Reframing the magnetotelluric phase tensor for monitoring applications: improved accuracy and precision in strike determinations

Ana G. Bravo-Osuna¹, Enrique Gómez-Treviño¹, Olaf J. Cortés-Arroyo¹, Nestor F. Delgadillo-Jauregui¹ and Rocío F. Arellano-Castro¹

¹ Departamento de Geofísica Aplicada, División de Ciencias de la Tierra, CICESE, Ensenada, Baja California, México. 22860.

2 Presently at Federal Institute for Geosciences and Natural Resources (BGR), Berlin, Germany.

ABSTRACT

The magnetotelluric method is increasingly being used to monitor electrical resistivity changes in the subsurface. One of the preferred parameters derived from the surface impedance is the strike direction, which is very sensitive to changes in the direction of the subsurface electrical current flow. The preferred method for estimating the strike changes is that provided by the phase tensor because it is immune to galvanic distortions. However, it is also a fact that the associated analytic formula is unstable for noisy data, something that limits its applicability for monitoring purposes, because in general this involves comparison of two or more very similar data sets. One of the issues is that the noise complicates the distribution of estimates between the four quadrants. This can be handled by sending all values to the same quadrant by adding or subtracting the appropriate amount. This is justified by showing that the analytic formula is also a least squares solution. This is equivalent to define penalty...
functions for the matrix of eigenvalues and then select the minima numerically. Contrary to the analytic formula this numerical approach can be generalized to compute strikes using windows of any number of periods, thus providing tradeoffs between variance and resolution. The performance of the proposed approach is illustrated by its application to synthetic data and to real data from a monitoring array in the Cerro Prieto geothermal field, México.

**Keywords**: Magnetotelluric monitoring Phase tensor Swift strike Galvanic distortion

**INTRODUCTION**

The evolution of the magnetotelluric (MT) method, from its scalar to its tensorial version, brought about issues regarding the best ways to process the impedance tensor to obtain information about the subsurface. In many instances what is required is to identify the directionality of the electric currents because this is related to lateral changes in electrical resistivity. In the case of two-dimensional (2D) models this translates into finding the strike angle for which the measured tensor reduces in some manner to a special case. When one of the axes of the coordinate system is parallel to the strike, the diagonal elements of the impedance tensor are zeroes. Swift’s (1967) approach to find the appropriate angle is to assume a rotated version of the measured tensor, which in general is not aligned along strike, and then look for the angle that best fits the 2D criterion.
The approach leads to an analytical formula. Another analytic formula that converts the elements of the impedance tensor into a strike angle was developed by Bahr (1988) by imposing the condition that the elements of the columns of the impedance tensor must have the same phases in the case of 2D structures. Groom and Bailey (1989) proposed an approach that numerically fits, in a least square sense, the measured impedances with an appropriate model that includes the strike angle. In this case the solution is not an analytic formula. Finally, there is the phase tensor of Caldwell et al. (2004) from which an analytic formula is derived for the strike angle.

The approaches described above for strike determinations differ from each other in several ways, and are also similar in other ways. Let us consider first the similarities between Bahr’s and the phase tensor approaches. Both are devised with the explicit purpose in mind of avoiding the effect of galvanic distortions. Also, they impose rigorous criteria in the sense that, strictly speaking, they can be met only by error-free data. A third similarity is that in both cases the approaches lead to analytic formulae. Let us now consider the similarities between the Swift’s and Groom-Bailey’s approaches. Both impose somewhat relaxed criteria as opposed to the exact requirements in the other two options. Instead of an exact fit a best fit is sought in the least squares sense. In Swift’s case, the anti-diagonal elements are not forced to be exactly zeroes but only to have their minimum possible amplitude. The Groom-Bailey approach also complies with a least squares criterion.
Presently it is almost a requirement to include phase tensor ellipses as part of the interpretation of any data set. Usually they are drawn over maps of the study area to reveal directionality of shallow or deep structures depending on the period of interest (e.g. Martí et al., 2020; Comeau et al., 2020). It is not uncommon for the axes of the ellipses to point in inconsistent directions because of random noise. Any improvement in this respect would be certainly welcome. However, where better determinations of strikes are most needed is in the recent application of the magnetotelluric method in monitoring applications. The preferred formula for monitoring purposes is that derived from the phase tensor (e.g. Peacock et al. 2012; 2013). One of the reasons is that there is no assumption about dimensionality. All other approaches assume 2D structures. Another reason is that it is immune to galvanic distortions, something that Swift’s formula is not. However, its drawback is that it is unstable for noisy data (Jones, 2012). This limits its applicability for monitoring purposes to very precise data, to ensure a precise estimation of strike. The question then arises of whether we can improve the estimation of strikes beyond the application of the analytic formula. All the analytic formulae provide strike estimates period by period. That is, each estimate and its variance are independent from the others. However, it is possible to link contiguous periods by assuming smooth variations of strike over period and produce a stable profile in the manner of the Occam philosophy (e.g. Constable et al., 1987). Muñiz et al. (2017) explored this path for the phase tensor using as seeds the estimations that are better constrained. Another possibility for stabilization is exemplified by the Groom-Bailey approach as
generalized by McNeice and Jones (2001); the estimates can be made period by period, for a given number of periods or for all periods together. The variance of the estimates generally diminishes as the number of periods increases. In a way, this follows the philosophy of Backus and Gilbert (1968) in the sense that the variance can be improved at the cost of resolution. In this paper we adhere to this viewpoint and look for estimates that can be made period by period, for a given number of periods, or for all periods at the same time.

