

# LOCATING AND DETERMINING DIMENSIONALITY OF UXO USING TIME DOMAIN ELECTROMAGNETIC INDUCTION

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## Abstract

We assume that the Time Domain Electromagnetic (TEM) response of a buried axisymmetric metallic object can be modelled as the sum of two dipoles centered at the midpoint of the body. The strength of the dipoles depends upon the relative orientation between the object and the source field, and also upon the shape and physical properties of the body. Upon termination of the source field, each dipole is assumed to decay as  $k(t + \alpha)^{-\beta} e^{-t/\gamma}$ . The parameters  $k$ ,  $\alpha$ ,  $\beta$  and  $\gamma$  depend upon the conductivity, permeability, size and shape of the object, and these can be extracted from the measurements by using a nonlinear parametric inversion algorithm. Investigations carried out using an analytic solution for a sphere and laboratory measurements of steel and aluminum rectangular prisms, suggest the following two-step methodology: (1) The value of  $\beta$  is first used as a diagnostic to assess whether the metallic object is non-magnetic or magnetic, (2) the ratios of  $k_1/k_2$  and  $\beta_1/\beta_2$  are then diagnostic indicators as to whether the geometry is plate-like (uninteresting) or rod-like (a high candidate for being a UXO). Results from the application of this algorithm to a TEM field data set acquired at the United States Army Corps of Engineers Environmental Research and Development Centre UXO Test Site have successfully identified a UXO to be magnetic and rod-like.

## Introduction

An explosive ordnance is a munition that is either launched or fired with the intent of detonation at a specified target. An unexploded ordnance (UXO) is an explosive ordnance that, due to some malfunction, remains undetonated. As a result, the ordnance can be found at the ground surface, partially buried, or buried at a depth of up to 8 m beneath the surface. The remediation of UXO-contaminated land has been made a high priority by the United States Department of Defense in order to either maintain safe usage for continuing military operations or to permit land transfer to the private sector. Practical and cost-effective strategies for remediation require both detection of possible targets and the ability to discriminate between UXO and contaminating scrap metal.

The detection of buried metallic objects can be accomplished with a variety of geophysical sensing techniques. Time domain electromagnetic induction (TEM) surveys have been successful in detecting both ferrous and non-ferrous metallic objects near the soil surface, and are a mainstay amongst technologies currently utilized in UXO clearance projects. In the TEM method a time varying magnetic field is used to illuminate a

conducting target. This primary field induces surface currents on the target which then generate a secondary magnetic field that can be sensed above ground. With time, the surface currents diffuse inwards, and the observed secondary field consequently decays. The rate of decay, and the spatial behavior of the secondary field, is determined by the target's conductivity, magnetic permeability, shape, and size.

Identification of a UXO from electromagnetic sensor data remains a major hurdle in reducing the high costs of remediation projects. It has been reported that approximately 70% of remediation costs are currently being used to excavate non-ordnance items (Butler *et al.*, 1998). The development of discrimination algorithms can be roughly categorized as either model-based or data-based. Data-based algorithms are pattern recognition procedures that compare a library of catalogued responses from various UXO items to measured responses (for example Damarla and Ressler, 2000). Model-based algorithms use either an exact or approximate forward modeling algorithm to determine a set of model parameters needed to replicate the measured responses, and subsequently relating the model parameters to physical parameters (Khadr *et al.*, 1998). One such model-based technique that has been the focus of much recent research is the determination of the time constants of the TEM response, or equivalently the poles of the frequency domain signal, to identify the buried target (Snyder *et al.*, 1999; Baum, 1997; Collins *et al.*, 1999). A method that represents a hybrid of the model-based and data-based algorithm is under development at Blackhawk Geometrics (Grimm, 2000). In that approach, a spheroid modeller, working jointly with a model-based inversion algorithm, generates a library of model parameters which can then be operated upon by a neural network classifier for comparison with parameters derived from the raw sensor signal.

In this paper we present a model-based TEM data interpretation algorithm which estimates the basic shape (rod-like or plate-like) and magnetic character (ferrous or non-ferrous) of a buried metallic object. We first present an approximate forward model that represents the time domain response of a metallic object as a pair of perpendicular dipoles located at the center of the buried target. This form of model was suggested to us in a personal communication from J.D. McNeill. The strengths of these dipoles decay with time, and the parameters that govern the time decay behavior are related to the conductivity, permeability, shape, and size of the buried target. Our parameterization is simple, and thus convenient to use in data fitting procedures. We next describe an inversion scheme to recover the model parameters from TEM data. Since these parameters encapsulate information about the physical attributes of the target, we can attempt to use them to determine if the target is ferrous and if the geometry is rod-like (most likely a UXO) or plate-like (most likely a non-ordnance item). Empirical relationships are developed that link the model parameters to the physical parameters of the target and these relationships form the basis of our algorithm. We conclude with the application of the algorithm to a synthetic data set contaminated with noise, and field data sets taken over a buried UXO and buried metallic scrap.

### **Development of Approximate Forward Modelling**

In order to invert measured TEM data for the physical parameters of the target, it is necessary to have a forward model to describe the TEM response for a buried metallic object. We can restrict our search for response solutions to axi-symmetric metallic targets, since this geometric subset adequately describes all UXO and the majority of buried metallic scrap encountered in a remediation survey. Unfortunately analytic expressions for the time domain response are restricted to a metallic sphere, and even an expression for a permeable

and conducting non-spherical axi-symmetric body is not available. Numerical solutions of Maxwell's equations, under continual development, are promising (e.g. Haber, 2000; Carin, 2000; Hiptmair, 1998), however, the computational time requirements for obtaining a solution still make them impractical for use as part of a rigorous inversion procedure. Our approach, therefore, is to use an approximate forward modelling that can adequately reproduce the measured electromagnetic anomaly in a minimal amount of time. The validity of this reduced modelling still needs further testing but the empirical tests carried out here suggest that it can be useful in practice.

The development of the approximate forward modelling is presented in four steps. We begin with the response of a sphere, so that the magnetic polarization dyadic  $\bar{\bar{\mathbf{M}}}$  is introduced. This dyadic is then altered so that it is applicable to an axi-symmetric body. This generates the "two-dipole" model mathematically. Next we introduce a parameterization for the time decays of each of the two dipoles and finally, we combine everything to generate our approximate forward modelling.

### 1. Response for a Spherical Body

Consider a permeable and conducting sphere of radius  $a$  illuminated by a uniform primary field  $\mathbf{B}^{\mathbf{P}}$  (fig. 1(a)). At a time  $t = 0$  the primary field is terminated, and eddy currents are induced in the sphere; they subsequently decay due to the finite conductivity of the sphere. The secondary field  $\mathbf{B}^{\mathbf{S}}$  generated by the decaying currents is dipolar:

$$\mathbf{B}^{\mathbf{S}}(t) = \frac{\mu_o}{4\pi r^3} \mathbf{m}(t) \cdot (3\hat{\mathbf{r}}\hat{\mathbf{r}} - \bar{\bar{\mathbf{I}}}) \quad (1)$$

where  $\mathbf{m}(t)$  is the dipole moment induced at the center of the sphere at time  $t$ ,  $r$  is the distance between the observation point and the sphere center,  $\hat{\mathbf{r}}$  is the unit vector pointing from the sphere center to the observation point  $P$ , and  $\bar{\bar{\mathbf{I}}}$  is the identity dyadic. The dipole moment is

$$\mathbf{m}(t) = \frac{2\pi}{\mu_o} \mathbf{B}^{\mathbf{P}} L^{\mathbf{B}}(t) \quad (2)$$

where

$$L^{\mathbf{B}}(t) = 6a^3 \mu_r \sum_{s=1}^{\infty} \frac{e^{-\frac{q_s t}{\tau}}}{q_s + (\mu_r - 1)(\mu_r + 2)} \quad (3)$$

where  $\tau = \sigma \mu a^2$ , and  $\mu_r = \mu / \mu_o$  is the relative permeability (Kaufmann, 1985). In general the magnetic permeability of highly permeable materials is a function of many parameters, including the strength of the incident magnetic field, temperature, and magnetic history. However, calculated TEM responses assuming a constant permeability of  $\mu_r = 150$  for steel and  $\mu_r = 1$  for aluminum compared well with laboratory TEM measurements of steel and aluminum targets (Pasion, 1999). Therefore we feel that eqs. (1) to (3) are suitable for the analysis that follows here. The values  $q_s$  are roots to the transcendental equation

$$\tan q_s = \frac{(\mu_r - 1) q_s}{q_s^2 + (\mu_r - 1)}. \quad (4)$$

Equations (1) to (4) reveal that the  $\mathbf{B}$ -field of a sphere in a uniform primary field is equivalent to the  $\mathbf{B}$ -field of a single magnetic dipole located at the center of the sphere and oriented parallel to the primary field.

For convenience we write the relationship between the induced dipole and the primary field as  $\mathbf{m} = \bar{\bar{\mathbf{M}}} \cdot \mathbf{B}^P$ , where  $\bar{\bar{\mathbf{M}}}$  is the magnetic polarizability dyadic. For a sphere,

$$\bar{\bar{\mathbf{M}}} = \frac{2\pi}{\mu_o} L^{\mathbf{B}}(t) \bar{\bar{\mathbf{I}}} = \frac{2\pi}{\mu_o} \begin{bmatrix} L^{\mathbf{B}}(t) & 0 & 0 \\ 0 & L^{\mathbf{B}}(t) & 0 \\ 0 & 0 & L^{\mathbf{B}}(t) \end{bmatrix}. \quad (5)$$

Baum (1999) details the characteristics of the magnetic polarizability dyadic, and notes that the triple degeneracy of the magnetic polarizability dyadic reflects the symmetry of the sphere.

