returns fits with a negative bias, and for the most part (between around 2500 and 3500 $\mu$Hz) the modified fitting code improves upon this. However, there is evidence of a small positive bias in results from the modified code which increases at lower frequencies and for lower-degree modes. This is believed to be a resolution issue. It has previously been shown that when fitting the linewidth of the modes in the power spectrum, the estimates returned are actually better matches to the true inputs plus the binwidth (see Chaplin et al 1997). Since the binwidth is larger for a power spectrum made from shorter times series, this effect is more easily observed in the 796-day data. The non-horizontal dotted lines in Fig. 2(c) give, for the $\ell = 0$ modes, the difference between the natural logarithm of the linewidth plus the binwidth and the natural logarithm of the linewidth only. As such, this gives the line we would expect the $\ell = 0$ points to fall upon (assuming all effects other than this resolution bias have been removed) and the plot shows that over much of the frequency range this is indeed the case for the modified fitting code. The overestimation is reduced by the square root of the number of peaks being fitted which explains why the effect is reduced for the higher-degree modes.

When fitting the power spectrum, one will nearly always find a very strong correlation between the estimated linewidths and heights of the modes. Therefore, much the same pattern as was seen for the linewidths will be seen for the heights but with the opposite bias. This is indeed seen to be the case as shown in Fig. 2(d). Again the modified code is seen to reduce the overall extent of the bias especially with the longer 3456-day times series.

The results of fitting the rotational splittings are shown in Fig. 2(e) In this plot the differences between averages of the fitted splittings and the true input values are divided by the error on the mean. This has been done to more easily display the splittings at all frequencies and angular degrees on the same scale. It has been well documented that at high frequencies the fitted splittings will tend to overestimate the true values (e.g., Chaplin et al. 2001). This is a result of the strong mode blending that occurs as the linewidths of the modes increase at higher frequencies. This problem tends to be larger for lower-degree modes as the separation in frequency between the outermost components is smaller. However, this effect is less obvious when plotting the differences and dividing by the error, since the uncertainties on the lower-degree splittings are also very large.

Fig. 2(e) clearly shows a distinct overestimate of the fitted splittings at high frequencies with a number of points showing values larger than 3 sigma. As with the previous parameters, over the medium to high-frequency ranges, there appears to be a significant reduction in the bias when us-
Fig. 2 Parameters averaged from fits to 60 sets of simulated 3456-day time series and 240 sets of simulated 796-day time series, plotted as a function of input frequency. Fitted backgrounds and asymmetries are plotted directly, whereas frequencies, linewidths, heights and splittings are plotted in relation to input values (in the sense fitted - input). Note that for the linewidths and heights it is the difference in the natural logarithms that are plotted. Open symbols signify fits returned from the standard fitting code, whereas solid symbols give results of the modified code. For the standard code in plots (a) and (f) the fitted parameters for the \(\ell = 0/2\) window are given by diamonds and by triangles for the \(\ell = 1/3\) window. In (b) to (e) diamonds signify \(\ell = 0\), triangles \(\ell = 1\), squares \(\ell = 2\) and circles \(\ell = 3\). The dashed lines give 0.5-sigma values about the true inputs for a single fit while the error bars in (a) give 1-sigma errors on the mean. (0.5-sigma is used as opposed to 1-sigma to give a better scaling for the plots.) Dotted lines show true input values. In (c) the dash-dot lines show the expected overestimation of the fits taking into account the finite resolution of the spectra (see text).
ing the modified fitting code. However, the extremely large overestimates that are seen when fitting the highest frequencies are not significantly reduced.

The final parameter investigated was the peak asymmetry. Even though no asymmetry was included in the simulations, it is still important to investigate in order to check that the model returns estimates consistent with zero. Fig. 2 shows the average fitted asymmetries as a function of frequency. Unlike the previous parameters the asymmetry was kept constant throughout the fitting window, meaning there is no ℓ-dependence and thus fewer points in the plot. While the intrinsic values of the average fitted asymmetries are actually very small, they are significant and hence not consistent with zero. For the standard fitting code there are two sets of values for the asymmetries, one for the ℓ = 0/2 pairs and one for the 1/3 pairs. This is the same scenario as for the backgrounds, and as for that parameter, there is a significant disparity between the average fitted asymmetries for the two different sets of mode pairs, with the ℓ = 0/2 generally showing a much larger bias than the ℓ = 1/3. In fact the disparities seen in the background and asymmetry are most likely related.

The modified code seems to reduce the overall bias in the fits, although in this case, it is not eliminated entirely. Also, the plots show that the extent of the bias has a fairly smooth response as a function of frequency and this trend is similar for all cases (although somewhat exaggerated for the fits to the ℓ = 0/2 pairs). These two facts would suggest that there is some underlying cause for the bias in the fitted asymmetries that is not being correctly addressed, even with the modified fitting technique.

5 Conclusions

We have introduced a new peak-bagging fitting technique that was designed in order to gain some of the advantages that fitting the entire ‘Sun-as-a-star’ p-mode spectrum simultaneously might bring. Results on Monte-Carlo simulations demonstrated that this modified fitting method enabled accurate estimates of the true background level in the vicinity of the p-modes to be determined, something that was not possible using previous peak-bagging techniques. The Monte-Carlo simulations also showed that small biases present in other parameters (e.g. heights and widths) can be reduced using the modified code.

If the modified fitting code is indeed giving the true background level then it will enable investigations of the noise characteristics to be carried out within the frequency range of the p-modes. Previously, study of noise characteristics (which include both solar and instrumental sources as well as atmospheric sources in the case of ground based instruments) was limited to measurements of the background at frequencies well above and below the main part of the p-mode spectrum (e.g. Chaplin et al. 2003).

The modified code has also been tested using data collected by the Global Oscillations at Low Frequencies (GOLF) instrument on board the Solar and Heliospheric Observatory (SOHO) spacecraft. The results of this analysis will be presented in a subsequent paper. Additionally the modified code has been adapted for use with time series that contain breaks, such as those collected by the ground based BiSON group. Analysis of these type of data (both simulated and real) is the subject of current work.

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