A QUANTITATIVE NONRADIAL OSCILLATION MODEL FOR THE SUBPULSES IN PSR B0943+10

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ABSTRACT

In this paper we analyze time series measurements of PSR B0943+10 and fit them with a nonradial oscillation model. The model we apply was first developed for total intensity measurements in an earlier paper, and expanded to encompass linear polarization in a companion paper to this one. We use PSR B0943+10 for the initial tests of our model because it has a simple geometry, it has been exhaustively studied in the literature, and its behavior is well documented. As a prelude to quantitative fitting, we have reanalyzed previously published archival data of PSR B0943+10 and uncovered subtle but significant behavior that is difficult to explain in the framework of the drifting-spark model. Our fits of a nonradial oscillation model are able to successfully reproduce the observed behavior in this pulsar.

Subject headings: polarization — pulsars: general — pulsars: individual (PSR B0943+10) — stars: neutron — stars: oscillations

1. INTRODUCTION

The regular subpulse behavior and simple morphology of PSR B0943+10 have made this pulsar a fiducial for testing pulsar models, especially those models that incorporate drifting subpulses. Most drifting-subpulse models are based on a generic model proposed by Ruderman & Sutherland (1975). They hypothesized that a vacuum gap forms above the surface of the magnetic polar cap and along the corotating magnetosphere. This vacuum gap results from the depletion of charge due to particle emission from the star. Because the vacuum gap cannot grow indefinitely, sparks discharge across the gap between the magnetosphere and the stellar surface. These spark regions are fixed in relation to each other. Because they occur in the region of the magnetosphere that does not corotate with the star, the sparks rotate around the magnetic pole at a period incommensurate with the spin period of the star. In this model the drifting subpulses are the observable manifestation of the rotating sparks. The drifting- (or rotating-) spark model, as it is commonly known, is the foundation for many current pulsar models (Komesaroff 1970; Backer 1976).

In contrast to the drifting-spark model, we have proposed a nonradial pulsation model to explain not only drifting subpulses in pulsars, but also a wide range of other pulsar behavior (Clemens & Rosen 2004, 2008). Pulsations are a widely observed phenomenon in normal stars and in compact objects; white dwarf stars, rapidly oscillating AP stars, delta Scuti stars, and even our Sun are known to have oscillation modes. Nonradial oscillations were previously proposed as an explanation for drifting subpulses (Gold 1968; van Horn 1980; Strohmayer 1992), but no phenomenological model has been developed to explain the range of behavior seen in pulsars with drifting subpulses. Our nonradial oscillation model was in its preliminary stages when Edwards et al. (2003) published their analysis of PSR B0320+39, shown in left panel of Figure 1. The phase behavior of the subpulses in the data they presented suggested that the modulations we see are a combination of two different manifestations of nonradial oscillations: timelike variations and nodal lines sweeping past our sight line.

In nonradial oscillations, a nodal line is a boundary of zero (modulated) emission that separates two regions of opposite phase. The changing pulsation phase associated with rotating nodal structure is familiar from studies of rapidly oscillating AP stars. For example, the rapidly oscillating AP star HR 3831 shows phase changes associated with rotation because it has a pulsation axis aligned to the magnetic axis of the star but misaligned to the rotation axis. The right panel in Figure 1 shows pulsational phase changes in HR 3831 as the star’s rotation carries a nodal line around the star. Compare this to the phase behavior of the subpulses in PSR B0320+39 in the left panel. PSR B0320+39 has a minimum in the subpulse amplitude envelope that corresponds to a 180° shift in the phase, as expected for a nodal line and as seen in HR 3831. Edwards (2004) has since published more complex observed phase behaviors that are challenging both to drifting-spark models and pulsation models. We will defer the analysis of these more complex behaviors until our model has been demonstrated on a simple case, PSR B0943+10.

Because it has simple subpulse phase behavior, PSR B0943+10 is an ideal candidate for testing our model. It has a simple average pulse shape and a single subpulse frequency. It has been interpreted previously in the context of the drifting-spark model (Deshpande & Rankin 2001), and in this paper we use the same data (Suleymanova et al. 1998) to fit PSR B0943+10 quantitatively using our nonradial oscillation model. As a prelude to fitting the data, we conducted an independent analysis. In this process, we found significant behavior not mentioned in previous publications and challenging to explain with any model. This behavior includes a splitting in the subpulse frequency and a bifurcation of the drift band in only part of the pulse profile. Detecting these behaviors requires Fourier transforms with higher frequency resolution and drift-band plots with finer amplitude resolution than previously published for this data set.

Time series analysis of PSR B0943+10 is difficult because of a large, apparently stochastic pulse amplitude distribution that is intrinsic to the star and not due to instrumental noise. It is possible to filter this in Fourier space as Edwards & Stappers (2002) have implemented, but this may alter the data in the time domain in ways that are not intuitive. Because most of the previous analysis of PSR B0943+10 was conducted in the time domain (Deshpande & Rankin 2001), we have chosen to do the same. It would be a profitable exercise, but beyond the scope of this paper, to repeat the analysis using the techniques of Edwards & Stappers (2002). To conduct our fitting in the time domain, we
have normalized the pulses as described in §4.1. The normalization process prevents our fit from being dominated by stochastic amplitude variations, but also removes any useful information that might be carried by pulse amplitudes. We are careful throughout this paper to distinguish between presentations of raw data and data that have been normalized. For instance, in §3, the data analysis section, we present plots based only on data without any alteration.

Our work in this paper is organized as follows: In §2 we present a detailed summary of the model that we fit. We then analyze the 430 MHz data of PSR B0943+10 in §3, highlighting differences in our analysis to those of Deshpande & Rankin (2001). We then, in §4, fit normalized data in the time domain to our pulsational model using gaussfit, a robust least-squares approximation package. Using gaussfit we are able to fit quantitative values for all of the relevant parameters in our model. We find that the parameters can be separated into geometrical and pulsational parameters and the geometrical parameters are mostly independent from the pulsational parameters. To test the quality of our fits we produce simulated data and fitted parameters and compare them to the data in §5. Our model successfully reproduces the essential features of PSR B0943+10. In §6 we discuss the results of our fit and the comparison of our model to data.

