

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

2 Chen Wang¹, Andrew Binley² and Lee D. Slater¹

3 ¹Department of Earth and Environmental Sciences, Rutgers University, Newark, New Jersey,
4 USA

5 ² Lancaster Environment Centre, Lancaster University, Lancaster, UK

6

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

8

9 **Corresponding author:**

10 Chen Wang

11 Phone: +1 973-353-5100; Fax: +1 973-353-1965;

12 Email: cw701@scarletmail.rutgers.edu

13

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 (ρ^*) or its inverse, the complex conductivity (σ^*):

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

86 where $|\rho^*|$ is the complex resistivity magnitude, φ is the complex resistivity phase ($\varphi \leq 0$) and i is
 87 the imaginary unit with $i^2 = -1$. Both ρ^* and σ^* 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_{app}^* (with magnitude
 94 $|Z_{\text{app}}^*|$ and φ_{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 ω 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 ρ_{app}^* (with magnitude $|\rho_{\text{app}}^*|$
 98 and the same phase φ_{app} as that of Z_{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 ρ_{app}^* is the complex resistivity of a homogeneous space equivalent to the value of Z_{app}^* resulting
 101 from application of Eq. (3). Eqs. (2) and (3) show that ρ_{app}^* , Z_{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 ρ^* (i.e., ρ^* 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 δ 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 ρ_{app}^* of four-electrode measurements
 111 acquired over a heterogeneous ρ^* subsurface.

112 To investigate the occurrence of negative IP in FDIP (i.e., a positive φ_{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 ρ_j^* (with magnitude $|\rho_j^*|$ and phase φ_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 φ_j decreases slightly to a negative value (i.e., becomes polarizable), φ_{app} will
 132 decrease to be < 0 if this polarizable cell is located in a zone of positive S_j . However, φ_{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 φ_{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 φ_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 (φ_{app} in this case), F is the forward modeling operator, \mathbf{m} is a vector of the model

145 parameters (φ_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 φ_j and S_j from all the cells. Eq. (12)
 152 explicitly shows that even when all $\varphi_j \leq 0$ ($j = 1, 2, \dots, M$), φ_{app} can be positive when relatively
 153 more negative φ_j cells concurrently have $S_j < 0$. The polarity of φ_{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_{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_{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_{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_{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_{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 φ_{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 0.25×0.25 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 φ_{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 ρ_{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 φ_{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 φ_{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 φ_{app} plotted against the sensitivity of the location with the
224 polarizable cell for the corresponding measurement array. The polarity of φ_{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 φ_{app} and the polarizable cell placed in negative sensitivity zones results in positive φ_{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 φ_{app}
232 polarity pattern. A 3×3 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 φ_{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 φ_{app} within the zones indicated by the
240 dashed triangles. These φ_{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 φ_{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 φ_{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 φ_{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 Ω m, φ_{app} increases from -14 mrad to 9 mrad for
249 the 100 Ω m block and to 33 mrad for the 200 Ω m block. This increase of φ_{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 $|\rho_{\text{block}}^*| = 100$ Ω m scenario, where $|\rho_{\text{block}}^*| = 50$ Ω m and $|\rho_{\text{block}}^*| = 200$ Ω 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 φ_{app} by changing the
256 sensitivity distribution.

257 So far, we have shown that the polarity of φ_{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 φ_{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 Ω m and φ_{block} varying from -5 to -120 mrad

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

274 The shape of the polarizable block also determines the φ_{app} change under various $|\rho_{\text{block}}^*|$
275 and φ_{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 φ_{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 φ_{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 φ_{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 φ_1) and
299 Z_2^* (with magnitude $|Z_2^*|$ and phase φ_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_{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 φ_{app} also has important,
 332 hitherto unrecognized, implications for the interpretation of spectral IP datasets. The φ_{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 $|\rho_{\text{bgk}}^*|$, φ_{bgk} and frequency-dependent $|\rho_{\text{block}}^*|$
337 and φ_{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 Ω m (from low to high frequency). The φ_{block} ranges from -21 mrad to $-$
342 174 mrad, with the peak occurring at ~ 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 Ω 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 φ_{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 φ_{app} curves when only the resistivity contrast
348 between the target and the background changes between the simulations. For the highest
349 background resistivity, $|\rho_{\text{bgk}}^*| = 55$ Ω m, the φ_{app} spectra are negative and display a negative peak
350 similar to the φ_{block} spectrum. When $|\rho_{\text{bgk}}^*|$ is reduced to 30 Ω m, φ_{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 $|\rho_{\text{bgk}}^*| = 10$ Ω m, all φ_{app} values
353 become positive and a peak of φ_{app} toward more positive values is observed.

