Target reconstruction with a reference point scatterer using phaseless far field patterns

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Abstract

An important property of the phaseless far field patterns with incident plane waves is the translation invariance. Thus it is impossible to reconstruct the location of the underlying scatterers. By adding a reference point scatterer into the model, we design a novel direct sampling method using the phaseless data directly. The reference point technique not only overcomes the translation invariance, but also brings a practical phase retrieval algorithm. Based on this, we propose a hybrid method combining the novel phase retrieval algorithm and the classical direct sampling methods. Numerical examples in two dimensions are presented to demonstrate their effectiveness and robustness.

Keywords: Phaseless data; phase retrieval; stability; sampling method; far field pattern;

AMS subject classifications: 35P25, 45Q05, 78A46, 74B05

1 Introduction

The inverse scattering theory has been a fast-developing area for the past thirty years. Applications of inverse scattering problems occur in many areas such as radar, nondestructive testing, medical imaging, geophysical prospection and remote sensing. Due to their applications, the inverse scattering problems have attracted more and more attention, and significant progress has been made for both the mathematical theories and numerical approaches [3, 4, 11, 15, 16].

In many cases of practical interest, it is very difficult and expensive to obtain the phased data, while the phaseless data is much easier to be achieved. Unfortunately, the reconstructions with phaseless data are highly nonlinear and much more severely ill-posed [1]. By adding a reference point scatterer into the scattering system, we introduce a direct sampling method using the corresponding phaseless far field data directly. Using at most three different scattering strength, we propose a novel phase retrieval scheme. We show that, if the point scatterer is far away from the unknown scatterers, such a phase retrieval scheme is Lipschitz stable with respect to the...
measurement noise. Based on this, we propose certain fast and robust algorithms for scatterer reconstructions by combining classical sampling methods.

We begin with the formulations of the acoustic scattering problems. Let \( k = \omega/c > 0 \) be the wave number of a time harmonic wave where \( \omega > 0 \) and \( c > 0 \) denote the frequency and sound speed, respectively. Let \( D \subset \mathbb{R}^n (n = 2, 3) \) be an open and bounded domain with Lipschitz-boundary \( \partial D \) such that the exterior \( \mathbb{R}^n \setminus \overline{D} \) is connected. Furthermore, let the incident field \( u^i \) be a plane wave of the form

\[
u^i(x) = u^i(x, \hat{\theta}) = e^{ikx \cdot \hat{\theta}}, \quad x \in \mathbb{R}^n, \tag{1.1}\]

where \( \hat{\theta} \in S^{n-1} \) denotes the direction of the incident wave and \( S^{n-1} := \{ x \in \mathbb{R}^n : |x| = 1 \} \) is the unit sphere in \( \mathbb{R}^n \). Then the scattering problem for the inhomogeneous medium is to find the total field \( u = u^i + u^s \) such that

\[
\Delta u + k^2 (1 + q) u = 0 \quad \text{in} \quad \mathbb{R}^n, \tag{1.2}
\]

\[
\lim_{r := |x| \to \infty} r^{n-1} \left( \frac{\partial u^s}{\partial r} - iku^s \right) = 0, \tag{1.3}
\]

where \( q \in L^\infty(\mathbb{R}^n) \) such that the imaginary part \( \Im(q) \geq 0 \) and \( q = 0 \) in \( \mathbb{R}^n \setminus \overline{D} \), the Sommerfeld radiating condition \( \tag{1.3} \) holds uniformly with respect to all directions \( \hat{x} := x/|x| \in S^{n-1} \). If the scatterer \( D \) is impenetrable, the direct scattering is to find the total field \( u = u^i + u^s \) such that

\[
\Delta u + k^2 u = 0 \quad \text{in} \quad \mathbb{R}^n \setminus \overline{D}, \tag{1.4}
\]

\[
\mathcal{B}(u) = 0 \quad \text{on} \quad \partial D, \tag{1.5}
\]

\[
\lim_{r := |x| \to \infty} r^{n-1} \left( \frac{\partial u^s}{\partial r} - iku^s \right) = 0, \tag{1.6}
\]

where \( \mathcal{B} \) denotes one of the following three boundary conditions

(1) \( \mathcal{B}(u) := u \quad \text{on} \quad \partial D; \quad (2) \mathcal{B}(u) := \frac{\partial u}{\partial \nu} \quad \text{on} \quad \partial D; \quad (3) \mathcal{B}(u) := \frac{\partial u}{\partial \nu} + \lambda u \quad \text{on} \quad \partial D
\]

corresponding to the case when the scatterer \( D \) is sound-soft, sound-hard, and of impedance type, respectively. Here, \( \nu \) is the unit outward normal to \( \partial D \) and \( \lambda \in L^\infty(\partial D) \) is the (complex valued) impedance function such that \( \Im(\lambda) \geq 0 \) almost everywhere on \( \partial D \). The well-posedness of the direct scattering problems \( \tag{1.2} - \tag{1.3} \) and \( \tag{1.4} - \tag{1.6} \) have been established and can be found in \[3\, [4],[16],[31],[32],[33],[35].

Every radiating solution of the Helmholtz equation has the following asymptotic behavior at infinity \[16, 29\]

\[
u^s(x, \hat{\theta}) = \frac{e^{ikx \cdot \hat{\theta}}}{\sqrt{8k\pi}} \left( e^{-i\frac{\pi}{2}} \sqrt{\frac{k}{2\pi}} \right) e^{ikr} \left\{ u_D^\infty(\hat{x}, \hat{\theta}) + O \left( \frac{1}{r} \right) \right\} \quad \text{as} \quad r := |x| \to \infty, \tag{1.7}\]

uniformly with respect to all directions \( \hat{x} := x/|x| \in S^{n-1} \). The complex valued function \( u_D^\infty(\hat{x}, \hat{\theta}) \) defined on \( S^{n-1} \) is known as the scattering amplitude or far field pattern with \( \hat{x} \in S^{n-1} \) denoting the observation direction. A wealth of results have been obtained on determining \( D \) from the knowledge of the far field pattern \( u_D^\infty \). We refer to the standard monographs \[3\, [4],[11],[15],[16].