The sphere solution possesses several characteristics that we retain in the formulation of our approximate solution for an axi-symmetric target. Firstly, the secondary field due to the induced currents generated in a sphere, illuminated by a uniform, step-off primary field, is dipolar at all points outside the sphere. We will also represent the secondary field for more general shapes as a dipolar field (eq. (1)). A dipolar field approximation is reasonable for any observation point far enough away from any localized current distribution (Jackson, 1975), and it has been reported that for observation points greater than 1 to 2 times the target length, a dipolar field assumption is adequate (Casey, 1999 or Grimm et al., 1997). Indeed, higher order multipoles induced in a target will decay at early times (Grimm et al., 1997).

Secondly, the induced dipole moment in the center of a sphere is given by the dyadic product  $\bar{\bar{\mathbf{M}}} \cdot \mathbf{B}^P$ . This form indicates that the induced dipole is proportional to the projection of the primary field along the direction of the induced dipole. The components of  $\bar{\bar{\mathbf{M}}}$  scale the strengths of the dipoles. The magnetic polarizability dyadic, in the case of the sphere, contains the function  $L^{\mathbf{B}}(t)$  that contains all the information about the time decay of the sphere and it depends upon the material properties, shape, and size of the target. Our hypothesis is that more general metallic shapes can also be approximately modelled with an induced dipole equal to the dyadic product  $\bar{\bar{\mathbf{M}}} \cdot \mathbf{B}^P$ . However, choosing the right functional form of  $\bar{\bar{\mathbf{M}}}$  will be crucial.

## 2. Approximating $\bar{\bar{\mathbf{M}}}$ for an Axi-Symmetric Body

Analytic expressions for  $\bar{\bar{\mathbf{M}}}$  for the time domain response of a permeable and conducting non-spherical axi-symmetric body are not available. Therefore we base our form of  $\bar{\bar{\mathbf{M}}}$  on the magnetostatic polarizability for a spheroid. Recall that the time domain response of a sphere the structure of  $\bar{\bar{\mathbf{M}}}$  is identical to the structure the polarizability dyadic of a magnetostatic sphere. The analytic solution for the magnetostatic response of a magnetic prolate spheroid is equivalent to the field of a magnetic dipole induced at the spheroid center (Das *et al.*, 1990):

$$\begin{aligned} \mathbf{m}^{spheroid} &= \mathbf{m}_1 + \mathbf{m}_2 \\ &= k_1 [(\hat{\mathbf{z}}' \cdot \mathbf{B}^P) \hat{\mathbf{z}}'] + k_2 [(\hat{\mathbf{y}}' \cdot \mathbf{B}^P) \hat{\mathbf{y}}' + (\hat{\mathbf{x}}' \cdot \mathbf{B}^P) \hat{\mathbf{x}}'] = \begin{bmatrix} k_2 & 0 & 0 \\ 0 & k_2 & 0 \\ 0 & 0 & k_1 \end{bmatrix} \cdot \mathbf{B}^P \end{aligned} \quad (6)$$

where  $k_1$  and  $k_2$  are the polarizability constants, which are functions of the conductivity, permeability, shape, and size. Eq. (6) reveals that the total induced dipole can be written as the sum of two orthogonal dipoles  $\mathbf{m}_1$  and  $\mathbf{m}_2$ . The first dipole moment  $\mathbf{m}_1$  is parallel to the major axis ( $\hat{\mathbf{z}}'$  in fig. 1(b)) of the spheroid, and its strength is proportional to the product of the primary field along that direction and the polarizability  $k_1$ . The second dipole moment is perpendicular to the major axis, and its magnitude is proportional to the component

of the primary field along that direction and the polarizability  $k_2$ . A consequence of  $k_1$  and  $k_2$  being functions of the spheroid's shape and size is that the orientation of the effective dipole will not be solely determined by the direction of the primary field, as is the case for a sphere. In addition, the orientation of  $\mathbf{m}^{spheroid}$  will be influenced by the aspect ratio of the spheroid.

The polarization dyadic in eq. (6) suggests a magnetic polarization dyadic for the TEM problem of the form

$$\bar{\mathbf{M}} = \begin{bmatrix} L_2(t) & 0 & 0 \\ 0 & L_2(t) & 0 \\ 0 & 0 & L_1(t) \end{bmatrix} \quad (7)$$

where we have simply replaced  $k_1$  and  $k_2$  in eq. (6) with the dipole decay functions  $L_1(t)$  and  $L_2(t)$ . The resultant induced dipole moment for this definition of the magnetic polarization dyadic is then

$$\begin{aligned} \mathbf{m}(t) &= \mathbf{m}_1(t) + \mathbf{m}_2(t) \\ &= L_1(t) [(\hat{\mathbf{z}}' \cdot \mathbf{B}^p) \hat{\mathbf{z}}'] + L_2(t) [(\hat{\mathbf{y}}' \cdot \mathbf{B}^p) \hat{\mathbf{y}}' + (\hat{\mathbf{x}}' \cdot \mathbf{B}^p) \hat{\mathbf{x}}'] \end{aligned} \quad (8)$$

Therefore, our approximate forward model represents the TEM response of two orthogonal dipoles. The first dipole is parallel to the symmetry axis of the target, and the second dipole is perpendicular to the symmetry axis. These dipoles decay independently according to the decay laws  $L_1(t)$  and  $L_2(t)$ , respectively.

By choosing the appropriate parameters, this 'two-dipole' model produces TEM responses that are consistent with those observed field measurements of UXO. It has been noted that the shape anomaly of the measured response for UXO changes with time (Grimm et al., 1997). The physical phenomena that gave rise to the temporal changes in shape anomaly was explained in terms of the nature of the induced eddy currents. Eddy currents that circulate end-to-end in the UXO dominate at early time but decay away quickly, while eddy currents that circulate about the long axis extend later into time. This observed field behavior can be duplicated by letting the two orthogonal dipoles  $\mathbf{m}_1(t)$  and  $\mathbf{m}_2(t)$  decay independently of each other. The dipole  $\mathbf{m}_1(t)$  is parallel to the long axis and it simulates the magnetic fields that arise from currents circulating about the axis. The dipole  $\mathbf{m}_2(t)$  is perpendicular to the long axis and it simulates the magnetic fields that arise from currents circulating end-to-end. By assigning a different decay characteristic (governed by its decay parameters) to each dipole, the relative contribution by each dipole to the secondary field can vary with time.

### 3. Time Decay Functions $L_1(t)$ and $L_2(t)$

The time decay for a sphere is determined by the sum of exponentials. This result generalizes to the case of a conductive body of arbitrary size and shape in an insulating medium illuminated by a step-off primary field (Kaufman, 1994). Thus the form for  $L(t)$  should, at least, be able to duplicate the time decay features observed for the sphere. Plots of the  $\mathbf{B}$ -field and  $\partial\mathbf{B}/\partial t$  response for both a magnetically permeable (e.g. steel) and non-permeable (e.g. aluminum) sphere can be found in fig. 2.

An appropriate form of the decay law for the  $\mathbf{B}$ -field is

$$L(t) = k (t + \alpha)^{-\beta} e^{\frac{-t}{\gamma}}. \quad (9)$$

The parameter  $k$  controls the magnitude of the modelled response. The three parameters  $\alpha$ ,  $\beta$ , and  $\gamma$ , control the duration and characteristics of the three different stages of the time decay curve. The duration of the relatively flat early time stage is proportional to the parameter  $\alpha$ . The linear decrease of response observed during the intermediate time stage is determined by  $t^{-\beta}$ . The exponential decay characterizing the late time stage is controlled by the parameter  $\gamma$ . Fig. 2(a) demonstrates the ability to reproduce the secondary  $\mathbf{B}$ -field. This form of the decay law, with the  $\alpha$  parameter absent, was suggested to us in a personal communication from J.D. McNeill.

The time derivative  $\partial\mathbf{B}/\partial t$ , which is measured directly with most TEM receivers can also be modelled with eq. (9). Fig. 2(b) includes plots of the  $\partial\mathbf{B}/\partial t$  curves for a steel and an aluminum sphere. The early time behaviour for the non-permeable sphere follows a  $t^{-1/2}$  decay and so these curves are different from those of  $\mathbf{B}$  in fig. 2(a). Nevertheless the curves are still represented by early time turn-overs, and linear and exponential decays that can be accommodated by eq. (9). The suitability is demonstrated by the fit between the laboratory measured response and a predicted response obtained by evaluating eq. (9).

In the following section we generically denote the TEM response as  $\xi(\mathbf{r}, t)$  where  $\xi$  can be the magnetic field or its time derivative. The time dependent decay of  $\xi$  is given by eq. (9).

