Solar east-west flow correlations that persist for months at low latitudes are dominated by active region inflows

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ABSTRACT

\textbf{Context.} Giant-cell convection is believed to be an important component of solar dynamics. For example, it is expected to play a crucial role in maintaining the Sun’s differential rotation.

\textbf{Aims.} We reexamine early reports of giant convective cells detected using a correlation analysis of Dopplergrams. We extend this analysis using 19 years of space- and ground-based observations of near-surface horizontal flows.

\textbf{Methods.} Flow maps are derived through the local correlation tracking of granules and helioseismic ring-diagram analysis. We compute temporal auto-correlation functions of the east-west flows at fixed latitude.

\textbf{Results.} Correlations in the east-west velocity can be clearly seen up to five rotation periods. The signal consists of features with longitudinal wavenumbers up to $m \approx 9$ at low latitudes. Comparison with magnetic images indicates that these flow features are associated with magnetic activity. The signal is not seen above the noise level during solar minimum.

\textbf{Conclusions.} Our results show that the long-term correlations in east-west flows at low latitudes are predominantly due to inflows into active regions and not to giant convective cells.

\textbf{Key words.} Sun: helioseismology – Sun: oscillations – Sun: interior – Sun: activity – waves – convection

1. Introduction

The transport of heat from the solar interior to the surface is achieved through convective motions in the outer 30\% of the Sun. From observations, we currently know of two well-defined scales of convection: granulation and supergranulation. Granulation cells have lifetimes of $\sim 10$ min, have scales of 1–2 Mm (spherical harmonic degree $\ell \geq 1000$), and are well reproduced by numerical studies (see review by Nordlund et al. 2009). Supergranulation is a larger scale of convection at $\sim 30$ Mm ($\ell \sim 120$), with lifetimes of one to two days and a depth structure that is debated (see Rincon & Rieutord 2018, for a review). The nature of convective motions at larger scales ($\ell < 100$) is not well understood, with a number of studies reporting different amplitudes (see review by Hanasoge et al. 2016) and different geometries (Beck et al. 1998; Hathaway et al. 2013). Here, we reexamine the earliest reported detection of giant convective cells by Beck et al. (1998), who used a 16-month time series of Dopplergrams from space, by utilizing the subsequent two decades of observations.

In examining surface Doppler images from the Mount Wilson Observatory, Labonte et al. (1981), and later Snodgrass & Howard (1984), concluded that there was no giant cell signal above 1–10 m s$^{-1}$ per wavenumber. It was not until the availability of data from the Michelson Doppler Imager (MDI) on board the Solar and Heliospheric Observatory spacecraft that Beck et al. (1998) detected long-lived large-scale velocity features. These features were found to have an e-folding lifetime of 126 days and to be highly elongated in longitude (up to $44^\circ$), with an aspect ratio $> 4$ on the surface. Beck et al. (1998) attributed this signal to the presence of giant convective cells. This result has been in contradiction with many numerical studies, which suggest that large-scale convection should form “banana cells” that are elongated in latitude (e.g., Miesch et al. 2008). Ulrich (2001) disputed the convective nature of the observed east-west flow correlations and proposed that they are Rossby or inertial waves.

Recent studies based on granulation tracking, supergranulation tracking, or local helioseismology have identified several components of near-surface large-scale motions: (1) surface inflows into active regions (Gizon et al. 2001; Gizon 2004; Haber et al. 2004); (2) vortical motions due to equatorial Rossby waves below $\sim 30^\circ$ latitude with longitudinal wavenumbers $m < 15$ (Löptien et al. 2018; Hanasoge & Mandal 2019; Liang et al. 2019; Proxauf et al. 2020; Hanson et al. 2020); and (3) east-west velocity features with $m = 1$ above 50$^\circ$ latitude (Hathaway et al. 2013; Bogart et al. 2015; Hathaway & Upton 2020).

