Bivariate Instantaneous Frequency and Bandwidth

Jonathan M. Lilly, and Sofia C. Olhede,

Abstract

The generalizations of instantaneous frequency and instantaneous bandwidth to a bivariate signal are derived. These are uniquely defined whether the signal is represented as a pair of real-valued signals, or as one analytic and one anti-analytic signal. A nonstationary but oscillatory bivariate signal has a natural representation as an ellipse whose properties evolve in time, and this representation provides a simple geometric interpretation for the bivariate instantaneous moments. The bivariate bandwidth is shown to consists of three terms measuring the degree of instability of the time-varying ellipse: amplitude modulation with fixed eccentricity, eccentricity modulation, and orientation modulation or precession. A application to the analysis of data from a free-drifting oceanographic float is presented and discussed.

Index Terms

Amplitude and Frequency Modulated Signal, Analytic Signal, Instantaneous Frequency, Instantaneous Bandwidth, Multivariate Time Series.

I. INTRODUCTION

The representation of a nonstationary real-valued signal as an amplitude- and frequency-modulated oscillation has proven to be a powerful tool for univariate signal analysis. With the analytic signal [1] as the foundation, the instantaneous frequency [2], [3] and instantaneous bandwidth [4] are time-varying quantities which measure, respectively, the local frequency content of the signal and its lowest-order local departure from a uniform sinusoidal oscillation. The instantaneous quantities are intimately connected with the first two moments of the spectrum of the analytic signal, and indeed can be shown to decompose the corresponding frequency-domain moments across time.

In the past few years there has been a great deal of interest in nonstationary bivariate signals. A variety of methods have been proposed for their analysis, including: examining pairs of time-frequency representations [5], [6], i.e. the dual frequency spectrum, the Rihaczek distribution, and relatives; creating suitable coherences based on these objects [7]; a bivariate version of the empirical mode decomposition [8]; local analysis of wavelet time-scale ellipse properties [9]; and an extension of wavelet ridge analysis for modulated oscillatory signals [10], [11] to the bivariate case [12]. The statistical and kinematic properties of stationary bivariate signals have also been recently revisited in a number of studies [13]–[16].

As an example of the importance of bivariate nonstationary signals, Fig. 1a shows the trajectory of a freely drifting oceanographic instrument [17] as it follows the ocean circulation for hundreds of kilometers. Acoustically-tracked subsurface “floats” such as this one [18] are a prominent data source for studying the physics of the turbulent ocean. There exist hundreds of such records from all over the world, and their analysis constitutes an important and active branch of oceanographic research.

The analysis of this data will illustrate the great utility of extending the concept of instantaneous moments to the bivariate case.

The goal of this paper is to extend the main tools for analyzing univariate modulated oscillations—the instantaneous moments—to the bivariate case. A key ingredient for the analysis of a multivariate signal is the specification of a suitable structural model, the purpose of which is to condense the information from a set of disparate signals into a smaller number of relevant and intuitive parameters. Work thirty years ago in the oceanographic literature [19]–[23] demonstrated the utility of considering a stationary bivariate signal to be composed of a random ellipse at each frequency. Here the recently introduced notion of a modulated elliptical signal [12] will be used as the foundation for our analysis.

The structure of the paper is as follows. The univariate instantaneous moments are reviewed in Section II-A Section III introduces the modulated elliptical signal, and expressions for pairs of instantaneous moments are derived. This suggests, for a multivariate signal, defining a joint instantaneous frequency and a joint instantaneous bandwidth which integrate, respectively, to the frequency and bandwidth of the average of the individual Fourier spectra. The joint instantaneous moments are derived for general multivariate signals in Section IV, and the bivariate case is then examined in detail. These quantities are shown to be independent of unitary transformations on the analytic signals, including in particular real rotations of the coordinate axes. An application to the oceanographic data of Fig. 1 is presented in Section V and the paper concludes with a discussion.

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J. M. Lilly is with Earth and Space Research, 2101 Fourth Ave., Suite 1310, Seattle, WA 98121, USA (e-mail: lilly@esr.org).
S. C. Olhede is with the Department of Statistical Science, University College London, Gower Street, London WC1 E6BT, UK (e-mail: s.olhede@ucl.ac.uk).
The most important results of this paper concern the instantaneous bandwidth, a fundamental quantity which accounts, together with variations of the instantaneous frequency, for the spread of the Fourier spectrum about its mean [4]. The correct generalization of the univariate bandwidth to the multivariate case has a surprising but intuitive form. In particular, the bivariate bandwidth admits an elegant geometric interpretation as a fundamental measure of the degree of instability of a time-varying ellipse. It consists of three terms: root-mean-square amplitude modulation, eccentricity modulation or distortion, and orientation modulation or precession, all of which contribute to the Fourier bandwidth.

All data, numerical algorithms, and scripts for analysis and figure generation, are distributed as a part of a freely available package of Matlab functions. This package, called JLAB, is available at the first author’s website, [http://www.jmlilly.net](http://www.jmlilly.net).

II. BACKGROUND

This section gives a compact review of the theory of univariate instantaneous moments, drawing in particular from [4], and gives definitions which will be used throughout the paper.

A. The Analytic Signal

A powerful model for a real-valued nonstationary signal \(x(t)\), assumed deterministic and square-integrable herein, is the modulated oscillation

\[
x(t) = a_x(t) \cos \phi_x(t)
\]

where \(a_x(t) \geq 0\) and \(\phi_x(t)\) are here defined to be particular unique functions called the instantaneous canonical amplitude and canonical phase, respectively [24]. It should be recognized that, in general, the choice of an amplitude and phase pair of functions in (1) for a given signal \(x(t)\) is not unique [see e.g. 25]. However, the canonical pair is of fundamental importance because it separates the signal variability into \(a_x(t)\) and \(\cos \phi_x(t)\), so that for suitable classes of signals these concur with a specified amplitude and phase (see Bedrosian’s theorem).

The canonical pair \([a_x(t), \phi_x(t)]\) is defined in terms of the analytic signal \(x_+(t)\), defined as [3], [26]

\[
x_+(t) = 2A[x](t) = x(t) + i \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{x(u)}{t-u} du
\]  

(2)

where “\(\int\)” is the Cauchy principal value integral and \(A\) is called the analytic operator. This creates a unique complex-valued object \(x_+(t)\) from any real-valued signal \(x(t)\), and permits \(x(t)\) to be associated with a unique amplitude and phase through

\[
a_x(t)e^{i\phi_x(t)} = x_+(t), \quad x_+(t) \neq 0
\]  

(3)

which defines the canonical pair \([a_x(t), \phi_x(t)]\). The original signal is recovered from the analytic signal via \(x(t) = \Re \{x_+(t)\}\), where “\(\Re\)” denotes the real part.
The action of the analytic operator is more clear in the frequency domain. $X(\omega)$ is the Fourier transform of $x(t)$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) e^{i\omega t} \, d\omega$$

and likewise $X_+(\omega)$ is the Fourier transform of $x_+(t)$. The time-domain operator in (2) becomes in the frequency domain simply [27]

$$X_+(\omega) \equiv 2U(\omega)X(\omega)$$

where $U(\omega)$ is the unit step function. Thus in the frequency domain, the analytic signal is formed by doubling the amplitudes of the Fourier coefficients at all positive frequencies, while setting those of all negative frequencies to zero.