#### 4. The Approximate Forward Model

With the above work, we can write an approximate expression for the secondary field response of an axi-symmetric target. First, let us switch from the body-fixed (primed) coordinate system to a space-fixed coordinate system, which is more amenable to the definitions of target and sensor location of a typical field survey (fig. 3). A vector  $\mathbf{v}'$  in the body-fixed system co-ordinate system is related to a vector  $\mathbf{v}$  in the space-fixed co-ordinate system via the Euler rotation tensor  $\mathbf{A}(\psi, \theta, \phi)$  by (Arfken, 1985)

$$\mathbf{v}' = \mathbf{A} \mathbf{v}. \quad (10)$$

Due to the axial symmetry of the problem  $\psi = 0$ , and the Euler rotation tensor can be written

$$\mathbf{A} = \begin{bmatrix} \cos \theta \cos \phi & \cos \theta \sin \phi & -\sin \theta \\ -\sin \phi & \cos \phi & 0 \\ \sin \theta \cos \phi & \sin \theta \sin \phi & \cos \theta \end{bmatrix}, \quad (11)$$

where  $\theta$  is the angle between the symmetry axis of the target ( $\hat{\mathbf{z}}'$  in fig. 1(b)) and the vertical axis in the space-fixed coordinate system ( $\hat{\mathbf{z}}$  in fig. 3), and  $\phi$  is the angle between the projection of  $\hat{\mathbf{z}}'$  onto the horizontal plane and  $\hat{\mathbf{x}}$ .

Our approximate forward modelling is written by substituting the definition of the induced dipole of eq. (8) into the expression for a dipole field (1), and carrying out the dyadic product. Let us consider a target whose center is located at  $\mathbf{R}$  in the space-fixed co-ordinate system. The secondary response  $\xi(\mathbf{r}, t)$  measured at a receiver/transmitter location  $\mathbf{r}$  and at a time  $t$  after the termination of the primary field, is then the sum of the responses of the two orthogonal dipoles:

$$\xi(\mathbf{r}, t) = \xi_1(\mathbf{r}, t) + \xi_2(\mathbf{r}, t) \quad (12)$$

where

$$\xi_i(\mathbf{r}, t) = \frac{\mu_o}{4\pi} \left( 3 [\mathbf{m}_i(t) \cdot (\mathbf{r} - \mathbf{R})] \frac{(\mathbf{r} - \mathbf{R})}{|\mathbf{r} - \mathbf{R}|^5} - \frac{\mathbf{m}_i(t)}{|\mathbf{r} - \mathbf{R}|^3} \right) \quad (13)$$

and

$$\mathbf{m}_1(t) = L_1(t) (\hat{\mathbf{z}}' \cdot \mathbf{B}^P) \hat{\mathbf{z}}' \quad (14)$$

$$\mathbf{m}_2(t) = L_2(t) [(\hat{\mathbf{x}}' \cdot \mathbf{B}^P) \hat{\mathbf{x}}' + (\hat{\mathbf{y}}' \cdot \mathbf{B}^P) \hat{\mathbf{y}}'] \quad (15)$$

are the dipole parallel and perpendicular to the axis of symmetry. The unit vectors are given by eq. (11).

In summary, the approximate response of buried metallic object given by eq. (12) can be generated from 13 parameters that describe the object. These model parameters are elements of the model vector

$$\mathbf{m} = [X, Y, Z, \phi, \theta, k_1, \alpha_1, \beta_1, \gamma_1, k_2, \alpha_2, \beta_2, \gamma_2]. \quad (16)$$

$X$  and  $Y$  denotes the surface projection of the centroid of the body, and  $Z$  is the depth of the object below the surface. The orientation of the target is described by the two angles  $\theta$  and  $\phi$ . The remaining parameters describe the decay characteristics of the two dipoles:  $k_1, \alpha_1, \beta_1,$  and  $\gamma_1$  describe the dipole parallel to the axis of symmetry ( $\mathbf{m}_1$ ), and  $k_2, \alpha_2, \beta_2,$  and  $\gamma_2$  describe the dipole perpendicular to the axis of symmetry ( $\mathbf{m}_2$ ). Thus the inversion for the model  $\mathbf{m}$  will immediately give estimates of target location and orientation. Information on the shape, size, and material parameters of the target may later be inferred from the remaining parameters.

### Non-Linear Parameter Estimation Procedure

In this paper we first assume that the response measured in a survey is due to a single body, and second, that the response of this single body can be accurately modelled with eq. (12). With these hypotheses, an inversion procedure can be developed that utilizes the approximate forward model.

The forward model can be expressed as

$$d_j = F_j[\mathbf{m}], \quad j = 1, 2, 3, \dots, N \quad (17)$$

This equation expresses the mapping of the model vector  $\mathbf{m}$  to a datum  $d_j$  by a functional  $F_j$ . The forward mapping  $F_j$  is defined by eq. (12) and it is a nonlinear functional of the 13 model parameters are given in eq. (16). In the inverse problem, these parameters are retrieved from a vector of observed data  $\mathbf{d}^{\text{obs}}$  by minimizing a least-squares objective function. Before proceeding to the details of the inversion there are two important practical aspects to be introduced. We need to ensure that selected parameters remain positive and we also need to scale the parameters to enhance stability in the iterative process.

In the approximate forward model the time decay parameters  $k_i, \alpha_i, \beta_i,$  and  $\gamma_i$  ( $i = 1, 2$ ) are defined as positive. In the inverse problem the positivity of these parameters can be maintained by solving the associated square-variable unconstrained problem (Gill et al., 1981). Following this formulation, each time decay

parameters  $m_i$  is replaced by the squared variable  $w_i$  such that  $m_i = w_i^2$ . A second transformation is a linear scaling that ensures that each component of  $\mathbf{m}$  is of order unity. This is done by dividing each parameter  $m_i$  by its typical value. In the following presentation of the inversion algorithm, the letter  $\mathbf{m}$  continues to denote the parameter vector to be inverted for, but now includes the positivity and scaling transformations.

## 1. Defining the Objective Function

If there are  $L$  time channels and  $K$  locations where TEM data are collected, then there will  $N = KL$  data points contained in the data vector  $\mathbf{d}^{\text{obs}}$ . Because data will be collected on several lines, with a number of stations per line, there will generally be far more data than model parameters ( $N \gg 13$ ). Therefore the inversion for  $\mathbf{m}$  involves solving an overdetermined system of non-linear equations, with the goal of finding the model that produces the data that best fits the observed data. This is a non-linear least squares problem and is solved by minimizing

$$\Phi(\mathbf{m}) = \frac{1}{2} \|\mathbf{W}_d (F[\mathbf{m}] - \mathbf{d}^{\text{obs}})\|^2 \quad (18)$$

where  $F[\mathbf{m}]$  is the forward modelled data,  $\mathbf{d}^{\text{obs}}$  is the observed data, and  $\Phi$  is the least squares objective function that measures how closely our predicted data matches the observed data.  $\mathbf{W}_d$  is the data weighting matrix. If the data are contaminated with unbiased Gaussian random noise, then  $\mathbf{W}_d$  is ideally a diagonal matrix whose elements are the reciprocals of the standard deviation of each datum. The noise arises from many sources, including sensor location errors, instrument noise, and inaccuracy of the forward modelling. It is unlikely that the Gaussian independent assumption is not valid, but it is essential to estimate a quantity that reflects the uncertainty in each datum. We assume that the errors can be characterized by a percentage of the datum value plus a threshold, that is

$$(\mathbf{W}_d)_{ii} = \frac{1}{\rho d_i + \epsilon} \quad (19)$$

where  $\rho$  is typically a percentage and  $\epsilon$  is a constant that characterizes ambient noise. The positive  $\epsilon$  ensures that small data points would have reasonable errors assigned to them, and thus prevents them from having undue influence on the solution.

## 2. Minimizing the Objective Function

For ease of notation the least squares problem is rewritten as

$$\text{minimize } \Phi(\mathbf{m}) = \frac{1}{2} \mathbf{r}(\mathbf{m})^T \mathbf{r}(\mathbf{m}) = \frac{1}{2} \sum_{i=1}^N r_i(\mathbf{m})^2 \quad (20)$$

where  $\mathbf{r}$  is the residual function

$$\mathbf{r}(\mathbf{m}) = \mathbf{W}_d (F[\mathbf{m}] - \mathbf{d}^{\text{obs}})$$

and  $r_i(\mathbf{m})$  is the  $i^{\text{th}}$  component of  $\mathbf{r}(\mathbf{m})$ . We adopt a modified Newton's method to minimize the objective function. The approach taken here is to first make an initial guess of the model parameters  $\mathbf{m}_0$ . Techniques for making this initial guess are outlined in Pasion (1999). The starting model  $\mathbf{m}_0$  is iteratively improved to



find the minimum of eq. (20). At each iteration a Newton's search direction is chosen that minimizes the local quadratic model about the current iterate  $\mathbf{m}_k$ . The Newton step  $\delta\mathbf{m}$  for the non-linear least squares problem is then given by

$$\mathbf{H}(\mathbf{m}_k) \delta\mathbf{m} = -\mathbf{J}(\mathbf{m}_k)^T \mathbf{r}(\mathbf{m}_k) \quad (21)$$

where the Jacobian matrix  $\mathbf{J}$  and the Hessian matrix  $\mathbf{H}$  are defined as

$$J_{ij}(\mathbf{m}) = \frac{\partial r_i}{\partial m_j} \quad \text{and} \quad \mathbf{H} = \mathbf{J}(\mathbf{m})^T \mathbf{J}(\mathbf{m}) + \mathbf{S}(\mathbf{m}),$$

where  $\mathbf{S}(\mathbf{m})$  is

$$\mathbf{S}(\mathbf{m}) = \sum_{i=1}^N r_i(\mathbf{m}) \nabla^2 r_i(\mathbf{m})$$

A new model  $\mathbf{m}_{k+1} = \mathbf{m}_k + \lambda\delta\mathbf{m}$  is then defined, where the positive scalar  $\lambda$  is chosen such that  $\Phi(\mathbf{m}_k + \lambda\delta\mathbf{m}) < \Phi(\mathbf{m}_k)$ . The sequence of iterations is terminated once the relative gradient measure is less than a tolerance level, or once there is insignificant change in the models of successive iterations (Dennis and Schnabel, 1983).