**THEORY**

**Phase tensor**

The history of the magnetotelluric method is plentiful of examples where the problem to be solved consists of avoiding something undesirable. For instance, consider the chaotic variations of the source strength when making telluric measurements. The variations were neutralized by the inclusion of the magnetic field into the telluric method (Cagniard, 1953). In turn, the also chaotic polarizations of the source were taken care of by the impedance tensor (Cantwell, T., 1960). The not less chaotic distribution of small, near surface heterogeneities that produce galvanic distortions was neutralized by the phase tensor (Caldwell et al., 2004). The present work also centers on an avoidance. In this case, mitigate the effect of random noise in the estimations of strike using the phase tensor.

Let us start by a brief account of the development of the phase tensor itself. In relation to distortions, small, near surface heterogeneities means that
depth, size and distance to the electric line must be smaller than the skin depth. All things considered, local electromagnetic induction is small so that only galvanic effects due to electric charges are important. Physically, the source of the charges is the undistorted electric field, so their strength must be proportional to the strength of the source. In short, that the measured or distorted electric field \( E_m \) can be modeled as \( E_m = CE \), where \( C \) is a distortion matrix and \( E \) is the undistorted electric field. In terms of \( x \) and \( y \) components

\[
\begin{pmatrix}
E_{xm} \\
E_{ym}
\end{pmatrix} = \begin{pmatrix}
C_{11} & C_{12} \\
C_{21} & C_{22}
\end{pmatrix} \begin{pmatrix}
E_x \\
E_y
\end{pmatrix}.
\]

(1)

The elements of the distortion matrix \( C \) are real on behalf of the large skin-depth assumption. The charges are in phase and follow the undistorted electric field. The diagonal elements account for the distortion produced by a component to the same the component, while the off diagonal for the orthogonal contributions. To see how the distorted electric fields affect the impedance tensor, consider the undistorted fields expressed in terms of the impedances

\[
\begin{pmatrix}
E_x \\
E_y
\end{pmatrix} = \begin{pmatrix}
Z_{xx} & Z_{xy} \\
Z_{yx} & Z_{yy}
\end{pmatrix} \begin{pmatrix}
H_x \\
H_y
\end{pmatrix},
\]

(2)

where \( H_x \) and \( H_y \) are the corresponding components of the magnetic field and \( Z_{ij} \) represents the elements of the impedance tensor. The distorted impedance tensor can be written as
\[
\begin{pmatrix}
Z_{xmx} & Z_{xym} \\
Z_{ymx} & Z_{yym}
\end{pmatrix}
= \begin{pmatrix}
C_{11} & C_{12} \\
C_{22} & C_{22}
\end{pmatrix}
\begin{pmatrix}
Z_{xx} & Z_{xy} \\
Z_{yx} & Z_{yy}
\end{pmatrix}
\] (3)

The real elements of the distortion matrix affect only the amplitude of the impedances, but this is solely for the individual products. The actual linear combination can produce distorted impedances that resemble none of the originals. This explains the importance of the subject and the attempts to neutralize the effects of the distortion matrix. The formula developed by Bahr (1988) for the strike of 2D structures might be considered a precursor to the more general formula derived from the phase tensor. It happens that for the strike angle the product of the distortion matrix and the impedance tensor has a special property: the elements of each column must have the same phase. Imposing this condition on the measured impedance Bahr (1988) developed his formula for the strike which holds regardless of the distortion matrix. The phase tensor of Caldwell et al. (2004) generalizes this formula and provides other parameters equally immune to distortions. Separating the impedance tensor \( Z \) in terms of its real \( X \) and imaginary \( Y \) parts such that \( Z = X + iY \), the phase tensor is defined, regardless of dimensionality, as

\[
\Phi = X^{-1}Y,
\] (4)

Explicitly

\[
\begin{pmatrix}
\Phi_{11} & \Phi_{12} \\
\Phi_{21} & \Phi_{22}
\end{pmatrix} = \frac{1}{\text{det}(X)}
\begin{pmatrix}
X_{22}Y_{11} - X_{12}Y_{21} & X_{22}Y_{12} - X_{12}Y_{22} \\
X_{11}Y_{21} - X_{21}Y_{11} & X_{11}Y_{22} - X_{21}Y_{12}
\end{pmatrix},
\] (5)
where \( \det(X) = X_{11}X_{22} - X_{21}X_{12} \) is the determinant of \( X \).