354 The differences among the three φ_{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 φ_{app} values are found when $|\rho_{block}^*|/|\rho_{bgk}^*|$ is relatively high,
357 being the case when $|\rho_{bgk}^*| = 10 \Omega \text{ m}$ for all frequencies and when $|\rho_{bgk}^*| = 30 \Omega \text{ m}$ at low frequencies.
358 The $|\rho_{app}^*|$ spectra also differ between the three simulations, exhibiting a frequency dependence
359 consistent with the polarity of φ_{app} . The percentage frequency effect ($\text{PFE} = (|\rho_{app}^*|_L - |\rho_{app}^*|_H) /$
360 $|\rho_{app}^*|_L$, where subscripts H and 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 $|\rho_{app}^*|$ with increasing frequencies) is always observed
363 when φ_{app} is positive. Just as with positive φ_{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 φ_{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. φ_{app} and M_{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 φ_{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 φ_{app} in Figure 8b is similar to the shape of the 0.1-200 Hz segment of the
384 simulated blue φ_{app} curve ($|\rho_{\text{bgk}}^*| = 30 \Omega \text{ m}$) in Figure 7b. The M_{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 φ_{app}
386 polarity at low frequencies. For the Wenner array measurement, a conventional negative φ_{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_{app} measured by the Wenner array is positive (27.8 mV/V),
389 being consistent with the negative φ_{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 φ_{app} recorded in surface four-
393 electrode FDIP measurements may be positive. The polarity of φ_{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 φ_{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 φ_{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 φ_{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 φ_{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 φ_{app} polarity can vary with frequency in FDIP
408 measurements, which results in φ_{app} spectra that are very different from the intrinsic phase
409 spectrum. This finding is confirmed by laboratory sandbox experiments where φ_{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 φ_{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.

425 **References**

- 426 Abdulsamad, F., Florsch, N., Schmutz, M. & Camerlynck, C. (2016) Assessing the high
427 frequency behavior of non-polarizable electrodes for spectral induced polarization
428 measurements. *J. Appl. Geophys.*, **135**, 449–455, Elsevier B.V.
429 doi:10.1016/j.jappgeo.2016.01.001
- 430 Bate, B., Cao, J., Zhang, C. & Hao, N. (2020) Spectral induced polarization study on enzyme
431 induced carbonate precipitations: influences of size and content on stiffness of a fine sand.
432 *Acta Geotech.*, **8**, Springer Berlin Heidelberg. doi:10.1007/s11440-020-01059-8
- 433 Binley, A. (2015) *Tools and Techniques: Electrical Methods. Treatise Geophys. Second Ed.*,
434 Vol. 11, Elsevier B.V. doi:10.1016/B978-0-444-53802-4.00192-5
- 435 Binley, Andrew & Kemna, A. (2005) DC resistivity and induced polarization methods. in
436 *Hydrogeophysics* eds. Rubin, Y. & Hubbard, S.S., pp. 129–156, Springer.
- 437 Blanchy, G., Saneiyan, S., Boyd, J., McLachlan, P. & Binley, A. (2020) ResIPy, an intuitive
438 open source software for complex geoelectrical inversion/modeling. *Comput. Geosci.*, **137**,
439 104423, Elsevier Ltd. doi:10.1016/j.cageo.2020.104423

440 Cole, K. s. & Cole, R.H. (1941) Dispersion and Absorption in Dielectrics I. Alternating Current
441 Characteristics. *J. Chem. Phys.*, **9**, 341–351.

442 Dahlin, T. & Loke, M.H. (2015) Negative apparent chargeability in time-domain induced
443 polarisation data. *J. Appl. Geophys.*, **123**, 322–332, Elsevier B.V.
444 doi:10.1016/j.jappgeo.2015.08.012

445 Flores, A., Kemna, A., Oberdörster, C., Zschornack, L., Leven, C., Dietrich, P. & Weiss, H.
446 (2012) Delineation of subsurface hydrocarbon contamination at a former hydrogenation
447 plant using spectral induced polarization imaging. *J. Contam. Hydrol.*, **136–137**, 131–144,
448 Elsevier B.V. doi:10.1016/j.jconhyd.2012.06.001