In practice, it is not always the case that the information about the full far field pattern is known,
but instead only its modulus might be given. Thus we are interested in the following inverse problem:

(IP1): **Determine $D$ from the knowledge of phaseless far field pattern $|u^\infty_D|$.**

A well known difficulty for (IP1) is that it is impossible to recover the location of a scatterer only from the phaseless far field pattern due to the translation invariance. Specifically, for the shifted obstacle $D_h := \{ x + h : x \in D \}$, or the shifted refractive index $n_h(x) := n(x - h)$ with a fixed vector $h \in \mathbb{R}^n$, the corresponding far field pattern $u^\infty_{D_h}$ satisfies the equality \[21, 24, 28]:

$$u^\infty_{D_h}(\hat{x}, \hat{\theta}) = e^{ikh \cdot (\hat{\theta} - \hat{x})} u^\infty_D(\hat{x}, \hat{\theta}), \quad \forall \hat{x}, \hat{\theta} \in \mathbb{S}^{n-1},$$  

(1.8)

i.e., the modulus of the far field pattern is invariant under translations. Therefore, only the shape rather than the location may be uniquely determined by the modulus of the far field pattern. In many corresponding uniqueness results with full far field patterns, the proofs heavily rely on the fact that the far field pattern $u^\infty_D$ uniquely determines the scattered wave $u^s$, i.e., Rellich’s lemma. If it is known a priori that the scatterer is a sound-soft ball centered at the origin, uniqueness is established to determine the radius of the ball by a single phaseless far field datum in [30]. Rellich’s lemma is avoided in this special case. By investigating the high frequency asymptotics of the far-field pattern, it was proved in [34] that the shape of a general smooth convex sound-soft obstacle can be determined by the modulus of the far-field pattern associated with one plane wave as the incident field. Up to now, no uniqueness results are available in determining general scatterers with the modulus of the far field pattern generated by one incident plane wave, $|u^\infty_D(\hat{x}, \hat{\theta})|$, $\hat{x}, \hat{\theta} \in \mathbb{S}^{n-1}$, even with the translation invariance taken into account. Initial effort was focused on the shape reconstruction numerically. Indeed, many efficient numerical implementations [1, 6, 14, 21, 12, 13, 25] imply that shape reconstruction from the phaseless far field pattern is possible. However, these methods are mainly iterative schemes based on the integral equations, and thus rely heavily on a priori information about the scatterer and are computationally expensive.

In recent years, considerable effort has been made to avoid using phaseless far field data or to break the translation invariance. Most of the works focus on the case that the point sources are scattered and the phaseless total/scattered fields are measured, where the translation invariance property does not hold \[2, 17, 18, 19, 20\]. The other possible way to break the translation invariance property is to consider superpositions of several plane waves rather than one plane wave as incident fields. The first breakthrough is given in [41], where the authors proved that the translation invariance property of the phaseless far field pattern can be broken if superpositions of two plane waves are used as the incident fields for all wave numbers in a finite interval. Further, a recursive Newton-type iteration algorithm in frequencies is also developed to numerically recover both the location and the shape of the scatterer $D$ simultaneously from multi-frequency phaseless far-field data. In a recent work [39], it was proved, under certain conditions on the scatterer, that the scatterer can be uniquely determined by the phaseless far-field patterns generated by infinitely many sets of superpositions of two plane waves with different directions at a fixed frequency. A fast imaging algorithm was also developed in [43] to numerically recover the scattering obstacles from the phaseless far-field data at a fixed frequency associated with infinitely many sets of superpositions of two plane waves with different directions. Recently, the a priori assumption on the scatterers introduced in [39] was removed in [40] by adding a known reference ball to the scattering system in conjunction with a simple technique based on Rellich’s lemma and Green’s representation formula for the scattering solutions. In addition, by adding a reference ball to the
scattering system uniqueness results were obtained in [44] for inverse scattering with phaseless far-field data corresponding to superpositions of an plane wave and point sources as the incident fields. Accordingly, a nonlinear integral equation method was also developed in [6] to reconstruct both the location and the shape of a scattering obstacle from such phaseless far-field data in two dimensions.

The reference ball technique dates back to [26] (see also [38]), where such a technique is used to avoid eigenvalues and choose a cut-off value for the linear sampling method. To enhance the interaction between the reference ball and the unknown target, the ball chosen in [26] cannot be too small or too far away from the unknown target. In this paper, we still consider scattering of a single plane wave given in (1.1), but add a known reference point scatterer into the scattering system, which is different from [40, 44, 6]. Also, different from [26], we expect the interaction between the reference point scatterer and the unknown target is as weak as possible and thus choose the point far away from the unknown target. Actually, numerical examples show that the interaction is very weak, even though the point scatterer is close to the target. The reference point technique not only breaks the translation invariance of the phaseless far-field patterns, but also gives a novel phase retrieval method. This makes it possible to determine the unknown target by combining the classical scatterer reconstruction methods with the phased data. We also want to strength that our methods proposed in the next sections are independent of any a priori geometrical or physical information on the unknown target.

The remaining part of the work is organized as follows. In Section 2, we first recall the scattering of plane waves by a point scatterer, and then consider the scattering of plane waves by a combination of the underlying scatterer D and a given point scatterer located at \( z_0 \in \mathbb{R}^n \setminus D \). Based on this, we propose a new inverse problem to determine the scatterer D with the corresponding phaseless far field data. A simple, fast and stable phase retrieval technique is then proposed. Section 3 is devoted to some direct sampling methods for scatterer reconstructions, which make no explicit assumptions on boundary conditions or topological properties of the scatterer D. These algorithms are then verified in Section 4 by extensive examples in two dimensions.

2 Preliminaries

2.1 Scattering of plane waves by a point scatterer

First, we recall the scattering of plane waves by a point scatterer [7]. We consider a point scatterer located at \( z_0 \in \mathbb{R}^n \) in the homogeneous space \( \mathbb{R}^n \). An incident plane wave \( u^i \) of the form (1.1) is scattered by the target at \( z_0 \). Recall that the fundamental solution \( \Phi(x, y), x, y \in \mathbb{R}^n, x \neq y, \) of the Helmholtz equation is given by

\[
\Phi(x, y) := \begin{cases} 
\frac{ik}{4\pi} h_0^{(1)}(k|x - y|) = \frac{e^{ik|x-y|}}{4\pi|x-y|}, & n = 3, \\
\frac{i}{4} H_0^{(1)}(k|x - y|), & n = 2,
\end{cases}
\]  

where \( h_0^{(1)} \) and \( H_0^{(1)} \) are, respectively, spherical Hankel function and Hankel function of the first kind and order zero. Then the scattered field \( u^s_{z_0} \) is given by

\[
u^s_{z_0}(x, \hat{\theta}, \tau) = \tau u^i(z_0, \hat{\theta})\Phi(x, z_0).
\]
Here, $\tau \in \mathbb{C}$ is the scattering strength of the target. From the asymptotic behavior of $\Phi(x, y)$ we deduce that the corresponding far field pattern is given by

$$u_{20}^{\infty}(\hat{x}, \hat{\theta}, \tau) = \tau u^{i}(z_0, \hat{\theta})e^{-ikz_0 \cdot \hat{x}} = \tau e^{ikz_0 \cdot (\hat{\theta} - \hat{x})}, \quad \hat{x}, \hat{\theta} \in \mathbb{S}^{n-1}, \tau \in \mathbb{C}. \quad (2.3)$$