**B. Global and Instantaneous Moments**

Using the analytic signal, one may describe both the local and the global behavior of any real-valued signal $x(t)$ as a modulated oscillation. A global description is afforded by the frequency-domain moments of the spectrum of the analytic signal. The *global mean frequency* and *global second central moment* of $x(t)$ are defined, respectively, by

$$\overline{\omega}_x = \frac{1}{2\pi \mathcal{E}_x} \int_{-\infty}^{\infty} \omega |X_+(\omega)|^2 \, d\omega$$

$$\overline{\sigma}_x^2 = \frac{1}{2\pi \mathcal{E}_x} \int_{-\infty}^{\infty} (\omega - \overline{\omega}_x)^2 |X_+(\omega)|^2 \, d\omega$$

where

$$\mathcal{E}_x = \frac{1}{2\pi} \int_{0}^{\infty} |X_+(\omega)|^2 \, d\omega = \int_{-\infty}^{\infty} |x_+(t)|^2 \, dt$$

is total energy of the analytic part of the signal. The second central moment $\overline{\sigma}_x^2$ characterizes the spread of the spectrum about the global mean frequency; thus $\overline{\sigma}_x$ is also known as the *bandwidth*.

It would be greatly informative to relate the global moments $\overline{\omega}_x$ and $\overline{\sigma}_x$ to the time evolution of the signal. This leads to the notion of *instantaneous moments*, time-varying quantities which recover the global moments through integrals of the form [4], [28]

$$\overline{\omega}_x = \mathcal{E}_x^{-1} \int_{-\infty}^{\infty} a_x^2(t) \omega_x(t) \, dt$$

$$\overline{\sigma}_x^2 = \mathcal{E}_x^{-1} \int_{-\infty}^{\infty} a_x^2(t) \sigma_x^2(t) \, dt.$$ 

It is a fundamental result that the derivative of the canonical phase

$$\omega_x(t) \equiv \frac{d}{dt} \phi_x(t) = \Im \left\{ \frac{d}{dt} \ln x_+(t) \right\}$$

satisfies (9), and is termed therefore the *instantaneous frequency* [2], [3]. It is important to point out that the instantaneous moments $\omega_x(t)$ and $\sigma_x^2(t)$ are not themselves moments; rather, they are instantaneous *contributions* to global moments.

The instantaneous second central moment $\sigma_x^2(t)$ is not uniquely defined by the integral (10), because more than one function of time may be shown to integrate to the corresponding global moment. However, the constraint that $\sigma_x^2(t)$ be nonnegative definite, like $\overline{\sigma}_x^2$, leads to the unique definition [4]

$$\sigma_x^2(t) = \frac{|d}{dt} x_+(t) - i \overline{\omega}_x x_+(t)|^2}{|x_+(t)|^2}.$$ 

The square root of the instantaneous second central moment, $\sigma_x(t)$, then has an intuitive interpretation as the average spread of the frequency content of the signal about the global mean frequency $\overline{\omega}_x$.

It has been shown by [4] that the fractional rate of amplitude modulation

$$\upsilon_x(t) = \frac{d}{dt} \ln a_x(t) = \Re \left\{ \frac{d}{dt} \ln x_+(t) \right\}$$

plays an important role within the instantaneous second central moment. After a simplification, (12) becomes

$$\sigma_x^2(t) = [\omega_x(t) - \overline{\omega}_x]^2 + \upsilon_x^2(t)$$

from which it is clear that variations of the instantaneous frequency $\omega_x(t)$ about the global mean frequency $\overline{\omega}_x$ do not account for the entirety of the global second central moment $\overline{\sigma}_x^2$. The remainder is accounted for by $\upsilon_x^2(t)$, and since $\overline{\sigma}_x$ is known as the *global bandwidth*, $\upsilon_x(t)$ is called the *instantaneous bandwidth* [4].
This section introduces a model for a modulated bivariate oscillation, extending and simplifying the development in [12].

A. Modulated Ellipse Model

Bivariate signals will be represented both in the complex-valued form $z(t) \equiv x(t) + iy(t)$, and in vector form $\mathbf{x}(t) \equiv [x(t) \ y(t)]^T$, with $x(t)$ and $y(t)$ both real-valued and where “$T$” denotes the matrix transpose. Our starting point for describing bivariate oscillations will be the modulated elliptical signal [12]

$$z(t) \equiv e^{i\theta(t)} \{a(t) \cos \phi(t) + ib(t) \sin \phi(t)\} \tag{15}$$

which represents $z(t)$ as the position traced out on the $[x, y]$ plane by a hypothetical particle orbiting a time-varying ellipse; see the sketch in Fig. 2. The two angles $\theta(t)$ and $\phi(t)$ are defined on the principal interval $(-\pi, \pi)$, while $a(t) \geq 0$; $b(t)$ may take on either sign for reasons to be see shortly.

The ellipse has instantaneous semi-major and semi-minor axes $a(t)$ and $|b(t)|$, and an instantaneous orientation of the major axis with respect to the $x$-axis given by $\theta(t)$. The angle $\phi(t)$, called the orbital phase, determines the instantaneous position of the particle around the ellipse perimeter with respect to the major axis. While $a(t)$ is nonnegative, the sign of $b(t)$ is chosen to reflect the direction of circulation around the ellipse. Note that the modulated univariate signal (1) is included as the special case $b(t) = 0, \theta(t) = 0$. Thus we can view the modulated elliptical signal as an extension of the amplitude/frequency modulated signal to the bivariate case.

The meaning of the “instantaneous” ellipse properties referred to in the previous paragraph is more clear if we separate the ellipse phase from the other variables. Introduce

$$z(t, t') \equiv e^{i\theta(t)} \{a(t) \cos [\phi(t) + \omega_\phi(t)t'] + ib(t) \sin [\phi(t) + \omega_\phi(t)t']\} \tag{16}$$

where $t'$ is seen as a “local” time. With $t$ fixed, $z(t, t')$ continually traces out the perimeter of a “frozen” ellipse as $t'$ varies; the period of orbiting the ellipse is $2\pi/\omega_\phi(t)$. It is these frozen ellipses which are plotted in Fig. 1b for the estimated modulated elliptical signal corresponding to the time series in Fig. 1a. Note that $z(t)$ itself will not in general trace out an ellipse if the ellipse geometry varies in time. However, if the parameters of the ellipse geometry $a(t), b(t)$, and $\theta(t)$ are slowly varying with respect to the phase $\phi(t)$, then $z(t)$ will approximate an ellipse.