449 Flores Orozco, A., Bückler, M., Steiner, M. & Malet, J.P. (2018) Complex-conductivity imaging
450 for the understanding of landslide architecture. *Eng. Geol.*, **243**, 241–252, Elsevier.
451 doi:10.1016/j.enggeo.2018.07.009

452 Garcia-Artigas, R., Himi, M., Revil, A., Urruela, A., Lovera, R., Sendrós, A., Casas, A., *et al.*
453 (2020) Time-domain induced polarization as a tool to image clogging in treatment wetlands.
454 *Sci. Total Environ.*, **724**. doi:10.1016/j.scitotenv.2020.138189

455 Kelter, M., Huisman, J.A., Zimmermann, E. & Vereecken, H. (2018) Field evaluation of
456 broadband spectral electrical imaging for soil and aquifer characterization. *J. Appl.*
457 *Geophys.*, **159**, 484–496, Elsevier B.V. doi:10.1016/j.jappgeo.2018.09.029

458 Kemna, A. (2000) *Tomographic inversion of complex resistivity – Theory and application*, PhD
459 thesis, Ruhr-University of Bochum.

460 Komarov, V. (1960) Bases of application of the induced polarization method for prospecting of
461 ore deposits. *Methodol. Tech. Explor.*, **23**, 7–17.

462 Komarov, V. (1980) *Electrical prospecting with Induced Polarization method*, Leningrad: Nedra
463 Press.

464 Lesmes, P. & Frye, M. (2001) Influence of pore fluid chemistry on the complex conductivity and
465 induced polarization responses of Berea sandstone. *J. Geophys. Res.*, **106**, 4079–4090.

466 Liu, W., Chen, R., Cai, H., Luo, W. & Revil, A. (2017) Correlation analysis for spread-spectrum
467 induced-polarization signal processing in electromagnetically noisy environments.
468 *Geophysics*, **82**, E243–E256. doi:10.1190/GEO2016-0109.1

469 Luo, Y. & Zhang, G. (1998) 2. Forward Theory of Spectral Induced Polarization. in *Theory and*
470 *Application of Spectral Induced Polarization*, pp. 13–90, Society of Exploration
471 Geophysicists. doi:10.1190/1.9781560801856.ch2

472 Mary, B., Saracco, G., Peyras, L., Vennetier, M., Mériaux, P. & Camerlynck, C. (2016) Mapping
473 tree root system in dikes using induced polarization: Focus on the influence of soil water
474 content. *J. Appl. Geophys.*, **135**, 387–396, Elsevier B.V. doi:10.1016/j.jappgeo.2016.05.005

475 Nabighian, M.N. & Elliot, C.L. (1976) NEGATIVE INDUCED-POLARIZATION EFFECTS
476 FROM LAYERED MEDIA. *GEOPHYSICS*, **41**, 1236–1255. doi:10.1190/1.2035915

477 Ntarlagiannis, D., Robinson, J., Soupios, P. & Slater, L. (2016) Field-scale electrical geophysics
478 over an olive oil mill waste deposition site: Evaluating the information content of resistivity
479 versus induced polarization (IP) images for delineating the spatial extent of organic
480 contamination. *J. Appl. Geophys.*, **135**, 418–426, Elsevier B.V.
481 doi:10.1016/j.jappgeo.2016.01.017

482 Pelton, W.H., Wards, S.H., Hallof, P.G., Sill, W.R. & Nelson, P.H. (1978) Mineral
483 Discrimination and Removal of Inductive Coupling with Multifrequency IP. *Geophysics*, **43**,

484 588–609.

485 Saneiyani, S., Ntarlagiannis, D., Ohan, J., Lee, J., Colwell, F. & Burns, S. (2019) Induced
486 polarization as a monitoring tool for in-situ microbial induced carbonate precipitation
487 (MICP) processes. *Ecol. Eng.*, **127**, 36–47, Elsevier. doi:10.1016/j.ecoleng.2018.11.010

488 Saneiyani, S., Ntarlagiannis, D., Werkema, D.D. & Ustra, A. (2018) Geophysical methods for
489 monitoring soil stabilization processes. *J. Appl. Geophys.*, **148**, 234–244, Elsevier B.V.
490 doi:10.1016/j.jappgeo.2017.12.008

491 Seigel, H.O. (1959) MATHEMATICAL FORMULATION AND TYPE CURVES FOR
492 INDUCED POLARIZATION. *GEOPHYSICS*, **24**, 547–565. doi:10.1190/1.1438625