2. OUR NONRADIAL OSCILLATION MODEL

In Clemens & Rosen (2004) we introduced an oblique pulsator model (Kurtz 1982) of high spherical degree ($\ell$) to explain the phenomena of drifting subpulses in radio pulsars. Originally this model only attempted to reproduce the behavior of the total intensity (Stokes $I$) of pulsars. In our companion to this paper (Clemens & Rosen 2008) we expand this model to encompass linear polarization by introducing two orthogonal polarization modes associated with pulsations. The first of these orthogonal polarization modes is modulated by pulsational displacements. These displacements have a transverse electric field vector that points toward the magnetic pole and follows the single-vector model of Radhakrishnan & Cooke (1969). We refer to this radiation as the “displacement polarization mode” and express the time-dependent amplitude of this radiation as the positive portion of the function:

$$A_{\text{DPM}}(t) = a_{0,\text{DPM}} + a_{1,\text{DPM}} \Psi_{\ell,m=0}(\theta_{\text{mag}}) \cos(\omega t - \psi_0 - \psi_{\text{delay}}),$$

where $\Psi_{\ell,m=0}$ is a spherical harmonic of high $\ell$ and $m = 0$. The variable $\theta_{\text{mag}}$ refers to the magnetic colatitude, because the pulsations in our model are aligned to the magnetic pole. The amplitudes $a_{0,\text{DPM}}$ and $a_{1,\text{DPM}}$ are to be fitted to the data.

The pulsational displacements and their associated velocities move toward and away from the magnetic pole. Thus, the induced electric field as a result of the velocities ($E = v \times B$) is naturally orthogonal to the Radhakrishnan & Cooke vector. We connect this induced electric field to the second orthogonal polarization mode, which we refer to as the “velocity polarization mode.” Mathematically, we express the velocity polarization mode as a time-varying amplitude:

$$A_{\text{VPM}}(t) = a_{0,\text{VPM}} \frac{\partial \Psi_{\ell,m=0}}{\partial \theta_{\text{mag}}} \sin(\omega t - \psi_0),$$

Fig. 1.—Left: Average pulse shape, subpulse amplitude (top), and phase (bottom) of PSR B0320+39 (Edwards et al. 2003). The subpulse amplitude envelope shows a minimum near zero at the same longitude as a 180° shift in the phase. A 60° slope has been removed from the phases. Right: Oscillation phase vs. magnetic rotation phase for rapidly oscillating AP star HR 3831 (Kurtz et al. 1990).
which incorporates the time derivative and the $\theta_{\text{mag}}$ derivative of equation (1), as appropriate for horizontal pulsation velocities. This equation is analogous to the $V_\theta$ in equation (3) of Dziembowski (1977).

To convert the amplitudes in equations (1) and (2) into quantities that can be directly compared to observations, we use the following transformations to calculate Stokes parameters in the frame of the star:

$$I = (A_{\text{DPM}})^2 + (A_{\text{VPM}})^2,$$

$$Q' = (A_{\text{DPM}})^2 - (A_{\text{VPM}})^2,$$

$$U' = 0.$$

We have followed Deshpande & Rankin (2001) in their use of prime notation to indicate measurements in the nonrotating frame of the star. Translating from the primed quantities into the observer’s frame requires incorporating the changing longitude we observe as the star spins and imposing rotation of the polarization angle so that the polarization of the displacement polarization mode follows the changing direction of the magnetic pole, given as

$$\chi_{\text{model}} = \chi_0 + \tan^{-1}\frac{\sin(\alpha)\sin(\phi - \phi_0)}{\sin(\alpha + \beta)\cos(\alpha) - \cos(\alpha + \beta)\sin(\alpha)\cos(\phi - \phi_0)}.$$

Because the total linear polarization does not change with the frame of reference,

$$L = \sqrt{Q'^2 + U'^2} = \sqrt{Q^2 + U^2},$$

then $Q'$ and $U'$ can be rotated into $Q$ and $U$ using the following transformation:

$$Q = L \cos(2\chi_{\text{model}}),$$

$$U = L \sin(2\chi_{\text{model}}).$$

And we have to make the rotational longitude, $\phi$, and the magnetic colatitude, $\theta_{\text{mag}}$, functions of time as follows:

$$\phi = \frac{t - t(\phi_0)}{P_1} \times 360,$$

$$\theta_{\text{mag}} = \cos^{-1}[\sin(\alpha)\cos(\phi - \phi_0)\sin(\alpha + \beta) + \cos(\alpha)\cos(\alpha + \beta)].$$

2.1. The Observed Pulse Window

Our model treats the pulsations as global oscillation modes of the neutron star, which means that they can modulate emission coming from any location on the star. However, pulsars are observed to emit radiation only from the region surrounding the magnetic poles. The theoretical explanation for this is generally framed around the Goldreich & Julian (1969) aligned rotator model, in which charged particles can escape only from the polar regions.

In order to make quantitative comparisons between our model and observations, we have to impose an “emission window,” analogous to the observed “pulse window,” that is separate from the pulsation model and limits the effects of pulsations to the regions that are observed to emit. For this purpose we have imposed on $I$, $Q$, and $U$ a “window function” that is zero in those portions of the pulse spin when the star is off, and is a Gaussian function with a maximum of unity in the emitting region. Observations show that this window is not necessarily centered on the longitude of the magnetic pole (Johnston et al. 2005). The window we use has a maximum $(\phi_{\text{mean}})$ and a width $(\sigma)$ equal to the best Gaussian fit to the average pulse shape. In all of the presentations of data that follow, zero in longitude refers to the center of this window rather than $\phi_0$, the longitude of the magnetic pole.

3. DATA ANALYSIS OF PSR B0943+10

In this section we discuss our analysis of archival data of PSR B0943+10. We highlight features in the data not mentioned in Deshpande & Rankin (2001). These include a small change in the subpulse frequency, a splitting in the drift band, and effects of the stochastic pulse height distribution. Most significantly, Deshpande & Rankin (2001) detect a modulation of the subpulses in a small portion of the data that bolsters support for their drift-spark model. We show in this section that the amplitude modulation, while present in that specific subsection of the data, is probably the result of the large stochastic pulse amplitude variations intrinsic to the pulsar. These amplitude variations make the data difficult to model and in a subsequent section, § 4.1, we discuss our strategy for circumventing these problems.

