

1 **On negative induced polarization in frequency domain measurements**

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7 **Abbreviated title:** Frequency domain negative IP effects

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## 14 **Summary**

15 Induced polarization (IP) has been widely used to non-invasively characterize electrical  
16 conduction and polarization in the subsurface resulting from an applied electric field. Earth  
17 materials exhibit a lossy capacitance defined by an intrinsic negative phase in frequency-domain  
18 IP (FDIP) or positive intrinsic chargeability in time-domain IP (TDIP). However, error-free  
19 positive apparent phase or negative apparent chargeability (i.e., negative IP effects) can occur in  
20 IP measurements over heterogeneous media. While negative IP effects in TDIP datasets have been  
21 discussed, no studies have addressed this topic in detail for FDIP measurements. We describe  
22 theory and numerical modeling to explain the origin of negative IP effects in FDIP measurements.  
23 A positive apparent phase may occur when a relatively high polarizability feature falls into  
24 negative sensitivity zones of complex resistivity measurements. The polarity of the apparent phase  
25 is determined by the distribution of subsurface intrinsic phase and resistivity, with the resistivity  
26 impacting the apparent phase polarity via its control on the sensitivity distribution. A physical  
27 explanation for the occurrence of positive apparent phase data is provided by an electric circuit  
28 model representing a four-electrode measurement. We also show that the apparent phase polarity  
29 will be frequency dependent when resistivity changes significantly with frequency (i.e. in the  
30 presence of significant IP effects). Consequently, negative IP effects manifest themselves in the  
31 shape of apparent phase spectra recorded with multi-frequency (spectral IP) datasets. Our results  
32 imply that positive apparent phase measurements should be anticipated and should be retained  
33 during inversion and interpretation of single frequency and spectral IP datasets.

34

35 **Key words:** Electrical properties; Hydrogeophysics; Electromagnetic theory

## 36 **1. Introduction**

37 Induced polarization (IP), a non-invasive electrical geophysical technique for subsurface  
38 characterization, has been widely used in various fields including hydrogeology, engineering,  
39 mining exploration and environmental problems (e.g., Pelton *et al.* 1978; Slater & Lesmes 2002;  
40 Flores *et al.* 2012; Saneiyan *et al.* 2019). IP measures both electrical conduction (i.e., resistivity)  
41 and polarization in a porous medium, therefore providing additional information beyond the direct  
42 current resistivity method. The polarization is quantified by either a chargeability in time-domain  
43 IP (TDIP) or a phase in frequency-domain IP (FDIP) measurements (Binley & Kemna 2005). The  
44 intrinsic capacitive properties of Earth materials are characterized by a positive intrinsic  
45 chargeability (in TDIP) or a negative intrinsic phase when expressed in impedance or complex  
46 resistivity space (in FDIP). One would therefore expect a positive apparent (measured)  
47 chargeability, or equivalently a negative apparent (measured) phase, which we define here as the  
48 normal (or positive) recorded IP response (Ward 1988).

49 In field data acquisition, a negative IP response, i.e., a negative apparent chargeability or a  
50 positive apparent phase, is sometimes observed in the measurements. Such negative IP  
51 measurements are often treated as errors and deleted during the data inversion or interpretation  
52 (e.g., Mary *et al.* 2016; Ntarlagiannis *et al.* 2016; Kelter *et al.* 2018; Garcia-Artigas *et al.* 2020).  
53 While negative IP responses may indeed reflect measurement artifacts, they can also result from  
54 the distortion of the electric field for certain types of heterogeneity close to the electrodes. Negative  
55 IP effects in TDIP measurements resulting from such effects have been investigated (Nabighian &  
56 Elliot 1976; Sumner 1976; Komarov 1980; Dahlin & Loke 2015). Dahlin & Loke (2015) conclude  
57 that negative apparent chargeability results when highly polarizable features fall within zones of

58 negative resistivity measurement sensitivity for the utilized electrode configuration. They found  
59 that the resistivity distribution influences the occurrence and magnitude of negative apparent  
60 chargeability data. Such negative IP measurements provide information about the distribution of  
61 features in the subsurface and should not simply be removed during data processing (Binley 2015;  
62 Dahlin & Loke 2015).

63 Negative IP effects in FDIP have seldom been reported and studied. Luo and Zhang (1998)  
64 presented analytical solutions that predict a positive apparent phase for a buried polarizable sphere  
65 measured by a dipole-dipole array. Some recent complex resistivity imaging studies (Flores  
66 Orozco *et al.* 2018; Liu *et al.* 2017) reported positive apparent phase measurements and included  
67 them in the inversion per recommendations of Dahlin & Loke (2015) for TDIP datasets. Although  
68 frequency and time domain signals are in principle equivalent via the Fourier transform when the  
69 frequency/time range is adequately large, the two commonly measured IP parameters, FDIP  
70 apparent phase and TDIP apparent chargeability, are not directly equivalent.

71 The apparent chargeability equation developed by Seigel (1959), extended by V. Komarov  
72 and colleagues in Russia shortly after Seigel's publication (Komarov 1960), provides a theoretical  
73 explanation for negative IP in TDIP measurements. To the knowledge of the authors, no equivalent  
74 formulation to explain the existence of negative IP in FD measurements has been presented.  
75 Considering the commonly established approximate proportionality between phase and  
76 chargeability (e.g., Van Voorhis *et al.* 1973; Lesmes & Frye 2001), we might expect similarities  
77 in the behavior of negative IP in FDIP measurements to that observed in TDIP reported by Dahlin  
78 & Loke (2015). However, the significance of negative IP effects in FDIP measurements remains  
79 poorly understood, especially with respect to spectral IP where the frequency dependence of IP

80 measurements is recorded. In this study, we integrate theory, numerical modeling, equivalent  
 81 electric circuits and laboratory measurements to comprehensively investigate negative IP effects  
 82 in FDIP, including single frequency and spectral IP measurements.

## 83 **2. Theory of negative IP effects**

84 The intrinsic electrical properties of the subsurface are described by a complex resistivity  
 85 ( $\rho^*$ ) or its inverse, the complex conductivity ( $\sigma^*$ ):

$$\rho^* = |\rho^*|e^{i\varphi} = \frac{1}{\sigma^*}, \quad (1)$$

86 where  $|\rho^*|$  is the complex resistivity magnitude,  $\varphi$  is the complex resistivity phase ( $\varphi \leq 0$ ) and  $i$  is  
 87 the imaginary unit with  $i^2 = -1$ . Both  $\rho^*$  and  $\sigma^*$  can also be presented in terms of real and imaginary  
 88 components that are directly related to the physical (e.g., pore geometry) and chemical properties  
 89 of the subsurface.

90 Field scale FDIP data are most commonly acquired using a four-electrode arrangement at  
 91 the Earth surface. Two electrodes inject a known sinusoidal alternating electrical current ( $\tilde{I}_0$ ) at  
 92 various frequencies, while the other two electrodes record the resultant sinusoidal voltage (or  
 93 potential difference,  $\Delta\tilde{U}$ ). According to Ohm's Law, the measured impedance  $Z_{\text{app}}^*$  (with magnitude  
 94  $|Z_{\text{app}}^*|$  and  $\varphi_{\text{app}}$ ) is determined as,

$$Z_{\text{app}}^* = |Z_{\text{app}}^*|e^{i\varphi_{\text{app}}} = \frac{\Delta\tilde{U}}{\tilde{I}_0} = \frac{|\Delta\tilde{U}| \sin(\omega t + \varphi_{\Delta U})}{|\tilde{I}_0| \sin(\omega t)} = \frac{|\Delta\tilde{U}|}{|\tilde{I}_0|} e^{i\varphi_{\Delta U}}, \quad (2)$$

95 where  $\omega$  is the angular frequency,  $t$  is time,  $|\tilde{I}_0|$  is the current amplitude,  $|\Delta\tilde{U}|$  is the voltage  
 96 amplitude and  $\varphi_{\Delta U}$  is the phase shift of the voltage sinusoid relative to the current sinusoid  $\tilde{I}_0$

97 (defined as the zero phase reference). The apparent complex resistivity  $\rho_{\text{app}}^*$  (with magnitude  $|\rho_{\text{app}}^*|$   
 98 and the same phase  $\varphi_{\text{app}}$  as that of  $Z_{\text{app}}^*$ ) is determined using the geometric factor of the applied  
 99 electrode array  $K$ ,

$$\rho_{\text{app}}^* = |\rho_{\text{app}}^*| e^{i\varphi_{\text{app}}} = K Z_{\text{app}}^* = K |Z_{\text{app}}^*| e^{i\varphi_{\text{app}}}. \quad (3)$$

100  $\rho_{\text{app}}^*$  is the complex resistivity of a homogeneous space equivalent to the value of  $Z_{\text{app}}^*$  resulting  
 101 from application of Eq. (3). Eqs. (2) and (3) show that  $\rho_{\text{app}}^*$ ,  $Z_{\text{app}}^*$ , and  $\Delta\tilde{U}$  are linearly related  
 102 parameters with differing magnitude but the same phase value.