B. Analytic Signal Pairs

The parameters of the modulated elliptical signal are made unique for a given pair of real-valued signals by referring to a pair of analytic signals, just as the univariate modulated signal model is made unique by referring to a single analytic signal.
In vector notation the modulated elliptical signal \( \mathbf{z}(t) \) becomes

\[
\begin{bmatrix}
x(t) \\
y(t)
\end{bmatrix} = \mathbf{J}[\theta(t)] \Re \left\{ e^{i\phi(t)} \begin{bmatrix} a(t) \\ -ib(t) \end{bmatrix} \right\}
\]  

(17)

where

\[
\mathbf{J}(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}
\]

(18)
is the counterclockwise rotation matrix. Thus we may write

\[
e^{i\phi(t)} \mathbf{J}[\theta(t)] \begin{bmatrix} a(t) \\ -ib(t) \end{bmatrix} \equiv \begin{bmatrix} x_+(t) \\ y_+(t) \end{bmatrix} = \begin{bmatrix} 2A|x(t)| \\ 2A|y(t)| \end{bmatrix}
\]

(19)

and this equation is taken as our definition of the ellipse parameters in terms of the analytic portions of \( x(t) \) and \( y(t) \). In terms of the analytic signals \( x_+(t) \) and \( y_+(t) \),

\[
z(t) = \Re \{ x_+(t) \} + i\Re \{ y_+(t) \}
\]

(20)

recovers the original complex-valued signal.

We emphasize that \( a(t) \), \( b(t) \), \( \theta(t) \), and \( \phi(t) \) appearing in (15) and (17) are not arbitrary functions which one is free to specify. They are, instead, unique properties of a given bivariate signal. While any bivariate signal can be represented as a modulated ellipse, just as any univariate signal can be represented as a modulated oscillation via the analytic signal, it is not always sensible to do so. If one begins with a pair of real-valued signals \( x(t) \) and \( y(t) \) which are, say, finite samples of a noise process, the ellipse parameters would likely not yield any sensible information; this would inform you that the modulated ellipse representation is not appropriate.

One may also associate a second pair of analytic signals with \( z(t) \), namely

\[
\begin{bmatrix} z_+(t) \\ z_-(t) \end{bmatrix} = \begin{bmatrix} A[z](t) \\ A[z^*](t) \end{bmatrix}
\]

(21)

the components of which are the analytic parts of \( z(t) \) and of its complex conjugate. Then

\[
z(t) = z_+(t) + z_-^*(t)
\]

(22)
decomposes \( z(t) \) into counterclockwise and clockwise rotating contributions, respectively. Note that for convenience we have defined \( z_-(t) \) to be an analytic, rather than an anti-analytic, signal. Making use of the unitary matrix

\[
\mathbf{T} \equiv \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & i \\ 1 & -i \end{bmatrix}
\]

(23)

one finds

\[
\begin{bmatrix} z_+(t) \\ z_-(t) \end{bmatrix} = \frac{1}{\sqrt{2}} \mathbf{T} \begin{bmatrix} x_+(t) \\ y_+(t) \end{bmatrix}
\]

(24)
as the relationship between the two different pairs of analytic signals. In order to distinguish between these two pairs, we refer to \( x_+(t) \) and \( y_+(t) \) as the Cartesian pair of analytic signals, and to \( z_+(t) \) and \( z_-(t) \) as the rotary pair (borrowing a term from the oceanographic community [19], [20]).

The ellipse parameters have simple expressions in terms of the amplitude and phases of the rotary analytic signals. Letting

\[
z_+(t) = a_+(t)e^{i\phi_+(t)}
\]

(25)

\[
z_-(t) = a_-(t)e^{i\phi_-(t)}
\]

(26)

and comparing the rotary decomposition (22) with \( \mathbf{T} \) times the matrix form of the modulated ellipse model (19), one finds [12]

\[
a(t) \equiv a_+(t) + a_-(t)
\]

(27)

\[
b(t) \equiv a_+(t) - a_-(t)
\]

(28)

\[
\phi(t) \equiv \frac{[\phi_+(t) + \phi_-(t)]}{2}
\]

(29)

\[
\theta(t) \equiv \frac{[\phi_+(t) - \phi_-(t)]}{2}
\]

(30)
as expressions for the ellipse parameters in terms of the amplitude and phases of the rotary analytic signals. The ellipse parameters are thereby uniquely defined. Unique relations of the ellipse parameters to the parameters of the Cartesian analytic signals are then implied by (24), and are given in Appendix A.

\[1\text{The factor of two difference in defining analytic versions of real-valued and complex-valued signals, i.e. the last equality in (19) versus (21), prevents factors of two from appearing in the decomposition equations (20) and (22).} \]
C. Amplitude and Eccentricity

It is convenient to replace the semi-major and semi-minor axes \( a(t) \) and \( b(t) \) with the root-mean-square amplitude

\[ \kappa(t) \equiv \sqrt{\frac{a^2(t) + b^2(t)}{2}} \]  

(31)

together with a relative of the eccentricity [29], [12] \( \lambda(t) \)

\[ \lambda(t) \equiv r_z \frac{a^2(t) - b^2(t)}{a^2(t) + b^2(t)} \]  

(32)

which measures the signal’s departure from circularity. Here the direction of rotation of the “particle” around the time-varying ellipse is denoted by \( r_z = \text{sgn} \{ b(t) \} \)

(33)

which we assume henceforth to be constant over the time interval of interest. These definitions give

\[ a(t) = \kappa(t) \sqrt{1 + |\lambda(t)|} \]  

(34)

\[ b(t) = r_z \kappa(t) \sqrt{1 - |\lambda(t)|} \]  

(35)

as expressions for the semi-major and (signed) semi-minor axes. Note that \( r_z \) is well defined for purely circular signals for which \( a(t) = |b(t)| \), whereas \( \text{sgn} \{ \lambda(t) \} \) vanishes.

The magnitude of \( \lambda(t) \), like the eccentricity \( \text{ecc}(t) = \sqrt{1 - b^2(t)/a^2(t)} \), varies between \( |\lambda(t)| = 0 \) for purely circularly polarized motion and \( |\lambda(t)| = 1 \) for purely linearly polarized motion. We refer to \( |\lambda(t)| \) as the ellipse linearity and to \( \lambda(t) \) as the signed linearity. While not in common use as a measure of eccentricity, we prefer \( \lambda(t) \) to other such measures since it leads to simple forms for subsequent expressions.

D. Rates of Change

The rates of change of the ellipse parameters have special interpretations. The time derivative of the orbital phase, termed the orbital frequency \( \omega_\phi(t) \equiv \frac{d}{dt} \phi(t) \), gives the rate at which the particle orbits the ellipse, while the time derivative of the orientation angle is the precession rate \( \omega_\theta(t) \equiv \frac{d}{dt} \theta(t) \). Amplitude modulation of the ellipse involves a time derivative of \( \kappa(t) \), while variation in \( \lambda(t) \) means that the shape of the ellipse is distorting with time. However, the ellipse parameters and their derivatives do not have an immediately evident relationship to the global moments of \( z(t) \). Thus, while these are useful local descriptions of joint signal variability, they are not interpretable as instantaneous moments.

E. Instantaneous Moment Pairs

In Section [II-A] it was shown that the instantaneous moments of a univariate signal—instantaneous frequency and bandwidth, in particular—provide a powerful description of local signal variability with a direct relationship to the global signal moments. We aim to identify the analogous quantities for a bivariate signal.