### 3. Error Bounds of the Parameter Estimates

Once the model parameters  $\mathbf{m}_*$  which minimize the objective function  $\Phi(\mathbf{m})$  have been obtained, we can examine the reliability and precision of the estimated parameters via the model covariance matrix. Let  $\mathbf{m}_+$  be the best estimate of the model in the absence of noise, and  $\delta\mathbf{m}_* = \mathbf{m}_+ - \mathbf{m}_*$ . The model covariance matrix  $\mathbf{V}_m$  is defined as the expectation value of  $\delta\mathbf{m}_* \delta\mathbf{m}_*^T$  (Bard, 1994)

$$\mathbf{V}_m \equiv E(\delta\mathbf{m}_* \delta\mathbf{m}_*^T) \approx E\left(\mathbf{H}_*^{-1} \mathbf{J}_*^T \delta\mathbf{d}^{\text{obs}} \delta\mathbf{d}^{\text{obs}T} \mathbf{J}_* \mathbf{H}_*^{-1}\right)$$

The Hessian and Jacobian in the above expression are evaluated at  $\mathbf{m} = \mathbf{m}_*$ , and are therefore constants. As a result they can be taken outside of the expectation value expression:

$$\mathbf{V}_m = \mathbf{H}_*^{-1} \mathbf{J}_*^T \mathbf{V}_d \mathbf{J}_* \mathbf{H}_*^{-1} \quad (22)$$

where  $\mathbf{V}_d$  is the covariance matrix of the data. In the case when the observations have uncorrelated errors, the data covariance matrix reduces to a diagonal matrix, and an estimate of the standard deviation of the  $i^{\text{th}}$  model parameter  $m_i$  is then

$$\sigma_i^m = (\mathbf{V}_m)_{ii} = \sigma^2 \left( \mathbf{H}_*^{-1} \mathbf{J}_*^T \mathbf{J}_* \mathbf{H}_*^{-1} \right)_{ii} \quad (23)$$

Model variance estimates applied to non-linear problems are not as reliable as when implemented in linear least squares problems, and they should only be used as a very rough estimate (Bard, 1974; Dennis and Schnabel, 1983). Nevertheless, eq. (23) at least provides a minimum estimate to the uncertainties of the parameters.

## **Relating Model Parameters to Material and Geometric Properties**

The above inversion generates the parameters that characterize a target's TEM anomaly. The next step is to interpret these parameters. Recall that UXO are typically rod-like rather than plate-like, and are magnetically permeable. In order to extract these potentially UXO identifying features from the recovered model  $m^*$ , we use the inversion procedure to fit a series of decay curves from a range of axi-symmetric targets of different shape, geometry, and material properties. We then generate empirical relationships between the parameters and target characteristics. The data curves used for this analysis were either TEM measurements made in the Geonics Ltd. laboratory, or they were synthetically generated decay curves for a sphere using equation (12).

## 1. Lab Setup and Measurements

A series of TEM measurements of metallic targets was made by Geonics Ltd. A  $40m \times 40m$  square transmitter loop was used to provide a relatively uniform field at the center of the loop. A  $1m$  diameter receiver coil was placed coaxial and coplanar to the transmitter loop, and each target was located at the center of the receiver loop. The Geonics PROTEM 47 time domain equipment was used for producing the transmitting field and for recording the time domain measurement due to a step-off current. Measurements of the time decay response of these targets were recorded as plots of  $\log(\partial \mathbf{B} / \partial t)$  vs.  $\log(t)$ . Since values were not recorded by a data logger, the plots were subsequently digitized. Plots of the steel target responses were digitized by hand by J.D. McNeill at Geonics Ltd., and the aluminum target responses were digitized at UBC after scanning the plots into a computer. Analyses were performed on both the impulse ( $\partial \mathbf{B} / \partial t$ ) response measured as an induced voltage in the receiver, and also on the  $\mathbf{B}$ -field response. The  $\mathbf{B}$ -field response was obtained by integrating the induced voltage.

Two sets of targets were measured. The first set of measurements involved recording the TEM response for a series of steel and aluminum rectangular prisms of different dimensions. Each prism had at least one dimension of 8 inches, and the targets ranged from a thin rod ( $8 \times 1/4 \times 1/4$  inch) to a cube ( $8 \times 8 \times 8$  inch) to a thin plate ( $8 \times 8 \times 1/4$  inch). A second set of measurements was made on 24 sample UXO. These targets included various ordnance items used by NATO since World War II. The ordnance range in length from 18 to 85 cm, and in diameter from 6.05 to 15.92 cm. A diagram of all the ordnance, along with a table listing the dimensions of each ordnance, is included in Pasion (1999).

The axi-symmetric targets were placed in two orientations at the center of the receiver loop in two orientations. Each target was measured with the axis of symmetry perpendicular and parallel to the primary field. Since the strength of each induced dipole is proportional to the projection of the primary field onto the dipole direction, the two measurement orientations isolate the decay behaviour of each of the two dipoles. For example, consider a plate. When the primary field is perpendicular to the plane of the plate, the projection of the primary field onto dipole 2 is zero, thus the approximate forward model assumes the response can be modelled as a single dipole perpendicular to the plate. The decay parameters of dipole 1 ( $k_1$ ,  $\alpha_1$ ,  $\beta_1$ , and  $\gamma_1$ ) can then be estimated by fitting this curve to the decay law (eq. (9)). When the primary field is parallel to the plane of the plate, the response is due to dipole 2 and parameters  $k_2$ ,  $\alpha_2$ ,  $\beta_2$ , and  $\gamma_2$  can be recovered.

Relationships between the target characteristics and the model parameters were established, in the following manner. A scaled-down version of the non-linear least squares techniques outlined in the previous section was used to obtain the decay parameters  $k$ ,  $\alpha$ ,  $\beta$ , and  $\gamma$  for each of the target's two dipoles. Secondly, we observed how recovered values of model parameters, or combinations of parameters, changed with the

dimensions and magnetic properties of the measured prism. The patterns in the behaviour of the parameters then led to the shape and permeability discrimination diagnostics that are proposed in the following sections.

## 2. A Relationship Between $\beta$ and Magnetic Permeability

UXO are generally made of steel, which is a ferrous material. Therefore, the magnetic permeability is likely an identifying characteristic of UXO. To generate a link between magnetic permeability and model parameters, forward modelled responses were calculated for a series of spheres varying in size and permeability. Both  $\mathbf{B}$ -field and  $\partial\mathbf{B}/\partial t$  data were then inverted to generate decay parameters, and in particular, to produce estimates of the parameter  $\beta$ . The plots of  $\beta$  as a function of sphere radii and magnetic permeability, are provided in fig. 4.

Fig. 4 suggests that the value of  $\beta$  obtained for a sphere may be diagnostic in determining whether the sphere is permeable or non-permeable. Fig. 4(a) exhibits the relationship for the  $\partial\mathbf{B}/\partial t$  responses. For a steel sphere ( $\mu_r = 150$ ), we see, for spheres with radii between 5 to 15 cm, that  $\beta$  falls between 1.11 and 1.35, while for a non-permeable sphere ( $\mu_r = 1$ ),  $\beta$  has a value of approximately 0.5, which corresponds to the early time  $t^{-1/2}$  behavior that Kaufman (1994) predicted for a non-permeable sphere. Therefore, when applying our inversion to the time derivative of the field, a value of  $\beta^{\partial\mathbf{B}/\partial t}$  greater than about 0.8 indicates that the target is most likely permeable. This analysis is repeated on the forward modelled  $\mathbf{B}$ -field responses, and the results are plotted in fig. 4(b). A threshold value of  $\beta^{\mathbf{B}} = 0.3$  could be used such that targets with a  $\mathbf{B}$ -field response characterized by a  $\beta > 0.3$  indicates a permeable target.

The use of  $\beta$  as a diagnostic to determine permeability can be extended to non-spherical targets by looking at the recovered  $\beta$  values for the aluminum and steel prisms. The inversion produces two values of  $\beta$ , one for each of the excited dipoles, to describe a buried target. We suggest taking the average of the two recovered  $\beta$  values, which we label as  $\bar{\beta}$ . When analyzing the  $\partial\mathbf{B}/\partial t$  responses of the axi-symmetric aluminum targets,  $\bar{\beta}^{\partial\mathbf{B}/\partial t} = 0.52$  with a standard deviation of 0.07. For the steel targets  $\bar{\beta}^{\partial\mathbf{B}/\partial t} = 1.11$  with a standard deviation of 0.08. These averages fall on either side of the 0.8 threshold obtained by fitting sphere  $\partial\mathbf{B}/\partial t$  responses.

When analyzing the  $\mathbf{B}$ -field responses of the axi-symmetric aluminum targets,  $\bar{\beta}^{\mathbf{B}} = 0.17$  with a standard deviation of 0.03. For the steel targets  $\bar{\beta}^{\mathbf{B}} = 0.5$  with a standard deviation of 0.2. These averages fall on either side of the 0.3 threshold obtained by fitting sphere  $\mathbf{B}$ -field responses, and so again, a consistent criterion can be used.