493 Slater, L.D., Choi, J. & Wu, Y. (2005) Electrical properties of iron-sand columns: Implications
494 for induced polarization investigation and performance monitoring of iron-wall barriers.
495 *Geophysics*, **70**, G87. doi:10.1190/1.1990218

496 Slater, L.D. & Lesmes, D. (2002) IP interpretation in environmental investigations. *Geophysics*,
497 **67**, 77–88. doi:10.1190/1.1451353

498 Sumner, J.S. (1976) *Principles of induced polarization for geophysical exploration*, Elsevier.

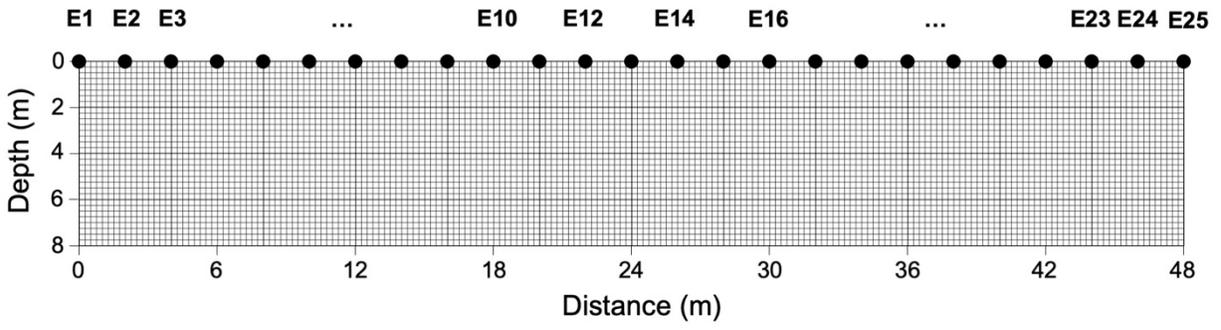
499 Voorhis, G.D. Van, Nelson, P.H. & Drake, T.L. (1973) COMPLEX RESISTIVITY SPECTRA
500 OF PORPHYRY COPPER MINERALIZATION. *GEOPHYSICS*, **38**, 49–60.
501 doi:10.1190/1.1440333

502 Ward, S.H. (1988) The Resistivity and Induced Polarization Methods. *Symp. Appl. Geophys. to*
503 *Eng. Environ. Probl. 1988*, pp. 109–250, Environment and Engineering Geophysical
504 Society. doi:10.4133/1.2921804

505 Wong, J. (1979) An electrochemical model of the induced-polarization phenomenon in

506 disseminated sulfide ores. *GEOPHYSICS*, **44**, 1245–1265. doi:10.1190/1.1441005

507



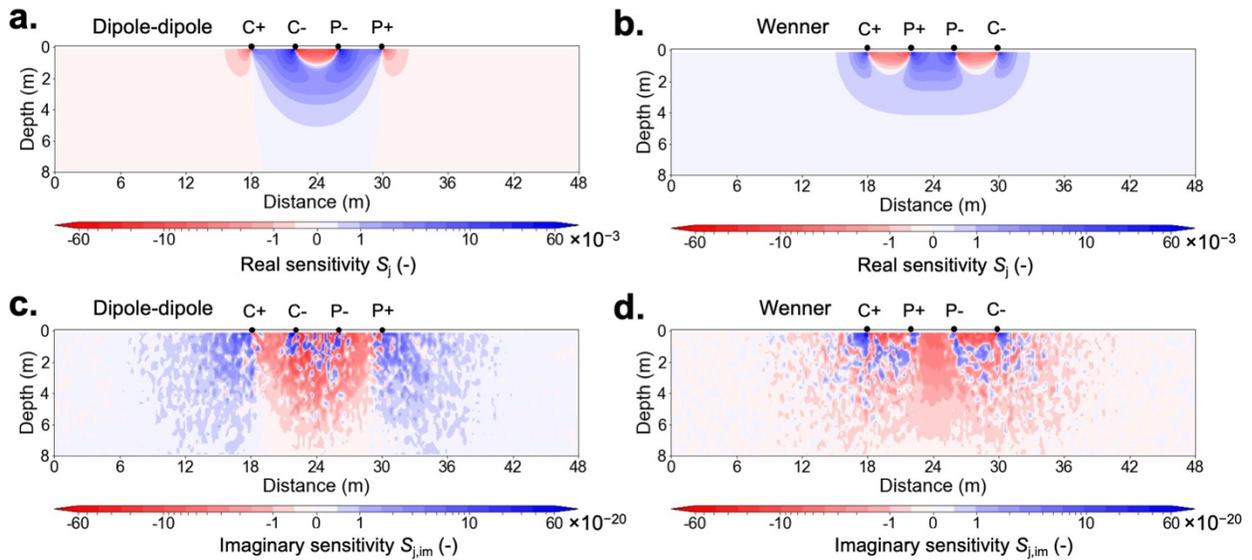
508

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.