The distorted impedance \( Z_m \) can be expressed as \( Z_m = CZ \). Separating the product into its real and imaginary parts \( Z_m = CX + iCY \), the distorted phase tensor can be written as \( \Phi_m = X^{-1}C^{-1}CY = \Phi \). That is, the galvanic distortions that so severely affect the impedance tensor have no effect whatsoever on the phase tensor. This property is largely responsible for its popularity in several applications of the magnetotelluric method. One of these applications is determining the strike direction. To this end, Caldwell et al. (2004) develop their analytic formula for the strike \( \theta \) as

\[
\theta = \alpha - \beta, \tag{6}
\]

where

\[
\alpha = \frac{1}{2} \tan^{-1} \left( \frac{\Phi_{12} + \Phi_{21}}{\Phi_{11} - \Phi_{22}} \right) \tag{7}
\]

and

\[
\beta = \frac{1}{2} \tan^{-1} \left( \frac{\Phi_{12} - \Phi_{21}}{\Phi_{11} + \Phi_{22}} \right). \tag{8}
\]

The formula for strike provided by the phase tensor reduces to that of Bahr’s (1988) when \( \beta = 0 \). It is in this sense that we mentioned above that the latter is a special case of the former. Our interest here is in the more general formula, particularly in relation to its vulnerability to random noise. Our curiosity
arose by analyzing the performance of several formulae for strike as presented by Jones (2012).

To illustrate step by step how the formula for strike is unstable we prepared a sequence of figures using synthetic data that corresponds to site 15 of the COPROD2S1 dataset (Varentsov, 1998). Figure 1 shows the elements of the impedance tensor which were distorted from the originals using the decomposition of $\mathbf{C}$ given by Groom and Bailey (1989). Table 1 summarizes this decomposition. The 2D undistorted impedance tensor $\mathbf{Z}_2$ is multiplied by the different operators to obtain the measured impedance $\mathbf{Z}_m$, which in this case is not actually measured but simulated. We chose the distorting parameters called twist and shear which along with the strike are expressed in degrees. The particular values are twist=20° and shear=30°, and an overall rotation of coordinates of -30°, so that the targeted strike is 30°. In addition, a relatively small random noise with normal distribution was added to the resulting impedances. The noise was computed as 1% of the off-diagonal impedances and added to all the elements. As stated earlier, the object of the exercise is to explore how these small errors amplify when computing the strike angle.
Table 1. The Groom and Bailey (1989) decomposition of the impedance tensor. $Z_m$ represents the measured impedance tensor, $R$ is the rotation matrix and $\theta$ is the strike angle. The distortion parameters are twist $t$, shear $e$ and the scaling factors $a$ and $b$. $Z_2$ is the undistorted 2D impedance tensor.
Figure 1. The elements of the impedance tensor correspond to site 15 of the COPROD2S1 dataset. The original impedances were distorted and random noise added as indicated.

a) Real parts. b) Imaginary parts. The values were computed as the mean of 30 realizations per period with their corresponding standard deviations. The continuous lines join the error-free values which in most cases are practically identical to the mean values. Black corresponds to $X_{11}$ and $Y_{11}$; blue to $X_{12}$ and $Y_{12}$; green to $X_{21}$ and $Y_{21}$; and red to $X_{22}$ and $Y_{22}$.
The first level of computations involves using the impedances in the formula for the phase tensor given by equation 5. The computed diagonal elements are shown in Figure 2a, where it can be observed that they are almost error-free for the first six periods and are noisier for the last six. The off-diagonal elements shown in Figure 2b behave very much the same. Notice that for the first six periods the elements are equal, indicating that the tensor is symmetric. In fact, they should be equal for all periods because the data were drawn from a 2D model. The fact that they are not for the last six periods is due entirely to the noise and to the distortions which introduce 3D effects.
Figure 2. a) Diagonal elements of the phase tensor derived from the impedance tensor shown in Figure 1. The color code is $\Phi_{11}$ green and $\Phi_{22}$ yellow. The black continuous line corresponds to error-free values. b) The corresponding off-diagonal elements. The color code is $\Phi_{12}$ green and $\Phi_{21}$ yellow. The black dashed line corresponds to error-free values.
We now turn to the ratios that define the angles $\alpha$ and $\beta$. The former involves the sum of the off-diagonal elements as numerator and the subtraction of the diagonal elements as denominator. Notice in Figure 3a that both quantities go through zero at a period of about 10 s. This will have implications when computing the ratio. Figure 3b illustrates the variations of the sum of the diagonal and the subtraction of the off-diagonal elements. This subtraction falls around zero for all periods, but no problem is expected when taking the ratio because for the angle $\beta$ the subtraction goes into the numerator. Notice that for the last six periods the subtraction has less scattering than the individual elements shown in Figure 2b.
Figure 3. a) Combination of the phase tensor elements that make up the ratio in the formula for the angle $\alpha$. The color code is $\Phi_{11} - \Phi_{22}$ green and $\Phi_{12} + \Phi_{21}$ yellow. The black continuous line corresponds to error-free values. b) Corresponding combination for the angle $\beta$. The color code is $\Phi_{11} + \Phi_{22}$ green and $\Phi_{12} - \Phi_{21}$ yellow. The black dashed line corresponds to error-free values.