## 3. Relationships Between Model Parameter Ratios and Target Shape

Empirical relationships were also established between the target shape and the ratios  $k_1/k_2$  and  $\beta_1/\beta_2$ . For space reasons, we present only the analysis of the  $\partial\mathbf{B}/\partial t$  response of the targets, and refer the reader to Pasion (1999) for the analysis of the  $\mathbf{B}$ -field data. There we show that the same target shape diagnostics, developed here for  $\partial\mathbf{B}/\partial t$  data, also apply to  $\mathbf{B}$ -field data.

The Ratio  $k_1/k_2$ . The recovered  $k$  values for targets ranging from a steel plate to a steel rod are shown in fig. 5(a), and the calculated  $k$ -ratios are shown in fig. 5(b). For a steel plate, the  $k$ -ratio  $k_1/k_2 < 1$ . For a steel bar the  $k$ -ratio  $k_1/k_2 > 1$ . The recovered  $k$  values for aluminum targets are shown in fig. 5(c). The opposite orientation effect was observed for an aluminum rod, that is  $k_1/k_2 < 1$  (fig. 5(d)).

The Ratio  $\beta_1/\beta_2$ . In addition to the relative strength of the dipoles being shape dependent, the slope of the time decay response (either  $\partial\mathbf{B}/\partial t$  or  $\mathbf{B}$ -field) during the intermediate time stage is dependent upon the target shape. This effect was seen in steel targets only. The steepness of the response during the intermediate time stage is reflected in the parameter  $\beta$ . The recovered  $\beta$  values for targets ranging from a steel plate to a steel rod are shown in fig. 6(a), and  $\beta$  values for aluminum targets are shown in fig. 6(c). A dipole that decays at a greater rate will have a larger  $\beta$ . The rate of decay of the  $\partial\mathbf{B}/\partial t$  response is greater when the plane of a steel plate is perpendicular to the primary field (dipole 1), than when the plane of a steel plate is parallel to the primary field (dipole 2). Thus, for a steel plate the  $\beta$ -ratio  $\beta_1/\beta_2 > 1$ . In the case of a rod, the  $\partial\mathbf{B}/\partial t$  response decays faster (and thus  $\beta$  is larger) when the main axis of the rod is perpendicular to the primary field (dipole 2). In the case of a steel rod the  $\beta$ -ratio  $\beta_1/\beta_2 < 1$  (fig. 6(b)).

For aluminum targets the response shape looks essentially the same for each of the targets. The  $\partial\mathbf{B}/\partial t$  response exhibits a power law decay of  $t^{-1/2}$  and is exponential at later times. The decay curves for aluminum targets are essentially the same regardless of target shape, and therefore there is no relationship between the  $\beta$ -ratio and the aspect ratio (fig. 6(d)).

#### 4. The Discrimination Algorithm Using $\partial\mathbf{B}/\partial t$ Data

The results from the previous section suggests the following algorithm for using  $\partial\mathbf{B}/\partial t$  data to help identify possible UXO targets:

1. Perform the non-linear inversion outlined in the previous section to recover model parameters for the two-dipole model.
2. Compute  $\bar{\beta} = \frac{1}{2}(\beta_1 + \beta_2)$ . If  $\bar{\beta} > 0.80$  then the target is most likely permeable.
3. Compute ratios  $\beta_1/\beta_2$  and  $k_1/k_2$ . There are two options:
  - $\bar{\beta} > 0.8 \Rightarrow$  *Ferrous Target*: If  $k_1/k_2 > 1$  and  $\beta_1/\beta_2 < 1$  then a permeable *rod-like* target was measured. If  $k_1/k_2 < 1$  and  $\beta_1/\beta_2 > 1$  then a permeable *plate-like* target was measured.
  - $\bar{\beta} < 0.8 \Rightarrow$  *Non-Ferrous Target*: If  $k_1/k_2 > 1$  then non-permeable *plate-like* target was measured. If  $k_1/k_2 < 1$  then the target is *rod-like*.  $\beta_1/\beta_2$  does not give supporting, or extra, information.

The above algorithm can be extended to the analysis of  $\mathbf{B}$ -field data simply by changing the  $\beta$  threshold to 0.3.

### **Synthetic Data Set**

The parameter estimation procedure is now tested on a synthetically generated field data set. The object of interest is a 75 mm anti-tank mortar. The primary decay curves for the axial and perpendicular orientations were obtained from measurements made by Geonics using the setup described in the previous section. These were inverted to recover the decay parameters for each dipole. For this simulation, the target is assumed to be buried at a depth of 67 cm ( $Z = 1m$ ), and located at  $(2mN, 2mE)$  on the survey grid. The mortar is oriented such that  $\phi = 30^\circ$  and  $\theta = 65^\circ$ . The data were forward modelled using eq. (12).

The survey consists of a  $2m \times 2m$  grid, containing 5 lines of data separated at 50 cm line spacing, with stations located at 20 cm intervals along each line. At each station the vertical component of the voltage is generated for 26 logarithmically spaced time channels. The time channels range from 0.01 ms to 100 ms. In order to make this example closer to a real TEM data set, 5% random Gaussian noise was added to the data and, since a real TEM instrument will have a finite measurement sensitivity to the secondary field, a data threshold of 0.001 is set. There are 1278 total data points exceeding the minimum threshold.

The inversion is carried out with a data weighting matrix in eq. (19) with  $\rho = 0.05$  and  $\epsilon = 0.001mV$ . The observed data, and data predicted by the recovered model, are compared in figures 7 and 8. Fig. 8 shows a plan view comparison for three of the 26 time channels. The difference maps exhibit a random distribution over the data, indicating the reluctance of the inversion to fit the noisy portion of the data. A comparison of the true model  $m_{true}$ , recovered model  $m_{rec}$  and the estimated model standard deviations  $\sigma^m$  are found in Table 1. The discrimination algorithm, when applied to the recovered decay parameters, yields the following. The value of  $\bar{\beta} = 1.07(> 0.8)$  indicates that the target is likely permeable. The ratios  $k_1/k_2 = 3.71(> 1)$  and  $\beta_1/\beta_2 = 0.71(< 1)$  indicate, for a magnetically permeable target, that the TEM response is likely from a rod-like target.

### Field Data Set

We now apply our algorithm to a TEM field data set acquired at the United States Army Corps of Engineer Environmental Research and Development Center UXO test site in Vicksburg, Mississippi. The Geonics EM63 instrument used for the survey is a multi-time channel time domain unit consisting of a  $1m \times 1m$  square transmitter coil and a coaxial horizontal circular receiver loop mounted on a two-wheel trailer. Measured voltages are averaged over 26 geometrically spaced time gates, spanning the range 180  $\mu sec$  to 25.14 ms.

A 105 mm projectile is placed in the ground with its center at 2.0 m East, 1.83 m North and at a depth of 0.44 m from the surface. The projectile was placed horizontal ( $\theta = 90^\circ$ ), with its tip pointing to the North ( $\phi = 0^\circ$ ). Once the target was placed in the ground, it was not covered in soil. The survey consisted of a  $2m \times 2m$  grid centered on the target, containing 5 lines running North-South separated at 50cm line spacing, with stations located at 5cm intervals along each line. A measured signal of less than 1mV is assumed to be indistinguishable from the noise. The resulting data set contains 1882 total data points.

The inversion is carried out with a data weighting matrix in eq. (19) with  $\rho = 0.05$  and  $\epsilon = 1mV$ . The first stage of the time decay evident in fig. 2 is not observed in the time window recorded by the EM63. Therefore, we invert this data by setting  $\alpha$  to be a small constant that does not affect the predicted data within the

EM63 time range. So only parameters  $k$ ,  $\beta$ , and  $\gamma$  for each dipole are recovered. The observed and predicted data are compared in figures 9 and 10. Fig. 10 shows a plan view comparison for five of the 26 time channels. At early times the anomaly has a single peak located approximately above the UXO center. This peak splits into two distinct peaks at late time. The recovered model predicts data that reflects this behavior. Fig. 9 compares the decay curve measured at four stations on the survey.

The recovered location and orientation parameters are listed in Table 2(a). The recovered easting of  $2.04m$  differs from the true value of  $2.00m$  by  $4cm$ . The recovered northing of  $1.77m$  differs from the true value of  $1.83m$  by  $6cm$ , placing the inducing dipole closer to the projectile tail. These errors are of the same magnitude as can be expected in spotting the station location in the field survey. In addition, the buried 105 mm projectile has a copper rotating band near the tail of the projectile. It has been suggested that the presence of the rotating band will shift the location of the induced dipole from the target center towards the tail (Miller, 2000). The recovered burial depth of  $0.47m$  is  $3cm$  deeper than the expected depths of  $0.44m$ . The orientation parameters  $\theta$  and  $\phi$  are well recovered. The recovered decay parameters are listed in Table 2(b) and the diagnostics applied to these parameters are listed in Table 2(c). The value of  $\bar{\beta} = 0.91 (> 0.8)$  indicates that the target is likely magnetically permeable. The ratios  $k_1/k_2 = 2.63 (> 1)$  and  $\beta_1/\beta_2 = 0.69 (< 1)$  indicate, for a magnetically permeable target, that the TEM response is likely from a rod-like target.