511

512



513

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

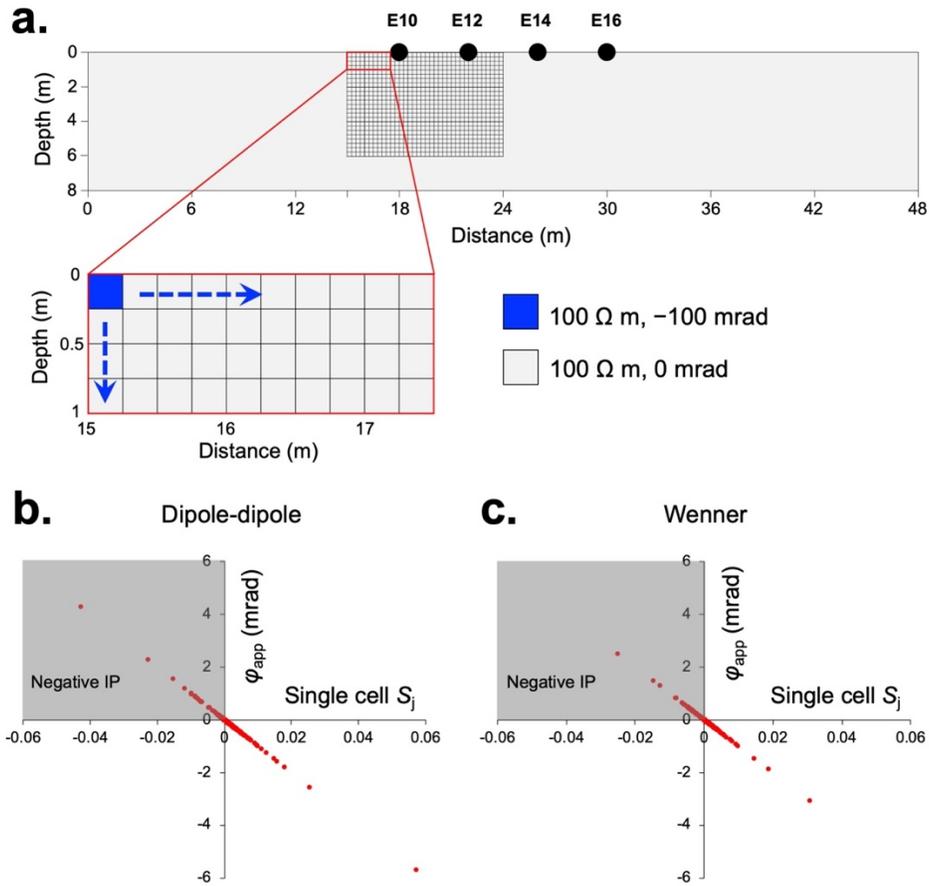
515 E14 and E16 for a $100 \Omega \text{ m}$ and -1 mrad homogeneous half space. (a) Real sensitivity of dipole-