The behavior of the ratios themselves is illustrated in Figure 4a. Both ratios are about uniform for all periods, as they should since they correspond to a uniform strike. The departures from uniformity are due to the effect of noise. The first observation about the ratio for the angle $\alpha$ is the relatively high noise for the period of 10 s and its two neighbors. This is where the diagonal elements $\Phi_{11}$ and $\Phi_{22}$ are very similar to each other, likely because of sensing a 1D structure. Both the numerator and denominator are close to zero for these periods and in particular the denominator. Turning now to the longest periods it can be observed that the ratio is very noisy for the last three periods. Again, this
is a reflection of the diagonal elements approaching each other, and likely in this case because of a diluting effect of the 2D structure in view of the increasing skin depth. Figure 4a also shows the ratio associated with the angle $\beta$. In this case the ratio is a lot less noisy mainly because the denominator is now the sum of the diagonal elements and not its difference as for the angle $\alpha$. Thus the strike angle $\theta = \alpha - \beta$ owes its sensitivity to noise mainly to the estimation of the $\alpha$ angle. The last step is to compute actual values of the strike and do the statistics. We compute the strikes for a number of realizations and take their arithmetic mean and its standard error for each period. The results are shown in Figure 4b. We present three scenarios: 1) random noise of 0.01 % using 30 realizations, 2) random noise of 1 % using 30 realizations, and 3) random noise of 1% using 1,000 realizations. The first essentially recovers the true strike because of the small error, the second recovers approximately the true strike only for the periods anticipated by the analysis of Figure 4a, and the third illustrates that increasing the number of realizations improves both accuracy and precision. However, one cannot but notice that for the period of 10 s the strike is underestimated in spite of the small error and the 1000 realizations. This point deserves further attention because it may be the tip of the iceberg of something deeper, in view that the ratios and the trigonometric function tangent are nonlinear combinations of the data and thus not easy to visualize.
Figure 4. a) The ratios that make up the formulae for the angles $\alpha$ and $\beta$. The color code is green for the angle $\alpha$ and yellow for the angle $\beta$. The black continuous line corresponds to error-free values. b) The final results for the strike $\theta = \alpha - \beta$. The red line corresponds to estimates assuming 1\% error and using 30 realizations. The black line corresponds to 1\% error and 1000 realizations. The dashed blue line corresponds to 0.01\% error and 30 realizations.
It is difficult just by inspection to predict the outcomes of a nonlinear formula. Consider for instance Figure 4a. There is a bias towards higher values for some of the realizations around the period of 10 s. One would think that increasing the number of realizations would simply fill out the empty spaces. In fact, this is what happens except for a small but significant difference. Figure 5a presents the estimates for the same 1 % error and 1000 realizations. It can be observed how some realizations are consistently negative for the period of 10 s. These values seem isolated outliers but actually they are not. Figure 5a also presents the case of 5 % error which illustrates that this is a systematic effect associated with how the strike is computed, and that it must be associated with a change in quadrant. At this level of error the results of using 30 or 1,000 realizations are comparable as illustrated in Figure 5b. This unconstrained use of the analytic formula cannot be used for doing statistics because of the bias introduced by the jump in quadrants.
Figure 5. The analysis of the formula for strike indicates that it is biased with respect to random noise: a) the individual strikes for each realization are plotted for two levels of error. The green lines correspond to 1% error and 1000 realizations. The yellow lines correspond to 5% error and 1000 realizations. The data were distorted assuming twist= 20° and shear =30°. The dashed line represents the true strike as computed with 0.01 % error. b) The black line represents the estimates of strikes using 5 % error and 30 realizations. The red line represents the estimates using 5 % error and 1,000 realizations, and the green dashed line corresponds to 0.01% and 30 realizations.
The overall analysis in this section applies only to site 15 of the COPROD2S1 data set. It does not pretend to be a general proof that one ratio or the other is most sensitive to random noise. We see it as a window into what is behind the unstable character of the formula for strike. In order to improve upon the performance of the formula we must deal with two issues. One is to neutralize the effect of the noise in relation to the change of quadrants. The other is to be able to use several contiguous periods at the same time which could further decrease the scattering of the estimations.