## Discussion and Conclusion

Efficient remediation of areas containing UXO first requires that purely conductive metal targets be distinguished from steel targets that are conductive and permeable. The second stage then focuses upon determining if the steel target is rod-like (and likely to be a UXO) or plate-like (and not of interest). To attack this problem we propose a modified parametric model from which TEM responses can be estimated. The TEM response of a buried axisymmetric metallic object is modelled as the sum of two dipoles located at the midpoint of the body. Non-linear inversion methods are used to extract the parameters from the field data and these parameters are subsequently used in a discrimination procedure which has two parts. First, the decision about whether the object is ferrous might be made by examining the size of the recovered  $\beta$ 's. Second, if the object is considered to be ferrous, then the ratios of  $k_1/k_2$  and  $\beta_1/\beta_2$  are diagnostic indicators of whether the geometry is plate-like or rod-like. These diagnostics were developed for both  $\mathbf{B}$ -field and  $\partial\mathbf{B}/\partial t$  TEM data.

This algorithm was applied to a synthetic data set as well as to a field data set collected by the Geonics EM63 time domain electromagnetic sensor over a 105 mm projectile. In both cases the diagnostics, applied to the recovered model parameters, correctly predicted that the TEM anomaly was produced by a magnetically permeable and rod-like metallic target. Although further testing will be required to fully evaluate our proposed technique, the results presented here are promising and may have a positive impact on the interpretation of UXO detection data.

## Acknowledgments

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a model representation of two perpendicular dipoles that independently decay as  $t^{-\beta} e^{-t/\gamma}$ . We also thank Dwain K. Butler and Jose Llopis at the USACE for their help in acquiring EM63 data at the WES UXO test site.

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**Locating and Determining Dimensionality of UXO Using Time Domain Electromagnetic Induction**

*Leonard R. Pasion and Douglas W. Oldenburg*

**FIGURES AND TABLES**

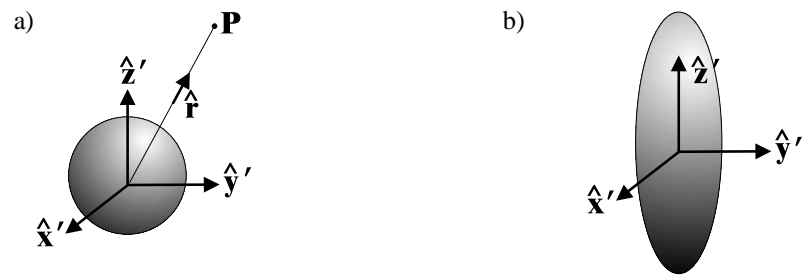


Figure 1: The body-fixed (primed) coordinate system for a sphere and a spheroid.

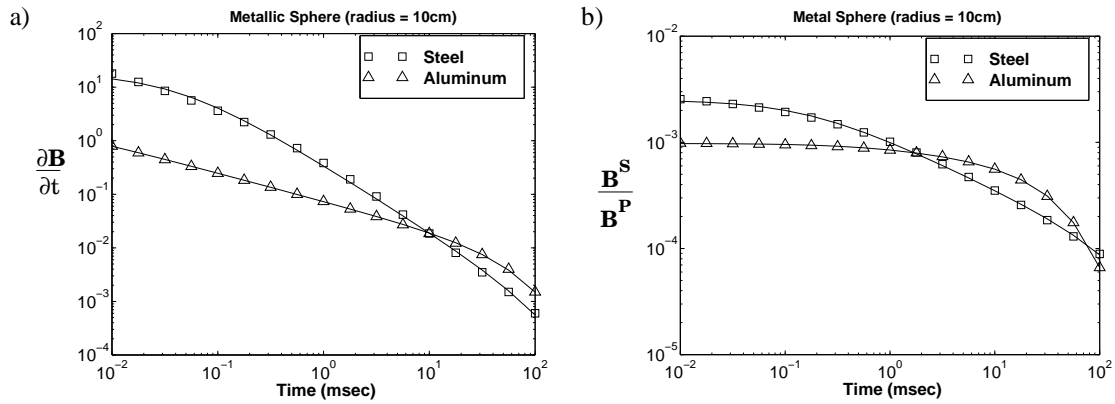


Figure 2: (a) The time decay behaviour of the time derivative of the magnetic field  $\partial\mathbf{B}/\partial t$ . (b) The time decay behaviour of the magnetic flux density  $\mathbf{B}$ . The  $\mathbf{B}$ -field response is normalized by the strength of the primary field. The solid lines are responses evaluated from eq. (9). The agreement supports the validity of this parametric representation of the time domain responses.

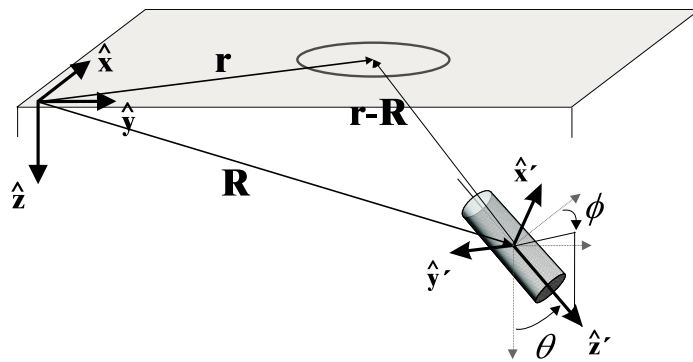


Figure 3: The field (unprimed) co-ordinate system for a buried target. The unit vectors  $\hat{\mathbf{x}}$ ,  $\hat{\mathbf{y}}$ , and  $\hat{\mathbf{z}}$  define the field co-ordinate system, and  $\hat{\mathbf{x}}'$ ,  $\hat{\mathbf{y}}'$ , and  $\hat{\mathbf{z}}'$  define the body-fixed co-ordinate system.

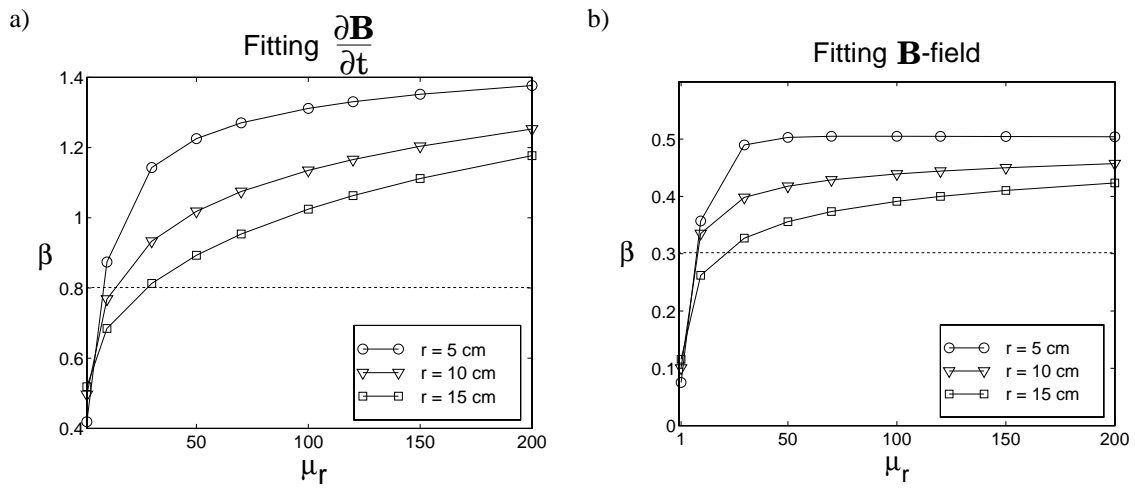


Figure 4: The behaviour of parameter  $\beta$  for various size spheres with varying permeability  $\mu$ . Panel (a) contains results of recovering  $\beta$  by fitting the  $\partial \mathbf{B} / \partial t$  data. Panel (b) contains results of recovering  $\beta$  by fitting the  $\mathbf{B}$ -field.