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

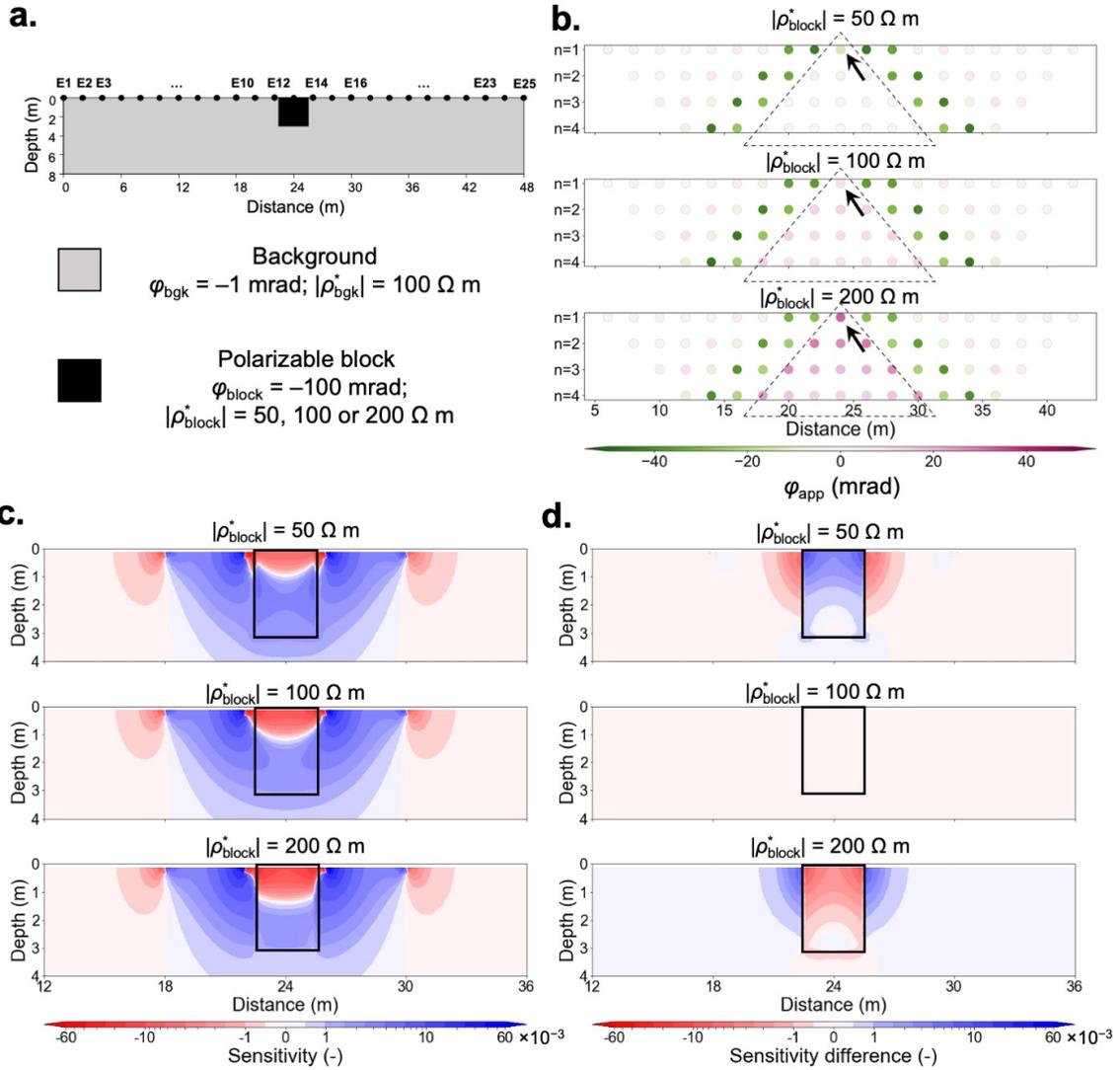
518 sensitivity of Wenner array.

519



521

522 Figure 3. Numerical modeling of the influence of sensitivity polarity on the φ_{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, φ_{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). φ_{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 φ_{app} polarity. (a). Synthetic model settings.

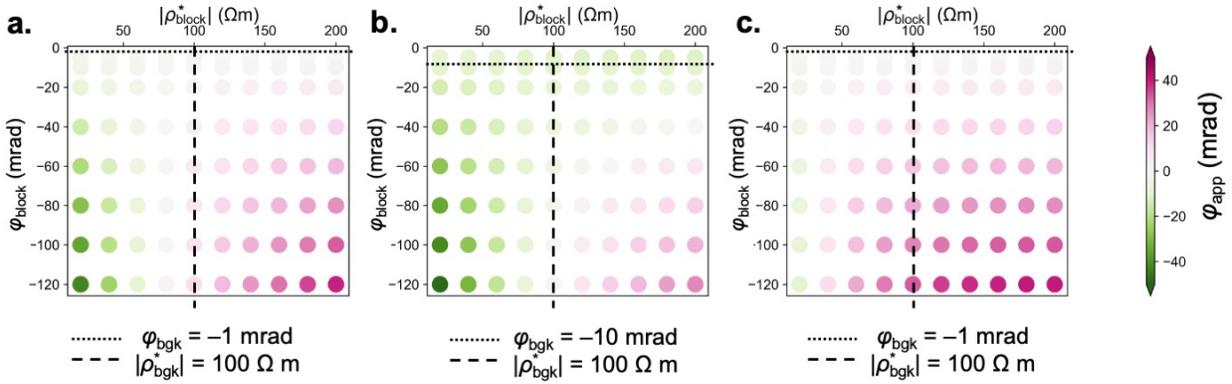
532 (b). Pseudosection of φ_{app} at various values of $|\rho_{\text{block}}^*|$ ($|\rho_{\text{block}}^*| = 100 \Omega \text{ m}$ is the homogeneous