The 430 MHz data of PSR B0943+10 that we examine in this paper are the same data analyzed by Deshpande & Rankin (2001) and include 986 pulses in all four Stokes parameters, sampled synchronously with the period of the star, $P_1 = 1.097608$ s, in millisecond increments. Therefore, each pulse sample contains 0.1559 s of measurements, corresponding to 50° of longitude. Of this, we designate a pulse window of about 20°, and the remaining portion of the sample is considered off-pulse, instrumental noise. Figure 2 shows a plot of the central 20° of single pulses of PSR B0943+10. The pulses clearly drift through the pulse window. Each pulse sample is not contiguous with the adjacent pulses, and there are not an integer number of time intervals, $t$, between each pulse. Our analyses and that of Deshpande & Rankin (2001) assume that the signal from the pulsar is zero during the off-pulse portion, but this is not shown in Figure 2. Figure 2 illustrates the
important periods in pulsar terminology: \( P_1 \), the spin period of the pulsar, the time between pulses, and the period at which the data are folded in Figure 2; and \( P_2 \), the spacing between adjacent subpulses. \( P_3 \) is ordinarily the length of time for a subpulse to return to a given longitude, but we have followed Deshpande & Rankin (2001) and used \( P_3 \) from the lowest alias of \( P_2 \) in the Fourier transform, specifically \( P_3 = 1.867P_1 \).

The data were originally acquired by Suleymanova et al. (1998), who describe the observing setup. The subpulses appear in their steady, highly organized bright “B-mode” state for the first 816 pulses, where the spacing between the subpulses, \( P_2 \), is about 31.78 ms. The remaining 170 pulses are disorganized, with generally lower amplitude, and the pulsar is considered to be in its “Q-mode” (quiescent) state. We do not present a full repetition of the analysis of Deshpande & Rankin (2001), nor do we apply all of their analysis techniques, e.g., folding the data at 20\( P_1 \), where \( P_3 (1.866P_1) \) is the length of time for a subpulse to return to a given longitude. Instead, we refer to the reader to their paper for most of the data analysis, except for those instances where we find significant behavior of the star not shown in their analysis.

Deshpande & Rankin (2001) use four basic tools in their analysis: synchronous folding of the data, the Fourier transform, the longitude resolved fluctuation spectrum, and the harmonic-resolved fluctuation spectrum. When synchronously folding the data at a given period, usually \( P_3 \), the subpulses drift through the pulse window at each successive spin until they repeat in longitude, thus creating a drift-band plot. For example, Deshpande & Rankin (2001) fold the data at 1.867\( P_1 \) and at 37.346\( P_1 \) (= 20\( P_1 \)), in their Figures 8 and 9, respectively. The longitude resolved fluctuation spectrum, first introduced by Backer (1973), is a Fourier transform calculated at each longitude. The harmonic-resolved fluctuation spectrum is a Fourier transform of the time series stacked at the spin frequency. Edwards & Stappers (2002) have shown that the combination of the longitude and harmonic-resolved fluctuation spectra present the same information as a two-dimensional Fourier spectrum. While we have also calculated the two-dimensional Fourier spectrum, we do not present it here because it is less familiar and not directly comparable to the analysis of Deshpande & Rankin (2001).

The most difficult aspect of modeling PSR B0943+10 is the large stochastic variation in the pulse amplitudes. The average signal-to-noise ratio of an integrated pulse is 4.93, where we used the off-pulse instrumental noise as the noise estimate. The standard deviation of the on-pulse data average to be 7.13 of the standard deviation of the noise but varies from 1.23 to 109.87. From this we conclude that the large variations in pulse height, as shown by the pulse height distribution given in Figure 3, are due to stochastic variations of the pulsar, not instrumental noise. Indeed, Cordes (1978) has argued that pulsar subpulses are amplitude modulation of shot noise-like emission. Our analysis of the data in this section is tempered by the possible effects of the stochastic variation, but we do not attempt to remove this variation by any normalization process until § 4.

### 3.1. Drift Bands

We begin our analysis by folding the data at \( P_3 \), a technique used by Deshpande & Rankin (2001) to show changes in the shape of individual subpulses as a function of longitude. Deshpande & Rankin (2001) remove an “aperiodically fluctuating base” prior to presenting their data. They do not describe their method for removing this base in enough detail for us to reproduce it. We have decided to present the unaltered data, which may not be exactly the same as in their presentation.

Our drift-hand plots in Figure 4 show broadening, or even splitting, on the right-hand side of the profile that is not evident in the lower resolution plot in Deshpande & Rankin (2001). To explore this, we folded shorter 100-pulse segments of the data at \( P_3 \) and noticed an obvious splitting on the right side of drift band, as shown in the left panel of Figure 4. While we only show the first 100 pulses here, drift bands of other 100-pulse segments of the data also display splittings on the right-hand side of the drift band. The washed-out appearance of the split in the drift band in the full data set (Fig. 4, right) is a result of the drifting in the subpulse frequency, which we discuss in § 3.2.

The splitting in the drift band is difficult to explain, because it is a longitude-dependent phenomenon. In the model of Deshpande & Rankin (2001) in particular, there is a pattern of 20 subbeams circling the magnetic pole. In each successive spin of the star, an individual subbeam moves a few degrees earlier in longitude. This means the same subbeam as seen on the right side of the pulse window will, in the next spin of the star, have moved to the left side. The splitting in the drift band indicates that all subbeams must be split when they are present on the right side of the pulse window and that they recombine when they move to the left side. This behavior is difficult to explain with the rotating spark model without the multiplication of new quantities, but we show that it is a natural result of the phase difference between the displacement and velocity polarization modes in our model.

### 3.2. Fourier Transform

The Fourier transform is an indispensable tool for finding frequencies in noisy data. Deshpande & Rankin (2001) use a fast Fourier transform to determine the subpulse frequency, but they do not calculate a fully resolved transform of all 816 pulses analyzed in their paper. Instead they calculate fast Fourier transforms of 256-pulse sections and average them. This prevents them from seeing fine structure in the subpulse frequency that is only visible at high resolution. We show that our resolved Fourier transform of the entire 816 pulses shows splitting in the subpulse frequency. This splitting is not the same as the sidelobes uncovered by Deshpande & Rankin (2001); this spacing is significantly smaller and does not have the obvious symmetry to be sidelobes due to an amplitude modulation. Our analysis shows that it is impossible with the limited data to determine whether the two largest peaks...
are two closely spaced, independent frequencies or the result of instabilities of the subpulse frequency. The presence of two independent frequencies would be a significant complication for the drifting-spark model but would convey asteroseismological information if the variations are nonradial oscillations. Additional data is necessary for a definitive identification of frequency splitting.

To compute our Fourier transforms we use a tool written by Carl Hansen for the Whole Earth Telescope (Nather et al. 1990). It is optimized for discrete, contiguous chunks of data with gaps in between. Unlike the fast Fourier transform, this tool does not require an integer number of time intervals, \( t \), between the chunks and is therefore appropriate for the data from PSR B0943+10. Like the fast Fourier transform, this tool assumes zeros between chunks of data, but as a discrete Fourier transform, it does not require \( 2^n \) data points or a resampling the data into equally spaced time bins.