A bivariate oscillatory signal can be equivalently expressed in terms of two pairs of analytic signals, the Cartesian pair \( x_+(t) \) and \( y_+(t) \) or the rotary pair \( z_+(t) \) and \( z_-(t) \). Each of these, in turn, can be analyzed by their individual instantaneous moments, using the ideas described in Section [II-A]. In particular, these two pairs of analytic signals lead to two pairs of instantaneous frequencies \( [\omega_x(t), \omega_y(t)] \) and \( [\omega_+(t), \omega_-(t)] \) and two pairs of instantaneous bandwidths \( [v_x(t), v_y(t)] \) and \( [v_+(t), v_-(t)] \), which contribute to two corresponding pairs of global moments. This approach is, however, unsatisfactory. Rather than a unified description of signal variability, an analysis based on these pairs of instantaneous moments achieves a description of two disparate halves of the bivariate signal \( z(t) \) considered separately.

Between the Cartesian and rotary analytic signals, one can represent the signal as a pair of oscillations in two different ways. The difficulty of a pairwise decomposition is highlighted by considering a range of different values of the linearity \( |\lambda(t)| \). Generally speaking, the decomposition of \( z(t) \) into the rotary pair of analytic signals is more appropriate when one of \( z_+(t) \) and \( z_-(t) \) is much larger than the other, e.g. \( |z_+(t)| \gg |z_-(t)| \). On the other hand, when (in some rotated coordinate system) one has \( |x_+(t)| \gg |y_+(t)| \), then the Cartesian decomposition is more appropriate. These two cases correspond to \( |\lambda(t)| \approx 0 \) and \( |\lambda(t)| \approx 1 \), respectively. In between these two cases, or for a signal ranging over both extremes, there is a gray area in which neither decomposition is particularly appropriate. A unified description is therefore necessary.

Note that herein we always use the term “circular” to describe the shape traced out on the \([x, y]\) plane by a bivariate signal, rather than in the statistical sense of a circularly symmetric or proper complex-valued signal [e.g. 13].
To begin with, the instantaneous moments of the analytic signal pairs can be cast in terms of the ellipse parameters, to show how changes in the ellipse geometry are expressed in time variability of the instantaneous moments pairs. Inverting (27–30), and utilizing (34) and (35), leads to

$$\phi_{\pm}(t) = \phi(t) \pm \theta(t)$$
$$a_{\pm}(t) = \frac{\kappa(t)}{\sqrt{2}} \sqrt{1 \pm r_z \sqrt{1 - \lambda^2(t)}}$$

(36)

(37)

giving the parameters of the rotary analytic signals in terms of the ellipse parameters. (In these and subsequent expressions, the signs on the right-hand-side are understood to be chosen to match the signs on the left.) Note that the amplitudes satisfy $a_+^2(t) + a_-^2(t) = \kappa^2(t)$, and we may identify $\kappa^2(t)$ as the sum of the instantaneous power of the two rotary signals. When $r_z > 0$, and the ellipse rotates in the positive (counterclockwise) direction, we have $a_+(t) > a_-(t)$ as expected, whereas the opposite is true for negative rotation. Note that as $|\lambda(t)|$ approaches zero, and the signal becomes nearly circular, the smaller of $a_+(t)$ and $a_-(t)$ approaches zero while the larger approaches $\kappa(t)$.

Differentiating (36) and the logarithm of (37) leads to

$$\omega_{\pm}(t) = \omega_\phi(t) \pm \omega_\theta(t)$$
$$v_{\pm}(t) = \frac{d \ln \kappa(t)}{dt} \pm r_z \frac{1}{2} \frac{1}{\sqrt{1 \pm r_z \sqrt{1 - \lambda^2(t)}}}$$

(38)

(39)

as expressions for the rotary instantaneous frequencies and bandwidths. It is useful to note an asymmetry of the rotary bandwidths. For small $|\lambda(t)|$ with $r_z > 0$, the denominator of the second term in $v_-(t)$ is close to zero while that of $v_+(t)$ is close to unity; the situation is reversed for $r_z < 0$. This shows that for small $|\lambda(t)|$, the bandwidth of the weaker of the two rotary signals is much more sensitive to variations in the degree of eccentricity than is the bandwidth of the stronger signal.

As discussed above, these expressions are useful in the near-circular case $|\lambda(t)| \approx 1$. The same procedure carried out in terms of the Cartesian analytic signals $x_+(t)$ and $y_+(t)$, presented in Appendix A, leads to more complicated relationships between variations of the ellipse geometry and the instantaneous moments of the two real-valued signals $x(t)$ and $y(t)$. In particular, we may note in those relationships the explicit dependence on the ellipse orientation $\theta(t)$. This renders the Cartesian instantaneous moments quite problematic for a unified description of the signal variability, since a simple coordinate rotation will cause these quantities to change.

There is therefore a disconnect between the moment-based description of bivariate signal variability, grounded on the analytic signals, and the modulated ellipse model of joint structure. This motivates the development of the next section.

IV. JOINT INSTANTANEOUS MOMENTS

The definitions of instantaneous moments in Section II-A, which are standard, can be extended to accommodate multivariate oscillatory signals. Let

$$x_+(t) = [x_{+,1}(t) \ x_{+,2}(t) \ \ldots \ x_{+,N}(t)]^T$$

(40)

be a vector of $N$ analytic signals, and let $X_+(\omega)$ be the corresponding frequency-domain vector; these are related by the inverse Fourier transform

$$x_+(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{i\omega t} X_+(\omega) \, d\omega.$$  \hspace{2cm} (41)

Here we begin with the general multivariate case and subsequently examine the bivariate case, $N = 2$, in detail.

A. Joint Moments

It is presumed that the components of $x_+(t)$ are sufficiently closely related that we should seek a unified description of their variability. To this end it is reasonable to define the joint analytic spectrum as the normalized average of the spectra of its $N$ components,

$$S_\mathbf{x}(\omega) \equiv \mathcal{E}^{-1}_\mathbf{x} ||X_+(\omega)||^2$$

(42)

where $||\mathbf{x}|| \equiv \sqrt{\mathbf{x}^H \mathbf{x}}$ is the Euclidean norm of a vector $\mathbf{x}$, "$H$" indicating the conjugate transpose, and where

$$\mathcal{E}_\mathbf{x} \equiv \frac{1}{2\pi} \int_{0}^{\infty} ||X_+(\omega)||^2 d\omega = \int_{-\infty}^{\infty} ||x_+(t)||^2 dt$$

(43)

is the total energy of the multivariate analytic signal. One may then define the joint global mean frequency

$$\mathcal{F}_\mathbf{x} \equiv \frac{1}{2\pi \mathcal{E}_\mathbf{x}} \int_{0}^{\infty} \omega S_\mathbf{x}(\omega) \, d\omega$$

(44)
associated with the joint analytic spectrum, together with
\[
\sigma_x^2(t) \equiv \frac{1}{2\pi |x|} \int_0^\infty (\omega - \omega_x)^2 S_x(\omega) \, d\omega \tag{45}
\]
which is the joint global second central moment.