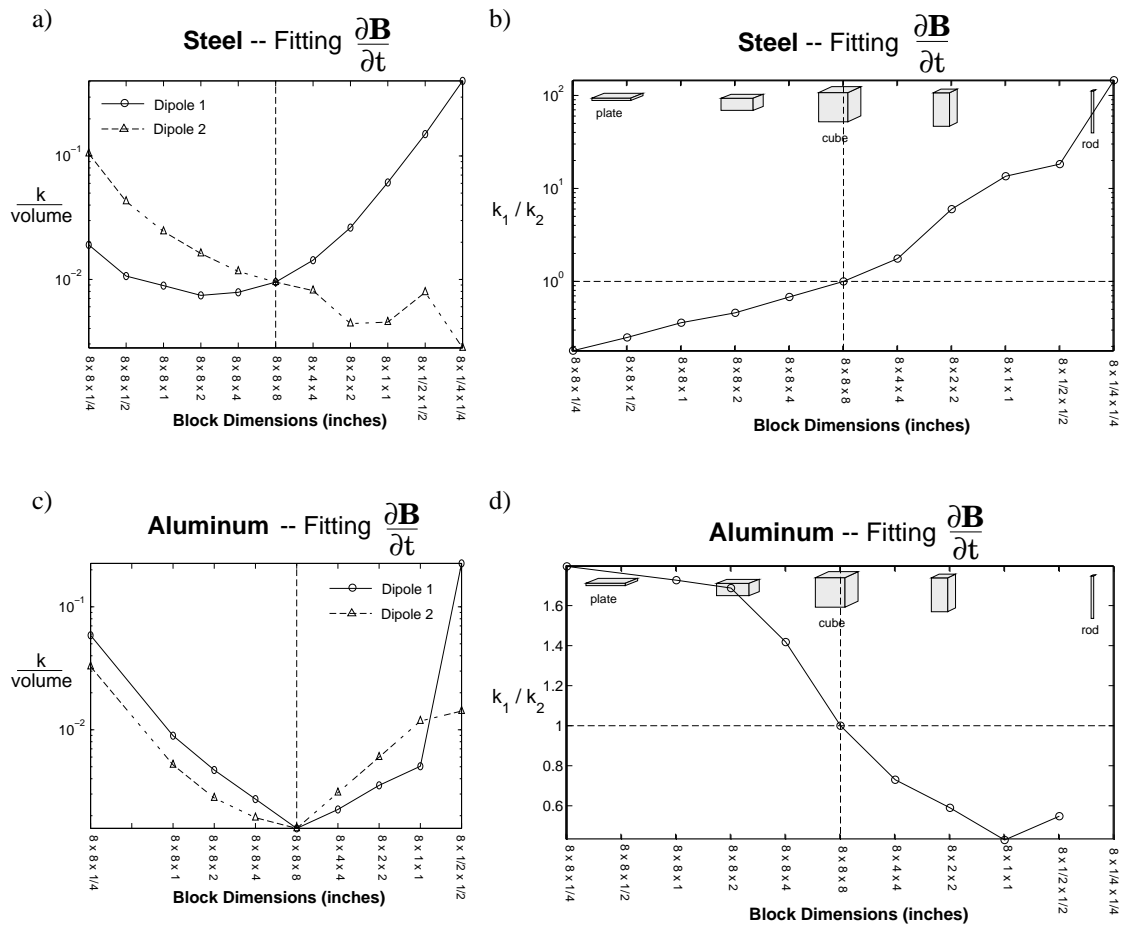


Figure 5: Relating the aspect ratio of a steel target with the ratio  $k_1/k_2$ . Plot (a) contains the recovered  $k$  parameter from fitting the measured  $\partial \mathbf{B} / \partial t$  response of steel axi-symmetric targets. Plot (b) illustrates the relationship between the  $k_1/k_2$  ratio derived from  $\partial \mathbf{B} / \partial t$  data and the shape of a steel target. Plot (c) contains the recovered  $k$  parameter from fitting the measured  $\partial \mathbf{B} / \partial t$  response of aluminum axi-symmetric targets. Plot (d) illustrates the relationship between the  $k_1/k_2$  ratio and the shape of an aluminum target.

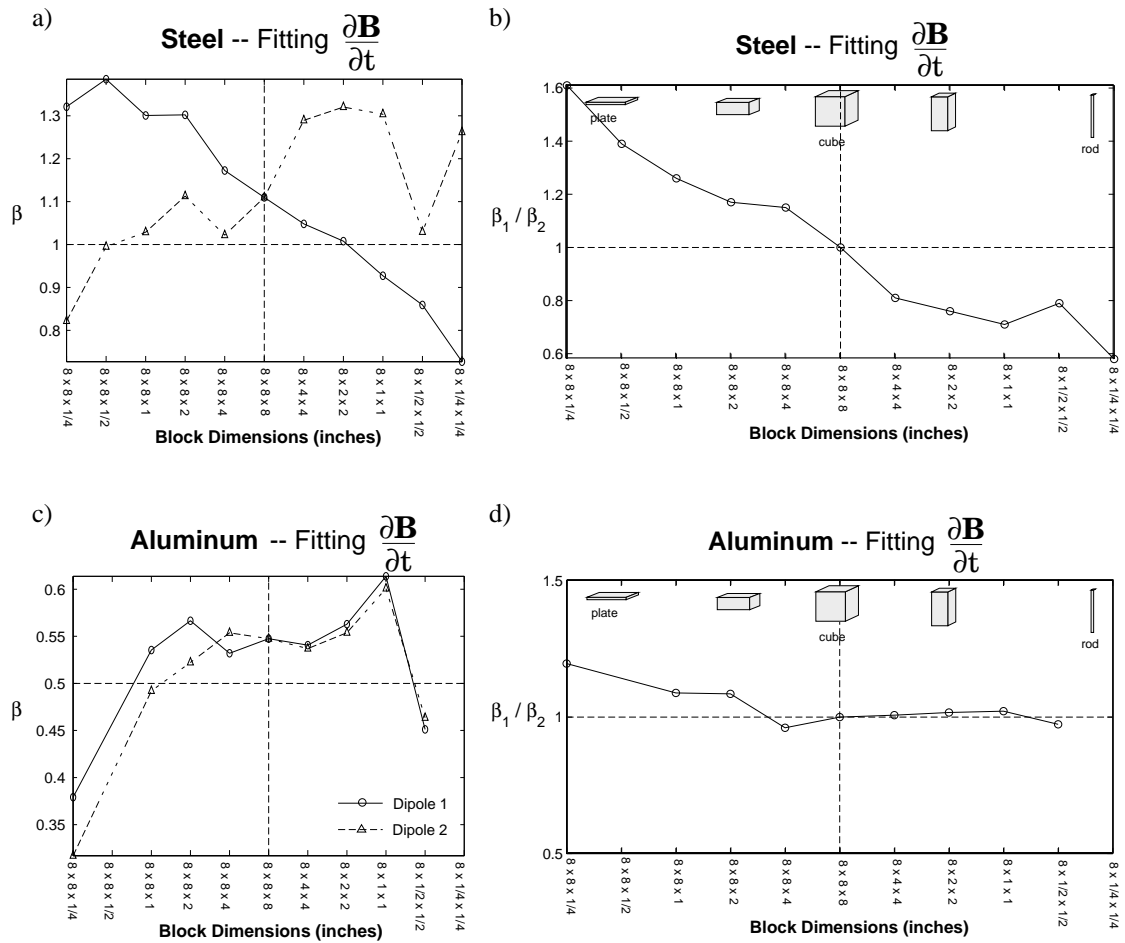


Figure 6: Relating the aspect ratio of a steel target with the ratio  $\beta_1/\beta_2$ . Plot (a) contains the recovered  $\beta$  parameter from fitting the measured  $\partial \mathbf{B}/\partial t$  response of steel axi-symmetric targets. Plot (b) illustrates the relationship between the  $\beta_1/\beta_2$  ratio derived from  $\partial \mathbf{B}/\partial t$  data and the shape of a steel target. Plot (c) contains the recovered  $\beta$  parameter from fitting the measured  $\partial \mathbf{B}/\partial t$  response of aluminum axi-symmetric targets. Plot (d) illustrates the relationship between the  $\beta_1/\beta_2$  ratio and the shape of an aluminum target.



(a) Location and Orientation

$m_i$	$m_o$	$m_{true}$	$m_{rec}$	$\sigma^m$
Northing ( $m$ )	1.90	2.00	2.00060	0.00254
Easting ( $m$ )	2.15	2.00	2.00105	0.00169
Depth from loop ( $m$ )	1.20	1.00	1.00051	0.00764
$\phi$ (degrees)	45	30	30.07	0.18
$\theta$ (degrees)	45	65	65.04	0.11

(b) Decay Parameters

$m_i$	$m_o$	$m_{true}$	$m_{rec}$	$\sigma^m$
$k_1$	7.07	12.02	12.064	0.18
$\alpha_1$	0.01	0.0076	0.00759	0.0030
$\beta_1$	1.00	0.89	0.890	0.0070
$\gamma_1$	3.16	17.65	17.635	0.16
$k_2$	7.07	3.30	3.252	0.071
$\alpha_2$	0.01	0.0077	0.0076	0.0027
$\beta_2$	1.00	1.25	1.252	0.014
$\gamma_2$	3.16	11.54	11.68	0.33

Table 1: Recovered parameters from the inversion of the synthetic data set.  $m_o$  is the starting model. The true model  $m_{true}$  and the recovered model  $m_{rec}$  are close. The difference between the true and recovered model falls within the estimated standard deviation.

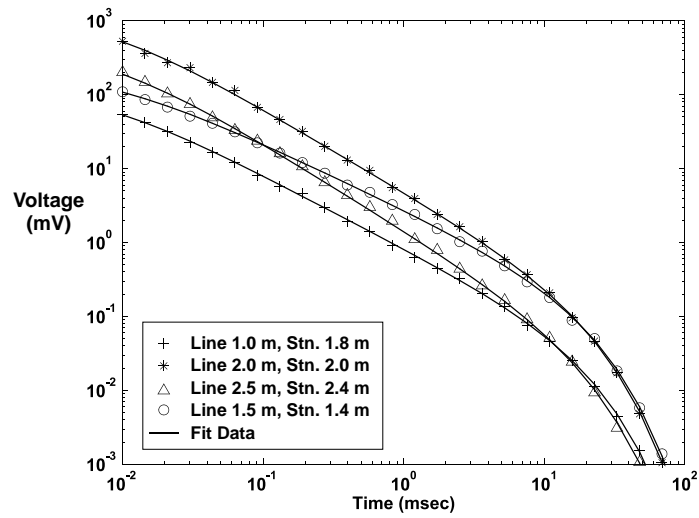


Figure 7: The observed and predicted decay curves for four stations in the synthetic data set inversion. The predicted vertical component of the response, represented by the solid lines, are a good match to the artificially generated noisy data set.