While the aforementioned studies focused on imaging distinct patterns of motion, a number of other studies have focused on constraining the amplitude of the flows. Hanasoge et al. (2012) used local helioseismic techniques to measure the horizontal flow velocities at a depth of $0.96 R_\odot$, finding upper limits on the amplitudes on the order of 0.1 m s$^{-1}$ per mode. This suppression of power at large scales compared to numerical simulations is puzzling (see Gizon & Birch 2012, for a commentary), considering that numerical studies had anticipated amplitudes
two orders of magnitudes larger (Miesch et al. 2008). Further complicating the debate, Greer et al. (2015) used a different local helioseismic technique and reported amplitudes larger than numerical studies for large-scale features ($\ell < 30$), but comparable amplitudes at smaller scales ($\ell > 30$). A careful assessment of the issues involved in these studies is available in Proxauf (2020).

In this study, we will address the observations of Beck et al. (1998). With an additional 19 years of observations (spanning nearly two solar cycles, see Fig. 1), we sought to differentiate signals associated with magnetic activity from those associated with non-magnetic processes (Rossby waves and possibly thermal convection). We find that the signals identified by Beck et al. (1998) are very likely due to the active region inflows, even near solar minimum.

2. Data analysis and results

We used the horizontal flow maps derived from observations of the ground-based Global Oscillation Network Group (GONG, Harvey et al. 1996) and the space-based Helioseismic and Magnetic Imager (HMI, Schou et al. 2012). From the former, we utilized 19 years of horizontal flows derived from the GONG++ ring-diagram analysis (RDA) pipeline from 2001–2020\(^1\). From the latter instrument, we used eight years of RDA data from 2010–2018\(^2\) and six years (2010–2016) of surface horizontal flow data derived from the local correlation tracking (LCT) of granules by Löptien et al. (2018)\(^3\). For both RDA data sets, we used the flows derived from 15° tiles. To remove small-scale features from the LCT maps, we smoothed them by convolving with a Gaussian of $\sigma = 7.2\text{°}$.

In total, we have three independent data sets. Corbard et al. (2003) details the GONG++ RDA pipeline, Bogart et al. (2011a,b) details the HMI RDA pipeline, and details of the LCT method are found in Welsch et al. (2004) and Fisher & Fisher & Welsch (2008). For each of these data sets, we have both of the horizontal flow components: in the direction of rotation $u_x$ and in the direction of the solar north pole $u_y$. We corrected for center-to-limb effects through the method described by Liang et al. (2019). For clarity in terminology, we refer to $u_x$ or $u_y$, as the measured flow within a single tile, and east-west (EW) or north-south (NS) as the flow signals averaged over longitude or temporal windows.

2.1. Coordinate system

We examine the flows in the Stonyhurst coordinate system, whereby the frame is rotating at the Earth’s orbital frequency. In this system, the latitude $\lambda$ and longitude $\phi$ are zero at the intersection of the equator and the central meridian on the visible disk. The latitude increases in the direction of the solar north pole, while the longitude increases in the direction of solar rotation. Accordingly, the longitudinal flows $u_\phi$ are positive if in the direction of rotation, and the latitudinal flows $u_\lambda$ are positive if in the direction of the north pole. The flow components are computed from small patches on the solar disk, assuming a Cartesian geometry $(x,y)$. As such, the approximations $u_x = u_\phi$ and $u_y = u_\lambda$ are taken within each small patch. In this study, we also analyze the horizontal divergence $\nabla \cdot u$ in spherical geometry, where $\nabla \cdot u = (r \cos \lambda)^{-1} \partial_r (r \cos \lambda \partial_r u_\phi) + \partial_\lambda u_\phi$.