103 For a heterogeneous subsurface with a two-dimensional distribution of intrinsic complex  
 104 resistivity  $\rho^*$  (i.e.,  $\rho^*$  varies in horizontal  $x$  and vertical  $z$  but constant in  $y$  direction), the potential  
 105  $U$  at coordinate  $(x, y, z)$  due to a point current source  $I$  is described by the Fourier transformed  
 106 Poisson's equation (e.g., Kemna 2000; Binley 2015),

$$\frac{\partial}{\partial x} \left( \frac{1}{\rho^*} \frac{\partial v^*}{\partial x} \right) + \frac{\partial}{\partial z} \left( \frac{1}{\rho^*} \frac{\partial v^*}{\partial z} \right) - \frac{v^* k^2}{\rho^*} = -I \delta(x) \delta(z), \quad (4)$$

$$U(x, y, z) = \frac{1}{\pi} \int_0^\infty v^*(x, k, z) \cos(ky) dk, \quad (5)$$

107 where  $\delta$  is the Dirac delta function,  $v^*$  is the Fourier transformed complex voltage and  $k$  is the wave  
 108 number. Eqs. (4) and (5) are solved numerically via discretization, for example using the finite  
 109 element method. The superposition of calculated potentials at the potential (voltage recording)  
 110 electrodes and application of Eqs. (2) and (3) yields the  $\rho_{\text{app}}^*$  of four-electrode measurements  
 111 acquired over a heterogeneous  $\rho^*$  subsurface.

112 To investigate the occurrence of negative IP in FDIP (i.e., a positive  $\varphi_{\text{app}}$ ), we consider a  
 113 subsurface modeled by a number of small cells with each cell  $j$  ( $j = 1, 2, \dots, M$ ) characterized by  
 114 an intrinsic complex resistivity  $\rho_j^*$  (with magnitude  $|\rho_j^*|$  and phase  $\varphi_j$ ). If we consider cells  
 115 parameterized in terms of the logarithms,  $\ln \rho_j^*$ , and measurements equivalently expressed as  
 116  $\ln \rho_{\text{app}}^*$ , then for a single four-electrode measurement, the sensitivity to the cell  $j$  ( $S_j^*$ ) quantifies how  
 117 the change in  $\ln \rho_j^*$  changes  $\ln \rho_{\text{app}}^*$ ,

$$S_j^* = \frac{\partial \ln \rho_{\text{app}}^*}{\partial \ln \rho_j^*} = \frac{\partial \ln(|\rho_{\text{app}}^*| e^{\varphi_{\text{app}} i})}{\partial \ln(|\rho_j^*| e^{\varphi_j i})} = \frac{\partial(\ln|\rho_{\text{app}}^*| + \varphi_{\text{app}} i)}{\partial(\ln|\rho_j^*| + \varphi_j i)} \quad (6)$$

118 Different from a conventional direct current (DC) resistivity measurement,  $S_j^*$  of the FDIP  
 119 measurement is a complex number. As the derivatives of the complex functions in Eq. (6) satisfy  
 120 the Cauchy-Riemann conditions (Kemna 2000), the following sensitivity components can be  
 121 expressed as the real part of  $S_j^*$ :

$$S_j = \frac{\partial \ln|\rho_{\text{app}}^*|}{\partial \ln|\rho_j^*|} = \frac{\partial \varphi_{\text{app}}}{\partial \varphi_j}. \quad (7)$$

122 The imaginary part of  $S_j^*$  is,

$$S_{j, \text{im}} = \frac{\partial \ln|\rho_{\text{app}}^*|}{\partial \varphi_j} = -\frac{\partial \varphi_{\text{app}}}{\partial \ln|\rho_j^*|}. \quad (8)$$

123 Although we mainly focus on the discussion of a single four-electrode measurement, it should be  
 124 noted that a matrix comprising of  $S_j$  from a sequence of four-electrode measurements is the  
 125 Jacobian matrix used, for example, in a gradient-based inverse problem. In Eq. (7), the sensitivity

126 expressed in terms of complex resistivity magnitude is equivalent to that obtained for DC  
 127 resistivity measurements, which can take either positive or negative values. An increase of  $|\rho_j^*|$  in  
 128 a positive  $S_j$  zone will increase  $|\rho_{app}^*|$ , whereas an increase of  $|\rho_j^*|$  in a negative  $S_j$  zone will decrease  
 129  $|\rho_{app}^*|$ . An equivalent pattern holds for the phase terms as shown in Eq. (7). To illustrate, assume  
 130 that the subsurface space has zero phase (i.e., is non-polarizable), and thus  $\varphi_{app} = 0$ . If the phase of  
 131 an arbitrary cell  $\varphi_j$  decreases slightly to a negative value (i.e., becomes polarizable),  $\varphi_{app}$  will  
 132 decrease to be  $< 0$  if this polarizable cell is located in a zone of positive  $S_j$ . However,  $\varphi_{app}$  may  
 133 increase to be  $> 0$  (i.e., negative IP signal) if this polarizable cell is in a zone of negative  $S_j$ . This  
 134 provides a theoretical basis for the presence of positive  $\varphi_{app}$  (negative IP effects) in FDIP  
 135 measurements, i.e.,  $\varphi_{app} > 0$  is possible although all  $\varphi_j \leq 0$ . The imaginary sensitivity (Eq. 8) plays  
 136 a negligible role as shown later.

137 While the above arguments are based on the analysis of a single cell  $\varphi_j$  and  $S_j$ , a more  
 138 generalized way is to consider the collective impacts from all the cells. Kemna (2000) exploited  
 139 the expression in Eq. (7) by forming a “final phase improvement” in the inversion of complex  
 140 resistivity data once satisfactory matching of the resistivity magnitudes was achieved. Building on  
 141 this, consider an expression for the inversion of phase angles using the Gauss-Newton approach  
 142 (neglecting any damping or regularization for simplicity) (e.g., Kemna 2000; Binley 2015),

$$[S^T S] \Delta \mathbf{m} = S^T [\mathbf{d} - F(\mathbf{m}_k)] \quad (9)$$

$$\mathbf{m}_{k+1} = \mathbf{m}_k + \Delta \mathbf{m} \quad (10)$$

143 where  $\mathbf{S}$  is the sensitivity matrix for a sequence of four-electrode measurements,  $\mathbf{d}$  is a vector of  
 144 measured data ( $\varphi_{app}$  in this case),  $F$  is the forward modeling operator,  $\mathbf{m}$  is a vector of the model



145 parameters ( $\varphi_j$  in this case),  $\mathbf{m}_k$  and  $\mathbf{m}_{k+1}$  are the model parameter set at iteration  $k$  and  $k+1$ ,  
 146 respectively,  $\Delta\mathbf{m}$  is the model parameter update at iteration  $k$ . Assuming that the inversion is  
 147 achieved with only one step from a starting model with all  $\varphi_j = 0$ , we have  $\mathbf{m}_k = \mathbf{0}$ ,  $F(\mathbf{m}_k) = \mathbf{0}$ ,  $\mathbf{m}_{k+1}$   
 148  $= \Delta\mathbf{m}$ . We can then write Eq. (9) as,

$$\mathbf{S}\Delta\mathbf{m} = \mathbf{d}. \quad (11)$$

149 In this simplified one-step inversion,  $\Delta\mathbf{m}$  is essentially the final model that matches  $\mathbf{d}$ . Again, if  
 150 we only consider a single four-electrode measurement, Eq. (11) gives,

$$\sum_{j=1}^M S_j \varphi_j = \varphi_{\text{app}}. \quad (12)$$

151 This approximation describes the collective impacts of  $\varphi_j$  and  $S_j$  from all the cells. Eq. (12)  
 152 explicitly shows that even when all  $\varphi_j \leq 0$  ( $j = 1, 2, \dots, M$ ),  $\varphi_{\text{app}}$  can be positive when relatively  
 153 more negative  $\varphi_j$  cells concurrently have  $S_j < 0$ . The polarity of  $\varphi_{\text{app}}$  will therefore depend on the  
 154 relative values of intrinsic phase and the sensitivity, where the latter is affected by the quadrupole  
 155 geometry and distribution of the intrinsic resistivity.

156 A similar association between negative IP signals and the sensitivity distribution is  
 157 recognized in TDIP data (Dahlin & Loke 2015). In TDIP, a unidirectional current is driven  
 158 between the current electrodes for a period of time and then abruptly switched off. The voltage  $V_p$   
 159 recorded right before switching off is used to obtain the apparent DC resistivity  $\rho_{\text{app}}^{(\text{DC})}$  (assuming  
 160 the current injection is long enough to approximate a DC condition). After switching off the current,  
 161  $V_p$  drops suddenly to a secondary voltage  $V_s$ , which then decays with time. Seigel (1959) defined  
 162 the apparent chargeability ( $m_{\text{app}}$ ) as the ratio of  $V_s$  to  $V_p$  to quantify the TDIP polarization strength.