The peaks in panel \( d \) are representative of all the aliased subpulse peaks. The larger of the two largest peaks (peak 2) is at 31.4647 Hz (\( P_2 = 31.7816 \) ms), which is not the largest subpulse alias in panel \( a \). We have chosen this subpulse alias as the subpulse frequency because this alias shows an integer relation to the second and third harmonics which are not shown in Figure 5. The location of the smaller peak (peak 1) is at 31.4618 Hz (\( P_2 = 31.7846 \) ms); the difference in the two peaks is 0.0029 Hz. This difference in frequency would be barely visible in a 256-pulse Fourier transform, especially if several transforms were averaged together, which could broaden the subpulse frequency.

To explore frequency and phase stability of the subpulses, we divided the data into sequential 50-pulse segments, where the number of pulses in each segment (50) was chosen arbitrarily. Since we were fitting only the subpulses in the pulse window, we needed enough pulses that the subpulse frequency could be accurately fit. The upper limit on our segment length was determined by attempting to minimize the amount of the frequency wander and having enough data segments to determine any trends. Fourier transforms of these smaller subsets of data can still clearly show the subpulse frequency, but they do not have the frequency resolution to separate the subpulse into the distinct peaks seen in the fully resolved Fourier transform. We used gaussfit, a robust least-squares approximation package, to fit the intensities (Stokes \( I \)) in each of the 50-pulse segments to a periodic function. Because the variance in the data is much higher than the off-pulse instrumental noise, we use the variance as an error estimate. This artificially forces the reduced \( \chi^2 \) to be near 1, so we have no absolute measure of the goodness of our fits.

We cannot use a sine wave alone for this fit because its mean would be zero, and our data are all positive. Removing the mean from the data, which would be the normal procedure for studying pulsations, does not help in this case because of the large amplitude variations. Thus, we have used for our fit the positive

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1 At http://clyde.as.utexas.edu/Software.html.
portion of our function $I = A \cos \left( \frac{2\pi t}{P_2} - \psi \right)$. This yields good convergence from gaussfit and permits us to measure the period and phase in each segment of the data. We have plotted the period and amplitude as a function of data segment in the left panels of Figure 6.

In Clemens & Rosen (2004) we pointed out that $P_2$, originally defined as the time between adjacent subpulses in a drift band, may not be a good estimate of the underlying pure subpulse frequency in our pulsation models. Consequently, we called the subpulse period $P_{time}$. Edwards (2006) has pointed out that the amplitude windowing which distorts measurements of $P_2$ in individual pulses does not affect measurements that rely on subpulse phase. In this paper all of our measurements of $P_2$ are either from Fourier transforms or linear fits using pure sine waves, which means our estimate of $P_2$ is a good measurement of $P_{time}$, and we are using the more familiar notation of $P_2$ for the measured subpulse period.

Our analysis of the smaller data segments reveals that at low resolution, $P_2$ wanders over the entire run. The change in $P_2$ with subset is shown in the bottom left panel of Figure 6 and the corresponding amplitudes ($A$) are shown in the top left panel. As shown in Figure 5, the fully resolved Fourier transform shows several distinct peaks. None of the subsets in Figure 6 is long enough to resolve these peaks individually; instead they appear to be a single peak with its maximum determined by the relative sizes of the unresolved peaks, causing the wander in $P_2$. This single peak is large compared to the noise and therefore gaussfit is able to fit it very accurately, resulting in the extremely small error bars in Figure 6. The width of the peak, however, is determined by the length of the time series and is larger than the error of the fit.

The pattern in the period change is more complex than we would expect from a pure frequency splitting, which would generate a regular pattern of period changes. We can explore things further by examining the subpulse phase. If the subpulse frequency were solely amplitude modulated we would expect symmetric sidelobes in the Fourier transform and our least-squares fit using a fixed frequency would show a stable subpulse phase. If the subpulse is actually two closely spaced frequencies, a fit with a fixed frequency would result in a periodic change in the subpulse phase.

To examine the phase behavior, we averaged the subpulse frequencies in the bottom left panel of Figure 6 and then fit each of the 50-pulse subsets with gaussfit with the same positive portion of the function $I = A \cos \left( \frac{2\pi t}{P_2} - \psi \right)$, but with fixed $P_2 = 31.7818 \text{ ms}$. The resulting phase as a function of time is shown in the bottom right panel of Figure 6, and the corresponding amplitudes are in the top panel. Because it appears that
the subpulse frequency and phase wander through only one or two cycles, we would require a longer data stream to conclusively determine whether more than one frequency is present. The amplitudes of the phase variations appears to increase with pulse number. Our data terminates with a transition from the stable B-mode to the disordered Q-mode, where the fits in the section were confined to the 816 pulses in the B-mode. Whether the transition into a disordered state is related to the increase in size of subpulse phase and period variations is a also question requiring more data.

3.3. Longitude and Harmonic-Resolved Fluctuation Spectra

Stochastic variation in the pulse amplitudes complicates attempts to extract any underlying signal for comparison to our model. While stochastic variation averages away in the drift bands and Fourier transforms of large segments of the data, in short sections the probability is greater that large stochastic variations can mimic an amplitude modulation. Within this data set Deshpande & Rankin (2001) have detected a periodic amplitude modulation in the region of pulses 129–384 of the data. This periodic amplitude modulation manifests itself as sidelobes around the lowest alias of the subpulse frequency in the longitude-resolved fluctuation spectra; these sidelobes are distinctly different that the peaks discussed in §3.2 and shown in Figure 6. The amplitude modulation is a key component for increasing their confidence in the circulating spark model; Deshpande & Rankin (2001) interpret it as a persistent pattern of subbeam brightness. They view it as a persistent pattern of subbeam brightness. In favor of the significance of amplitude-modulated sidelobes, Monte Carlo simulations in which we shuffled the amplitudes in pulses 129–384 of the data showed a very low probability (less that 1%) that two peaks of the size seen in the longitude-resolved fluctuation spectra would occur by chance. However, the probability that an amplitude modulation could appear in the longitude-resolved fluctuation spectrum and be masked by noise in the harmonic-resolved fluctuation spectrum is equally small. The sidelobes are also not present in the other sections of the data, neither in the longitude nor harmonic-resolved fluctuation spectrum. For comparison, we show the next 256 pulses (pulses 384–640) in the right panels of Figures 7 and 8.