2.2. Auto-correlation functions

Following Beck et al. (1998), we computed the auto-correlation functions of the flow maps. For each time segment of duration $N_t = 720$ days, we considered the centered and normalized velocity components

$$\bar{u}_i(\lambda, \phi, t) = \frac{u_i(\lambda, \phi, t) - \mu_i(\lambda, \phi)}{\sigma_i(\lambda, \phi)},$$

where $\mu_i(\lambda, \phi)$ is mean value of $u_i$ over the 720-day window and $\sigma_i(\lambda, \phi)$ is standard deviation. The longitude-averaged auto-correlation of the time series $\bar{u}_x$ at fixed latitude at time lag $\tau$ is

$$\mathcal{A}_{\text{EW}}(\lambda, \tau) = \frac{1}{N_t N_\phi} \sum_\phi \sum_\tau \bar{u}_x(\lambda, \phi, t) \bar{u}_x(\lambda, \phi, t + \tau).$$

The auto-correlation of $\bar{u}_x$ is

$$\mathcal{A}_{\text{NS}}(\lambda, \tau) = \frac{1}{N_t N_\phi} \sum_\phi \sum_\tau \bar{u}_x(\lambda, \phi, t) \bar{u}_x(\lambda, \phi, t + \tau).$$

Long-lived features will appear as peaks in the auto-correlation at integer values of the local rotation period. The auto-correlations were computed using a temporal window of $\sim 720$ days. Due to the different cadence of each of the data sets, this results in $N_t = 633$ for each window in the RDA data and $N_t = 720$ for the LCT data. For each consecutive sample, the window was shifted by $N_t/2$. In order to improve the signal-to-noise ratio, we then took the mean across longitude and temporal samples. When comparing LCT and RDA, we used the RDA auto-correlation for flows at a depth of $\sim 2$ Mm, which is the shallowest depth from the GONG++ full disk flow map pipeline\(^4\). Comparison between HMI RDA and LCT is possible at the surface, though Proxauf et al. (2020) notes that HMI RDA surface measurements are unreliable. The HMI RDA flow inversions that target the surface appear to have excess power at low $m$, relative to

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\(^1\) https://gong.nso.edu/

\(^2\) http://jsoc.stanford.edu/

\(^3\) Data available upon request.

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Fig. 1. Monthly sunspot number (SSN) for cycles 23 and 24. The observation period of the present study and references herein are highlighted for reference.
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Figure 2 shows the auto-correlations of the EW flows as a function of time lag and latitude. These results are consistent with Beck et al. (1998), showing a long-lived EW flow signal that persists for at least five solar rotations and can be seen at latitudes up to 45°. Due to the latitudinal differential rotation, the peaks in $A$ are at greater time lags at higher latitudes. In each data set, the peak in $A$ is preceded and followed in time by negative correlation sidelobes.

Figure A.2 shows similar results for the NS flow signal. However, in order to obtain such clear results, the solar Rossby waves (Löptien et al. 2018) were filtered out (see Appendix A). While similar to the EW signal, the NS signal seems to become consistent with noise within four rotations.

2.3. Spectral analysis

Here, we examine the spectral components of the EW auto-correlation in a similar manner to Ulrich (2001). For this section, we present the results for HMI RDA. The results of the other data sets are similar.

We performed a temporal Fourier transform on the EW auto-correlation function (Eq. (2)),

$$\hat{A}_{EW}(\lambda, \omega) = \sum_{\tau} A_{EW}(\lambda, \tau) e^{i\omega \tau},$$

where the sum is taken over all time lags $\tau$ and $\omega$ is the angular frequency. In the Stoneyhurst frame, a flow pattern proportional to $e^{i m \phi}$, which is rotating at the rotation rate $\Omega_\lambda$, will have a frequency $m \Omega_\lambda$, where $\Omega_\lambda$ is the local rotation rate at latitude $\lambda$. We have corrected the data for the Earth’s orbital frequency, 31.7 nHz.

Figure 3 shows $|\hat{A}_{EW}|^2$ as a function of frequency and latitude, at a depth of 2 Mm. These results show that $A_{EW}$ has components up to $m = 9$, after which the signal is consistent with noise. At higher latitudes ($\lambda > 40^\circ$), modes on the order of $m > 6$ are not seen. The mode $m = 2$ at the equator has the greatest amplitude. In the frequency domain, the deflections with latitude seen in Fig. 2 are toward lower frequencies and are qualitatively consistent with differential rotation.