Reframing the phase tensor

Let us begin by recalling Swift (1967) formula for strike in terms of the elements of the impedance tensor. The first thing is to consider a rotated version of the impedance tensor through a rotation matrix $\mathbf{R}$. That is

$$
\begin{pmatrix}
Z'_{11} & Z'_{12} \\
Z'_{21} & Z'_{22}
\end{pmatrix} = \mathbf{R}(\theta) \mathbf{Z} \mathbf{R}^T(\theta).
$$

(9)

The second step is to solve for the strike angle using a least squares criterion. That is, finding $\theta$ that minimizes the penalty function

$$
C_{L2}(\theta) = |Z'_{11}(\theta)|^2 + |Z'_{22}(\theta)|^2.
$$

(10)
This last step acknowledges that the data is not ideal, in the sense that it may have some errors. The penalty function $C_{l_2}(\theta)$ has a minimum at all quadrants, and given the periodicity of the rotation matrix they are 90 degrees apart from each other. The formula for the strike $\theta = \alpha - \beta$ has the same periodicity, but there is no warranty that in the presence of noise the computed values correspond to some sort of minima. Swift (1967) analytic formula is

$$\tan 4\theta = \frac{2\text{Real}[(Z_{yy} - Z_{xx})(Z_{xy}^* + Z_{yx}^*)]}{|Z_{xy} + Z_{yx}|^2 - |Z_{yy} - Z_{xx}|^2}. \quad (11)$$

The asterisk stands for the conjugate of a complex number. It is stated in articles and textbooks that this follows from minimizing $|Z'_{xx}|^2 + |Z'_{yy}|^2$ or maximizing $|Z'_{xy}|^2 + |Z'_{yx}|^2$ (e.g. Vozoff, 1972; Gamble et al., 1979; Clarke et al., 1983; Simpson and Bahr, 2005; Chave and Jones, 2012). A least squares criterion does not impose a rigid equality between the quantities involved. This results in flexible solutions that can accommodate contradictory data. On the other hand, exact solutions that impose equality conditions are presumably unstable in the presence of noise. However, what if an expression is at the same time a least squares and an exact solution? This seems to be the case of equation 11 because Sims and Bostick (1969) obtained the same formula without recurring to the least squares criterion. They consider a special combination of the tensor elements and show that the quantity describes an ellipse when rotated. Then they find the angle of the major axis of the ellipse. The result is
equation 11. In what follows we show that equation 6, which is considered an exact, analytic solution for strike can also be considered as a least squares solution.

The expressions for $\alpha$ and $\beta$ in equation 6 were obtained originally by Bibby (1977) decomposing the direct-current resistivity tensor into symmetric and skew-symmetric parts. The angles are used by Caldwell et al. (2004) to factorize the phase tensor as

$$\Phi = R^T(\alpha - \beta) \begin{pmatrix} \Phi_{\text{max}} & 0 \\ 0 & \Phi_{\text{min}} \end{pmatrix} R (\alpha + \beta) ,$$

(12)

where $\alpha - \beta$ and $R$ are the strike angle and the rotation matrix, respectively. $\Phi_{\text{max}}$ and $\Phi_{\text{min}}$ are the singular values of the phase tensor. Notice that the rotation on the left is for the difference of the angles and that that on the right is for the sum of them. To make this product symmetric multiply on the right by $R (2\beta)$ so that

$$\Phi = R^T(\alpha - \beta) \begin{pmatrix} \Phi_{\text{max}} & 0 \\ 0 & \Phi_{\text{min}} \end{pmatrix} R (\alpha - \beta) R (2\beta) .$$

(13)

We now solve for the matrix of eigenvalues as

$$\begin{pmatrix} \Phi_{\text{max}} & 0 \\ 0 & \Phi_{\text{min}} \end{pmatrix} = R (\alpha - \beta) \Phi R^T (2\beta) R^T (\alpha - \beta) .$$

(14)

Assuming that the strike $\theta = \alpha - \beta$ is an unknown, the matrix of singular or principal values is no longer diagonal unless it corresponds to the appropriate strike. In general
Posed in these terms, the analytic formula for strike obtained by Caldwell et al. (2004) would correspond to setting $\Phi_{12}' = \Phi_{21}' = 0$. Using Swift’s criterion this would correspond to minimize

$$C_{l_2}(\theta) = \Phi_{12}'(T_i, \theta)^2 + \Phi_{21}'(T_i, \theta)^2 .$$

In other words, for every period $T_i$, we would look for the angle $\theta$ that minimizes the size of the anti-diagonal elements of the matrix defined by equation (14). One way to do it is to proceed as for equation 10 and obtain a formula for the least square solution. This is done in the Appendix with the result that the formula is the same as the well known formula for strike given by equation 6. The other way to proceed is to compute the angle $\beta$ using equation 8, and sample the penalty function for different values of $\theta$ and select numerically the angle where the function has its minima. This we call reframing the phase tensor for strike determinations.

Evaluating the analytic formula or locating the minimum of the penalty function should provide the same strike. This is in fact so but the way they do it is entirely different. This is illustrated in Figure 6 for two of the realizations shown in Figure 5, but using impedances from a single period. In other words, the data is a single impedance tensor altered by random error in two different ways. The
target is a strike of 30 degrees. We plot the penalty function from -90 to 90 degrees and the strikes provided by the analytic function. It can be observed that the two values provided by the formula fall exactly at one of the minima of the penalty function in each case. The important thing here is that even if the estimate falls outside the target quadrant, it falls exactly at another minimum. There is a minimum at each quadrant distant 90 degrees from the contiguous one.