In Figure 7 the aliased subpulse frequency is reflected about the Nyquist frequency and appears at about 0.43 Hz. As the panel on the left shows, the subpulse signal has sidelobes indicative of amplitude modulation with a period of just over $37P_s$. The difference in our longitude resolved fluctuation spectrum (Fig. 7, left; and Fig. 7 in Deshpande & Rankin 2001) is likely due to their removal of the aperiodically fluctuating base, which we do not remove. In Figure 8 the aliased subpulse frequency at its true value of 0.49 Hz and shows no significant sidelobe structure. Visual inspection of the harmonic-resolved fluctuation spectrum in Figure 4 of Deshpande & Rankin (2001) does not show evidence of sidelobes, nor do we see them in our harmonic-resolved fluctuation spectrum of the entire 816 pulses. The sidelobes are also not present in the other sections of the data, neither in the longitude or harmonic-resolved fluctuation spectrum. For comparison, we show the next 256 pulses (pulses 384–640) in the right panels of Figures 7 and 8.

The amplitude modulation interpreted by Deshpande & Rankin (2001) can be seen in the longitude and harmonic-resolved fluctuation spectra of pulses 129–384 in the left panels of Figures 7 and 8. In Figure 7 the aliased subpulse frequency is reflected about the Nyquist frequency and appears at about 0.43 Hz. As the panel on the left shows, the subpulse signal has sidelobes indicative of amplitude modulation with a period of just over $37P_s$. The difference in our longitude resolved fluctuation spectrum (Fig. 7, left; and Fig. 7 in Deshpande & Rankin 2001) is likely due to their removal of the aperiodically fluctuating base, which we do not remove. In Figure 8 the aliased subpulse frequency at its true value of 0.49 Hz and shows no significant sidelobe structure. Visual inspection of the harmonic-resolved fluctuation spectrum in Figure 4 of Deshpande & Rankin (2001) does not show evidence of sidelobes, nor do we see them in our harmonic-resolved fluctuation spectrum of the entire 816 pulses. The sidelobes are also not present in the other sections of the data, neither in the longitude or harmonic-resolved fluctuation spectrum. For comparison, we show the next 256 pulses (pulses 384–640) in the right panels of Figures 7 and 8.

In favor of the significance of amplitude-modulated sidelobes, Monte Carlo simulations in which we shuffled the amplitudes in pulses 129–384 of the data showed a very low probability (less that 1%) that two peaks of the size seen in the longitude-resolved fluctuation spectra would occur by chance. However, the probability that an amplitude modulation could appear in the longitude-resolved fluctuation spectrum and be masked by noise in the harmonic-resolved fluctuation spectrum is equally small. The sidelobes do not appear in the harmonic-resolved fluctuation spectrum either in our plots or in the plots of Deshpande & Rankin (2001). There are a number of other reasons to suspect that it is something other than a pure amplitude modulation. Based on the models of the drifting carousel pattern of sparks, Edwards & Stappers (2002) show that an amplitude modulation caused by a persistent patterns in the subpulses ought to appear as a
Fig. 7.—Longitude-resolved fluctuation spectrum for two consecutive 256-pulse segments of the data from PSR B0943+10. **Left**: Subsection of the data (pulses 129–384) where Deshpande & Rankin (2001) discovered the tertiary modulation that supports the rotating spark model. **Right**: Longitude-resolved fluctuation spectrum for the next portion of the data (pulses 385–640); the tertiary modulation is not present. The contour levels are in mJy².

Fig. 8.—Harmonic-resolved fluctuation spectrum for the same two consecutive 256-pulse segments of the data from PSR B0943+10, as in Fig. 7. **Left**: The subsection of the data (pulses 129–384) where Deshpande & Rankin (2001) discovered the tertiary modulation. **Right**: The harmonic-resolved fluctuation spectrum for the next portion of the data (pulses 385–640). The contour levels are in mJy².
sidelobe of the DC component at zero frequency. In the same section of the data it does not appear as a low-frequency sidelobe. It is also not present in any of the other independent 256-pulse samples, nor has it reappeared in subsequent data.

Clearly it is of the highest importance to establish whether periodic amplitude modulation occurs in the drifting subpulses. It would be the strongest evidence in favor of the drifting subpulse model and against the model we present here. Its appearance in a single section of the data, like the appearance of frequency splitting and phase wandering in the subpulses, needs to be addressed with additional data before we can assess its consequences for our model.

4. MODEL FITTING AND SIMULATIONS

Departing now from new phenomena uncovered by our analysis, we address the central question of this work, which is whether or not our pulsational model can reproduce the stable features of the observations. We will address this question by first attempting to fit the data using our pulsational model. The purpose of the fitting is to select objective quantities for the model variables. Once we have completed the fitting, we will use the fitted parameters to generate synthetic pulsar light curves and see how well they reproduce the features of the drifting subpulses. As we show, the best fits to our model generate synthetic data that reproduces the drift bands with splitting, the polarization properties shown in Figure 9, and the essential features of the harmonic- and longitude-resolved fluctuation spectra.

4.1. Fitting Strategy

Before we can apply our general model to a specific star like PSR B0943+10, we must first determine whether orthogonal polarization mode emission is present in the star and whether the emission polarization angles are consistent with the single-vector model of Radhakrishnan & Cooke (1969). Figure 9 shows a histogram of the distribution of polarization angles in individual subpulses as first introduced by ; Stonebring et al. (1984a, 1984b). At each longitude it counts how many individual pulses show polarization angles in each bin. The polarization angles clearly group around two parallel polarization tracks. These tracks are only gently curved, indicating that our sight line misses the magnetic pole so that the entire pulse samples a small range of magnetic longitude.

We have identified the stronger track with the displacement polarization mode of our model and the weaker track with the velocity polarization mode because the reverse identification did not yield acceptable fits. Inspection of the bottom panel of Figure 9 suggests that the amplitude coefficient for the displacement polarization mode must be larger than the velocity polarization mode, i.e., the displacement polarization mode is the dominant mode in this star. By convention, we would call this the primary polarization mode, but we want to avoid attaching the displacement polarization mode to the primary polarization because it is not clear that the displacement polarization mode will dominate in every pulsar.