2.4. Mode parameters of the EW signal

In this section, we characterize the modes that contribute to the long-lived EW signal. We computed the power spectrum $P(\lambda, \omega)$ of the EW flow maps for the entire HMI observation time

$$P(\lambda, \omega) = \left| \sum_\phi \sum_{\tau} u_{\lambda}(\lambda, \phi, t) e^{i \omega \tau} \right|^2,$$

where the sum over $\phi$ is over the range $-45^\circ \leq \phi \leq 45^\circ$ for each snapshot. The resulting spectra are similar to Fig. 3 but have an exponential probability distribution function. The azimuthal modes $m$ at each latitude $\lambda$ were fit assuming a Lorentzian line profile,

$$L_m(\omega, \Lambda_m) = \frac{A_m}{1 + (\omega - \omega_m)^2/(\Gamma_m/2)^2} + B_m,$$

where $A_m$ is the mode amplitude, $\omega_m$ is the mode frequency, $\Gamma_m$ is the full width at half maximum, $B_m$ is the background noise of the spectrum, and $\Lambda_m = \{A_m, \omega_m, \Gamma_m, B_m\}$. We fit only up to $m = 6$ due to poor a signal-to-noise ratio for higher modes. Details of the fitting can be found in Appendix B. We opted to
Fig. 3. Plot of $|\hat{A}_{\text{EW}}(\lambda, \omega)|^2$ for HMI RDA as a function of latitude and frequency. The frequency is scaled by the rotational frequency at the equator $\Omega_{\text{eq}} (2\pi/27.275 \text{ days})$. At the equator, power can be seen near integer values of $\omega/\Omega_{\text{eq}}$, up to values of about nine. At latitudes above 30 deg, these higher-order modes are not visible.

Fig. 4. Difference between the rotational frequencies of the long-lived EW signals in HMI RDA and the synodic frequencies of Doppler features ($\Omega_D$, top) or magnetic features ($\Omega_M$, bottom). The Doppler and magnetic rotational frequencies are summarized in Snodgrass & Ulrich (1990). Error bars are one sigma.

fit each individual mode $m$ using a frequency window of width $\Delta \omega / \Omega_{\text{eq}} = 1$.

Having fit the power spectrum of the EW flows, we compared the rotational frequencies to those derived from Doppler and magnetic features (Snodgrass & Ulrich 1990). Snodgrass & Ulrich (1990) identified magnetic features from rebinned Mount Wilson magnetograms, with each bin covering approximately $6^\circ$ in latitude. This coarse grid smooths out small-scale activity and should be dominated by strong magnetic features. As such, the derived rotation rates are consistent with sunspot rotation rates below $40^\circ$ latitude (see review in Beck 2000). Figure 4 compares these rotation rates to the fit rotation frequencies of the HMI RDA EW power spectrum. These results show a mismatch between the frequencies of the EW signal and Doppler features (supergranules). However, when compared to the magnetic tracers, the EW frequencies are similar at all latitudes. This indicates that the EW flow rotation rate is consistent with the rotation rate of magnetic regions. These results are also in agreement with the HMI LCT and GONG++ RDA data sets.

Figure 5 shows the other fit parameters of the HMI RDA EW spectrum. The measured lifetimes ($2/\Gamma_m$) show that, within error estimates, the modes at all latitudes have a lifetime of 4–5 months. This is consistent with Beck et al. (1998). In the HMI RDA flow maps, the signal-to-noise ratio $S/N (A_m/B_m)$ is near or above parity (up to six) for modes from $m = 2$ up to $m = 5$. Finally, we compared the mean background power ($\sum_{m=-1}^{6} B_m/5$) in the three data sets and find values of $\sim 1–1.2 \text{ m s}^{-1}$, with the greatest power at the active latitudes. This gives a limit on the sensitivity of detecting giant flow cells.