The joint instantaneous frequency \(\omega_x(t)\) and joint instantaneous second central moment \(\sigma_x^2(t)\) of \(x_+(t)\) are then some quantities that decompose the corresponding global moments across time, i.e. which satisfy
\[
\begin{align*}
\bar{\omega}_x &= \mathcal{E}_x^{-1} \int_{-\infty}^{\infty} \|x_+(t)\|^2 \omega_x(t) \, dt \\
\bar{\sigma}_x^2 &= \mathcal{E}_x^{-1} \int_{-\infty}^{\infty} \|x_+(t)\|^2 \sigma_x^2(t) \, dt
\end{align*}
\tag{46, 47}
\]
noting that \(\|x_+(t)\|^2\) is the instantaneous total analytic signal power. Since the univariate instantaneous frequency \(\omega\) may be rewritten
\[
\omega_x(t) \equiv \Im \left\{ \frac{d \ln x_+(t)}{dt} \right\} = \Im \left\{ \frac{x_+(t) \, dx_+(t)}{|x_+(t)|^2} \right\}
\tag{48}
\]
we define the joint instantaneous frequency according to the same form,
\[
\omega_x(t) \equiv \frac{\Im \left\{ x_+^H(t) \frac{dx_+(t)}{dt} \right\}}{|x_+(t)|^2} 
\tag{49}
\]
and find that this does indeed satisfy \(46\).

Likewise, the second instantaneous central moment, which we define as
\[
\sigma_x^2(t) \equiv \frac{\| \mathbf{d} \bar{\omega}_x(t) \|_2}{\|x_+(t)\|^2} \tag{50}
\]
is a nonnegative-definite quantity satisfying \(47\). This gives the normalized departure of the rate of change of the vector-valued signal from a uniform complex rotation at the constant single frequency \(\bar{\omega}_x\). Equations \(49\) and \(50\) are clearly the natural generalizations of \(11\) and \(12\) to multivariate analytic signals.

**B. Joint Instantaneous Bandwidth**

Furthermore we may generalize the notion of instantaneous bandwidth to a multivariate signal. We define the *joint instantaneous bandwidth* via its relationship to the joint second central moment
\[
v_x^2(t) \equiv \sigma_x^2(t) - [\omega_x(t) - \bar{\omega}_x]^2 \tag{51}
\]
by comparison with \(14\) for the univariate case. That is, we *define* the squared instantaneous bandwidth to be that part of the instantaneous second central moment not accounted for by deviations of the instantaneous frequency from the global mean frequency.

This definition of the bandwidth leads, after some manipulation, to
\[
v_x^2(t) = \frac{\| \mathbf{d} \bar{\omega}_x(t) - i \omega_x(t) \bar{x}_+(t) \|^2}{\|x_+(t)\|^2} \tag{52}
\]
which is the normalized departure of the rate of change of the vector-valued signal from a uniform complex rotation at a single *time-varying* frequency \(\omega_x(t)\). For \(x_+(t)\) a 1-vector consisting of a single analytic signal, \(x_+(t)\), \(52\) becomes
\[
v_x^2(t) = \left( \frac{\mathbf{d} \bar{\omega}_x(t) - i \omega_x(t)}{|x_+(t)|^2} \right) \frac{\|x_+(t)\|^2}{\|x_+(t)\|^2} = \left. \frac{dx_+(t)}{dt} \ln a_x(t) \right| = v_x^2(t) \tag{53}
\]
and the joint instantaneous bandwidth correctly reduces to the univariate bandwidth defined in \(13\).

The contributions to the instantaneous second central moment and the instantaneous bandwidth are perhaps more clear if we write \(50\) and \(52\) out as summations
\[
\begin{align*}
\sigma_x^2(t) &= \sum_{n=1}^{N} a_n^2(t) \left\{ v_n^2(t) + |\omega_n(t) - \omega_x|^2 \right\} \\
v_x^2(t) &= \sum_{n=1}^{N} a_n^2(t) \left\{ v_n^2(t) + |\omega_n(t) - \omega_x|^2 \right\}
\end{align*}
\tag{54, 55}
\]
where \( a_n(t) \) is the amplitude of the \( n \)th analytic signal, and so forth. The first of these states that amplitude modulation, as well as departures of the instantaneous frequencies from the global mean frequency \( \overline{\omega} \), contribute to the second instantaneous central moment. The second states that amplitude modulations together with departures of the instantaneous frequencies from the time-varying joint instantaneous frequency \( \omega_x(t) \) contribute to the squared joint instantaneous bandwidth. In both cases contributions from different times and different signal components are weighted according to the instantaneous power \( \sigma^2(t) \).

C. Invariance

The joint instantaneous moments defined above have the important property that they are invariant to transformations of the form

\[
y_+(t) \equiv cUx_+(t)
\]

where \( c \) is some constant and \( U \) is an \( N \times N \) unitary matrix; these may be termed *scaled unitary transformations*. Since in the frequency domain we have also \( Y_+(\omega) = cUX_+(\omega) \), it is obvious that

\[
S_Y(\omega) \equiv \frac{Y_H^+(\omega)Y_+(\omega)}{2\pi \int_{-\infty}^{\infty} Y_H^+(\omega)Y_+(\omega) d\omega} = S_X(\omega)
\]

hence the joint analytic spectrum is unchanged, as are the joint global moments \( \overline{\omega} \) and \( \overline{\sigma}^2 \). The joint instantaneous frequency transforms as

\[
\omega_y(t) \equiv \frac{\Im \{ Y_H^+(t) \} y_+(t) }{\parallel y_+(t) \parallel^2} = \omega_x(t)
\]

and hence remains unchanged. Similarly it is easy to see that the joint instantaneous bandwidth \( \nu_x(t) \) and joint instantaneous second central moments \( \sigma^2(t) \) and are also all invariant to scaled unitary transformations. In particular, these joint instantaneous moments are invariant to coordinate rotations.

D. Joint Bivariate Moments

We now examine the joint instantaneous moments for bivariate signals by taking either \( x_+(t) = [z_+(t) \ z_-(t)] \) or \( x_+(t) = [x_+(t) \ y_+(t)] \). Since these two vectors are related by the scaled unitary transformation (24), from the invariance shown in the preceding subsection, the results will be identical in either case. The bivariate joint instantaneous moments integrate to the global moments of the joint analytic spectrum

\[
S_z(\omega) = \frac{|X_+(\omega)|^2 + |Y_+(\omega)|^2}{E_x + E_y} = |Z_+(\omega)|^2 + |Z_-(\omega)|^2
\]

where \( E_x = \frac{1}{2\pi} \int_0^\infty |X_+(\omega)|^2 d\omega \), and so forth. Note that we will use the subscript “\( z \)” rather than the subscript “\( x \)” to specifically denote bivariate quantities.