163 Considering the same scenario where the subsurface is modeled by  $M$  small cells with index  $j$ , the  
164 theoretical relationship between a single measure of  $\rho_{\text{app}}^{(\text{DC})}$  and  $m_{\text{app}}$ , and the intrinsic parameters  
165  $\rho_j^{(\text{DC})}$  and  $m_j$  making up the subsurface is (Seigel, 1959),

$$m_{\text{app}} = \sum_{j=1}^M \frac{\partial \ln \rho_{\text{app}}^{(\text{DC})}}{\partial \ln \rho_j^{(\text{DC})}} m_j = \sum_{j=1}^M S_j^{(\text{DC})} m_j. \quad (13)$$

166 where  $S_j^{(\text{DC})}$  is the sensitivity to a cell  $j$  in terms of DC resistivity, being analogous to the sensitivity  
167 in terms of complex resistivity magnitude (Eq. (7)). Eq. (13) has essentially the same structure as  
168 Eq. (12). With all  $m_j \geq 0$  ( $j = 1, 2, \dots, M$ ) for Earth materials, the polarity of  $m_{\text{app}}$  is decided by the  
169 polarity of  $S_j^{(\text{DC})}$  and the relative values of  $m_j$ . Eq. (13) predicts that negative  $m_{\text{app}}$  is possible when  
170 features with relatively high  $m_j$  fall into negative sensitivity zones, providing theoretical support  
171 for the negative IP effects in TDIP. In practice,  $m_{\text{app}}$  defined by Seigel (1959) is difficult to measure  
172 and an integral chargeability is instead commonly measured (Binley 2015), which can exhibit  
173 equivalent negative IP effects (Dahlin & Loke 2015).

174 We stress that laboratory measurements of intrinsic complex resistivity or chargeability on  
175 a core or soil sample (considered homogeneous at the measurement scale but in fact likely to  
176 contain small scale heterogeneity) can never exhibit negative IP effects when 1D current flow is  
177 maintained. Such negative IP effects sometimes reported in the literature (e.g., Abdulsamad *et al.*  
178 2016; Saneiyani *et al.* 2018; Bate *et al.* 2020) can only arise from measurement errors.

179

180

### 181 **3. Numerical modeling**

182 To investigate the behavior of the  $\varphi_{\text{app}}$  polarity, 2D forward modeling of synthetic intrinsic  
183 complex resistivity distributions was performed using cR2  
184 (<http://www.es.lancs.ac.uk/people/amb/Freeware/cR2/cR2.htm>) in its python wrapper ResIPy  
185 (Blanchy *et al.* 2020). The region of interest of the synthetic model contains 25 electrodes spaced  
186 2 m apart for a total length of 48 m and extends to 8 m depth (Figure 1). A quadrilateral mesh with  
187 each mesh cell of size  $0.25 \times 0.25$  m (i.e., 8 nodes per electrode) was used for the computations.  
188 In this case, each mesh cell corresponds to a small cell  $j$  described in Section 2. This mesh extends  
189 a large distance beyond the region of interest and incorporates boundary conditions that  
190 approximate an infinitely large model space. Different intrinsic resistivity and phase values were  
191 assigned to different regions to illustrate specific aspects of negative IP effects predicted by theory.  
192 Forward models were run to determine  $\varphi_{\text{app}}$  of either a single four-electrode measurement or to  
193 construct a pseudosection from a sequence of measurements. The four electrodes include a pair of  
194 electrodes (positive C+ and negative C-) for current injection and a pair of electrodes (positive P+  
195 and negative P-) for voltage (potential) measurements.

196 The sensitivity distribution for a single four-electrode measurement on a selected synthetic  
197 model was computed using cR2, with a vector of  $S_j^*$  corresponding to each mesh cell in the  
198 modeling space as the output (Eq. 6). No noise was added to the forward modeling and sensitivity  
199 distribution calculation so as to avoid the complicating effects of random errors on the modeling  
200 results.

#### 201 **3.1 Influence of sensitivity distribution**

202 The sensitivity distribution for a dipole-dipole array (E10=C+, E12=C-, E14=P-, E16=P+)  
203 and also for a Wenner array (E10=C+, E12=P+, E14=P-, E16=C-) was first computed for a  
204 homogeneous, low polarizability half-space ( $|\rho^*| = 100 \Omega \text{ m}$ ,  $\varphi = -1 \text{ mrad}$ ) (Figure 2). The  
205 imaginary sensitivity (Eq. (8), Figure 2c and 2d) exerts a negligible control on the measurements  
206 as its values are many orders of magnitude less than the real sensitivity (Eq. (7), Figure 2a and 2b).  
207 A simulation on a homogenous, high polarizability half-space ( $|\rho^*| = 100 \Omega \text{ m}$ ,  $\varphi = -100 \text{ mrad}$ )  
208 results in similar negligible response in the imaginary sensitivity distribution, again being many  
209 orders of magnitude less than the real sensitivity. We therefore refer to the real sensitivity in all  
210 future discussion of sensitivity patterns. Different patterns of positive and negative sensitivity are  
211 observed for the dipole-dipole (Figure 2a) and Wenner arrays (Figure 2b). The sensitivity of zones  
212 away from the electrode array is close to zero, therefore having a negligible effect on the  $\rho_{\text{app}}^*$   
213 measurement.

214 To illustrate the influence of the sensitivity distribution on the polarity of the measured  
215 phase, new forward models were run where  $\varphi_{\text{app}}$  of a single measurement using E10, E12, E14 and  
216 E16 was computed with a small polarizable cell ( $|\rho^*| = 100 \Omega \text{ m}$  and  $\varphi = -100 \text{ mrad}$ ) of the same  
217 size as a mesh cell ( $0.25 \times 0.25 \text{ m}$ ) placed at various locations in a background non-polarizing half  
218 space ( $|\rho^*| = 100 \Omega \text{ m}$ ,  $\varphi = 0 \text{ mrad}$ ) (Figure 3a). Starting from the first mesh cell, the polarizable  
219 cell was moved to the right and down one cell by one cell to cover the horizontal distance from 15  
220 to 24 m and the depth range from 0 to 6 m (containing the zone of enhanced sensitivity). With the  
221 polarizable cell at each mesh cell location, the apparent phase  $\varphi_{\text{app}}$  of a dipole-dipole array  
222 (E10=C+, E12=C-, E14=P-, E16=P+) and a Wenner array (E10=C+, E12=P+, E14=P-, E16=C-)

223 was computed. Figure 3 shows  $\varphi_{\text{app}}$  plotted against the sensitivity of the location with the  
224 polarizable cell for the corresponding measurement array. The polarity of  $\varphi_{\text{app}}$  is the inverse of the  
225 polarity of the sensitivity, i.e., the polarizable cell placed in positive sensitivity zones results in  
226 negative  $\varphi_{\text{app}}$  and the polarizable cell placed in negative sensitivity zones results in positive  $\varphi_{\text{app}}$ .  
227 The magnitude of the negative IP signal increases linearly with the magnitude of the negative  
228 sensitivity.

229

### 230 3.2 Influence of heterogeneity

231 We next investigate the effect of heterogeneity on the sensitivity and hence the  $\varphi_{\text{app}}$   
232 polarity pattern. A  $3 \times 3$  m polarizable block was located between 22.5 m and 25.5 m along the  
233 line and placed at a depth of 0 m to 3 m (Figure 4a). The background was set with  $\varphi_{\text{bgk}} = -1$  mrad  
234 (low polarization) and  $|\rho_{\text{bgk}}^*| = 100 \Omega \text{ m}$ , while the polarizable block was assigned  $\varphi_{\text{block}} = -100$   
235 mrad and  $|\rho_{\text{block}}^*|$  equal to either 50, 100 or 200  $\Omega \text{ m}$ . For each  $|\rho_{\text{block}}^*|$  scenario, a  $\varphi_{\text{app}}$  pseudosection  
236 was computed for a dipole-dipole array sequence with  $a = 4$  m and  $n = 1, 2, 3$  and 4 (i.e., electrodes  
237 placed in the order C+, C-, P-, P+ with spacing  $a, a \times n$  and  $a$  between C+ and C-, C- and P-, and  
238 P- and P+, respectively) (Figure 4b). The results show that the resistivity of the polarizable block  
239 has a significant influence on the polarity and magnitude of  $\varphi_{\text{app}}$  within the zones indicated by the  
240 dashed triangles. These  $\varphi_{\text{app}}$  values increase from negative to positive (i.e. negative IP effect) as  
241  $|\rho_{\text{block}}^*|$  increases from 50 to 100  $\Omega \text{ m}$ . Higher positive values of  $\varphi_{\text{app}}$  (i.e., enhanced negative IP  
242 effects) are observed for  $|\rho_{\text{block}}^*|$  equal to 200  $\Omega \text{ m}$ .

243            This control of the resistivity of the heterogeneity on the polarity of  $\varphi_{\text{app}}$  results from how  
244 the presence of the heterogeneity modifies the sensitivity distribution relative to a homogeneous  
245 resistivity medium. To illustrate this, a single  $\varphi_{\text{app}}$  measurement using E10=C+, E12=C-, E14=P-,  
246 E16=P+ (pointed out by arrows in Figure 4b) is used as an example. The corresponding sensitivity  
247 distribution for the three synthetic models with different  $|\rho_{\text{block}}^*|$  values is shown in Figure 4c. As  
248  $|\rho_{\text{block}}^*|$  increases from 50 to 100, and then to 200  $\Omega$  m,  $\varphi_{\text{app}}$  increases from -14 mrad to 9 mrad for  
249 the 100  $\Omega$  m block and to 33 mrad for the 200  $\Omega$  m block. This increase of  $\varphi_{\text{app}}$  toward more  
250 positive values with increasing  $|\rho_{\text{block}}^*|$  can be explained by the expansion of the negative sensitivity  
251 zones within the polarizable block boundary as  $|\rho_{\text{block}}^*|$  increases (Figure 4c). This change in the  
252 sensitivity pattern is highlighted by the difference in sensitivity referenced to the sensitivity for the  
253  $|\rho_{\text{block}}^*| = 100$   $\Omega$  m scenario, where  $|\rho_{\text{block}}^*| = 50$   $\Omega$  m and  $|\rho_{\text{block}}^*| = 200$   $\Omega$  m highlights increased  
254 and decrease sensitivity respectively within the block boundary (Figure 4d). This confirms that the  
255 resistivity heterogeneity has a significant influence on the polarity of  $\varphi_{\text{app}}$  by changing the  
256 sensitivity distribution.