We can fit the rotation axis inclination, $\alpha$, and impact parameter, $\beta$, using these polarization angle swings but the fits are ambiguous, yielding only the ratio of $\alpha$ to $\beta$. We discuss our results for these parameters in § 4.2. Deshpande & Rankin (2001) apply external information to constrain $\alpha$ independently and for consistency we have chosen values of $\alpha$ near theirs. The only pulsational parameter that interacts with these $(\alpha, \beta)$ is the spherical harmonic degree $\ell$ because the positions of the nodal lines that encircle the magnetic pole are related $\ell$, and the path our sight line threads through these nodal lines is related to $\beta$. Figure 10 is a scan of the spherical harmonic along a single magnetic longitude and shows the location of nodal lines for a particular $\ell$ ($\ell = 75$) and various choices of the impact parameter $\beta$. Using only the data on PSR B0943+10 analyzed in this paper, there is no way for us to know which nodal region our sight line passes through, nor is there anyway to constrain $\ell$ that is independent of $\beta$. In most emission models the divergence of the dipole field magnifies the structure at the stellar surface so the apparent $\ell$ we fit represents a surface $\ell$ approximately 7 times higher, as we explained in Clemens & Rosen (2004). Without a more complete model of emission, all information about the size of $\ell$ is surrogate information and can never be interpreted asteroseismologically.

Fitting the other pulsational parameters—amplitudes, periods, and phases—requires that we minimize an appropriate metric for the mismatch between the measured Stokes parameters and the model Stokes parameters. This is complicated by the lack of circular polarization in our model but its presence in the star. The measured circular polarization, while not as strong as the linear
polarization, does contribute to the intensity in each pulse and may or may not follow the shape of the pulse profile. Since we are only modeling the total and linear intensity, we remove the measured $V$ from the measured $I$ by constructing the Stokes parameter $I_{\text{new}} = (I^2 - V^2)^{1/2}$ for each pulse. Otherwise we ignore the measured Stokes parameter $V$.

A further complication arises from the appearance in the data of polarization vector changes. In our model the displacement and velocity polarization modes are orthogonal polarization states, which allows them to be described by a single Stokes parameter, e.g., $Q$, and leaves the other Stokes parameter zero. Deshpande & Rankin (2001) have shown that removal of the Radhakrishnan and Cooke vector rotation from the PSR B0943+10 data moves all of the subpulse behavior to one Stokes parameter, which they call $Q'$, and leaves only noise in the Stokes parameter $U'$, further confirming that the variations of PSR B0943+10 come from the superposition of orthogonal polarization modes. We have adopted the notation of Deshpande & Rankin (2001) and refer to the Stokes parameters calculated by our model before the incorporation of any vector rotation as primed quantities. This gives us the option to conduct our fits in either the observed Stokes parameters (unprimed space) or in the frame of the star where the theoretical values are calculated (primed space). We have chosen to do the fitting in observational (unprimed) space because it does not require any further alteration to the data. All the fits we present are the result of the simultaneous minimization of $I_{\text{model}} - I_{\text{new}}$, $Q_{\text{model}} - Q_{\text{data}}$, and $U_{\text{model}} - U_{\text{data}}$. The quantities $I_{\text{model}}, Q_{\text{model}}$, and $U_{\text{model}}$ are calculated using equations (3), (8), and (9).

The final complication, and the most vexing, is the large, apparently stochastic, pulse height variations. To minimize the effects of the pulse height variation on our fits, we have normalized the data to remove the pulse-to-pulse amplitude variations. To do this, we constructed an average pulse shape from the entire data set and normalized it to unit amplitude. We then normalized each individual pulse in the Stokes parameter $I$ by multiplying each pulse by a single scale factor. The scale factor was based on the maximum intensity in each individual pulse. This maximum does not always occur at the center of the pulse profile, so we calculated the scale factor for each pulse from the ratio of its maximum to the average pulse at the same longitude. We treated the linear polarization parameters in a similar manner by constructing the linear polarization $[L = (Q^2 + U^2)^{1/2}]$. We normalized this quantity to unit amplitude in the same way that was done for $I$. Then we scaled the individual pulses in $Q$ and $U$ by the maximum of the linear polarization for that pulse, effectively reducing both $Q$ and $U$ by the same normalization factor. This process leaves the ratio of $Q$ to $U$ the same, thus having no effect on the value of the polarization angle. Since the instrumental noise is normalized in the same manner as the rest of the pulse, an amplitude modulation can still be present as the ratio of pulse height to noise is preserved. However, the normalization process reduces any amplitude modulations while still retaining the structure in the individual pulses.

Normalization contained in the pulse-to-pulse variations makes the fitting exercise tractable but removes any information contained in the amplitude variations. Before this normalization process fits to the data did not consistently converge to solutions. Afterward the fits are always well behaved but can no longer yield any astrophysical information about pulse amplitudes. However, frequency and phase are only minimally affected. We have demonstrated this by recalculating and examining all of the figures in §3 for the normalized data. The appearance of the drift bands and longitude and harmonic fluctuation spectrum remain essentially unchanged.

![Fig. 11.—Polarization angle histogram of all 816 pulses of PSR B0943+10.](image)

The lines represent different values of $\alpha$, $\beta$, $\phi_0$, and $\chi_0$. The top line is the polarization angle calculated from the values determined by Deshpande & Rankin (2001), given in Table 1. The bottom line is the result of fits 1 and 2 in Table 1, which are indistinguishable.

### 4.2. Fits to the Polarization Angle

Fits to the polarization angle swing, depicted in Figure 9, require four parameters: $\alpha$, $\beta$, $\phi_0$ (the rotational longitude of the magnetic pole), and $\chi_0$ (the position angle of the linear polarization at $\phi_0$). As we have already mentioned, these parameters cannot be fit independently. For any given value of $\alpha$ and $\phi_0$, a corresponding $\beta$ and $\chi_0$ can be fitted, and there is no substantial difference in the quality of the fits. However, only one value of the ratio of $\alpha$ to $\beta$ yields good fits, and we have preserved the best ratio in all of our work. The fits showed that $\phi_0$ and $\chi_0$ are also highly correlated.

Figure 11 shows the Radhakrishnan & Cooke (1969) polarization angle calculated from equation (6) superimposed on the polarization angle histogram of the data using the values shown in Table 1. These values were constrained to keep $\alpha$ near the values given in Deshpande & Rankin (2001). Our fit has a different slope than that of Deshpande & Rankin (2001), which cannot be matched by only adjusting $\phi_0$. Inspection of Figure 11 shows that our fit is slightly better than that of Deshpande & Rankin (2001).