To see (59), it is helpful to relate the joint analytic spectrum to the two-sided spectrum of the complex-valued signal \( z(t) \), which has a Fourier transform \( Z(\omega) \). The Fourier transforms of \( z_+(t) \) and \( z_-(t) \) are, respectively,

\[
Z_+(\omega) = U(\omega)Z(\omega) = \frac{1}{2} [X_+(\omega) + iY_+(\omega)]
\]

\[
Z_-(\omega) = U(\omega)Z(-\omega) = \frac{1}{2} [X_+(\omega) - iY_+(\omega)]
\]

where \( Z(\omega) \) is the Fourier transform of \( z(t) \) and where \( U(\omega) \) is again the unit step function. Inserting these into the second and third expressions in (59), we note at once a cancelation of cross-terms, and the equality follows. Note that (60)–(61) show that \( z_+(t) \) and \( z_-(t) \) have Fourier coefficients drawn entirely from the positive-frequency and negative-frequency sides of \( Z(\omega) \), respectively (hence the notation “+” and “−”). The joint analytic spectrum simply consists of averaging the positive and negative frequency halves of \( |Z(\omega)|^2 \), followed by a normalization to unit energy.

1) Instantaneous Frequency: To obtain the bivariate instantaneous frequency, we begin with the definition (49) and insert expressions (37) and (38) for the rotary instantaneous frequencies and amplitudes. This leads to

\[
\omega_z(t) = \omega_x(t) + r_z \sqrt{1 - \lambda^2(t)} \omega_y(t)
\]

for the bivariate instantaneous frequency written in terms of the ellipse parameters. Alternatively, we could have used expressions (70)–(71) and (74)–(75) from Appendix A for the Cartesian analytic signals. Inserting these into (49) we obtain again (62), as we must account on the invariance proved in the last subsection. The relative difficulty of directly manipulating the cumbersome Cartesian expressions illustrates the importance of this general result.

The joint bivariate instantaneous frequency \( \omega_z(t) \) gives a correct measure the time-varying frequency content of a bivariate signal regardless of the polarization state. For \( |\lambda(t)| = 0 \), the signal is purely circular, and \( \omega_z(t) \) becomes \( \omega_x(t) + r_z \omega_y(t) \). Comparison with (38) shows that this is \( \omega_z(t) \) if \( r_z > 0 \) and \( \omega_z(t) \) if \( r_z < 0 \), that is, for circularly polarized motion \( \omega_z(t) \)
becomes the instantaneous frequency of the non-vanishing rotary component. On the other hand, in Appendix A it is shown that for a purely linear signal, \(|\lambda(t)| = 1\) or vanishing minor axis \(|b(t)| = 0\), the Cartesian instantaneous frequency along the coordinate axis aligned with the ellipse major axis is equal to \(\omega_\phi(t)\), but (62) shows that with \(|\lambda(t)| = 1\) then this is also equal to \(\omega_z(t)\). Thus the use of \(\omega_z(t)\) is quite desirable as it is appropriate for any polarization, unlike the partitioning into a rotary pair or a Cartesian pair.

2) *Instantaneous Bandwidth:* Similarly the bivariate bandwidth, expressed in terms of the ellipse parameters, is

\[
v_z^2(t) = \frac{d\ln\kappa(t)}{dt}^2 + \frac{1}{1 - \lambda^2(t)} \left( \frac{1}{2} \frac{d\lambda(t)}{dt} \right)^2 + \lambda^2(t)\omega_\phi^2(t)
\]

as follows from the definition (53) together with (39) or (79–80). This expression is quite illuminating. The first term in (63) measures the strength of amplitude modulation, the second is the rate of ellipse distortion on account of changing eccentricity, and the third is due to the precession of the ellipse. Note each of these quantities is independent of coordinate rotation, that is, the orientation angle \(\theta(t)\) does not explicitly appear. We define these three terms as

\[
v_z(t) = v_z^2(t) + v_\lambda^2(t) + v_\phi^2(t)
\]

which we call the squared *amplitude bandwidth, deformation bandwidth*, and *precession bandwidth*, respectively.

An illustration of the bivariate bandwidth is presented in Figure 3. Three different time-varying elliptical signals are plotted, all having constant orbital frequency \(\omega_\phi(t)\) and each with exactly one of the three terms in (63) being nonzero. The quantities \(v_\kappa(t), v_\lambda(t), v_\phi(t)\) are each set equal to the value of 0.025. The bold portion of the line in all three panels show an initial single orbit. A circle marks the beginning of each record and an “x” marks the end of the record.

Fig. 3. An ellipse with uniformly increasing relative amplitude (a), i.e. \(v_\kappa(t)/\omega_\kappa(t)\) constant, an ellipse with uniformly increasing eccentricity (b), i.e. \(v_\lambda(t)/\omega_\lambda(t)\) constant, and an ellipse uniformly precessing (c) with \(v_\phi(t)/\omega_\phi(t)\) constant. The three constants have been chosen to all be the same value of 0.025. The bold portion of the line in all three panels show an initial single orbit. A circle marks the beginning of each record and an “x” marks the end of the record.
appropriate for the near-circular case \(|\lambda(t)| \approx 0\) and the near-linear case \(|\lambda(t)| \approx 1\), respectively. In the near-circular case, the deformation bandwidth is due to the (small) departures of the linearity \(|\lambda(t)|\) from zero, while in the near-linear case it is due to the (small) departures of the linearity from unity. Another interpretation may be found by noting, using \([34,35]\),

\[
|\psi_\chi(t)| = \left| \frac{a(t) b(t)}{a^2(t) + b^2(t)} \frac{d}{dt} \ln \left| \frac{b(t)}{a(t)} \right| \right|
\]

(68)

which states that the deformation bandwidth is due to fractional changes in the ellipse aspect ratio \(b(t)/a(t)\), weighted in proportion to the ratio of the ellipse area \(\pi a(t)|b(t)|\) to the root-mean-square radius.

V. APPLICATION

This section presents an application of the bivariate instantaneous frequency and bandwidth to a typical signal from physical oceanography.

A. Data and Method

As discussed already in the introduction, Fig. 1 presents data from a subsurface, acoustically-tracked oceanographic float [18]. The data was downloaded from the World Ocean Circulation Experiment Subsurface Float Data Assembly Center (WFDAC) at http://wfdac.whoi.edu. This particular record, from an experiment in the eastern North Atlantic Ocean that is well-known among oceanographers [17], [30], [31], was recorded somewhat to the south and west of the Canary Islands. The float was drifting at a depth between 1000 m and 1300 m. Its horizontal position was recorded once per day, as inferred by triangulation of acoustic travel times from several nearby sound sources. The region shown ranges from 20°–25°N and 27°–22.5°W.

The loops in the trajectory are known to be the imprint of one of the large “eddies” or “vortices” [32] that are common in the ocean, a distant relative of the swirls one observes when stirring coffee; note that the loops seen in Fig. 1 are up to fifty kilometers in diameter. An instrument trapped in such a vortex will record both nearly circular motion from orbiting around the vortex center, which is well modeled as a modulated elliptical signal as shown by [12], as well as translational or advected motion of the vortex center itself. This suggests modeling the observed signal \(z^{(o)}(t) = x^{(o)}(t) + iy^{(o)}(t)\) via the unobserved components model [33]

\[
z^{(c)}(t) = z(t) + z^{(c)}(t)
\]

(69)

where \(z(t)\) is a modulated elliptical signal and where \(z^{(c)}(t)\) is a residual defined to include everything else. The residual \(z^{(c)}(t)\) is expected to include the turbulent background flow together some measurement noise, but may be considered “noise” at present since we are interested only in \(z(t)\).