257            So far, we have shown that the polarity of  $\varphi_{\text{app}}$  is determined by three major factors: (1) the  
258 location of polarizable objects relative to positive/negative sensitivity zones; (2) the intrinsic phase  
259 of the polarizable objects relative to the surrounding subsurface; (3) the subsurface resistivity  
260 heterogeneity that changes the sensitivity patterns. To illustrate the collective impacts of the  
261 intrinsic resistivity and intrinsic phase, we computed  $\varphi_{\text{app}}$  for a dipole-dipole array (E10=C+,  
262 E12=C-, E14=P-, E16=P+) using the same model structure and background settings as shown in  
263 Figure 4a, but with  $|\rho_{\text{block}}^*|$  varying from 20 to 200  $\Omega$  m and  $\varphi_{\text{block}}$  varying from -5 to -120 mrad

264 (Figure 5a). When  $|\rho_{\text{block}}^*| = 20, 40$  or  $60 \Omega \text{ m}$ , all  $\varphi_{\text{app}}$  are negative and become more negative with  
265  $\varphi_{\text{block}}$  changing from  $-5$  to  $-120$  mrad. When  $|\rho_{\text{block}}^*| = 80, 100, 120$  or  $140 \Omega \text{ m}$ ,  $\varphi_{\text{app}}$  is negative  
266 when  $\varphi_{\text{block}}$  is small ( $-5$  mrad), but becomes positive when  $\varphi_{\text{block}}$  is more negative. At  $|\rho_{\text{block}}^*|$  above  
267  $140 \Omega \text{ m}$ , all  $\varphi_{\text{app}}$  are positive even when  $\varphi_{\text{block}}$  is only  $-5$  mrad; again,  $\varphi_{\text{app}}$  becomes more positive  
268 as  $\varphi_{\text{block}}$  becomes more negative. A clear transition from negative  $\varphi_{\text{app}}$  to positive  $\varphi_{\text{app}}$  can be  
269 observed in Figure 5a, which shows that a higher  $|\rho_{\text{block}}^*|$  relative to  $|\rho_{\text{bgk}}^*|$  tends to result in positive  
270  $\varphi_{\text{app}}$ . The  $\varphi_{\text{app}}$  pattern will also be affected by other factors, for example the background phase  $\varphi_{\text{bgk}}$ .  
271 Figure 5b presents the  $\varphi_{\text{app}}$  change when  $\varphi_{\text{bgk}}$  is set to be  $-10$  mrad. In this situation, more points  
272 show negative  $\varphi_{\text{app}}$  with positive  $\varphi_{\text{app}}$  only occurring when  $|\rho_{\text{block}}^*|$  is sufficiently large and  $\varphi_{\text{block}}$  is  
273 sufficiently negative.

274 The shape of the polarizable block also determines the  $\varphi_{\text{app}}$  change under various  $|\rho_{\text{block}}^*|$   
275 and  $\varphi_{\text{block}}$  conditions. Figure 5c shows the simulation with the same model settings as that in Figure  
276 5a except that the vertical extent of the polarizable block is reduced to be between 0 to 1 m. In this  
277 case, most of the points show positive  $\varphi_{\text{app}}$  due to the increased portions of negative sensitivity  
278 zone in the polarizable block. For example, in the case of  $|\rho_{\text{block}}^*| = 100 \Omega \text{ m}$  in Figure 4c, when  
279 the vertical extent of the polarizable block is reduced to be between 0 to 1 m, most of the regions  
280 within the block would have negative sensitivity. In this situation, positive  $\varphi_{\text{app}}$  is more likely as  
281 per Eq. (12).

282 The above results were obtained from simple, heterogenous synthetic models. For a real  
283 subsurface, the interactions between complicated structures and zones may result in various  $\varphi_{\text{app}}$   
284 patterns, making it difficult to generalize about what situations will result in negative IP effects.

285 One important observation from Figure 5 is that even weakly polarizable objects (e.g.,  $\varphi_{\text{block}} = -5$   
286 and  $-10$  mrad) may produce negative IP signals, especially when the objects have high resistivity  
287 relative to the background (e.g., polarizable objects characterized by low water content, low  
288 porosity or high electrical resistivity pore fluids).

#### 289 **4. A physical explanation of negative IP effects using an electrical circuit**

290 We have so far explained the occurrence of negative IP signals using theory and numerical  
291 modeling. Next, we seek a more physical explanation as a positive phase implies that the electrical  
292 current lags the voltage, which is considered to be non-physical in the presence of IP effect. We  
293 use a simplified electrical circuit model to provide a physical explanation for negative IP effects.  
294 We consider a subsurface represented by a resistor/impedance network circuit (Figure 6a). A  
295 sinusoidal current  $\tilde{I}_0$  with fixed amplitude  $|\tilde{I}_0|$  and zero reference phase is injected between C+ and  
296 C-, while the resultant sinusoidal voltage  $\Delta\tilde{U}$  (with amplitude  $|\Delta\tilde{U}|$  and phase  $\varphi_{\Delta U}$ ) is measured  
297 between P+ and P- in the same manner as a dipole-dipole array. Comparing the relative locations  
298 of the circuit components in Figure 6a with Figure 2a,  $Z_1^*$  (with magnitude  $|Z_1^*|$  and phase  $\varphi_1$ ) and  
299  $Z_2^*$  (with magnitude  $|Z_2^*|$  and phase  $\varphi_2$ ) represent impedance components located in the positive and  
300 negative sensitivity zones, respectively. We next evaluate how changes of  $Z_1^*$  or  $Z_2^*$  alter the  
301 apparent measured impedance  $Z_{\text{app}}^*$  (i.e.,  $\Delta\tilde{U}/\tilde{I}_0$ ).

302 To make the analysis simple, we set all other circuit components to be pure resistors  
303 (represented by symbol 'R'). According to Figure 6a,  $\tilde{I}_0$  exits the network via 'C-' by passing  $Z_1^*$ ,  
304  $Z_2^*$ ,  $R_3$ ,  $R_4$  and  $R_5$ , which gives  $\tilde{I}_0 = \tilde{I}_1 + \tilde{I}_4 + \tilde{I}_2$  with  $\tilde{I}_2 = \tilde{I}_3 + \tilde{I}_5$ , where  $\tilde{I}_1$  to  $\tilde{I}_5$  are the currents flowing  
305 through the corresponding impedance/resistors. We simplify this network circuit to an equivalent



306 linear circuit that is easier to analyze (Figure 6b). In Figure 6b,  $R_{3s}$ ,  $R_{4s}$  and  $R_{5s}$  represent the  
 307 equivalent total resistances of the current path prior to  $R_3$ ,  $R_4$ , and  $R_5$  respectively, while other  
 308 components are identical to those shown in Figure 6a. The total impedance of this circuit is,

$$Z_{\text{tot}}^* = \frac{\tilde{U}_0}{\tilde{I}_0} = \frac{1}{\frac{1}{Z_1^*} + \frac{1}{R_{4s} + R_4} + \frac{1}{\frac{(R_{5s} + R_5)(R_{3s} + R_3)}{(R_{5s} + R_5) + (R_{3s} + R_3)} + Z_2^*}} = \frac{1}{\frac{1}{Z_1^*} + a + \frac{1}{b + Z_2^*}}, \quad (14)$$

309 where  $\tilde{U}_0$  is the total voltage between C+ and C- and  $a$  and  $b$  are real number constants as  
 310 resistances  $R_{3s}$ ,  $R_{3s}$ ,  $R_{4s}$ ,  $R_4$ ,  $R_{5s}$  and  $R_5$  do not change. According to the voltage divider rule,

$$\frac{\Delta\tilde{U}}{\tilde{U}_0} = \frac{R_3}{(R_{3s} + R_3)} \frac{b}{b + Z_2^*} = c \frac{b}{b + Z_2^*}, \quad (15)$$

311 where  $c$  is again a real number constant representing a constant resistance term. Combining Eq.  
 312 (2), Eq. (14) and (15) gives,

$$|Z_{\text{app}}^*| e^{\varphi_{\text{app}} i} = \frac{\Delta\tilde{U}}{\tilde{I}_0} = \frac{bc}{\frac{b + Z_2^*}{Z_1^*} + aZ_2^* + ab + 1}. \quad (16)$$