2.5. Relation to activity

The results in the previous sections confirmed the presence of long-lived signals in the near-surface horizontal flows in three independent data sets. However, these results do not confirm the convective or magnetic nature of such flows. Rossby waves rotate retrogradely at fractional values of $\omega/\Omega_{\text{eq}}$ (in the Stoneyhurst frame) and hence can be filtered from the data (see Appendix A). At such large scales and long lifetimes, there remain a few potential causes for the origin of these signals: giant-cell convection, active region inflows, or a side effect of magnetic activity. With the long observation window of...
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Fig. 6. Top panels: average flows around convergent features during the solar maximum of cycle 23. Each of the first three rows is a different component of the flow, from top to bottom: EW flows, NS flows, and divergence maps. The fourth row is the average unsigned magnetic field. Each column is the mean map at an integer rotation before and after the initial identification (0 CR, middle column). Bottom-left panel: monthly sunspot number as a function of year (blue), with the center of the one-year observation window overplotted (vertical black line) and window size shown (shaded gray region). Bottom-right panel: spatial correlation coefficient between the CR = 0 flow maps and the flow maps at time lag CR (EW: blue; NS: orange; $\nabla_h \cdot u$: green).

GONG++, we will exploit the data of two solar cycles to shed light on the nature of these flows.

To investigate this, we first performed further processing on the flow data sets and filtered out modes greater than $\omega/\Omega_1 = 8$ and less than $\omega/\Omega_1 = 1$. This was done with a filter that has a flat top from $\omega/\Omega_1 = 1$ to $\omega/\Omega_1 = 8$ and is tapered to zero by a cosine bell that goes from 1 to zero in a distance $\Delta \omega/\Omega_1 = 1$. This filter was applied to both negative and positive frequencies, and the flow maps were then reconstructed into the time domain. Applying such a filter to the data introduces some smoothing and removes frequencies lower than the equatorial rotation frequency.

For comparison with magnetic activity, we tracked MDI and HMI magnetograms, which overlap with the GONG++ data presented here. We utilized the tracking routine mTrack\(^5\), which is the same routine used to track Doppler images for the HMI RDA pipeline. The magnetograms were sampled at a cadence of 96 min. and tracked for $1/24$th of a Carrington rotation (CR), resulting in 17 magnetograms per data cube. Each magnetogram was projected onto a cylindrical equidistant map, tracked at the Snodgrass magnetic tracer rotation rate (Snodgrass & Ulrich 1990), and subsampled to a resolution of $0.6^\circ$ per pixel.

The flow and magnetic maps were collected into data cubes spanning one year in time. We then identified converging flow features with amplitudes $-\nabla_h \cdot u > 3 \times 10^{-8} \, s^{-1}$, considering these signals to be above the background noise in the GONG++ divergence data. Here, we focused on the convergent features since the analysis of the resulting maps is intuitive when considering magnetic activity. We performed the same analysis on EW and divergent flows, finding similar results. At fixed time, among the converging flow features that are within $30^\circ$ of one another, we selected the largest amplitude feature to avoid counting a large converging flow feature multiple times. Additionally, features outside of $40^\circ$ from disk center were rejected to avoid the disk-edge. On average, we identified 850 features within each one-year window. For each identified feature, the flow and magnetic maps were shifted such that the feature is positioned at image center. A mean of the images was then computed.

\(^5\)http://hmi.stanford.edu/rings/modules/mtrack.html
order to determine the evolution of these flows, this procedure was repeated at integer (up to 3) CRs before and after the initial identification. Differential rotation $\Omega_M^\lambda$ was accounted for in this averaging routine.