Furthermore, by the representation (2.3) it is easy to deduce that for $h \in \mathbb{R}^n$,

$$u_{20}^{\infty}(\hat{x}, \hat{\theta}, \tau) = e^{ikh \cdot (\hat{\theta} - \hat{x})}u_{20}^{\infty}(\hat{x}, \hat{\theta}, \tau), \quad \hat{x}, \hat{\theta} \in \mathbb{S}^{n-1}, \tau \in \mathbb{C}. \quad (2.4)$$

Then the translation relation for the phaseless data $|u_{20}^{\infty}|$ also holds, i.e., given $h \in \mathbb{R}^n$ we have

$$|u_{20}^{\infty+h}(\hat{x}, \hat{\theta}, \tau)| = |u_{20}^{\infty}(\hat{x}, \hat{\theta}, \tau)|, \quad \hat{x}, \hat{\theta} \in \mathbb{S}^{n-1}, \tau \in \mathbb{C}.$$

### 2.2 New scattering system with a given point scatterer

Let $z_0 \in \mathbb{R}^n \setminus \overline{D}$ be a fixed point outside $D$. By adding a point scatterer into the underlying scattering system, we consider the new scattering system by $D \cup \{z_0\}$. In the sequel, for an incident plane wave $u^i(x) = u^i(x, \hat{\theta}) = e^{ikx \cdot \hat{\theta}}$ we will indicate the dependence of the scattered field and its far field pattern on the incident direction $\hat{\theta}$ and the scattering strength $\tau$ by writing, respectively, $u_{D\cup\{z_0\}}^{\infty}(x, \hat{\theta}, \tau)$ and $u_{D\cup\{z_0\}}^{\infty}(x, \hat{\theta}, \tau)$. Since the point scatterer is given in advance, the inverse problem considered is modified as follows.

**IP2:** Determine $D$ from the knowledge of phaseless far field pattern $|u_{D\cup\{z_0\}}^{\infty}|$.

Note that if $\tau = 0$ then (IP2) is reduce to (IP1). Following the arguments given in [21, 24, 28], it is easy to check that the translation invariance property of the phaseless far field data $|u_{D\cup\{z_0\}}^{\infty}|$ also holds, i.e., for any $h \in \mathbb{R}^n$, we have

$$|u_{D\cup\{z_0\}}^{\infty+h}(\hat{x}, \hat{\theta}, \tau)| = |u_{D\cup\{z_0\}}^{\infty}(\hat{x}, \hat{\theta}, \tau)|, \quad \forall \hat{x}, \hat{\theta} \in \mathbb{S}^{n-1}, \tau \in \mathbb{C}.$$

However, the reference point $z_0$ is given in advance, and this makes it possible to determine $D$ from the phaseless data $|u_{D\cup\{z_0\}}^{\infty}|$.

It is well known that the nature of the scatterer $D$ can be uniquely determined by the phased far field patterns $u_{D}^{\infty}$ [4]. In our subsequent analysis, we try to retrieve these phased data from the phaseless far field patterns $|u_{D\cup\{z_0\}}^{\infty}|$. To do so, we expect that the interaction between the unknown target $D$ and the given point scatterer is as weak as possible. Such a fact can be achieved by choosing the reference point scatterer far away from the target. Actually, this is verified by the following Theorem (2.1).

For any $\varphi \in H^{-1/2}(\partial D)$ and $\psi \in H^{1/2}(\partial D)$, the single-layer potential is defined by

$$(S\varphi)(x) := \int_{\partial D} \Phi(x, y)\varphi(y)ds(y), \quad x \in \mathbb{R}^n \setminus \partial D,$$

and the double-layer potential is defined by

$$(K\psi)(x) := \int_{\partial D} \frac{\partial \Phi(x, y)}{\partial \nu(y)}\psi(y)ds(y), \quad x \in \mathbb{R}^n \setminus \partial D,$$
respectively. It is shown in [35] that the potentials $\mathcal{S} : H^{-1/2}(\partial D) \rightarrow \mathcal{H}_{1 \text{loc}}^{1}(\mathbb{R}^{n} \setminus \partial D)$, $\mathcal{K} : H^{1/2}(\partial D) \rightarrow \mathcal{H}_{1 \text{loc}}^{1}(\mathbb{R}^{n} \setminus \overline{D})$ are well defined. We also define the restriction of $\mathcal{S}$ and $\mathcal{K}$ to the boundary $\partial D$ by

\begin{align}
(S\varphi)(x) := \int_{\partial D} \Phi(x,y)\varphi(y)ds(y), \quad x \in \partial D,
\end{align}

\begin{align}
(K\psi)(x) := \int_{\partial D} \frac{\partial \Phi(x,y)}{\partial n(y)}\psi(y)ds(y), \quad x \in \partial D.
\end{align}

We refer to [35] for the properties of the boundary operators $\mathcal{S} : H^{-1/2}(\partial D) \rightarrow H^{1/2}(\partial D)$ and $\mathcal{K} : H^{1/2}(\partial D) \rightarrow H^{1/2}(\partial D)$.

**Theorem 2.1.** Let $z_0$ be a point outside $D$ such that the distance $\rho := \text{dist}(z_0, D)$ is large enough. Then we have

\begin{align}
u_{D\cup\{z_0\}}^\infty(x, \hat{\theta}, \tau) = u_D^\infty(x, \hat{\theta}) + u_{z_0}^\infty(x, \hat{\theta}, \tau) + O\left(\rho^{\frac{1-n}{2}}\right), \quad \forall \hat{x}, \hat{\theta} \in S^{n-1}, \tau \in \mathbb{C}.
\end{align}

**Proof.** The integral equation method is used in our subsequent analysis. For simplicity, we only prove the case when the scatterer $D$ is sound-soft. The other cases can be dealt with similarly.

We seek a solution in the form

\begin{align}
u_{D\cup\{z_0\}}^s(x, \hat{\theta}, \tau) = (K - i\eta S)\psi_{D\cup\{z_0\}} + u_{z_0}^s(x, \hat{\theta}, \tau), \quad x \in \mathbb{R}^n \setminus D \cup \{z_0\},
\end{align}

with a density $\psi_{D\cup\{z_0\}} \in H^{1/2}(\partial D)$ and a coupling parameter $\eta > 0$. Then from the jump relation of the double layer potential we see that the representation $u_{D\cup\{z_0\}}^s$ given in (2.8) solves the exterior Dirichlet boundary problem provided the density is a solution of the integral equation

\begin{align}
\left(I/2 + K - i\eta S\right)\psi_{D\cup\{z_0\}} = -(u^i + u_{z_0}^s) \quad \text{on} \quad \partial D.
\end{align}

Note that $I/2 + K - i\eta S$ is bijective and the inverse $(I/2 + K - i\eta S)^{-1} : H^{1/2}(\partial D) \rightarrow H^{1/2}(\partial D)$ is bounded [34, 35]. Therefore

\begin{align}
\psi_{D\cup\{z_0\}} = -(I/2 + K - i\eta S)^{-1}(u^i + u_{z_0}^s) := \psi_D - (I/2 + K - i\eta S)^{-1}u_{z_0}^s.
\end{align}