313 Considering that the intrinsic phase shifts of the earth materials are small negative values ( $-0.2 <$   
 314  $\varphi < 0$ ),  $\cos \varphi \approx 1$  and  $\varphi \approx \sin(\varphi) \approx \tan(\varphi) \approx \tan^{-1}(\varphi)$ . Any impedance term  $Z^*$  can then be written in  
 315 rectangular form as  $Z^* = |Z^*| \cos(\varphi) + i|Z^*| \sin(\varphi) \approx |Z^*| + i|Z^*| \varphi$ . When  $Z_1^*$  (located in the positive  
 316 sensitivity zone of the array) is polarizable (i.e.,  $\varphi_1 < 0$ ) and  $Z_2^*$  (located in the negative sensitivity  
 317 zone) is non-polarizable (i.e.,  $\varphi_2 = 0$ ), Eq. (16) gives,

$$\varphi_{\text{app}} \approx \varphi_1 \frac{b + |Z_2^*|}{b + |Z_2^*| + a|Z_1^*||Z_2^*| + ab|Z_1^*| + |Z_1^*|}, \quad (17)$$

318 which explicitly shows that  $\varphi_{\text{app}} < 0$ , being a measurement signal with normal polarity. On the  
 319 contrary, if  $Z_2^*$  is polarizable (i.e.,  $\varphi_2 < 0$ ) and  $Z_1^*$  is non-polarizable (i.e.,  $\varphi_1 = 0$ ), Eq. (16) results  
 320 in,

$$\varphi_{\text{app}} \approx -\varphi_2 \frac{|Z_2^*| + a|Z_1^*||Z_2^*|}{b + |Z_2^*| + a|Z_1^*||Z_2^*| + ab|Z_1^*| + |Z_1^*|}, \quad (18)$$

321 which gives  $\varphi_{\text{app}} > 0$ , being a measurement signal with negative IP polarity. It can be concluded  
 322 that the negative IP signals originate from the fact that the impedance is determined from dividing  
 323 the recorded voltage  $\Delta\tilde{U}$  by the input current  $\tilde{I}_0$  instead of by the current flowing through the  
 324 impedance across which  $\Delta\tilde{U}$  is recorded, i.e.,  $\tilde{I}_3$  in our case. It is the phase difference between  $\tilde{I}_3$   
 325 and measured  $\tilde{I}_0$  that gives the non-physical impression of the current lagging the voltage as  
 326 implied by a positive phase. The circuit model analogy also explains the impact of sensitivity on  
 327 the resistivity measurements (i.e., resistance measurement in the circuit model). Considering  $Z_1^*$   
 328 and  $Z_2^*$  as pure resistors (i.e., zero phase), Eq. (16) shows that  $|Z_{\text{app}}^*|$  increases with the increase of  
 329  $|Z_1^*|$ , whereas it decreases with the increase of  $|Z_2^*|$ .

## 330 5. Frequency dependence

331 The influence of resistivity and phase variability on the polarity of  $\varphi_{\text{app}}$  also has important,  
 332 hitherto unrecognized, implications for the interpretation of spectral IP datasets. The  $\varphi_{\text{app}}$  polarity  
 333 can vary with frequency if the resistivity of polarizable features changes significantly with  
 334 frequency, e.g., as observed for electronically conducting materials (e.g., Pelton *et al.* 1978; Wong

335 1979). We examine this effect using the same synthetic model structure shown in Figure 4a but  
336 assigning various values of frequency independent  $|\rho_{\text{bgk}}^*|$ ,  $\varphi_{\text{bgk}}$  and frequency-dependent  $|\rho_{\text{block}}^*|$   
337 and  $\varphi_{\text{block}}$ . We define the frequency dependence of the polarizability of the block using a Cole-  
338 Cole type model (Cole & Cole 1941; Pelton *et al.* 1978) with parameters previously found to fit  
339 laboratory experimental data obtained on a zero valent iron-sand mixture (50% iron by volume)  
340 (Slater *et al.* 2005) (Figure 7a). The spectra cover frequencies from  $10^{-3}$  to  $10^4$  Hz, with  $|\rho_{\text{block}}^*|$   
341 decreasing from 41 to 14  $\Omega$  m (from low to high frequency). The  $\varphi_{\text{block}}$  ranges from  $-21$  mrad to  $-$   
342  $174$  mrad, with the peak occurring at  $\sim 1$  Hz. The frequency independent background half-space  
343 was assigned  $\varphi_{\text{bgk}} = -1$  mrad, with the  $|\rho_{\text{bgk}}^*|$  set to either 10, 30 or 55  $\Omega$  m in order to simulate  
344 scenarios with  $|\rho_{\text{bgk}}^*|$  lower, close to or higher than  $|\rho_{\text{block}}^*|$  (Figure 7a).

345 Figure 7b shows the apparent parameters  $|\rho_{\text{app}}^*|$  and  $\varphi_{\text{app}}$  from the single measurement for a  
346 dipole-dipole array (E10=C+, E12=C-, E14=P-, E16=P+) at various frequencies. Three  
347 simulations result in completely different shapes of  $\varphi_{\text{app}}$  curves when only the resistivity contrast  
348 between the target and the background changes between the simulations. For the highest  
349 background resistivity,  $|\rho_{\text{bgk}}^*| = 55$   $\Omega$  m, the  $\varphi_{\text{app}}$  spectra are negative and display a negative peak  
350 similar to the  $\varphi_{\text{block}}$  spectrum. When  $|\rho_{\text{bgk}}^*|$  is reduced to 30  $\Omega$  m,  $\varphi_{\text{app}}$  is negative at high frequencies  
351 but increases to be positive below around 20 Hz. Peaks are observed in both positive and negative  
352 apparent phase domains. For the lowest background resistivity  $|\rho_{\text{bgk}}^*| = 10$   $\Omega$  m, all  $\varphi_{\text{app}}$  values  
353 become positive and a peak of  $\varphi_{\text{app}}$  toward more positive values is observed.

354 The differences among the three  $\varphi_{\text{app}}$  curves can be explained by the difference in  
355 resistivity of  $|\rho_{\text{block}}^*|$  relative to  $|\rho_{\text{bgk}}^*|$  and how this difference affects the sensitivity distribution, as

356 demonstrated in Section 3.2. Positive  $\varphi_{\text{app}}$  values are found when  $|\rho_{\text{block}}^*|/|\rho_{\text{bgk}}^*|$  is relatively high,  
357 being the case when  $|\rho_{\text{bgk}}^*| = 10 \Omega \text{ m}$  for all frequencies and when  $|\rho_{\text{bgk}}^*| = 30 \Omega \text{ m}$  at low frequencies.  
358 The  $|\rho_{\text{app}}^*|$  spectra also differ between the three simulations, exhibiting a frequency dependence  
359 consistent with the polarity of  $\varphi_{\text{app}}$ . The percentage frequency effect ( $\text{PFE} = (|\rho_{\text{app}}^*|_{\text{L}} - |\rho_{\text{app}}^*|_{\text{H}}) /$   
360  $|\rho_{\text{app}}^*|_{\text{L}}$ , where subscripts  $\text{H}$  and  $\text{L}$  refer to a high and low measurement frequency, respectively) is  
361 another measure of the IP effect that was popular in mineral exploration (Ward 1988). Figure 7b  
362 shows that a negative PFE (i.e., increasing  $|\rho_{\text{app}}^*|$  with increasing frequencies) is always observed  
363 when  $\varphi_{\text{app}}$  is positive. Just as with positive  $\varphi_{\text{app}}$  values, a negative PFE is non-physical from the  
364 perspective of IP mechanisms and another representation of negative IP effects in frequency  
365 domain IP measurements.

366 In summary, this simulation of frequency dependent data demonstrates the possibility of a  
367 wide range of  $\varphi_{\text{app}}$  spectra, which can be very different from the spectra of an intrinsic polarizable  
368 target. This has significant implications with respect to the interpretation of field-measured phase  
369 curves.

## 370 **6. Sandbox experiments**

371 Laboratory sandbox experiments were conducted to verify the observations from numerical  
372 modeling (Figure 8a). A sandbox 36 cm wide, 15 cm high and 55 cm long was filled with sand  
373 fully saturated with tap water (resistivity of  $40 \Omega \text{ m}$  at  $25 \text{ }^\circ\text{C}$ ). Four electrodes were deployed in  
374 the central area of the sandbox with a 5 cm spacing. The distance between the electrodes and the  
375 box wall was large enough to ignore boundary effects on the measurements. FDIP (from 0.1 to  
376 100 Hz) and TDIP data (1 Hz waveform) were measured using an Ontash & Ermac PSIP

377 instrument and an IRIS Syscal Pro instrument, respectively.  $\varphi_{\text{app}}$  and  $M_{\text{app}}$  of the background sand  
378 was  $-2$  mrad and  $2$  mV/V respectively, providing a low polarizability background matrix.