Figure 6. Typical migration of quadrants due to errors in the data when using the analytic formula. The continuous lines represent the penalty functions $C_{i,j}(\theta)$ for a single period for two realizations. The circles correspond to the estimates using the analytic formula. The black circle falls on the first quadrant exactly at the minimum of the penalty function and the same happens for the red circle at the fourth quadrant. The dashed lines indicate that the periodicity of the penalty functions is $90^\circ$. 
To do proper statistics using either the analytic formula or the minima of the penalty function we need to place all estimates in the same quadrant. In the case of the formula this can be done by identifying the quadrant of a given estimate and adding or subtracting the proper amount to send it to the desired quadrant. This is illustrated in Figure 7a for seven realizations where all are sent to the first quadrant. This we call a constrained application of the analytic formula, as opposed to unconstrained when simply accepting the computed values. In the case of the penalty function the moving around of the estimates is not necessary. It is sufficient to limit the sampling to any of the quadrants, in this case the first. Figure 7b compares the performance of both approaches using 30 realizations. It can be observed that the difference between the corresponding estimates and their variances are not significant. In fact, they should be identical but are not because they were not computed using exactly the same distributions of 30 samples.
Figure 7. a) Penalty functions and concentration into quadrant I of the strike estimated using the analytic formula. The continuous lines represent the penalty functions for one period and seven realizations. The red circles represent the strikes from the analytic formula. b) The black curve represents the estimates using least squares and the red curve using the analytic formula. The green dashed line corresponds to the true strike.
The numerical approach does not add much to the constrained application of the formula, except that it saves the rearrangement of the estimates into a given quadrant. However, it opens the door to the possibility of using in the estimation of strike more than one period, something that the analytic formula cannot handle by itself. To use more than one period we minimize

\[ C_{L_2}(\theta) = \sum_i \left( \Phi'_{12}(T_i, \theta) \right)^2 + \left( \Phi'_{21}(T_i, \theta) \right)^2. \]  

(17)

The summation is over any number of periods, depending on how wide we desire a windows to be. Figure 8 illustrates how the performance of the phase tensor improves when using windows of several periods. In general, if \( n_T \) is the total number of periods and \( n_p \) is the number of periods of the window, there will be \( n_T - n_p + 1 \) windows over which the strike can be estimated. Figure 8a illustrates the case for \( n_T = 12 \) and \( n_p = 2 \) so that there are 11 windows. The estimates are plotted at the geometric mean of the first and last period. When \( n_p = 1 \) there are 12 windows, just as when applying the analytic formula. This we call windows of one period. It can be observed in Figure 8b that wider windows perform even better, both in accuracy and precision. We present an example of uniform windows but the application is open for combination of sizes, either fixed beforehand or using nested estimations controlled by their variance. Figure 8 also illustrates that the estimates using windows of several periods are not arithmetic averages of the values provided by the analytic formula shown in
Figure 7b. The joint optimization of several periods performs better than individual estimations or combinations of them.

![Graph showing strike estimates with error bars for two and four periods.]

**Distortion:** Twist $20^\circ$, Shear $30^\circ$. Random error 5%.

**Graph a)**: Least squares estimates assuming windows of two periods. The green dashed line represents the true strike.

**Graph b)**: Least squares estimates assuming windows of four periods.

Figure 8. a) Least squares estimates assuming windows of two periods. The green dashed line represents the true strike. b) Least squares estimates assuming windows of two periods. The green dashed line represents the true strike.
Another improvement that can be easily implemented is the change of norm in equation 17. Analytically, switching from a $L_2$ to a $L_1$ norm is a much elaborated mathematical process. However, numerically this is a trivial change. Instead of minimizing the penalty function given in equation 17 we can minimize

$$\mathcal{L}_{L_1}(\theta) = \sum_i \left( |\Phi'_{12}(T_i, \theta)| + |\Phi'_{21}(T_i, \theta)| \right).$$

(18)

In the presence of outliers it is well known the property of the $L_1$ norm of providing better estimates, because the $L_2$ norm by its very nature amplifies the effect of outliers.

At this point it is worth recalling that even though we are applying the same criterion to impedances and to elements of the phase tensor, there are qualitative differences between the two types of data. Swift’s formula operates directly on the elements of the impedance tensor. On the other hand, the condition imposed on the phase tensor is applied to combinations of ratios of the impedances. Figure 9a illustrates that working with impedances produces strikes that are more stable than those obtained with the elements of the phase tensor. We used in this case data with no galvanic distortions to isolate the precision aspect of the comparison. Figure 9b illustrates the performance when the data are affected by galvanic distortions. It can be observed that while the size of the errors is still somewhat smaller for impedances, their mean values are way off the target. On the other hand, the strikes from the phase tensor are more
accurate but have less precision. This lose of precision seems to be the penalty for being immune to galvanic distortions.

Figure 9. Relative performance of the least squares approach as applied to the impedance tensor and the phase tensor. a) With no galvanic distortions. The green dashed line represents
the true strike. The black continuous line represents the estimates using the impedance tensor. The red line corresponds to the estimates using the analytic formula derived from the phase tensor. b) With galvanic distortions. The green dashed line represents the true strike. The black continuous line represents the estimates using the impedance tensor. The red line corresponds to the estimates using the analytic formula derived from the phase tensor.