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 $|\rho_{\text{block}}^*|$ and ϕ_{block} on the modeled ϕ_{app} under various conditions. (a) and (b).

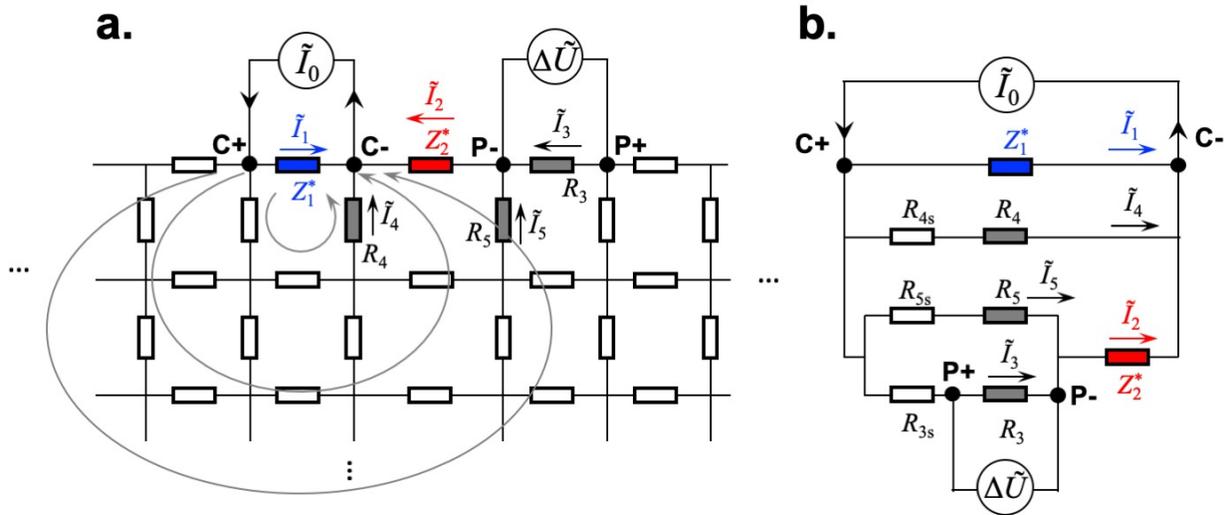
539 ϕ_{app} modeled using a dipole-dipole array (E10=C+, E12=C-, E14=P-, E16=P+) for the synthetic

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

541 lines). (c). ϕ_{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.