379 To simulate a scenario similar to the synthetic model in Section 5, a piece of the iron  
380 mineral magnetite (dimensions approximately 8 cm length, 4 cm height and 5 cm width) was  
381 buried between the middle two electrodes at 2 cm depth. The  $\varphi_{\text{app}}$  collected using the dipole-dipole  
382 array is negative at high frequencies and then increases to positive values below 4 Hz (Figure 8b).  
383 The spectral shape of  $\varphi_{\text{app}}$  in Figure 8b is similar to the shape of the 0.1-200 Hz segment of the  
384 simulated blue  $\varphi_{\text{app}}$  curve ( $|\rho_{\text{bgk}}^*| = 30 \Omega \text{ m}$ ) in Figure 7b. The  $M_{\text{app}}$  measured with the dipole-dipole  
385 array is  $-42.5$  mV/V, also indicating a negative IP response. Its polarity is consistent with the  $\varphi_{\text{app}}$   
386 polarity at low frequencies. For the Wenner array measurement, a conventional negative  $\varphi_{\text{app}}$   
387 spectrum is observed (Figure 8c) as the polarizable magnetite falls within the positive sensitivity  
388 zones of this array (Figure 2b). The  $M_{\text{app}}$  measured by the Wenner array is positive ( $27.8$  mV/V),  
389 being consistent with the negative  $\varphi_{\text{app}}$  recorded in the frequency domain. These laboratory  
390 experiments therefore confirm the observations from numerical modeling and theory.

## 391 7. Conclusions

392 In a heterogenous polarizable subsurface the apparent phase  $\varphi_{\text{app}}$  recorded in surface four-  
393 electrode FDIP measurements may be positive. The polarity of  $\varphi_{\text{app}}$  is associated with the  
394 sensitivity distribution of a four-electrode measurement layout and is determined by the intrinsic  
395 phase and resistivity of the subsurface. Considerations of the sensitivity patterns of complex  
396 resistivity measurements theoretically confirm the occurrence of positive  $\varphi_{\text{app}}$ , i.e., for a non-  
397 polarizable subsurface, placing a small, highly polarizable object in the negative and positive

398 sensitivity zones will result in positive and negative  $\varphi_{\text{app}}$ , respectively. This is consistent with a  
399 simplified electric circuit model, which physically explains the negative IP (i.e., the paradox of  
400 current appearing to lag voltage) to result from the measured voltage drop across the potential  
401 electrodes being divided by the input current at the current electrodes instead of the current flowing  
402 through the impedance across the potential electrodes.

403 Numerical modeling shows the  $\varphi_{\text{app}}$  polarity is dictated by the relative values of both the  
404 intrinsic phase and the intrinsic resistivity of a polarizable heterogeneity compared to the  
405 background medium. The control of the relative strength of the intrinsic resistivity on  $\varphi_{\text{app}}$  results  
406 from its influence on the sensitivity distribution of a measurement. In the case that the intrinsic  
407 resistivity varies significantly with frequency, the  $\varphi_{\text{app}}$  polarity can vary with frequency in FDIP  
408 measurements, which results in  $\varphi_{\text{app}}$  spectra that are very different from the intrinsic phase  
409 spectrum. This finding is confirmed by laboratory sandbox experiments where  $\varphi_{\text{app}}$  of a dipole-  
410 dipole array on a buried piece of magnetite is negative from 100 to 4 Hz and then becomes positive  
411 below 4 Hz. Our results emphasize the need to accurately quantify error sources in FDIP  
412 measurements as positive  $\varphi_{\text{app}}$  measurements should be expected, are likely to be common in  
413 heterogeneous systems and should not simply be discarded prior to further data processing e.g.  
414 inversion. This observation is consistent with previously studied negative apparent chargeability  
415 data in TDIP measurements.

## 416 **Acknowledgements**

417 This research was partly funded by the U.S. Department of Energy under grant DE-SC0016412  
418 and a Rutgers University-Newark Graduate School Dissertation Fellowship award to C. Wang.

419 Supplemental funding for this project was provided by the Rutgers University-Newark  
420 Chancellor's Research Office. C. Wang thanks Sina Saneiyan (Rutgers University-Newark) for  
421 guidance on the use of ResIPy. We thank Andreas Hördt, Konstantin Titov, Timothy Johnson and  
422 an anonymous reviewer for their valuable comments that improved the quality of the paper.

### 423 **Data availability**

424 The data from this work is available upon request from the corresponding author.

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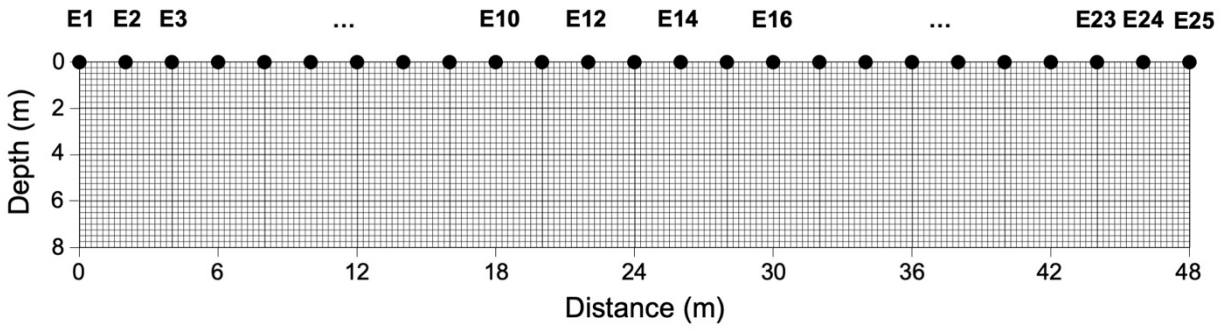
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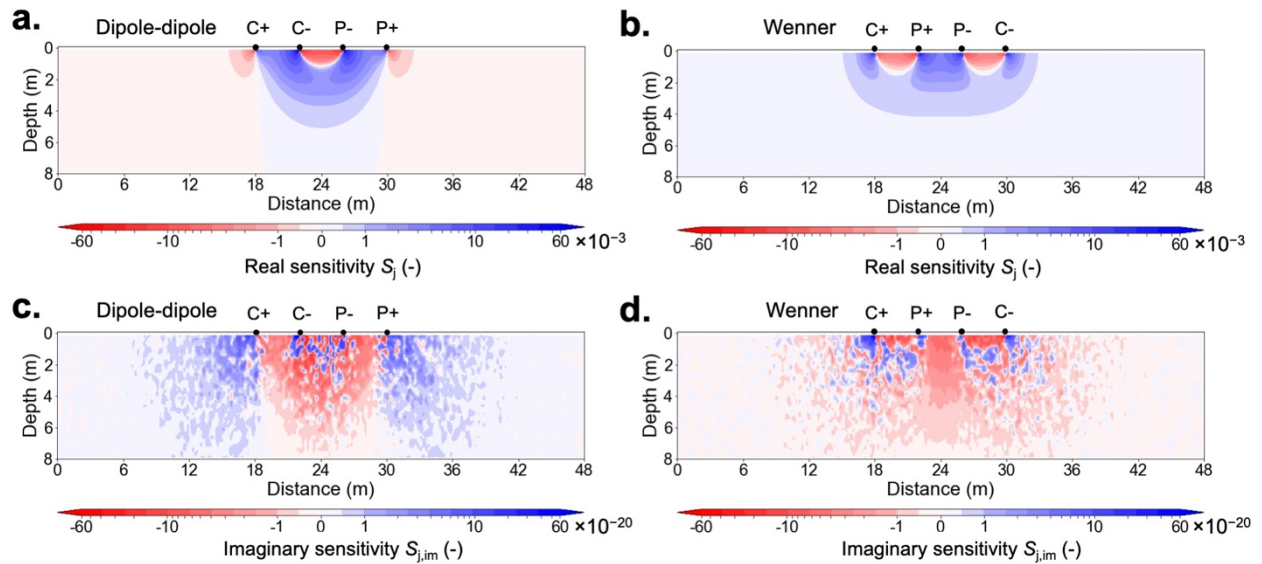
508

509 Figure 1. Numerical modeling set up with 25 electrodes (E1 to E25) on a model space using 0.25

510 × 0.25 m mesh cells.

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513

514 Figure 2. Sensitivity distribution of complex resistivity measurements using electrodes E10, E12,

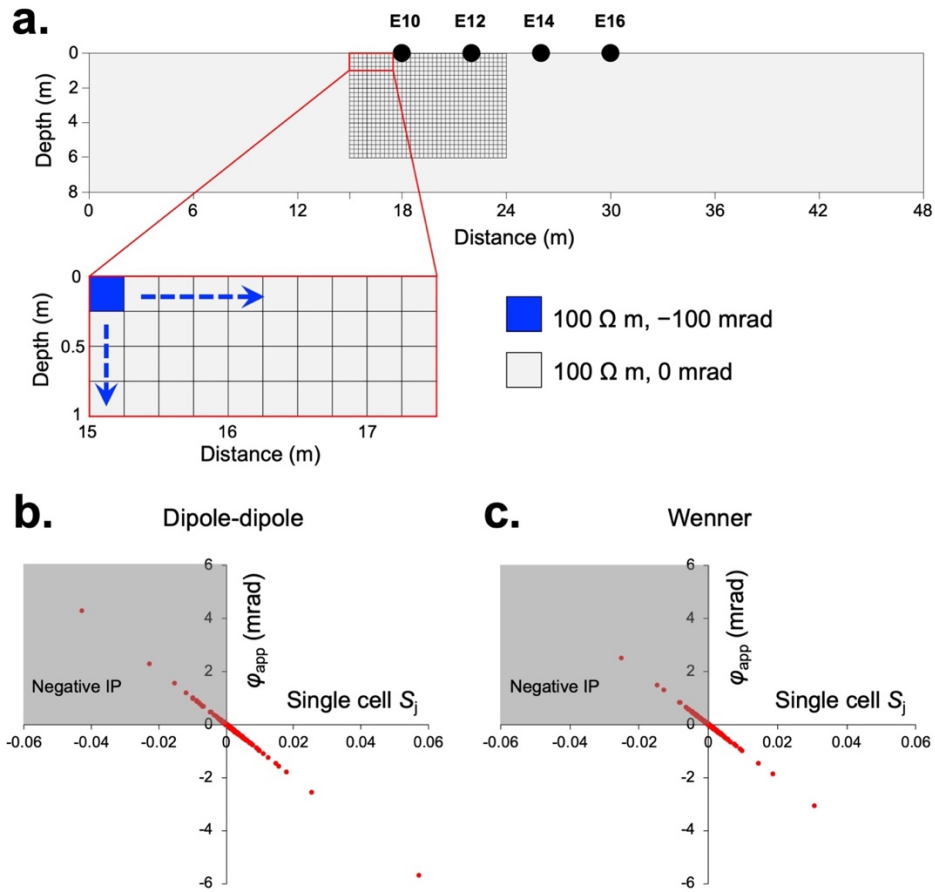
515 E14 and E16 for a  $100 \Omega \text{ m}$  and  $-1 \text{ mrad}$  homogeneous half space. (a) Real sensitivity of dipole-