A means of estimating \(z(t)\) using an extension of wavelet ridge analysis [10], [11] was developed by [12]. The details of this method are not particularly important here, except to note that it is expected to give a reasonable estimate \(\hat{z}(t)\) of the presumed (unobserved) modulated elliptical signal \(z(t)\). We apply this method, with parameter choices noted in Appendix B, to the record in Fig. 1. This leads to the estimated bivariate oscillation \(\hat{z}(t)\) shown in Fig. 1b, and—by subtraction, \(\hat{z}^{(c)}(t) \equiv z^{(o)}(t) - \hat{z}(t)\)—to the estimated residual signal \(\hat{z}^{(c)}(t)\) shown in Fig. 1c. In Fig. 1d the signal \(\hat{z}(t)\) is represented by snapshots of the modulated signal at different times, by letting the orbital phase vary with the ellipse geometry held fixed, as described earlier in the discussion of (16). Note that in Fig. 1c virtually all the looping motions have been removed, leaving behind a large-scale meander plus small-scale irregularities.

B. Instantaneous Amplitude and Frequency

Having achieved this decomposition, we now analyze in more detail the properties of the estimated bivariate oscillatory signal \(\hat{z}(t)\) using the joint instantaneous moments. The estimated signal, shown in Fig. 4a together with the instantaneous root-mean-square amplitude \(\hat{r}(t)\), exhibits both substantial amplitude as well as frequency modulations; note that we will denote all properties of \(\hat{z}(t)\) with a hat, “\(^\wedge\)”, to distinguish them from the properties of the unobserved true signal \(z(t)\). In particular, during the last third of the record, say after yearday 250, the signal amplitude is greatly reduced compared to the earlier time period. This accounts for the transition from large to small ellipses seen in Fig. 1b.

This signal rotates in a clockwise fashion: the negative rotary component \(\hat{z}_-(t)\) dominates the positive component \(\hat{z}_+(t)\), thus the signed linearity \(\hat{\lambda}(t)\) (not shown) is everywhere negative; this sense of rotation can be inferred from the phase shift between the real and imaginary parts of \(\hat{z}(t)\) in Fig. 4b. The linearity \(|\hat{\lambda}(t)|\) itself (Fig. 4b) is generally very small, corresponding to nearly circular motion, apart from a few excursions to higher values. Thus only a handful of the ellipses shown in Fig. 1b exhibit substantial eccentricity.

The instantaneous frequency content of the signal is shown Fig. 4c. Here, the bivariate instantaneous frequency of \(\hat{z}(t)\), \(\hat{\omega}_z(t)\), is presented together with its orbital frequency \(\hat{\omega}_\phi(t)\) and precession rate \(\hat{\omega}_\theta(t)\). The precession rate is seen to be much smaller than the orbital frequency, and consequently, it follows from (62) that the bivariate instantaneous frequency \(\hat{\omega}_z(t)\) should be close to the orbital frequency \(\hat{\omega}_\phi(t)\). A transition is seen around yearday 250 to higher-frequency oscillations,
corresponding to the transition to smaller amplitudes seen in Fig. 4. Interestingly, in the higher-frequency portion of the record, the precession rate exhibits a tendency for negative rotation, but establishing whether or not this result is statistically significant is beyond the scope of the present paper. The precession rate exhibits a tendency for negative rotation, but establishing whether or not this result is statistically significant is beyond the scope of the present paper. The precession rate is expected to be more robust against the effects of noise than the negative rotary signal that would be observed along the coordinate axes. This suggests that although $\omega(t)$ and $\theta(t)$ are of great interest from the point of view of the modulated ellipse model, $z(t)$ provides a superior estimate of the time-varying frequency content of a noisy bivariate oscillatory signal. The preceding discussion has also brought up the importance of understanding the effects of noise on the estimation procedure, a task which is currently underway.

C. Instantaneous Bandwidth

The three terms in the bivariate instantaneous bandwidth of the estimated signal $z(t)$—$\nu_x(t)$, $\nu_\omega(t)$, and $\nu_\theta(t)$—are shown in Fig. 4. During most of the record, the amplitude bandwidth $\nu_x(t)$ is larger than the other two. Since these contribute as their
squares to the bivariate bandwidth \( \tilde{\nu}_z(t) \), we see that in fact the amplitude bandwidth \( \tilde{\nu}_a(t) \) accounts for the large majority of the estimated joint instantaneous bandwidth \( \tilde{\nu}_z(t) \) during most of the record. Thus the lowest-order departure of the bivariate signal from a uniform oscillation at frequency \( \omega_z(t) \) is generally due to modulation of the root-mean-square ellipse amplitude, with its eccentricity and orientation held fixed. This result could have different important physical interpretations; however, such questions are outside the scope of the present paper.

The three bandwidth quantities are again shown in Fig. 4d, but this time each has been divided by \( \tilde{\omega}_z(t) \), thus comparing the contributions to the instantaneous bandwidth with the instantaneous frequency. Whereas Fig. 4d suggests that the instantaneous frequency increases after yearday 250, Fig. 4e shows that this transition is offset by the corresponding increase in instantaneous bandwidth. This means that the variability of the modulated elliptical, on time scales proportional to the local period of oscillation, is in fact similar during the different periods of the record. If anything, the “relative bandwidth” is somewhat lower during the last third of the record then in the remainder.

VI. DISCUSSION

This paper has examined the extension of instantaneous moments to bivariate signals. To accomplish this, “joint instantaneous moments” were defined for a general multivariate signal. These are time-varying quantities that integrate to the global moments of the average analytic spectrum, that is, the spectrum of the analytic part of the signal components averaged over the number of components. While the joint instantaneous frequency is simply the power-weighted average of the instantaneous frequencies of the components, the joint instantaneous bandwidth has an unexpected but very intuitive form. It measures the extent to which a multivariate oscillation does not evolve simply by oscillating at the single, time varying frequency given by the joint instantaneous frequency.

These joint instantaneous moments together with the notion of a modulated elliptical signal represent a powerful means for analyzing bivariate signals. The “modulated elliptical signal” is a representation of a bivariate oscillation, alternately regarded as a pair of analytic signals, as a single time-varying structure. The joint moments describe the nature of the variability of this structure, combining the partial descriptions achieved from the individual moments of the members of a signal pair. Expressing the bivariate instantaneous frequency and bandwidth in terms of the time-varying ellipse parameters leads to illuminating expressions. In particular, the instantaneous bandwidth was shown to consist of three terms: amplitude modulation, deformation or eccentricity modulation, and orientation modulation or precession. These three quantities thus express the three basic ways ellipse geometry can change with time; the bivariate bandwidth is therefore seen as the degree of instability of a modulated elliptical signal.

An application of the instantaneous moments to a bivariate time series from oceanography leads to a wealth of information regarding the time variability of the signal, which in this case could reflect time variations in an underlying oceanic vortex structure. In addition to pursing physical questions raised here, the most important outstanding tasks involve quantifying the errors involved in estimating bivariate oscillatory signals. There are both “deterministic” errors due to the time variability of the signal, which have recently been examined for univariate wavelet ridge analysis [34], as well as random errors due to background noise. These topics are the subjects of ongoing research.