Note that $u_{z_0}^s := (K - i\eta S)\psi_D$ is the radiating solution to the original scattering system with a sound-soft obstacle $D$. A straightforward calculation shows that the fundamental solution $\Phi$ satisfies Sommerfeld’s finiteness condition

\begin{align}
\Phi(x, y) = O\left(||x - y||^{\frac{1-n}{2}}\right), \quad |x - y| \rightarrow \infty.
\end{align}

Inserting this into the representation (2.2) of $u_{z_0}^s$ we see that

\begin{align}
u_{z_0}^s|_{\partial D} = O\left(\rho^{\frac{1-n}{2}}\right), \quad \rho \rightarrow \infty.
\end{align}

This implies that

\begin{align}
\psi_{D\cup\{z_0\}} = \psi_D + O\left(\rho^{\frac{1-n}{2}}\right), \quad \rho \rightarrow \infty.
\end{align}

Inserting this into (2.8), we find that

\begin{align}
u_{D\cup\{z_0\}}^s(x, \hat{\theta}, \tau) = u_D^s(x, \hat{\theta}) + u_{z_0}^s(x, \hat{\theta}, \tau) + O\left(\rho^{\frac{1-n}{2}}\right), \quad x \in \mathbb{R}^n \setminus D \cup \{z_0\}, \quad \rho \rightarrow \infty.
\end{align}

Then (2.7) follows by letting $|x| \rightarrow \infty$. \qed
2.3 Phase retrieval

The following lemma may have its own interest.

**Lemma 2.2.** Let \( z_j := x_j + iy_j, j = 1, 2, 3 \), be three different complex numbers such that they are not collinear. Then the complex number \( z \in \mathbb{C} \) is uniquely determined by the distances \( r_j = |z - z_j|, j = 1, 2, 3 \).

**Proof.** Denote by \( Z_j = (x_j, y_j) \) the point in the plane corresponding to the given complex number \( z_j, j = 1, 2, 3 \). Define \( Z = (x, y) \) to be the point corresponding to the unknown complex number \( z \).

Then \( Z \) locates on the spheres \( \partial B_{r_j}(Z_j) \) centered at \( Z_j \) with radius \( r_j, j = 1, 2, 3 \). Note that there are at most two points \( Z_A \) and \( Z_B \) located simultaneously on the two spheres \( \partial B_{r_j}(Z_j), j = 1, 2 \), i.e.

\[
|Z_A - Z_j| = |Z_B - Z_j| = r_j, j = 1, 2. \tag{2.9}
\]

If \( Z_A = Z_B \), then we just take \( Z = Z_A \); otherwise, we claim that only one of the two points \( Z_A \) and \( Z_B \) is the point \( Z \) pursued. On the contrary, we have

\[
|Z_A - Z_3| = |Z_B - Z_3| = r_3.
\]

This, together with (2.9), implies that the three points \( Z_j, j = 1, 2, 3 \), are located on the perpendicular bisector of the line segment \( Z_AZ_B \). This contradicts to the assumption that \( z_j, j = 1, 2, 3 \), are not collinear. The proof is complete. \( \square \)

Actually, Lemma 2.2 provides a novel phase retrieval technique which can be implemented easily. Using the same notations in Lemma 2.2, we have the following phase retrieval scheme.

![Figure 1: Sketch map for phase retrieval scheme.](image)

**Phase Retrieval Scheme.** (Numerical simulation for Lemma 2.2)

1. Collect the distances \( r_j := |z - z_j| \) with given complex numbers \( z_j, j = 1, 2, 3 \). If \( r_j = 0 \) for some \( j \in \{1, 2, 3\} \), then \( Z = Z_j \); otherwise, go to the next step.
(2) Look for the point $M = (x_M, y_M)$. As shown in Figure 1, $M$ is the intersection of circle centered at $Z_2$ with radius $r_2$ and the ray $Z_2Z_1$ with initial point $Z_2$. Denote by $d_{1,2} := |z_1 - z_2|$ the distance between $Z_1$ and $Z_2$. Then

$$x_M = \frac{r_2}{d_{1,2}} x_1 + \frac{d_{1,2} - r_2}{d_{1,2}} x_2, \quad y_M = \frac{r_2}{d_{1,2}} y_1 + \frac{d_{1,2} - r_2}{d_{1,2}} y_2. \quad (2.10)$$

(3) Look for the points $Z_A = (x_A, y_A)$ and $Z_B = (x_B, y_B)$. Note that $Z_A$ and $Z_B$ are just two rotations of $M$ around the point $Z_2$. Let $\alpha \in [0, \pi]$ be the angle between the rays $Z_2Z_1$ and $Z_2Z_A$. Then, by the law of cosine we have

$$\cos \alpha = \frac{r_1^2 - r_2^2 - d_{1,2}^2}{2r_2d_{1,2}}. \quad (2.11)$$

Noting that $\alpha \in [0, \pi]$ and $\sin^2 \alpha + \cos^2 \alpha = 1$, we deduce that $\sin \alpha = \sqrt{1 - \cos^2 \alpha}$. Then

$$x_A = x_2 + \Re\{[(x_M - x_2) + i(y_M - y_2)]e^{-i\alpha}\}, \quad (2.12)$$
$$y_A = y_2 + \Im\{[(x_M - x_2) + i(y_M - y_2)]e^{-i\alpha}\}, \quad (2.13)$$
$$x_B = x_2 + \Re\{[(x_M - x_2) + i(y_M - y_2)]e^{i\alpha}\}, \quad (2.14)$$
$$y_B = y_2 + \Im\{[(x_M - x_2) + i(y_M - y_2)]e^{i\alpha}\}. \quad (2.15)$$

(4) Determine the point $Z$. $Z = Z_A$ if the distance $|Z_AZ_3| = r_3$; otherwise, $Z = Z_B$.

**Remark 2.3.** Actually, the above scheme provides a stable phase retrieval algorithm. Indeed, let $\epsilon > 0$ and assume that

$$|r_j^\epsilon - r_j| \leq \epsilon, \quad j = 1, 2, 3.$$  

Here, and throughout the paper, we use the subscript $\epsilon$ to denote the polluted data. From (2.10), we deduce that

$$|x_M^\epsilon - x_M| = \frac{|x_1 - x_2|}{d_{1,2}} |r_2^\epsilon - r_2| \leq \epsilon \quad \text{and} \quad |y_M^\epsilon - y_M| = \frac{|y_1 - y_2|}{d_{1,2}} |r_2^\epsilon - r_2| \leq \epsilon. \quad (2.16)$$

Similarly, (2.11) implies the existence of a constant $c_1 > 0$ depending on $Z_j, j = 1, 2, 3$, such that

$$|e^{i\alpha^\epsilon} - e^{i\alpha}| \leq c_1 \epsilon.$$  

Combing this with (2.16) and (2.12)-(2.15), we find that there exists a constant $c_2 > 0$ depending on $Z_j, j = 1, 2, 3$, such that

$$|x_{ii}^\epsilon - x_{ii}| \leq c_2 \epsilon \quad \text{and} \quad |y_{ii}^\epsilon - y_{ii}| \leq c_2 \epsilon, \quad ii = A, B.$$  

Therefore, we have

$$|Z^\epsilon - Z| \leq \sqrt{2}c_2 \epsilon.$$  

This implies that our phase retrieval scheme is Lipschitz stable with respect to the measurement noise level $\epsilon$. 