516 dipole array (E10=C+, E12=C-, E14=P-, E16=P+). (b) Real sensitivity of Wenner array (E10=C+,

517 E12=P+, E14=P-, E16=C-). (c) Imaginary sensitivity of dipole-dipole array. (d) Imaginary

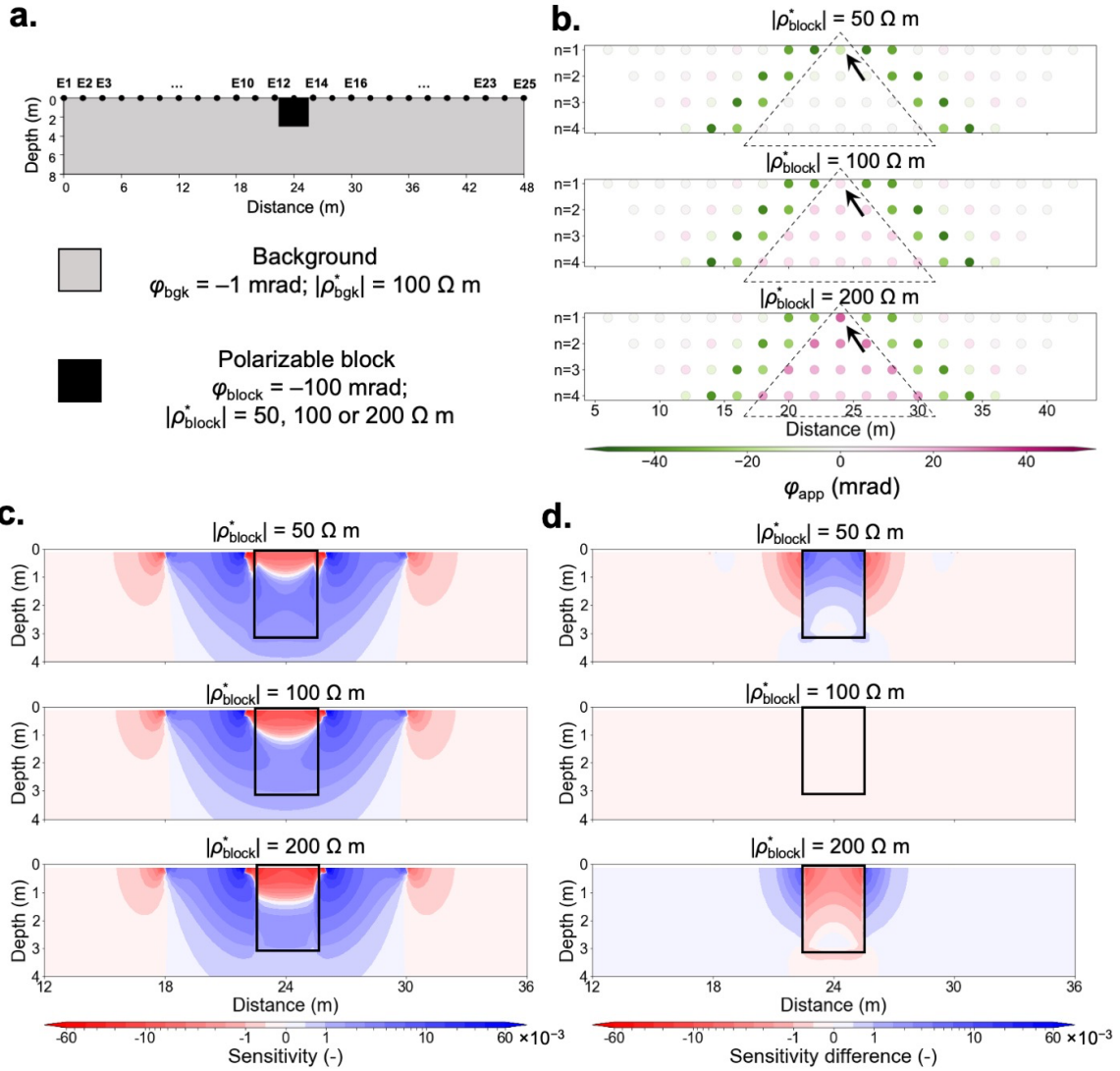
518 sensitivity of Wenner array.

519



521

522 Figure 3. Numerical modeling of the influence of sensitivity polarity on the  $\varphi_{app}$  polarity. (a).  
 523 Illustration of the model configuration; a polarizable cell (blue) moves to the right and down one  
 524 cell by one cell (in the zoomed in figure) to cover the region of 15 to 24 m distance and 0 to 6 m  
 525 depth (meshed region in the zoomed out figure); with the polarizable cell in each location,  $\varphi_{app}$  for  
 526 a dipole-dipole array (E10=C+, E12=C-, E14=P-, E16=P+) and a Wenner array (E10=C+, E12=P+,  
 527 E14=P-, E16=C-) were computed; (b) and (c).  $\varphi_{app}$  for a dipole-dipole (b) and Wenner (c) array  
 528 versus the sensitivity  $S_j$  (unitless) of the single cell containing the polarizable cell (blue cell in 3a);  
 529 Grey shaded quadrants highlight the negative IP responses.



530

531 Figure 4. Influence of resistivity heterogeneity on the  $\varphi_{\text{app}}$  polarity. (a). Synthetic model settings.

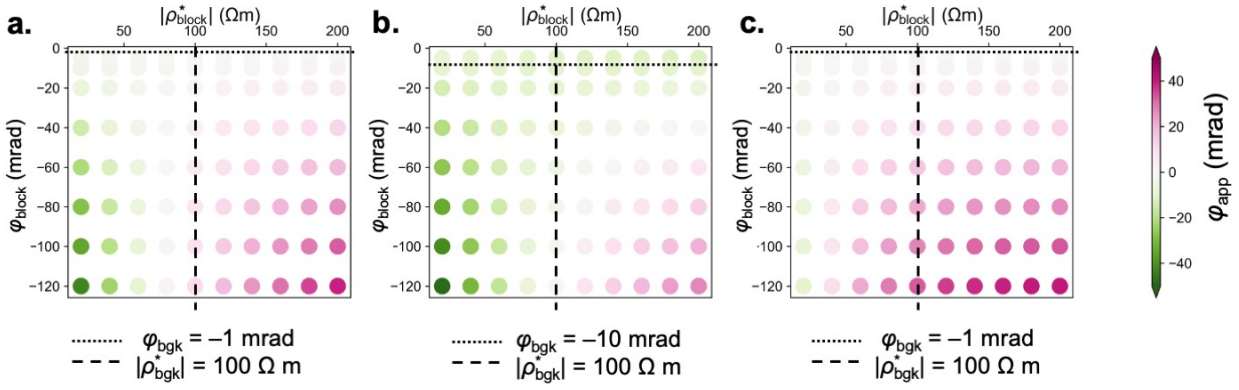
532 (b). Pseudosection of  $\varphi_{\text{app}}$  at various values of  $|\rho_{\text{block}}^*|$  ( $|\rho_{\text{block}}^*| = 100 \Omega \text{ m}$  is the homogeneous

533 resistivity condition); data within the dashed triangles are influenced by  $|\rho_{\text{block}}^*|$ . (c). Sensitivity

534 distribution of the single four-electrode measurement pointed out by the arrow in (b) with various

535  $|\rho_{\text{block}}^*|$  corresponding to the pseudosections. (d). Sensitivity difference relative to that of  $|\rho_{\text{block}}^*|$

536  $= 100 \Omega \text{ m}$ .



537

538 Figure 5. Impacts of  $|\rho_{block}^*|$  and  $\varphi_{block}$  on the modeled  $\varphi_{app}$  under various conditions. (a) and (b).