### 4.3. Fits to the Pulsation Model

Once we have settled on values for parameters $\alpha$ and $\beta$ from geometric arguments, it is straightforward to fit our pulsational model to the data, but we must first choose a value for $\ell$. As we have already discussed, the features we observe in pulsar beams are spatially magnified, so that value of $\ell$ we use is a surrogate for a larger value at the stellar surface. We can constrain the apparent $\ell$ by constraining the allowed pulsar geometry in the model. As presented by Deshpande & Rankin (2001), the subpulses in PSR B0943+10 drift all the way across the pulse window without large phase jumps or even drift-band curvature, which is consistent with the identification of this pulsar as a conal signal pulsar (Rankin 1990). In our model, where the subpulses are generated by a single frequency oscillation, continuous drift implies that our sight line does not cross a nodal line. If we only allow geometries in which our sight line passes through one of the first three nodal regions, as plotted in Figure 10, and if we further exclude the possibility that our sight line passes through the central region of the star (which would be analogous to the core type
classification of Rankin 1990), then \( \ell \) is constrained by the absence of nodal lines in the average profile. For this geometry, the lower and upper bounds on \( \ell \) are 55 and 125, respectively. We have chosen \( \ell \) to be in the midrange of these bounds, 75, in all the fits that follow. In addition to fixing \( \ell \), we fix \( P_1 \) to be the known period of the pulsar. We choose \( \phi_0 \), the closest approach to the magnetic pole, to be near the center of the average pulse profile based on the observations of the pulse profile of PSR B0943+10 at other frequencies (Smits et al. 2006).

Once \( \alpha, \beta, \ell, \) and \( P_1 \) are fixed, we use the following procedure to fit the remaining parameters:

1. Choose trial values for all the parameters based on observed phases and amplitude ratios.
2. Generate a time series in \( I, Q, \) and \( U \) in equations (3), (8), and (9), sampled in the same way as the data.
3. Impose a Gaussian window function on each pulse where the function is \( g = e^{-((\gamma - \gamma_{\text{mean}})^2) / (2\gamma^2)} \), where \( \gamma = 3.25^\circ \) and \( \gamma_{\text{mean}} = 0^\circ \).
4. Calculate the metrics \( I_{\text{model}} - I_{\text{new}}, Q_{\text{model}} - Q_{\text{data}}, \) and \( U_{\text{model}} - U_{\text{data}} \) and repeat until they are minimized.

Because our Fourier analysis in \( \ell \) showed that the frequency and phase of the subpulse wandered slightly over the 816 pulses, we fit the data in 100-pulse segments. We implement the procedure above using gaussfit. Like our analysis of the multiple peaks near the subpulse frequency in \( \ell \), the variance in the data is much higher than the off-pulse instrumental noise, so we use the variance as an error estimate. This artificially forces the reduced \( \chi^2 \) to be near 1, so we have no absolute measure of the goodness of our fits, nor will it ever be possible to do better for this pulsar unless we understand the physics behind the pulse height distribution.

One other complication in our fit is that our model includes no unpolarized component. This does not mean that our model radiation remains fully polarized, but rather that any unpolarized component arises from the superposition of the orthogonal polarization modes, i.e., if \( Q_{\text{model}} \) and \( U_{\text{model}} \) are equal, we have complete depolarization. Deshpande & Rankin (2001) have shown, and we concur, that PSR B0943+10 contains a substantial additional unpolarized component, and the absence of this component from our model makes amplitudes we fit a compromise with reality.

The best results of this fitting process are summarized in Table 2. To be certain the fitted parameters do not depend on how our sight line slices through the nodal region, we repeated these fits for all the values of \( \beta \) shown in Figure 10. We performed a similar experiment by varying \( \phi_0 \). Except for the expected 180° difference in phase between the odd and even nodal regions, the results were the same as what we have tabulated, indicating our fit is robust against changes in geometrical parameters. Because our \( \chi^2 \) is only useful for guiding us to the best fit and is not able to tell us the likelihood that the data are consistent with our model, we cannot claim to have confirmed or ruled out a pulsational model. The value of a quantitative fit is that it allows us to attempt to reproduce the data in an objective way, as opposed to the subjective attempts in Clemens & Rosen (2004). In the next section we show the results of the simulations. Fits to the remaining 100-pulse segments were not substantially different from the ones shown here.

We find that the fitted parameters can be divided into two categories: pulsational and geometrical parameters. The pulsational parameters (e.g., the pulsational period, \( P_2 \), and the spherical harmonic, \( \Psi_{l,m=0} \)) describe the surface variations that are centered on the magnetic pole. The geometrical parameters (e.g., the offset between the rotation and magnetic axes, \( \alpha \), and the distance between the magnetic pole and our line of sight, \( \beta \)) are highly correlated, and it is their ratios rather than their precise values that are significant in matching the observations. Therefore, our fits do not yield the full geometry of the star, and external information has to be applied to know, for example, the inclination of the rotation axis. Fortunately, the geometrical parameters are mostly independent from the pulsational parameters, and thus the degeneracy of the geometric parameters does not prevent us from constraining the pulsational properties, a prerequisite for learning about the physics of pulsating neutron stars. The simple average pulse profile and single frequency of PSR B0943+10 are reflected in a single set of pulsational parameters that would correspond in our model to a single pulsation mode. In the more complex pulsars our model may demand multiple oscillation modes whose pulsational parameters would yield physical properties of neutron stars.

5. SYNTHETIC PULSAR LIGHT CURVES

We now use our pulsational model and the best parameters fitted to the first 100 pulses of PSR B0943+10 to generate and analyze synthetic pulsar light curves. In this way, we learn whether our model has captured the essence of the variations in PSR B0943+10.
As in §3, we begin with the drift band. Figure 12 shows a comparison between the first 100 normalized pulses of the data (left) and synthetic light curves of PSR B0943+10. In the middle panel we have added random noise to the model to increase its variance to be approximately that of the data. Our drift band shows the splitting on the right-hand side of the profile as observed in the star. In the model it comes about because the timelike maxima in the velocity polarization mode are offset from the displacement polarization mode by an amount related to $\psi_{\text{delay}}$. Note that this means that the parallel track has a polarization different from the displacement polarization track, an effect reproduced in our polarization angle histogram, which we discuss later.

In our model the amplitudes and the spatial envelopes of the velocity and displacement polarization modes are also different, because $\Psi_{l,m=0}$ has its maxima at antinodes and $\partial \Psi_{l,m=0}/\partial \theta_{\text{mag}}$ has its maxima at nodes. This is why the velocity polarization mode track does not extend through the whole drift band. This track is also asymmetric, appearing on the right-hand side of the drift band and not the left. In our model this asymmetry arises from an offset between a maximum in the pulse window, $\phi_{\text{mean}}$, and the longitude of the magnetic pole, $\phi_0$. While the difference between $\phi_{\text{mean}}$ and $\phi_0$ is not related to pulsation parameters, it may lead to interesting insight in the pulsar emission mechanism.