8
Using the phase retrieval scheme, we wish to approximately reconstruct \( u^\infty_D \) from the knowledge of the perturbed phaseless data \( u_{D\cup\{z_0\}}^{\infty,\varepsilon} \) with a known error level

\[
\left| u_{D\cup\{z_0\}}^{\infty,\varepsilon} \right| - \left| u_{D\cup\{z_0\}}^{\infty} \right| \leq \varepsilon.
\]

(2.17)

**Theorem 2.4.** Let \( \tau_j \in \mathbb{C}, j = 1, 2, 3 \) be three scattering strengths with different principle arguments and \( \rho \) be the distance between the point \( z_0 \) and the unknown target \( D \). Under the measurement error estimate (2.17), we have

\[
\left| u_{D}^{\infty,\varepsilon} - u^{\infty}_D \right| \leq c_3 \varepsilon + O \left( \rho^{\frac{1-n}{2}} \right),
\]

for some constant \( c_3 > 0 \) depending only on \( \tau_j, j = 1, 2, 3 \).

**Proof.** Let

\[
Z_j := u_{z_0}^{\infty} (\hat{x}, \hat{\theta}, \tau_j) = \tau_j e^{ikz_0(\hat{\theta} - \hat{x})}, \quad j = 1, 2, 3.
\]

Then the assumption on the strengths implies that the three points \( Z_j, j = 1, 2, 3 \) are not collinear. Define \( r_j^{\varepsilon} := \left| u_{D\cup\{z_0\}}^{\infty,\varepsilon} \right| \). Using Theorem 2.1, we have

\[
r_j^{\varepsilon} = \left| u_{D}^{\infty,\varepsilon} + u_{z_0}^{\infty} + O \left( \rho^{\frac{1-n}{2}} \right) \right|
\]

where \( \rho \) is the distance between the point \( z_0 \) and the unknown target \( D \). Then, following the arguments in Remark 2.3, we have

\[
\left| u_{D}^{\infty,\varepsilon} - u^{\infty}_D + O \left( \rho^{\frac{1-n}{2}} \right) \right| \leq c_3 \varepsilon,
\]

for some constant \( c_3 > 0 \) depending only on \( \tau_j, j = 1, 2, 3 \). The stability estimate (2.18) now follows by using the triangle inequality.

Finally, we want to remark that Theorem 2.4 implies that our phase retrieval scheme provides a stable method for the phase reconstruction. This will also be verified by the numerical example in Section 4.

### 2.4 Stability estimates for the inverse problems

Stability of recovery of the scatterer is crucial for numerical algorithms. It has been proved in several papers that the inverse problems with phased far field patterns are ill-posed. Stability was first considered by Isakov [9, 10] for the determination of a sound-soft obstacle. We refer to Potthast [36], Cristo and Rondi [5] for the extension to both the sound-soft and sound-hard obstacles. Hähner and Hohage [8] considered the stability estimate for the inhomogeneous medium case.

In this subsection, let \( D_1, D_2 \) be two scatterers, \( \rho \) be the distance between the point \( z_0 \) and the unknown obstacles \( D_1 \cup D_2 \) and \( \tau_j \in \mathbb{C}, j = 1, 2, 3 \), be three scattering strengths with different principal arguments. Let \( |u_{D_1\cup\{z_0\}}^{\infty}|, |u_{D_2\cup\{z_0\}}^{\infty}| \) be the corresponding phaseless far field patterns. Following the same arguments in Theorem 2.4, we have the following stability estimate.
Theorem 2.5. If
\[
\left| u_{D_1 \cup \{z_0\}}^\infty (\hat{x}, \hat{\theta}, \tau) - u_{D_2 \cup \{z_0\}}^\infty (\hat{x}, \hat{\theta}, \tau) \right| < \epsilon, \quad \forall \hat{x}, \hat{\theta} \in S^{n-1}, \tau \in \{\tau_1, \tau_2, \tau_3\},
\]
then for sufficiently large \( \rho \) we have
\[
\left| u_{D_1}^\infty (\hat{x}, \hat{\theta}) - u_{D_2}^\infty (\hat{x}, \hat{\theta}) \right| < c_3 \epsilon + O\left(\frac{1}{\rho^{1-n}}\right), \quad \forall \hat{x}, \hat{\theta} \in S^{n-1}.
\]
where \( c_3 > 0 \) is a constant depending only on \( \tau_j, j = 1, 2, 3 \).

Combining Theorem 2.5 and Theorem 1 in [9], Theorem 15 in [36] and Theorem 1.2 in [8], respectively, we immediately obtain the following stability estimate for scatterer reconstruction with phaseless far field patterns.

Theorem 2.6. Denote by \( \delta = \delta(\epsilon, \rho) \) the right-hand side of (2.20). The following scatterer reconstruction stabilities hold.

(1) Assume that \( D_m = \{ x \in \mathbb{R}^3 : |x| < r_m(\theta) \} \) is star-shaped with
\[
\| r_m \|_{C^{2,\alpha}(S^2)} < 1/R_1, \quad 1/R_1 < r_m < R_0, \quad m = 1, 2.
\]
For \( k < \pi/R_0 \), if (2.19) holds for any fixed \( \hat{\theta} \in S^2 \), then the Hausdorff distance between \( D_1 \) and \( D_2 \) satisfies that
\[
\text{dist}(D_1, D_2) < C \left( \ln \left( -\ln(\delta) \right) \right)^{1/C},
\]
where \( C \) is a constant depending only on \( R_1 \).

(2) Assume that \( D_m \subset B_R(0), m = 1, 2 \), are sound-soft or sound-hard obstacle with \( C^2 \) boundary satisfying the exterior cone condition with angle \( \beta \). If (2.19) holds for all \( \hat{x}, \hat{\theta} \in S^{n-1} \), then the Hausdorff distance between the convex hulls \( \mathcal{H}(D_1) \) and \( \mathcal{H}(D_2) \) satisfies the estimate
\[
\text{dist}(\mathcal{H}(D_1), \mathcal{H}(D_2)) \leq \frac{C}{|\ln(\delta)|^\alpha},
\]
where the constants \( C > 0 \) and \( 0 < \alpha < 1 \) uniformly depend only on \( R \) and \( \beta \).