APPENDIX A

THE CARTESIAN ANALYTIC SIGNALS

Here we derive expressions for the Cartesian instantaneous moments, that is, the instantaneous moments of \( x_+(t) = a_x(t) \) and \( y_+(t) = a_y(t) \) for a bivariate signal represented as \( z(t) = a_x(t) + iy_y(t) \). In terms of the parameters of the modulated elliptical signal of Section III the amplitudes and phases of the Cartesian analytic signals are

\[
\alpha_x(t) = \kappa(t) \sqrt{1 + |\lambda(t)| \cos 2\theta(t)}
\]
\[\alpha_y(t) = \kappa(t) \sqrt{1 - |\lambda(t)| \cos 2\theta(t)}\]
\[\phi_x(t) = \phi(t) + \Im \{\ln \{a(t) \cos \theta(t) + ir_z |b(t)| \sin \theta(t)\}\}\]
\[\phi_y(t) = \phi(t) - r_z \pi/2 + \Im \{\ln \{|b(t)| \cos \theta(t) + ir_z a(t) \sin \theta(t)\}\}\]

which simplifies the presentation of [12]. Note that the combination \( \Im \{\ln \{z\}\} \) is used to implement the four-quadrant inverse tangent function, with \( \tan^{-1} \{\Im \{\ln \{z\}\}\} = \Im(z)/\Re(z) \). As \( |\lambda(t)| \) approaches zero, and the signal is nearly circular, both the Cartesian amplitudes \( a_x(t) \) and \( a_y(t) \) approach \( \kappa(t) \). Meanwhile the phases \( \phi_x(t) \) and \( \phi_y(t) + r_z \pi/2 \) both approach \( \phi(t) + r_z \theta(t) \), which has been previously identified as the phase of whichever rotary component, \( z_+(t) \) or \( z_-(t) \), has the larger amplitude.
Differentiating the phase expressions (72) one obtains the following, rather complicated, forms for the Cartesian instantaneous frequencies

\[
\omega_x(t) \equiv \frac{d\phi_x(t)}{dt} = \omega_\phi(t) + r_z \frac{\kappa^2(t)}{a_x^2(t)} \times \\
\left[ \omega_\theta(t) \sqrt{1 - \lambda^2(t)} - \frac{1}{2} \frac{\sin 2\theta(t)}{1 - \lambda^2(t)} \frac{d|\lambda(t)|}{dt} \right]
\]

(74)

\[
\omega_y(t) \equiv \frac{d\phi_y(t)}{dt} = \omega_\phi(t) + r_z \frac{\kappa^2(t)}{a_y^2(t)} \times \\
\left[ \omega_\theta(t) \sqrt{1 - \lambda^2(t)} + \frac{1}{2} \frac{\sin 2\theta(t)}{1 - \lambda^2(t)} \frac{d|\lambda(t)|}{dt} \right].
\]

(75)

To derive these, we first note that the derivatives of (72) become

\[
\omega_x(t) = \omega_\phi(t) + 3 \left\{ \frac{d}{dt} \ln \left[ 1 + i \frac{b(t)}{a(t)} \tan \theta(t) \right] \right\}
\]

(76)

\[
\omega_y(t) = \omega_\phi(t) + 3 \left\{ \frac{d}{dt} \ln \left[ 1 + i \frac{a(t)}{b(t)} \tan \theta(t) \right] \right\}
\]

(77)

after pulling out the real parts of the quantities inside the natural logarithms. Then using

\[
\frac{d}{dt} \ln \left[ \frac{b(t)}{a(t)} \right] = \frac{d}{dt} \ln \left[ \frac{\sqrt{1 - |\lambda(t)|}}{\sqrt{1 + |\lambda(t)|}} \right] = - \frac{1}{1 - \lambda^2(t)} \frac{d|\lambda(t)|}{dt} = - \frac{d}{dt} \ln \left[ \frac{a(t)}{b(t)} \right]
\]

(78)

we find (74), (75) follow in a few straightforward lines of algebra.

The Cartesian frequencies have three terms: the orbital frequency \(\omega_\phi(t)\), a weighted version of the precession rate \(\omega_\theta(t)\), and an orientation-dependent term involving the deformation rate. The last term makes an exactly opposite contribution to the two power-weighted instantaneous frequencies \(a_x^2(t)\omega_x(t)\) and \(a_y^2(t)\omega_y(t)\). Owing to cancelation of these orientation-dependent terms, we readily obtain (62) for the bivariate instantaneous frequency using the definition (49) applied to the Cartesian analytic signal vector \([x(t) \quad y(t)]^T\).

Now for the instantaneous bandwidths, differentiating the logarithm of (70) gives

\[
v_x(t) = \frac{d}{dt} \ln \left[ \frac{\sqrt{1 - |\lambda(t)|}}{\sqrt{1 + |\lambda(t)|}} \right] = - \frac{1}{1 - \lambda^2(t)} \frac{d|\lambda(t)|}{dt} = - \frac{d}{dt} \ln \left[ \frac{a(t)}{b(t)} \right]
\]

(79)

\[
v_y(t) = \frac{d}{dt} \ln \left[ \frac{\sqrt{1 - |\lambda(t)|}}{\sqrt{1 + |\lambda(t)|}} \right] = - \frac{1}{1 - \lambda^2(t)} \frac{d|\lambda(t)|}{dt} = - \frac{d}{dt} \ln \left[ \frac{a(t)}{b(t)} \right]
\]

(80)

as expressions for Cartesian bandwidths \(v_x(t) \equiv \frac{d}{dt} \ln a_x(t)\) and \(v_y(t) \equiv \frac{d}{dt} \ln a_y(t)\), respectively. In each of these there are three terms, due to amplitude modulation, precession, and deformation, respectively; the contributions from each of the latter two terms are orientation-dependent. These expressions together with the multivariate instantaneous bandwidth definition (53) give (63) for the form of the bivariate bandwidth.

**Appendix B**

**Numerical Method**

This appendix describes the means of generating the estimated elliptical signal \(\tilde{z}(t)\) shown in Fig. 1 and Fig. 4. The reader is asked to refer to [12] for details; here we only give a brief overview together with parameter settings. The wavelet transform using a generalized Morse wavelet [35], [36] with parameter choices \(\beta = 3\) and \(\gamma = 3\) is performed on two real value time series \(x(t)\) and \(y(t)\). This transform is performed at fifty logarithmically spaced frequency bands ranging from one cycle per 53 days to one cycle per 2.6 days.

Wavelet ridge analysis consists of locating a maximum of the transform modulus across scale at each time, then connecting these points across time into a continuous chain [10], [11], [34]. The wavelet transform along this so-called “ridge curve” constitutes an estimate of a (univariate) modulated oscillatory signal. It was shown by [12] that way the wavelet ridge curves of the \(x(t)\) and \(y(t)\) time series, constructed separately, can be combined to form estimates of a modulated elliptical, or bivariate oscillatory, signal. Application of this method to the time series shown in Fig. 1 gives the estimated elliptical signal \(\tilde{z}(t)\).
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