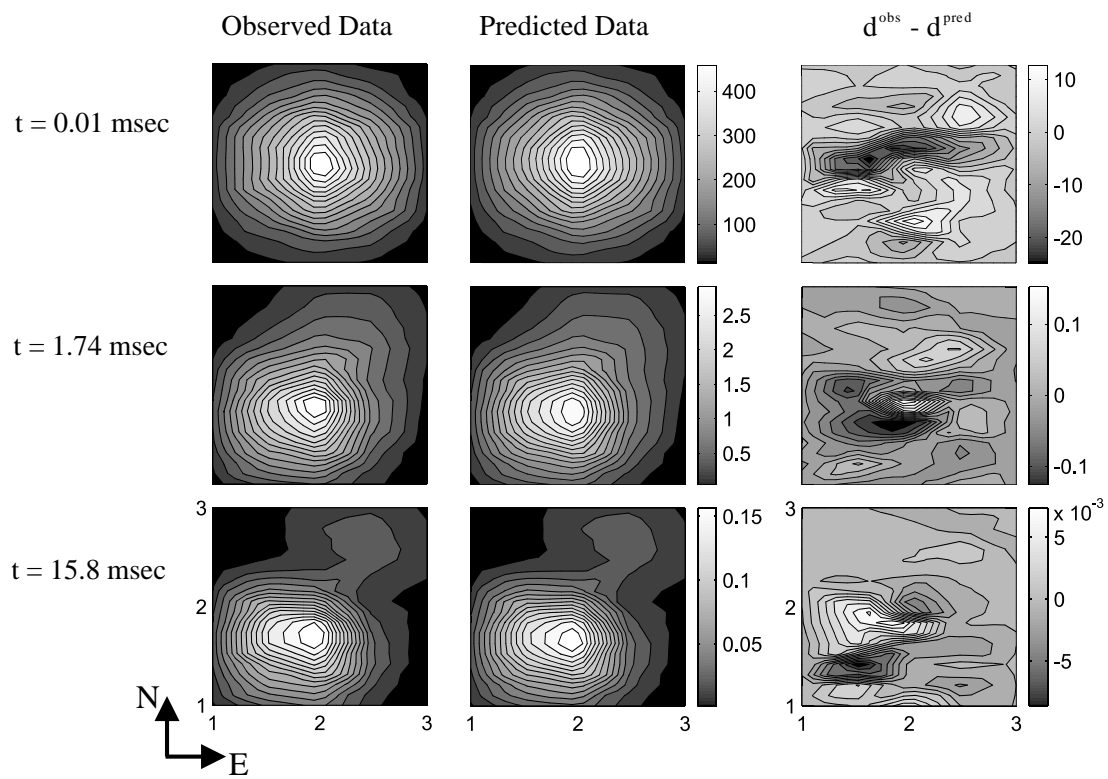


Figure 8: Plan view plots of the observed and predicted data for 3 of the 26 time channels in the synthetic data set inversion.

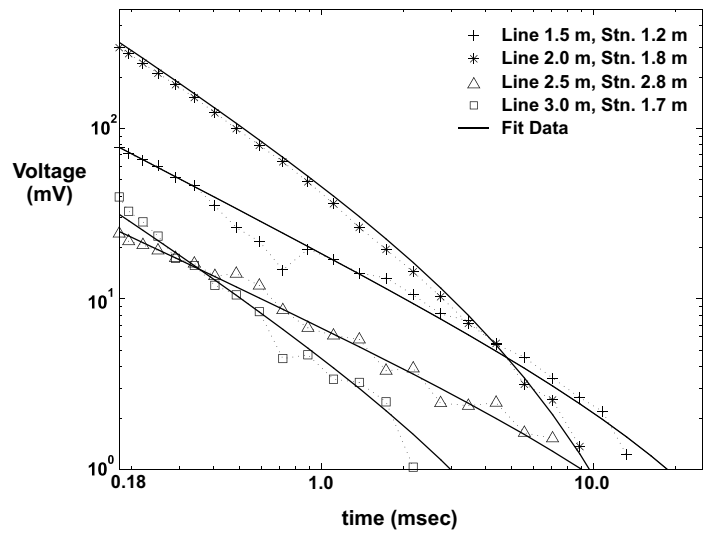


Figure 9: The observed and predicted decay curves for four stations in the 105 mm projectile UXO field data set inversion. The predicted decay of the vertical component of the measured voltages are represented by the solid lines, and the symbols represent the Geonics EM63 field measurements.

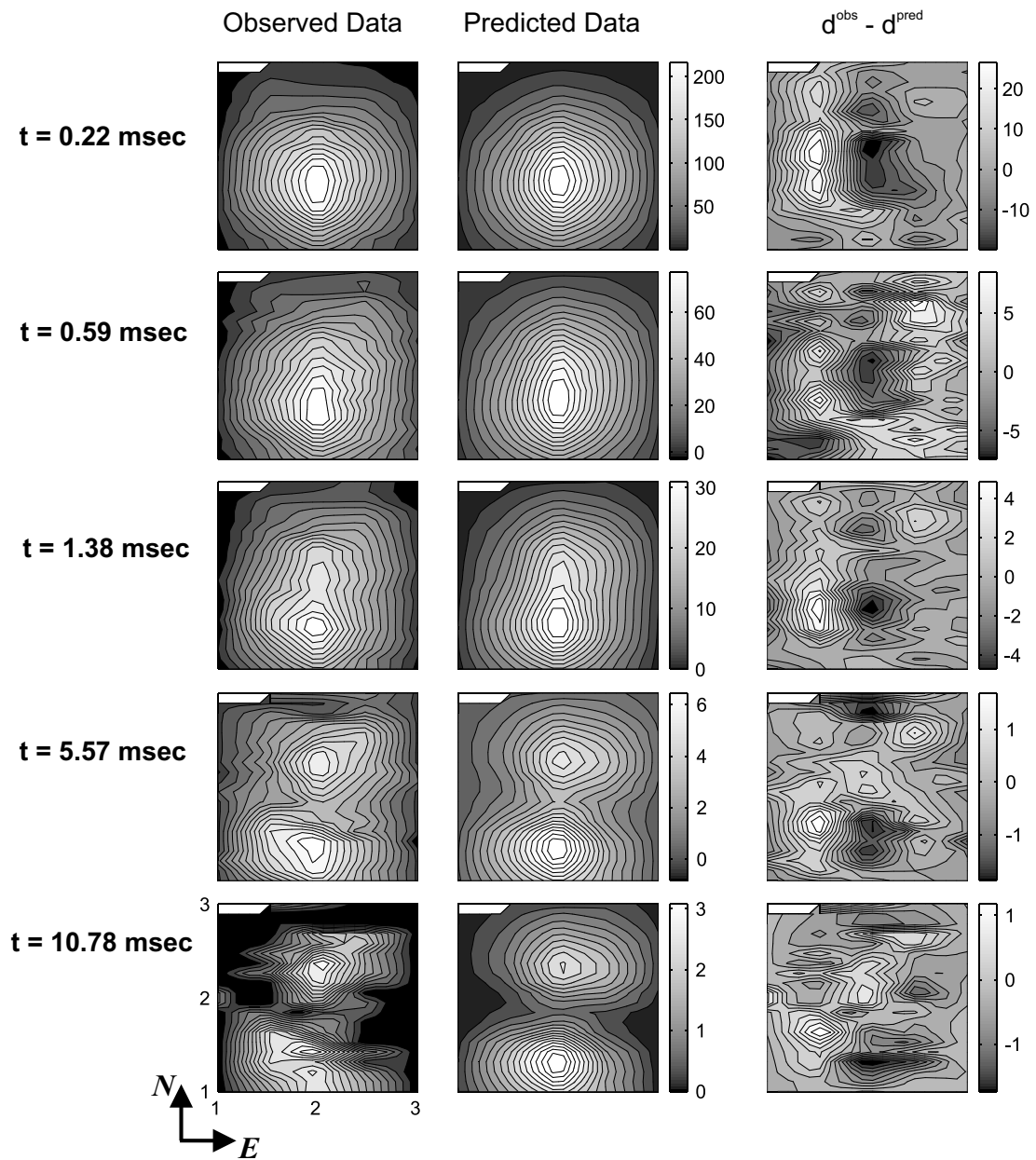


Figure 10: Plan view plots of the observed and predicted data for 5 of the 26 time channels in the 105 mm projectile UXO field data set inversion. The predicted data provide a reasonable match to the TEM response measured by the Geonics EM63.

(a) Location and Orientation

	$m_o$	$m_{rec}$	Expected Parameters	$\sigma^m$
Northing ( $m$ )	1.9	1.77	1.83	0.008
Easting ( $m$ )	2.15	2.04	2.00	0.005
Burial Depth ( $m$ )	0.6	0.47	0.44	0.01
$\phi$ (degrees)	45	10.1	$\sim 0$	0.9
$\theta$ (degrees)	45	84.7	$\sim 90$	0.13

(b) Decay Parameters

$m_i$	$m_o$	$m_{rec}$	$\sigma^m$
$k_1$	7.07	76.8	2.5
$\beta_1$	1.00	0.74	0.04
$\gamma_1$	3.16	31.8	8.3
$k_2$	7.07	29.2	3.7
$\beta_2$	1.00	1.08	0.07
$\gamma_2$	3.16	6.1	1.3

(c) Diagnostics

Diagnostic	Result	Conclusion
$\bar{\beta}$	0.91	permeable
$k_1/k_2$	2.63	rod-like
$\beta_1/\beta_2$	0.69	rod-like

Table 2: Recovered parameters for the field data inversion. Table (a) demonstrates that the inversion was successful in obtaining the approximate location and orientation of the target. Table (b) lists the recovered decay parameters of the two dipoles. Table (c) lists the results of applying the identification diagnostics to the recovered decay parameters. Application of the diagnostics indicates that the buried target is permeable and rod-like and therefore a candidate for UXO.