(3) Assume that \( q_m \in H^s(\mathbb{R}^3) \) for some fixed \( s > 3/2 \), \( \text{supp}(q_m) \subset B_1 \) and \( \| q_m \|_{H^s} < C_q \) for some fixed constant \( C_q > 0 \), \( m = 1, 2 \). For any fixed constant \( \epsilon_0 \in (0, (2s-3)/(2s+3)) \), the maximum norm of \( q_1 - q_2 \) can be estimated as
\[
\| q_1 - q_2 \|_{\infty} \leq C \left( -\tilde{\ln}(16\pi^2\delta) \right)^{\epsilon_0-(2s-3)/(2s+3)},
\]
where \( C \) depends only on \( C_q, \epsilon_0 \), and \( \tilde{\ln}(t) := \ln(t) \) for \( t < 1/e \) and \( \tilde{\ln}(t) := -1 \) otherwise.
3 Direct sampling methods

In this section, we investigate the numerical method for reconstruction of $D$ by using phaseless far field data $|u^\infty_{D \cup \{z_0\}}|$. We will focus on designing a direct sampling method which do not need any a priori information on the geometry and physical properties of the obstacle. Roughly speaking, a direct sampling method chooses an appropriate indicator function $I(z)$, $z \in \mathbb{R}^n$, such that its value has an obvious change across the boundary of the scatterers.

We first introduce two auxiliary functions

$$G(z, \hat{\theta}) := \int_{S^{n-1}} u^\infty_D(\hat{x}, \hat{\theta}) e^{ik\hat{x} \cdot z} ds(\hat{x})$$

$$A(z) := \int_{S^{n-1}} G(z, \hat{\theta}) e^{-ik\hat{\theta} \cdot z} ds(\hat{\theta}), \quad z \in \mathbb{R}^n. \quad (3.1)$$

By the well-known Riemann-Lebesgue Lemma, both $G$ and $A$ tend to 0 as $|z| \to \infty$. For the scattering problems $[1.2]-(1.3)$ and $[1.4]-(1.6)$, it is well known that the far field pattern $u^\infty_D$ has the following form (cf. [16])

$$u^\infty_D(\hat{x}, \hat{\theta}) = \int_{\partial D} \left\{ u^s(y, \hat{\theta}) \frac{\partial e^{-ik\hat{x} \cdot y}}{\partial \nu(y)} - \frac{\partial u^s(y, \hat{\theta})}{\partial \nu} e^{-ik\hat{x} \cdot y} \right\} ds(y), \quad (\hat{x}, \hat{\theta}) \in S^{n-1}. \quad (3.2)$$

Inserting this into (3.1), integrating by parts and using the well-known Funk-Hecke formula [4, 29], we deduce that

$$G(z, \hat{\theta})$$

$$= \int_{S^{n-1}} \int_{\partial D} \left\{ u^s(y, \hat{\theta}) \frac{\partial e^{-ik\hat{x} \cdot (y-z)}}{\partial \nu(y)} - \frac{\partial u^s(y, \hat{\theta})}{\partial \nu} e^{-ik\hat{x} \cdot (y-z)} \right\} ds(y) ds(\hat{x})$$

$$= \int_{\partial D} \left\{ -iku^s(y, \hat{\theta}) \nu(y) \cdot \int_{S^{n-1}} \hat{x} e^{-ik\hat{x} \cdot (y-z)} ds(\hat{x}) - \frac{\partial u^s(y, \hat{\theta})}{\partial \nu} \int_{S^{n-1}} e^{-ik\hat{x} \cdot (y-z)} ds(\hat{x}) \right\} ds(y)$$

$$= \int_{\partial D} \left\{ -ik \mu_1 u^s(y, \hat{\theta}) \nu(y) \cdot \frac{y-z}{|y-z|} f_1(k|y-z|) - \mu_0 \frac{\partial u^s(y, \hat{\theta})}{\partial \nu} f_0(k|y-z|) \right\} ds(y), \quad (3.2)$$

where

$$\mu_\alpha = \begin{cases} 2\pi i^{-\alpha}, & n = 2, \\ 4\pi i^{-\alpha}, & n = 3 \end{cases} \quad \text{and} \quad f_\alpha(t) = \begin{cases} J_\alpha(t), & n = 2, \\ j_\alpha(t), & n = 3 \end{cases}$$

with $J_\alpha$ and $j_\alpha$ being the Bessel functions and spherical Bessel functions of order $\alpha$, respectively. This implies that $G$ is a superposition of the Bessel functions $f_0$ and $f_1$. We thus expect that $G$ (and therefore $A$) decays like Bessel functions as the sampling points away from the boundary of the scatterer.

Then one may look for the scatterers by using the following indicators [27, 29, 37] with phased far field patterns,

$$I_2(z) = |A(z)| \quad \text{and} \quad I_3(z, \hat{\theta}) = |G(z, \hat{\theta})|, \quad (3.3)$$

where $A$ and $G$ are given in (3.1). In [29], it has been showed that the indicator $I_2$ has a positive lower bound for sampling points inside the scatterer, and decays like Bessel functions as the sampling points away from the boundary. If the size of the scatterer $D$ is small enough
(compared with the wavelength), \( I_3 \) takes its local maximum at the location of the scatterer \(^{27, 37}.\)

Consider now the case of phaseless far field measurements. Using (2.3) and (2.7), we have

\[
\mathcal{F}(\hat{x}, \hat{\theta}, z_0, \tau) := |u^\infty_D(\hat{x}, \hat{\theta}, \tau)|^2 - |u^\infty_D(\hat{x}, \hat{\theta})|^2 - |\tau|^2
\]

\[
= |u^\infty_D(\hat{x}, \hat{\theta}) + u^\infty_D(\hat{x}, \hat{\theta}, \tau) + O\left(\rho^{\frac{1-n}{2}}\right)|^2 - |u^\infty_D(\hat{x}, \hat{\theta})|^2 - |\tau|^2
\]

\[
= u^\infty_D(\hat{x}, \hat{\theta}) e^{-i k \xi \cdot \hat{x}} + u^\infty_D(\hat{x}, \hat{\theta}) e^{i k \xi \cdot \hat{x}} + O\left(\rho^{\frac{1-n}{2}}\right), \quad (\hat{x}, \hat{\theta}) \in \mathbb{S}^{n-1}, \tau \in \mathbb{C}. (3.4)
\]

Denote by \( \Theta \) a finite set with finitely many incident directions as elements. Then, for any fixed \( \tau \in \mathbb{C}\setminus\{0\} \) and \( z_0 \in \mathbb{R}^n \setminus \mathcal{D} \), we introduce the following two indicators

\[
I_{z_0}^{\Theta}(z) := \left| \sum_{\theta \in \Theta} \int_{\mathbb{S}^{n-1}} \mathcal{F}(\hat{x}, \hat{\theta}, z_0, \tau) \cos[k \hat{x} \cdot (z - z_0)] ds(\hat{x}) \right|, \quad z \in \mathbb{R}^n, \tag{3.5}
\]

\[
I_{z_0}(z) := \left| \int_{\mathbb{S}^{n-1}} \int_{\mathbb{S}^{n-1}} \mathcal{F}(\hat{x}, \hat{\theta}, z_0, \tau) \cos[k (\hat{x} - \hat{\theta}) \cdot (z - z_0)] ds(\hat{x}) ds(\hat{\theta}) \right|, \quad z \in \mathbb{R}^n. \tag{3.6}
\]