We now turn to explore the recovery of a profile that varies with period. We use this same profile later when comparing recoveries that are very close to each other. The target strikes are 20, 30 and 40 degrees, assuming error-free data and with small to moderate errors. Figure 10a illustrates the behavior of the penalty function for the error-free case. The minima occur exactly at the 20, 30 and 40 degrees targets, as they should. Although the data has no errors, inspecting the penalty function it is possible to predict which strike value will be the least well determined when including errors. This will be the strike of 20 degrees because it has the flattest penalty function. Then follows 40 degrees and the best determined will be 30 degrees because it has the sharpest penalty functions. This is confirmed by Figure 10b which corresponds to data assuming a relatively small error of 1% in the impedances. We present penalty functions for only 5 realizations per period to avoid crowding, but somehow these are sufficient to corroborate the predictions mentioned before. It can be observed that the wider scattering of the minima occurs around 20 degrees. Then it is followed by those for 40 degrees and finally by the less scattered around 30 degrees. This is the same sequence anticipated by the relative flatness of the
error-free penalty functions. This is in turn reflected in the size of the error bars in Figure 11. The largest errors are for the strike of 20 degrees followed by 40 degrees, and 30 degrees being the best constrained strike. In this case we used 30 realizations.

Figure 10. Variation of the penalty functions for a profile of strikes of 20, 30 and 40 degrees. a) No random error added to the impedances. The minima of the penalty function fall, as
they should, exactly at the assumed strikes. b) With random errors of 1%. The minima in this case spread around the assumed values. The degree of dispersion in each case reflects the accuracy and precision of the estimates of strike.

Figure 11. The estimates of strike correspond to the penalty functions shown in Figure 10b, except that in this case we used 30 realizations. See the main text for a discussion on how the dispersion of the penalty function relates to the accuracy and precision of the final estimates.

APPLICATION TO DIFFERENTIAL SYNTHETIC DATA

We now turn to the target of detecting changes using the profile discussed above. We place as target a one degree change in strike, using the same site 15 of the COPROD2S1 synthetic dataset, distorted with a twist of 20 degrees and a shear of 30 degrees. Random noise of 5% is added to the distorted impedances.
We begin with the unconstrained analytic formula. It can be observed in Figure 12a that only for the strike of 30 degrees we have the possibility of hitting the target of one degree difference. The other two have too large variances and most estimates are way off the desired targets. Of course, we know that this is not the proper way to use the formula. Figure 12b illustrates the performance of the reframed version using windows of one period. It can be observed that for the group of 20 degrees the estimations are almost as bad as those provided by the unconstrained analytic formula. For the group of 30 degrees the estimations are comparable for the two approaches. This means that for these four periods there are no shifts of quadrants in the analytical solutions. The shifts return for the last four periods. We include this comparison between the unconstrained analytic formula and estimates using windows of one period just for the sake of completeness. In fact, as discussed earlier, the analytic formula when constrained to a single quadrant produces the same estimate as the windows of one period. It is when using wider windows that the variance can be actually reduced.
Figure 12. Recovery of strikes for two profiles that differ by one degree over all periods: a) unconstrained application of the analytic formula. The black line corresponds to the estimates for the base profile of 20, 30 and 40 degrees. The red line corresponds to the estimates of the perturbed profile of 21, 31 and 41 degrees. b) Using the reframed version for windows of one period. The black line corresponds to the estimates for the base profile of 20, 30 and 40 degrees. The red line corresponds to the estimates of the
perturbed profile of 21, 31 and 41 degrees. In both cases the dashed lines represents the corresponding true values.

We present results using windows of even number of periods from four to ten. It is difficult to predict for a non-uniform profile the effect of jumps of strikes when using windows of several periods because of possible incompatibilities. We don’t know how the least square criterion can handle very different values at the same time, and what the effect would be on the variances, which are crucial for detecting changes. The estimations using windows of four and six periods are shown in Figure 13a. We will only point out that the estimations improve for most of the windows except for the first and possibly for the second. Further improvements can be observed in Figure 13b for windows of eight and ten periods. Considering these sequence of results, the overall lesson of the exercise is that we should experiment with windows of different number of periods to obtain changes that are statistically significant.
Figure 13. Recovery of strikes using windows of different widths. The dashed lines correspond to the base profile of 20, 30 and 40 degrees and the continuous ones to the altered or perturbed profile of 21, 31 and 41 degrees. a) Windows of four and six periods. b) Windows of eight and ten periods.
APPLICATION TO FIELD DATA

To illustrate the performance of the approach when using field data we selected a site from a monitoring network installed around the Cerro Prieto geothermal field, located in northern Baja California, México. Beginning in 2015 with one station the network grew to include up to eighteen by 2018, as new stations were built at the facilities of CICESE. Electric fields were measured using 25-m dipoles with a common central Pb-PbCl$_2$ electrode in “L” array, while magnetic fields used BF-4 type coils. Details about the Pb-PbCl$_2$ electrodes can be found in Petiau (2000) and Booker and Burd (2006) and about the BF-4 type coils in the operation manual of the MT-1 system (Electromagnetic Instruments Inc., 1996). The sampling frequency was 18.18 Hz. Only two stations were equipped with induction coils but all were synchronized for later processing of the time series using remote reference techniques (Gamble et al., 1979). Specifically, we used the RRRMT8 algorithm developed by Chave et al. (1987) and Chave and Thomson (1989). More details about the network are described by Cortés-Arroyo (2018) and Cortés-Arroyo et al. (2018).