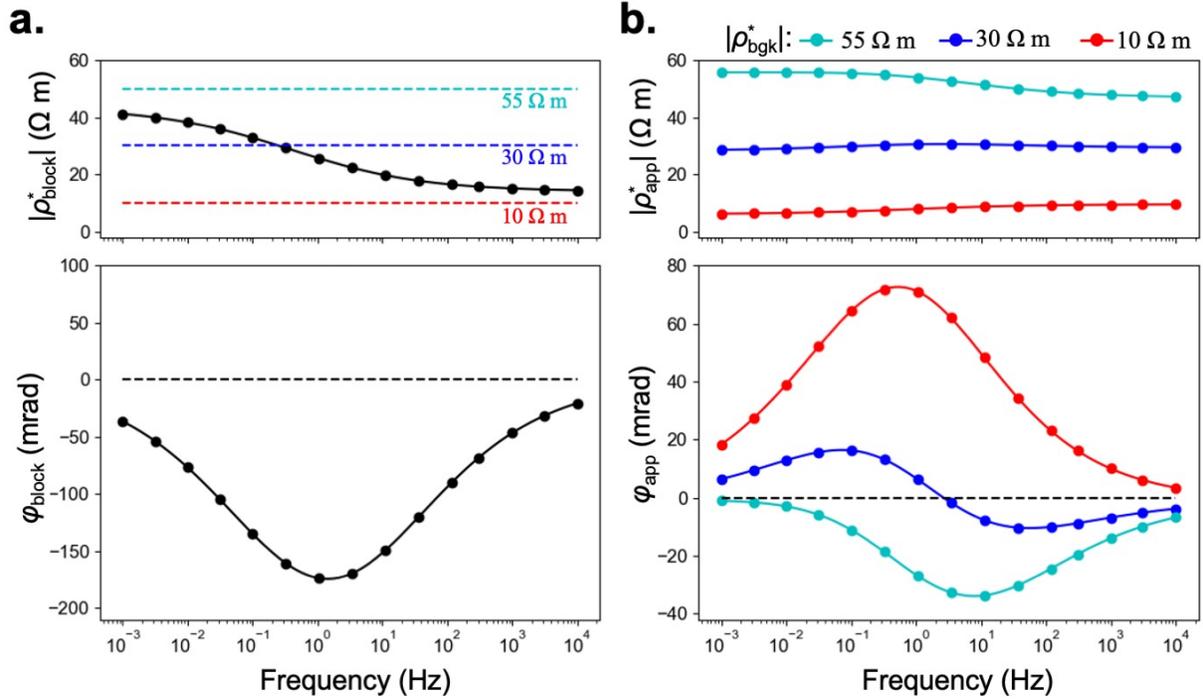
543



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

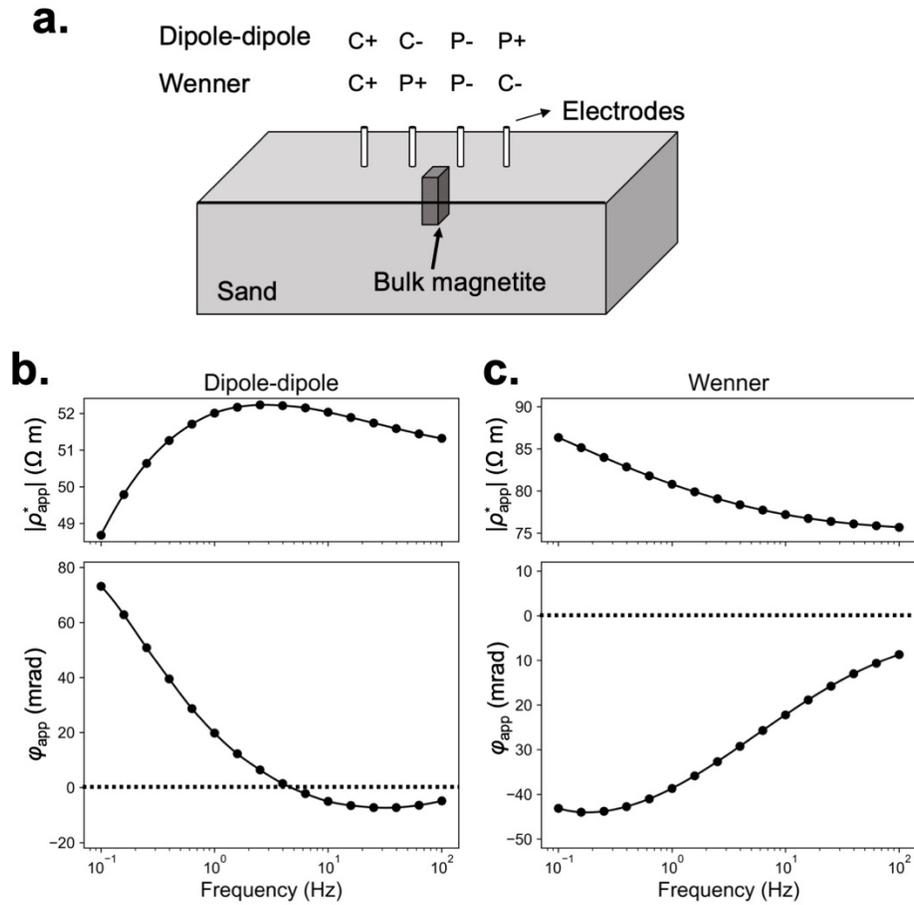
552 using model structure shown in Figure 4a (a). Intrinsic resistivity and phase spectra of the

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 φ_{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 φ_{app} spectra measured by dipole-dipole array; black dashed line represents $\varphi_{app} = 0$ mrad (c)
 560 $|\rho_{app}^*|$ and φ_{app} spectra measured by Wenner array; black dashed line represents $\varphi_{app} = 0$ mrad.