Insert (3.4) into (3.5) or (3.6). Then a straightforward calculation shows that

\[
I_{z_0}^{\Theta}(z) = \left| \sum_{\theta \in \Theta} (V_{z_0}(z, \hat{\theta}) + \overline{V_{z_0}(z, \hat{\theta})}) \right| + O\left(\rho^{\frac{1-n}{2}}\right) \quad \text{and} \quad I_{z_0}(z) = |V_{z_0}(z) + \overline{V_{z_0}(z)}| + O\left(\rho^{\frac{1-n}{2}}\right)
\]

with

\[
V_{z_0}(z, \hat{\theta}) := \frac{\tau e^{-i k \hat{\theta} \cdot \hat{x}}}{2} \left[ G(z, \hat{\theta}) + G(2z_0 - z, \hat{\theta}) \right] \quad \text{and} \quad W_{z_0}(z) := \frac{\tau}{2} \left[ A(z) + A(2z_0 - z) \right], \quad z \in \mathbb{R}^n.
\]

Let \( D(z_0) \) be the point symmetric domain of \( D \) with respect to \( z_0 \). If the size of the scatterer \( D \) is small enough, from (3.2) we expect that the indicator \( I_{z_0}^{\Theta} \) takes its local maximum on the locations of \( D \) and \( D(z_0) \). For extended scatterer \( D \), from the behavior of the indicator \( A \) we expect that the indicator \( I_{z_0} \) takes its maximum on or near the boundary \( \partial D \cup \partial D(z_0) \).

Note that the indicator \( I_{z_0}^{\Theta}/I_{z_0} \) produces a false scatterer \( D(z_0) \). However, since we have the freedom to choose the point \( z_0 \), we can always choose \( z_0 \) such that the false domain \( D(z_0) \) located outside our interested searching domain. One may also overcome this problem by considering another indicator \( I_{z_1}^{\Theta}/I_{z_1} \) with \( z_1 \in \mathbb{R}^n \setminus \mathcal{D} \) and \( z_1 \neq z_0 \).

**Scatterer Reconstruction Scheme One.**

1. Collect the phaseless data set \( \{ |u^\infty_D(\hat{x}, \hat{\theta}, \tau)| : (\hat{x}, \hat{\theta}) \in \mathbb{S}^{n-1}, \tau \in \{0, \tau_1\} \} \).
2. Select a sampling region in \( \mathbb{R}^n \) with a fine mesh \( \mathcal{T} \) containing the scatterer \( D \).
3. Compute the indicator functional \( I_{z_0}(z) \) (or \( I_{z_0}^{\Theta} \) in the case of small scatterers) for all sampling point \( z \in \mathcal{T} \).
4. Plot the indicator functional \( I_{z_0}(z) \) (or \( I_{z_0}^{\Theta} \) in the case of small scatterers).
Using the Phase Retrieval Scheme proposed in the previous section, we obtain the approximate phased far field pattern \( u_D^\infty \). Then we have the second scatterer reconstruction algorithm.

**Scatterer Reconstruction Scheme Two.**

1. **Collect the phaseless data set** \( \{ |u_D^\infty(\hat{x}, \hat{\theta}, \tau)| : (\hat{x}, \hat{\theta}) \in S^{n-1}, \tau \in \{ \tau_1, \tau_2, \tau_3 \} \} \).

2. **Use the Phase Retrieval Scheme** to obtain the phased far field patterns \( u_D^\infty(\hat{x}, \hat{\theta}) \) for all \( (\hat{x}, \hat{\theta}) \in S^{n-1} \).

3. **Select a sampling region** in \( \mathbb{R}^n \) with a fine mesh \( T \) containing \( D \).

4. **Compute the indicator functional** \( I_2(z) \) (or \( I_3(z, \hat{\theta}) \) in the case of small scatterers with a fixed incident direction \( \hat{\theta} \)) for all sampling point \( z \in T \).

5. **Plot the indicator functional** \( I_2(z) \) (or \( I_3(z, \hat{\theta}) \) in the case of small scatterers).

### 4 Numerical examples and discussions

Now we present a variety of numerical examples in two dimensions to illustrate the applicability and effectiveness of our sampling methods. There are totally nine groups of numerical tests to be considered, and they are respectively referred to as \( I_{\theta_0}-\text{Soft}, I_{\theta_0}-\text{Multiple}, I_{\theta_0}-\text{MultiScalar}, I_{\theta_0}-\text{Small}, \text{PhaseRetrieval}, I_2-\text{Soft}, I_2-\text{Multiple}, I_2-\text{MultiScalar}, \) and \( I_3-\text{Small} \). The boundaries of the scatterers used in our numerical experiments are parameterized as follows

- **Kite**: \( x(t) = (a, b) + (\cos t + 0.65 \cos 2t - 0.65, 1.5 \sin t), \quad 0 \leq t \leq 2\pi \), \hspace{1cm} (4.1)
- **Peanut**: \( x(t) = (a, b) + 2\sqrt{3} \cos^2 t + 1 (\cos t, \sin t), \quad 0 \leq t \leq 2\pi \), \hspace{1cm} (4.2)
- **Pear**: \( x(t) = (a, b) + (2 + 0.3 \cos 3t)(\cos t, \sin t), \quad 0 \leq t \leq 2\pi \), \hspace{1cm} (4.3)
- **Circle**: \( x(t) = (a, b) + r (\cos t, \sin t), \quad 0 \leq t \leq 2\pi \), \hspace{1cm} (4.4)

with \((a, b)\) be the location of the scatterer which may be different in different examples and \(r\) be the radius of the circle.

![Kite, Peanut, Pear, Circle](image.png)

**Figure 2:** Different shapes to be used in the later examples.

Define \( \theta_m := 2\pi m/N, m = 0, 1, \cdots, N-1 \), let \( \hat{\theta}_l = (\cos \theta_l, \sin \theta_l) \) and \( \hat{x}_j = (\cos \theta_j, \sin \theta_j) \) for \( j, l = 0, 1, \cdots, N-1 \). In our simulations, we use the boundary integral equation method to compute the far field patterns \( u_D^\infty(\hat{x}_j, \hat{\theta}_l, \tau), \) \( j, l = 0, 1, \cdots, N-1 \), for \( N \) equidistantly...
distributed incident directions and $N$ observation directions. We further perturb this data by random noise

$$
|u_{D \cup \{ z_0 \}}^\infty \delta (\hat{x}_j, \hat{\theta}_l, \tau)| = |u_{D \cup \{ z_0 \}}^\infty (\hat{x}_j, \hat{\theta}_l, \tau) |(1 + \delta \ast e_{rel}), \quad j, l = 0, 1, \cdots, N - 1,
$$

where $e_{rel}$ is a uniformly distributed random number in the open interval $(-1,1)$. The value of $\delta$ used in our code is the relative error level. We also consider absolute error in Example PhaseRetrieval. In this case, we perturb the phaseless data

$$
|u_{D \cup \{ z_0 \}}^\infty (\hat{x}_j, \hat{\theta}_l, \tau)| = \max \left\{ 0, |u_{D \cup \{ z_0 \}}^\infty (\hat{x}_j, \hat{\theta}_l, \tau) | + \delta \ast e_{abs} \right\}, \quad j, l = 0, 1, \cdots, N - 1,
$$

where $e_{abs}$ is again a uniformly distributed random number in the open interval $(-1,1)$. Here, the value $\delta$ denotes the total error level in the measured data.