Figures 6 and 7 show these stacked maps for the solar maximum of cycle 23 and the minimum between cycles 23 and 24, respectively. For better visualization, we linearly interpolated the RDA flow maps to improve resolution by a factor of four. In both figures, large-scale flows can be observed at the initial identification time ($CR = 0$). During the maximum, the EW and NS flows extend $30^\circ$ (from center) in the longitude direction and $15^\circ$ in latitude. These flow patterns are seen at least 3 CR before and after initial identification. The computed spatial correlation coefficients show that the flows have a correlation of $\sim 0.45$ with the initial identified features at $+3$ CR. The corresponding magnetic maps show that these flows are related to activity. We have not included the vorticity here as the Rossby waves and Rossby filters make it difficult to obtain a clean signal at higher latitudes (see Appendix A for a discussion).

During the minimum, the identified flows extend approximately $15^\circ$ from the center. The associated magnetic maps show very weak magnetic fields. Unlike the maximum, the identified flows in the minimum do not persist beyond a local rotation period. This is further evidenced in the correlation coefficients, which drop to near zero by $\pm 1$ CR. The maps of the flows at integer rotations share no visual correlations with the identified feature.

Figure 8 shows the spatial correlation of the flow maps as a function of CR and time. These results further demonstrate that these large-scale flows persist for a number of rotations during solar maximum. As the solar cycle begins its descending phase, the flows persist for shorter periods of time. At the solar minima, 2008 and 2019, the flows persist for no more than one rotation.

3. Discussion and conclusions
We have investigated the long-lived large-scale EW flow signal reported by Beck et al. (1998). Using nearly 20 years of flow maps derived from observations by both GONG and HMI, we confirm the presence of this signal in the near-surface horizontal flows. However, with careful comparisons with magnetic field maps, we find that these signals are consistent with magnetic active region inflows. Specifically, the rotation rate of these signals is consistent with the measured rotation rate of magnetic features (Snodgrass & Ulrich 1990). Furthermore, we find that there are no long-lived features, above the noise level, observed to persist beyond a rotation during the solar minimum.
Beck et al. (1998) observed giant long-lived velocity features and attributed these features to convection. This conclusion was based on the assumption that their data set, consisting of MDI Dopplergrams from 1996 to 1997, would not be significantly contaminated by magnetic activity during a minimum. While active region inflows were not discovered at this time, the authors were still aware that magnetic activity would affect the results. Nevertheless, during this observation window, cycle 23 was ascending and the monthly sunspot number was growing from 10 to 45 (see Fig. 1). We investigated a similar period in cycle 24 (2009, one year after Fig. 7) and found inflows into active regions that persisted for a number of months. These inflows are at a weaker level than in Fig. 6 and, after three rotations, have a spatial correlation coefficient of 0.25. The EW component of these inflows is what is observed in the correlation functions of Beck et al. (1998). We note that there is a very small window (one year centered at the minimum) in which inflows do not dominate the low-latitude flow field. Interestingly, the studies of Labonte et al. (1981) and Snodgrass & Howard (1984) did not report on these long-time correlations, despite the data covering cycles 20 and 21 and the supposed sensitivity of ~1 m s\(^{-1}\) per wavenumber.

The reported observations from Beck et al. (1998) are consistent with ours, with the exception that they incorrectly identified the large velocity cells as convective in origin. Specifically, the EW correlations have an e-folding lifetime of approximately four months, and the extent of the flows is greater in the longitude direction. Beck et al. (1998) reported EW flows elongated in longitude with an aspect ratio greater than four. We also find EW flows elongated in longitudes, however with an aspect ratio around two (Fig. 6). We attribute this difference to the averaging scheme. We were careful to select large converging flow features multiple times to avoid smearing, thus our aspect ratio is smaller. The elongation of the spatial correlation of EW flows in the longitudinal direction is not surprising given the geometry of active regions.