We can also make a visual comparison of our pulsational model to the data in Fourier space. We used the model parameters in Table 2 to create 816 pulses matching the resolution of the data. The bottom panel of Figure 13 shows a fully resolved Fourier transform of our simulated data where we have added random noise to our simulation. The Fourier transform of our simulation does not show the splitting in the subpulse frequency, as discussed in §3, because we have created our model only using a single subpulse frequency. To reproduce the closely spaced peaks, our model would require the introduction of additional, incommensurate subpulse frequency or the subpulse frequency would have to wander as a function of time. For comparison, the top panel of Figure 13 shows the fully resolved Fourier transform of the normalized data.

While the unfolded Fourier transform does not reproduce the amplitude of the harmonic structure well, this is not surprising because we are modeling the subpulses, not the low frequency pulse window. Despite this, both the harmonic- and longitude-resolved fluctuation spectra, shown in Figures 14 and 15, respectively, closely reproduce the data. In both figures we show the

![Figure 12](image1.png)

**Fig. 12.** Left: First 100 pulses of PSR B0943+10 folded at $P_1 = 1.8708 P_1$, using the fitted value of $P_1$, compared to the left panel of Fig. 4, where the data are folded at a value of $P_1$ calculated from the Fourier transform. Middle: Our model of the data, using the values of the parameters in Table 2 calculated from gaussfit. We have added random noise to the simulation so the variance in the simulation approximately matches the variance in the data. Right: Our same simulation, without noise. The phase in the bottom panels were calculated using all 816 pulses in both the data and simulations.

![Figure 13](image2.png)

**Fig. 13.** Top: Fully resolved Fourier transform all 816 normalized pulses from the archival data of PSR B0943+10. Bottom: Fourier transform of our simulated data to which we have added random noise. The peaks centered around 30 Hz are the subpulse frequency and its aliases. The harmonic of the subpulse frequency and its aliases are centered around 65 Hz. The values of the parameters used in our model are listed in Table 2.
Fig. 14.—Top panels: Longitude-resolved fluctuation spectra of the first 100 pulses of data. The left-hand side shows the raw data, while the right-hand side shows the data after our normalization process. Bottom panels: Longitude-resolved fluctuation spectra of 100 pulses of simulated data using the values of the parameters listed in Table 2 (left) and with the addition of noise (right). The contour levels are in mJy$^2$.

Fig. 15.—Same as Fig. 14, but for harmonic-resolved fluctuation spectra.
spectra of first 100 pulses of data that were fit using *gaussfit*, both unaltered and normalized. The bottom panels show the fluctuation spectra from our simulation with and without noise.

We can further compare our model to the data using plots similar to those in Stinebring et al. (1984a, 1984b). We recreate plots of the polarization angle, polarization fraction, and total intensity. Figure 16 shows the normalized data of all 816 pulses in the left panel and all 816 pulses from our simulation in the right panel. We have added noise to our simulation in the same way as we did in the drift-band plot in Figure 12. Our model pulse shape reflects the Gaussian window imposed and does not match the data in the wings of the profile. Our polarization fraction is more tightly grouped at 0% and 100%, in part because we do not include an unpolarized component in our model, as we discussed in §3. Our velocity polarization mode is weaker than the corresponding track in the angle histogram of the data. Nonetheless, Figure 16 represents a dramatic improvement in our ability to model pulsar data. It is based on an objective fit that was not constrained by the quantities in this plot but by the light curve directly. The striking similarities between our model and the data give us confidence that our pulsation model can reproduce significant features in PSR B0943+10.

6. SUMMARY AND CONCLUSIONS

Our objective in this paper has been first to fit single pulse data of PSR B0943+10 using a pulsational model and, second, to create simulated pulsar light curves from the best fit. The model we use was developed in Clemens & Rosen (2004) and in a companion paper to this one (Clemens & Rosen 2008). It is founded on pulsational displacements and their associated surface velocities, which result in the emission of distinct polarization modes orthogonal to each other.

We chose PSR B0943+10 as an initial target for our study because it has a simple morphology and is well studied in the literature. We were surprised in our analysis of published data to find undocumented behaviors in the form of drift-band splitting, frequency splitting, and phase wandering of the subpulses. The drift-band splitting is a stable feature of the data, while the frequency and phase wandering are more erratic and in need of further observational study. The model we use is based on a single pulsational frequency and cannot reproduce the erratic behavior, but it has been able to reproduce the drift-band splitting.

Our fitting exercise has been successful; our least-squares fitting algorithm converged to a solution for an amplitude normalized data set from which we had removed circular polarization. The fitted parameters for this converged model represent an objective application of our pulsation model to PSR B0943+10. We have used them to create synthetic data that reproduce the stable features of the observations, including the pulse shape, the polarization angles, and the orthogonal mode morphology. Our model cannot reproduce the pulse amplitude distribution, and therefore we cannot quantitatively access the validity of our model beyond its ability to reproduce the more stable features of the data.

If we accept a pulsational model for the drifting subpulses in PSR B0943+10, then some of our fitted parameters have physical significance. $P_2$ represents the eigenperiod of a single pulsation mode. The spherical harmonic degree of this mode, $\ell$, is unknown for reasons we have explained. The amplitudes we measured and the value of $\ell$ are not meaningful without better understanding of the radio emission mechanism. The fitted quantity $\psi_{\text{delay}}$ reflects
an adjustment between the phase of the pulsational displacements and pulsational velocities, which would be 90° for an adiabatic pulsation. Positive values of $\psi_{\text{delay}}$ mean that maximum heating follows maximum compression and requires some mechanism for delaying the flux changes (van Kerkwijk et al. 2000). Curiously, the sign and value of $\psi_{\text{delay}}$ in our fits to PSR B0943+10 are similar to the measured value for pulsating white dwarfs in which the flux is delayed by a surface convection zone (Goldreich & Wu 1999). Our value of $\psi_{\text{delay}}$ is consistent with the identification of the displacement polarization mode as a surface flux phenomena related to pulsations.

From an astroseismological perspective, PSR B0943+10 is not the most interesting target for study. The simplicity that makes it attractive for testing our model limits the useful physics that can be extracted. There are other pulsars, e.g., PSR B0031−07, that show multiple subpulse frequencies and others that appear to show correlated oscillations between the magnetic poles (Weltevrede et al. 2007). It will be an interesting challenge to apply our pulsational model to objects like these. Edwards (2006) has proposed the phase behavior of subpulses as a critical test of the drifting-spark model, and his argument applies equally well to our model. It is possible that broader studies of more complex pulsars will be able to rule out nonradial oscillations as the origin of their subpulses, but for now, PSR B0943+10 remains an encouraging example of a pulsar whose subpulse behavior can be reproduced by stellar pulsation model.

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