In the simulations, we use $0.05$ as the sampling space and $N = 512, k = 8$. If not otherwise stated, we take $z_0 = (12,12)$.

In the first four examples, we consider the indicators $I_{z_0}$ and $I_{z_0}^\Theta$ given by (3.6) and (3.5), respectively, with $\tau = 1$.

**Example I_{z_0}-Soft.** This example checks the validity of our method for scatterers with different reference points. For simplicity, we impose Dirichlet boundary condition on the underlying scatterer. The scatterer is a kite with $(a, b) = (0, 0)$. Figure 3 shows the results with 10% noise and three reference points $z_0 = (2, 4), z_0 = (4, 4)$ and $z_0 = (12, 12)$. As expected, the indicator $I_{z_0}$ takes a large value on $\partial D \cup \partial D(z_0)$, where $D(z_0)$ is the symmetric domain of $D$ about the reference point $z_0$. The symmetric domain of $z_0 = (12, 12)$ is outside of the sampling space. Note that $D(z_0)$ changes as the reference point $z_0$ changes, thus it is very easy to pick the correct domain $D$ by considering the indicator $I_{z_0}$ with different reference points, or we can just choose $z_0$ far enough. As shown in Figures 3, the left hand scatterer should be the one searched.

![Figure 3](image-url)

**Example I_{z_0}-Multiple.** We consider the scattering by a scatterer with two disjoint components. The scatterer is a combination of a sound-soft peanut shaped domain with $(a, b) = (0, 0)$ and a sound-hard kite shaped domain with $(a, b) = (6, 0)$. Figure 4 shows the reconstructions with different noises.
Example $I_{z_0}$-Multiple. In this example, the underlying scatterer is a combination of a big pear domain centered at $(0, 0)$ and a mini disk with radius $r = 0.1$ centered at $(a, b) = (2, 2)$. We impose Dirichlet boundary condition on both of them. The reconstructions are shown in Figure 5. We observe that both parts can be reconstructed clearly. In particular, the mini disk is also exactly located, even with 30% noise.

Example $I_{z_0}$-Multiscalar. In this example, the underlying scatterer is a combination of two mini disks, one with radius 0.05 centered at $(a, b) = (3, 3)$ and the other with radius 0.15 centered at $(a, b) = (1, 1)$. We impose Dirichlet boundary condition on the smaller disk and Neumann boundary condition on the bigger one. Figure 6 shows the reconstructions by $I_{z_0}^\Theta$ with the same reference points as in the Example $I_{z_0}$-Soft.

In the next example, we consider the effectiveness and robustness of the novel phase retrieval proposed in Section 2.3. After this, we check the validity of the Scatterer Reconstruction Scheme Two. In the following examples, we take $\tau = -1, 1, i$.

Example PhaseRetrieval. This example is designed to check the phase retrieval scheme proposed in Section 2.3. The underlying scatterer is chosen to be a kite shaped domain. For comparison, we consider the real part of far field pattern at a fixed incident direction $\hat{\theta} = (1, 0)$. Figure 7 shows the results without measurement noise by using three different reference points $(2, 2)$, $(3, 3)$ and $(4, 4)$. In particular, the reference point $(2, 2)$ is very close to the kite shaped
Figure 6: **Example $I_{z_0}^\Theta$-Small.** Reconstruction of two small disks by using $I_{z_0}^\Theta$ with 10% noise at different reference points. Here, $\Theta := \{(1,0),(0,1),(-1,0),(0,-1)\}$.

domain. However, Figure 7(a) shows that the multiple scattering is very week. Of course, Figures 7(b)-(c) show that the interaction between the reference point and the kite shaped domain decreases as the reference point away from the target. Figures 8-9 show the results with relative error and absolute error considered, respectively. We find that our phase retrieval scheme is quite robust with respect to noise. This also verifies the theory provided in Theorem 2.4.

![Figure 7](image-url)

Figure 7: **Example Phase Retrieval.** Phase retrieval for the real part of the far field pattern without error at a fixed incident direction $\hat{\theta} = (1,0)$ using different reference points.

**Example $I_2$-Soft.** The scatterer is the same as the **Example $I_{z_0}$-Soft.** For comparisons, we choose the same reference points $z_0 = (2,4), (4,4), (12,12)$. Figure 10 gives the results with 10% noise. Different to the **Example $I_{z_0}$-Soft**, no false domain appears in the reconstructions.

**Example $I_2$-Multiple.** The scatterer is the same as the **Example $I_{z_0}$-Multiple.** Figure 11 gives the results with 10%, 30% noise.

**Example $I_2$-Multiscalar.** The scatterer is the same as the **Example $I_{z_0}$-Multiscalar.** Figure 12 gives the results with 10%, 30% noise.

**Example $I_3$-Small.** The scatterer is the same as the **Example $I_{z_0}$-Small.** Figure 13 shows the reconstructions by $I_3(z)$ with different incident directions $(1,0)$ and $(0,1)$. 
Figure 8: **Example PhaseRetrieval.** Phase retrieval for the real part of the far field pattern with relative error at a fixed incident direction $\hat{\theta} = (1, 0)$.

(a) without noise.
(b) 10% noise.
(c) 30% noise.

Figure 9: **Example PhaseRetrieval.** Phase retrieval for the real part of the far field pattern with absolute error at a fixed incident direction $\hat{\theta} = (1, 0)$.

(a) without noise.
(b) 0.1 noise.
(c) 0.3 noise.

Figure 10: **Example I2-Soft.** Reconstruction of kite shaped domain with 10% noise and different reference points.

(a) $z_0 = (2, 4)$.
(b) $z_0 = (4, 4)$.
(c) $z_0 = (12, 12)$. 
Figure 11: **Example I₂-Multiple.** Reconstruction of mixed type scatterers with different noise.

Figure 12: **Example I₂-Multiscalar.** Reconstruction of multiscalar scatterers with different noise.

Figure 13: **Example I₃-Small.** Reconstruction of two small mixed type disks with 10% noise and different incident waves.
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