In an argument against the convective nature of these cells, Ulrich (2001) suggested that the long-lived signals in the EW flows might be due to Rossby or inertial waves. The solar equatorial Rossby waves, discovered and subsequently characterized, are not seen in the EW flow maps at low latitudes and thus cannot cause the signals originally observed by Beck et al. (1998) and Ulrich (2001). We find that the active region inflows generate a flow pattern with azimuthal modes up to \(m = 9\), which is consistent with the spectral analysis of Ulrich (2001). Ulrich (2001) found that while the rotational frequencies of these modes were consistent with magnetic features, a couple modes were slower. However, in our study, we find that, at each latitude, all modes have frequencies consistent with the rotational frequency of magnetic features and that they are absent during the solar minimum, thus indicating a magnetic association for these signals. Aware of the potential affect in Doppler images, Ulrich (2001) defined quiet Sun as latitudes without magnetic activity. When studying solar dynamics, the definition of quiet Sun should be extended to exclude the latitudes that contain the inflows around active regions (Gizon et al. 2001). We have shown that the inflows extend 15° in latitude from the center of active regions.

Proxauf (2020, his Fig. 4.2) reported that during periods of high magnetic activity, the EW flow amplitude at the surface increases by ~3 m s\(^{-1}\) for \(f < 10\). He attributed this change in the power spectrum to inflows into active regions, which is consistent with the present study.

Meanwhile, Hanasoge et al. (2020) examined toroidal flow power as a function of depth, latitude, and wave number and compared observations to simulations. These authors showed that the observed toroidal flow is higher in low-latitude regions than at high latitudes, which is in contrast to simulations. We note that the analysis of Hanasoge et al. (2020) was performed from 2010 to 2018, covering cycle 24 (see Fig. 1). Our analysis shows that large-scale active region inflows will contribute to the toroidal (EW) component of the flow field during this period.

In future work, we intend to extend this study beyond 50° latitudes in order to investigate the dependency of the high-latitude flows (Hathaway et al. 2013; Bogart et al. 2015; Hathaway & Upton 2020) with solar activity.

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Appendix A: North-south auto-correlation maps

![Graph of Power spectra of GONG++ flow components and synthetic Rossby waves in the Stoneyhurst coordinate system at the equator. From the top to bottom panels are: the $u_y$ flows, radial vorticity, and $u_r$ component of synthetic Rossby waves (e.g., Hanson et al. 2020). Modes rotating at frequencies of multiples of $\Omega_{eq}$ are indicated by the vertical gray lines. The vertical red lines show the rotational frequencies of the sectoral Rossby waves. We have considered only the $m > 2$ sectoral Rossby waves seen in the Sun.](image1)

Fig. A.1. Power spectra of the GONG++ flow components and synthetic Rossby waves in the Stoneyhurst coordinate system at the equator. From the top to bottom panels are: the $u_y$ flows, radial vorticity, and $u_r$ component of synthetic Rossby waves (e.g., Hanson et al. 2020). Modes rotating at frequencies of multiples of $\Omega_{eq}$ are indicated by the vertical gray lines. The vertical red lines show the rotational frequencies of the sectoral Rossby waves. We have considered only the $m > 2$ sectoral Rossby waves seen in the Sun.

It is well established that Rossby modes are present within the HMI LCT (Loétié et al. 2018), HMI RDA (Proxauf et al. 2020), and GONG++ RDA (Hanson et al. 2020) data. In the Stoneyhurst frame, Rossby waves will have a frequency of $\omega = \Omega_{eq}(m - 2/(m + 1))$. Without accounting for Rossby waves, the Rossby modes will contribute to the $u_y$ and radial vorticity maps. Figure A.1 shows how the Rossby waves and the long-lived modes of this paper are both present in the NS and vorticity spectra. In order to remove the Rossby components, we remapped the flows onto a longitude-latitude frame that rotates at the equatorial rotation rate ($\Omega_{eq}/2\pi = 453.1$ nHz). The flow maps within the equatorial rotation frame $u_{eq} = \{u_{eq}^x, u_{eq}^y\}$ were then Fourier transformed in longitude and time,