539  $\varphi_{app}$  modeled using a dipole-dipole array (E10=C+, E12=C-, E14=P-, E16=P+) for the synthetic

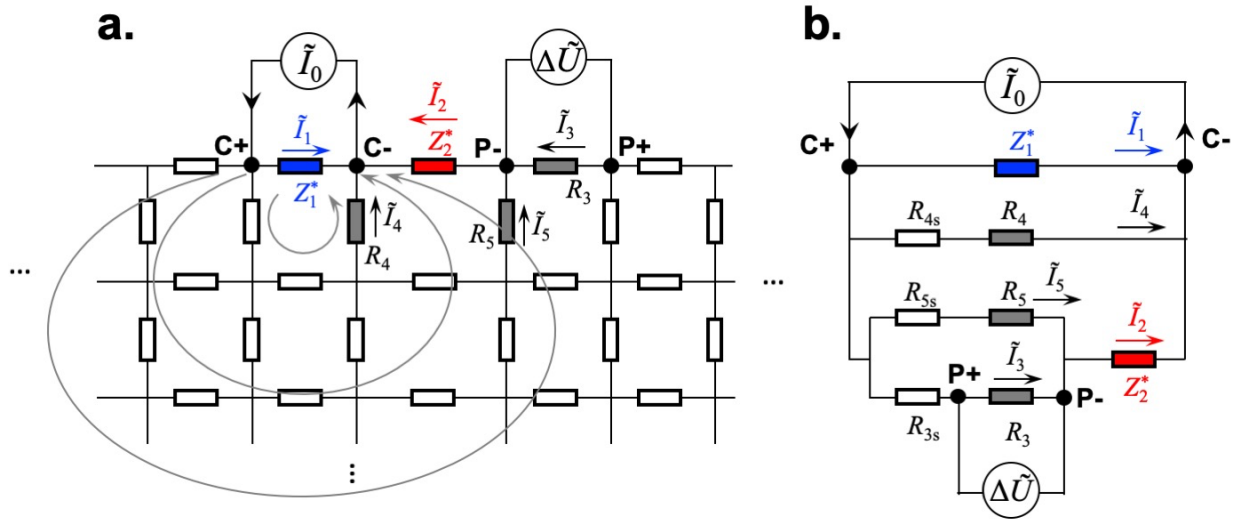
540 structure shown in Figure 4a under various background settings (indicated by dotted and dashed

541 lines). (c).  $\varphi_{app}$  modeled with the same settings as (a) but with vertical extent of polarizable block

542 in Figure 4a reduced to be between 0 and 1 m.

543

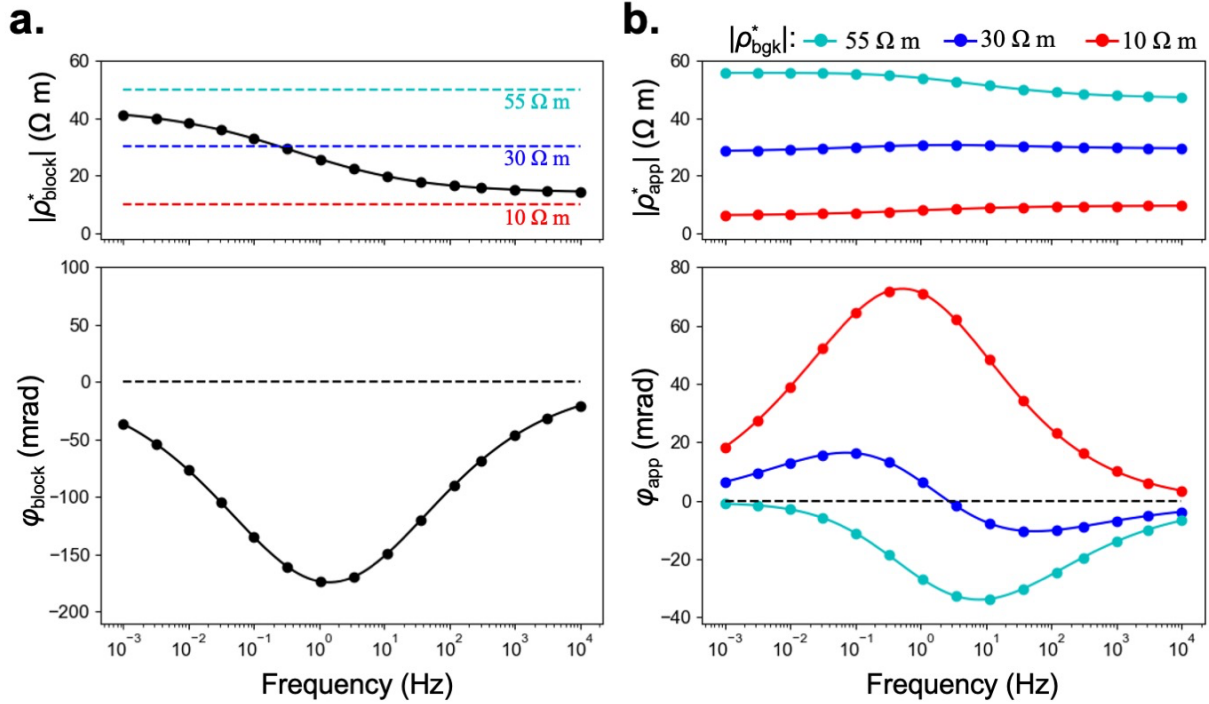




544

545 Figure 6. (a). Electrical conduction through the subsurface modeled as a resistor/impedance  
 546 network circuit; grey arrows illustrate idealized current flow directions in a real subsurface space  
 547 for comparison;

548 (b) A simplified equivalent linear electrical circuit of the circuit conceptualized in  
 549 (a).



550

551 Figure 7. Simulation based on a polarizable block with frequency-dependent complex resistivity

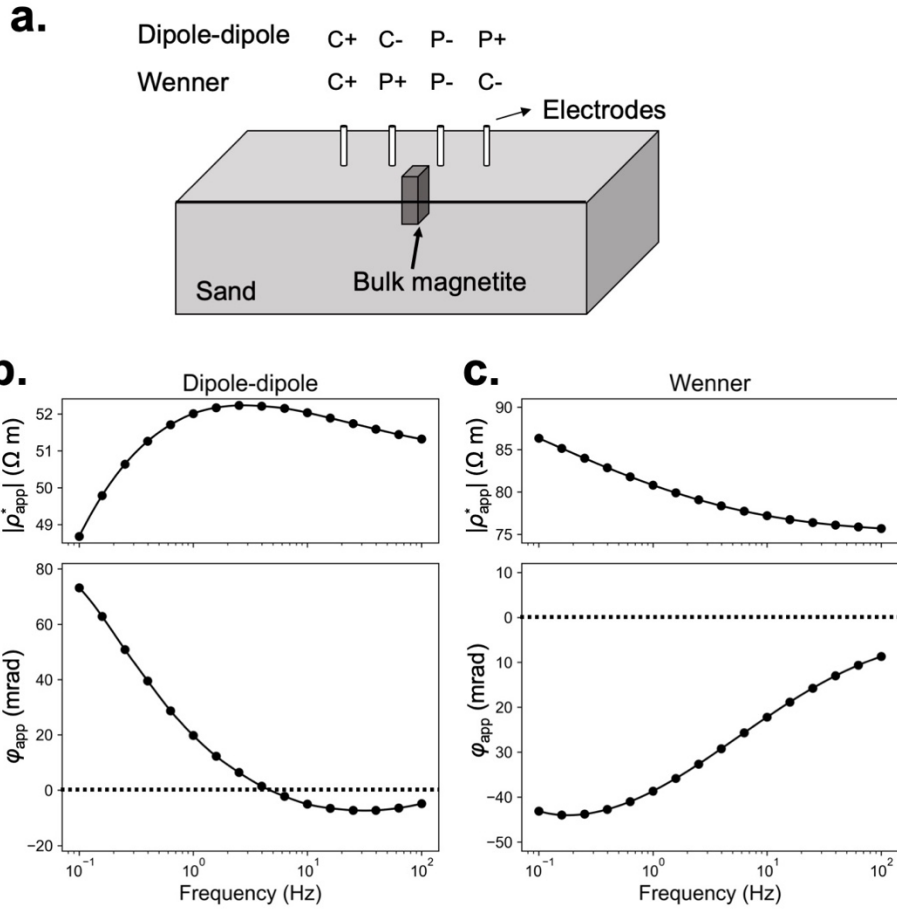
552 using model structure shown in Figure 4a (a). Intrinsic resistivity and phase spectra of the

553 polarizable block and the selection of frequency-independent background resistivity (colored

554 dashed lines); black dashed line represents  $\varphi_{\text{block}} = 0$  mrad; (b).  $|\rho_{\text{app}}^*|$  and  $\varphi_{\text{app}}$  spectra under

555 different  $|\rho_{\text{bkg}}^*|$  conditions; black dashed line represents  $\varphi_{\text{app}} = 0$  mrad.

556



557

558 Figure 8. Sandbox experiments. (a) Schematic diagram of sandbox experimental set-up. (b).  $|\rho_{app}^*|$   
 559 and  $\varphi_{app}$  spectra measured by dipole-dipole array; black dashed line represents  $\varphi_{app} = 0$  mrad (c)  
 560  $|\rho_{app}^*|$  and  $\varphi_{app}$  spectra measured by Wenner array; black dashed line represents  $\varphi_{app} = 0$  mrad.