We chose to compare two-week data from 2017 with also two-week data from 2018 for a station labeled E-4. The comparison is shown in Figure 14 using windows of one period. In all cases we assumed a 5 % error in the data to compute the errors in the strikes. It can be observed that the estimations are very stable for periods between 10 and 100 seconds. In fact it can be stated that the strike didn’t change from one year to the other in that range of periods. However,
for shorter and longer periods the story is very different because of the large oscillations in the strike estimates for both years.

![Figure 14](image)

Figure 14. Strike estimations using field data from two consecutive years using windows of one period. The red lines correspond to the year 2017 and the blue lines to the year 2018.

The strikes are a lot smoother when we use windows that include several periods. There is a total of 36 periods, so choosing windows of 6, 12 and 18 would provide 31, 25 and 19 estimates, respectively. The resulting strikes for these windows are shown in Figure 15. The estimates for the shortest window already show a definite trend for the difference between the two years as illustrated in Figure 15a. This happens for periods shorter than 10 seconds, which for the area corresponds to depths of penetration of less than 3-4 km. This would locate the changes within the depth range of the Cerro Prieto geothermal
field, with no appreciable change for the deeper and more regional background. It is intriguing that the curves cross each other, implying that the strike went a few degrees to one side for the deeps and to the opposite side for the shallows. This trend is reproduced for the wider windows of 12 and 18 periods as shown in Figures 15b and 15c, respectively. However, notice that the negative changes tend to disappear for the widest window, as indicated by the lowering of the red curve of 2017 towards the blue of 2018. Notice also the stability and consistency of the strikes for the longer periods, implying a stable background with a consistent strike for the two years.

Figure 15. Strike estimations using field data from two consecutive years. a) Using windows of 6 periods. b) Using windows of 12 periods. c) Using windows of 18 periods. The red lines correspond to the year 2017 and the blue lines to the year 2018.
CONCLUSION

The unstable character of the formula for strike derived from the phase tensor is not intrinsic to the phase tensor. It is possible to keep the assets of the phase tensor in regards to dimensionality and to its immunity to galvanic distortions by reframing how the strike is estimated. One of the problems to overcome is the effect of random noise which may change the quadrants of the estimated strike. This can be addressed by sending the estimates from any quadrant to a chosen one, which we call the constrained analytical formula. This is justified because the analytic formula for strikes is also a least squares solution. Regardless of the quadrant any estimate falls on a minimum of the least squares penalty function. This means that choosing the minima of the penalty function over any quadrant is equivalent to sending to the same quadrant the estimates of the analytic formula. Using the minima of the penalty function allows the estimation to be made using several periods at the time. This estimation over windows of several periods, something intrinsically impossible using the analytic formula, allows the recovery of strikes with lower variances. Thus the least squares approach certainly brings something new to the estimation of strike. Regarding uncertainties we didn’t explore beyond the standard deviation assuming normal distributions. The subject is open for future work using more elaborated methods as stochastic, Bayesian or Markov chains approaches for sampling and estimating parameters and their uncertainties. Another extension would be the implementation of nested algorithms for either imposing continuity of strikes over period, or getting the optimal combination of window sizes for a given variance.
DECLARATIONS

Availability of data and materials

The time series of electric and magnetic fields from which Figures 15 and 16 were obtained are available at CEMIE-Geo’s repository and belong to CICESE. In order to have access to the data, it would be necessary to make an agreement with CICESE. Please contact Dr. José Romo (jromo@cicese.mx). A Matlab code for the computation of strikes over windows of several periods is available from the corresponding author upon request.

Abbreviations

MT: Magnetotelluric
2D: Two-dimensional
3D: Tree-dimensional
TE: Transverse electric
TM: Transverse magnetic

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Author information

Affiliations

Departamento de Geofísica Aplicada, División de Ciencias de la Tierra, CICESE, Ensenada, Baja California, México. 22860.

Ana G. Bravo-Osuna, Enrique Gómez-Treviño, Nestor F. Delgadillo-Jauregui and Rocío F. Arellano-Castro

Federal Institute for Geosciences and Natural Resources (BGR), Berlin, Germany. Olaf J. Cortés-Arroyo

Contributions

All authors contributed about equally to this work.

All authors read and approved the final manuscript.
Corresponding author
Ana G. Bravo-Osuna

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