$$u_{eq,m}^r(r, \theta, \omega) = \int_0^T \int_0^{2\pi} u_{eq}^r(r, \theta, \phi, t) \exp(-im\phi+i\omega t) \, d\phi, \quad (A.1)$$

where $T$ is the total observation time of the respective data set. A cosine bell filter was then applied to each $m$, centered on the theoretical dispersion $\omega_r = -2\Omega_{eq}/(m + 1)$ with a tapering length of 35 nHz. The width $W(m)$ of the filter follows,

$$W(m) = \begin{cases} 75 \text{ nHz} & m \leq 10, \\ 50 \text{ nHz} & m > 10, \end{cases} \quad (A.2)$$

where the width narrows for $m > 10$ in order for the filter to not cross the zero frequency. We performed this filtering on $\pm 3$ aliases from the central ridge (which arise from the window function), as well as in the $-m$ half-space.

With the solar equatorial Rossby modes removed, we reconstructed the flow maps by performing the inverse Fourier transform. The maps were then remapped onto the Stoneyhurst coordinate system. Figure A.2 shows the NS auto-correlation maps after the removal of the Rossby waves.

Fig. A.2. Top: temporal auto-correlation function of $u_y$ for HMI RDA, LCT, and GONG++ RDA. A signal can be seen to persist for at least four solar rotations. Statistical error bars are comparable to line thickness and hence are not shown. Bottom: 2D map of the $u_y$ auto-correlation maps, as a function of latitude and time lag. The deflection with latitude is consistent with differential rotation. The dashed vertical lines in each image indicate an integer equatorial rotation.

While this filtering works near the equator, at higher latitudes the signal of the long-lived flows also becomes filtered. This is because the Rossby wave frequencies are independent of the differential rotation (latitude), while the advected flows are not. At higher $m$, the flow frequencies cross over into the

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Rossby mode filters at mid to high latitudes. This explains the lack of deflection in Fig. A.1. The vorticity of large-scale flows changes sign at the equator. The inflow vorticity is weaker than the divergence (factor 5, Hindman et al. 2009) and comparable to the Rossby waves. This means that, with or without filtering, obtaining a mean image of the inflow vorticity through our methodology (in the Northern or Southern Hemisphere) is difficult.

Appendix B: Mode fitting methodology

The estimates on the parameters $\Lambda_m$ in Eq. (6) are determined by minimizing the negative of the log-likelihood function,

$$J_{\lambda,m}(\Lambda_m) = \sum_{k=1}^{K} \ln L_{\lambda}(\omega_k, \Lambda_m) + P(\lambda, \omega_k)/L_{\lambda}(\omega_k, \Lambda_m), \quad (B.1)$$

where $\omega_k$ is the $k$th frequency bin within the frequency window from $(m - 1/2)\Omega \lambda$ to $(m + 1/2)\Omega \lambda$ consisting of $K$ frequencies.

The minimization of $J_{\lambda,m}$ and the estimates on the error of the fits were computed from samples drawn by Markov chain Monte Carlo (MCMC) methods. The priors on the frequencies are $\omega_m \in [m\Omega \lambda - \Omega \lambda/2, m\Omega \lambda + \Omega \lambda/2]$, where $\Omega \lambda$ is specified by the Snodgrass rotation rate (Snodgrass & Ulrich 1990). The priors on $A_m$ and $B_m$ are such that the MCMC samples were drawn from the set $[0, 5 \text{ max } P(\lambda, \omega)]$. The prior on the line width are $\Gamma_m \in [2(\omega_{k+1} - \omega_k), \Omega \lambda/2]$. We utilized the python library emcee (Foreman-Mackey et al. 2013) with 100 walkers and 1000 steps. The first 200 steps were taken as burn-in, and we computed the 16th, 50th, and 84th percentiles from the remaining 800 steps. We verified these results with the semi-analytical method in Toutain & Appourchaux (1994). The results obtained through MCMC are presented here since the error calculations are more numerically stable for fits at higher $m$ and $\lambda$. 