I. INTRODUCTION

Statistical information about the roughness of a surface is contained in its rms height and in its normalized-surface-height autocorrelation function. Efforts to obtain these properties of rough surfaces from measurements of light scattered into the far field from them are of interest because of the contactless nature of this approach, and because measurements in the far field are easier to carry out than measurements in the near field.

This problem has been studied in the past by several authors. In the case of a one-dimensional randomly rough dielectric surface, Chakrabarti et al. [1] inverted by a Fourier transformation an expression for the contribution to the mean differential reflection coefficient, obtained in the Kirchhoff approximation, from light scattered diffusely from the surface. The incident light was s-polarized and the plane of incidence was perpendicular to the generators of the surface. Good agreement with numerically generated scattering data was obtained for weakly rough surfaces.

The case of a two-dimensional randomly rough surface was studied by Chandley [2], and by Marx and Vorburger [3]. Chandley used scalar diffraction theory and a thin random phase screen approximation [4] to model the interaction of light with the randomly rough surface. A thin phase screen may be regarded as a layer of negligible thickness that alters the phase of the wave scattered from it but does not change its magnitude. It is derived from simple optical path length and geometrical optics arguments. He used the angular dependence of the mean intensity of the scattered light in the far field as the experimental quantity to be inverted. The nature of his scattering model allowed this inversion to be carried out by means of a Fourier transformation. The dielectric constant of the scattering medium does not appear explicitly in Chandley's theory which means that it is impossible to use it to recover the dielectric constant of the scattering medium from the experimental scattering data if it is not known in advance.

In their study of this problem, Marx and Vorburger applied the Kirchhoff approximation for the scattering of a scalar plane wave from a two-dimensional randomly rough perfectly conducting surface to obtain the mean intensity of the scattered field. The determination of the rms height of the surface and the normalized-surface-height autocorrelation function was achieved by assuming an expression for the latter function of a particular analytic form and by the determination of the parameters defining it by a least-squares fit of the theoretical mean intensity to the experimental result.

In contrast to these studies, in this paper we present an approach to the determination of the rms height and the normalized-surface-height autocorrelation function of a two-dimensional randomly rough penetrable surface, in particular a dielectric surface, from the inversion of optical scattering data. It is based on a vector theory of rough surface scattering rather than on a scalar theory, namely phase perturbation theory [5]. The dielectric constant of the medium is taken into account in this approach. This version of rough-surface-scattering theory was chosen in this study because in a recent comparison between experimental data and the predictions of three perturbation theories for the scattering of electromagnetic radiation from two-dimensional randomly rough metal surfaces, it produced the best results [5]. We expect it to be equally accurate in describing the scattering of visible light from a two-dimensional randomly rough dielectric surface. Specifically, we use the expression for the contribution to the mean differential reflection coefficient from the in-plane, co-polarized component of the light scattered incoherently when the dielectric surface is illuminated at normal or non-normal incidence by s-polarized light.

This expression is evaluated with the use of an expression for the normalized-surface-height autocorrelation function that contains adjustable parameters. The values of these parameters...
are then determined by a least-squares fit of the resulting expression to the corresponding experimental scattering data. The reconstruction of the parameters is performed using scattering data for both a single and several angles of incidence, and the sensitivity to noise of the reconstructed parameters is investigated. We note that the contribution to the mean differential reflection coefficient from the in-plane co-polarized component of the light scattered incoherently when the surface is illuminated by \( p \)-polarized light can also be used for this purpose. However, the expression one works with to effect this inversion is somewhat simpler in \( s \) polarization than in \( p \) polarization. In addition, there is no Brewster effect in \( s \) polarization, so that a smoother function of the scattering angle is being inverted in \( s \) polarization than in \( p \) polarization. It is for these reasons that we have chosen to work with \( s \) polarization.

This paper is organized as follows: First, the scattering system is presented (Sec. II) followed by elements of scattering theory (Sec. III) that will be useful for the subsequent discussion. Then in Sec. IV, we present the inversion scheme that will be used to reconstruct the surface-height autocorrelation function. The results obtained by the use of this procedure are presented in Sec. V for a set of different correlation functions and scattering geometries. Section VI presents discussions of these results and the conclusions that can be drawn from this study. The paper ends with an appendix detailing the derivation of expressions, central to the present work, for the first few moments of the scattering matrix for \( s \)-to-\( s \) scattering obtained on the basis of phase perturbation theory.

II. THE PHYSICAL SYSTEM STUDIED

The physical system we study in this paper consists of vacuum in the region \( x_3 > \zeta(x_3) \), and a dielectric medium, characterized by a dielectric constant \( \varepsilon \) that is real, positive, and frequency independent, in the region \( x_3 < \zeta(x_3) \) (Fig. 1). Here \( x_i = (x_1, x_2, 0) \) is a position vector in the plane \( x_3 = 0 \). The surface profile function \( \zeta(x_3) \) is assumed to be a single-valued function of \( x_3 \) that is differentiable with respect to \( x_1 \) and \( x_2 \). It is also assumed to constitute a stationary, zero-mean, isotropic, Gaussian random process defined by

\[
\langle \zeta(x_3) \rangle = \delta^2 W(|x_3 - x_3'|),
\]

where the angle brackets denote an average over the ensemble of realizations of \( \zeta(x_3) \), \( \delta \) is the rms height of the surface, and \( W(|x_3|) \) is the normalized-surface-height autocorrelation function, with the property that \( W(0) = 1 \).

The surface profile function has a Fourier integral representation,

\[
\hat{\zeta}(Q_3) = \int d^2 x_1 \zeta(x_1) \exp(-i Q_3 \cdot x_1).
\]

We also introduce the notation

\[
\hat{\zeta}(Q_3) = \int d^2 x_1 \zeta(x_1) \exp(-i Q_3 \cdot x_1).
\]

The Fourier coefficient \( \hat{\zeta}(Q_3) \) is also a zero-mean Gaussian random process defined by

\[
\langle \hat{\zeta}(Q_3) \rangle = (2\pi)\delta(Q_3 + Q_4)\delta^2 g(|Q_4|),
\]

where \( g(|Q_4|) \) is the power spectrum of the surface roughness, is defined by

\[
W(|x_3|) = \int d^2 x_1 \exp(-i Q_3 \cdot x_1).
\]

It follows from Eqs. (1) and (5) that \( g(|Q_4|) \) is normalized to unity,

\[
\int \frac{d^2 Q_3}{(2\pi)^2} g(|Q_3|) = 1.
\]

III. SCATTERING THEORY

The surface \( x_3 = \zeta(x_3) \) is illuminated from the vacuum by an electromagnetic field of frequency \( \omega \). The electric field in the vacuum above the surface is the sum of an incident and a scattered field, \( E(x; t) = E^{(i)}(x; t) + E^{(s)}(x; t) \), where

\[
E^{(i)}(x; t) = \left\{ \frac{\varepsilon \omega}{\varepsilon_0} [\hat{k}_p \delta_0(k_3) + \hat{s}_3 k_3] B_p(k_3) + (\hat{s}_3 \times \hat{k}_3) \right. \times B_s(k_3) \exp[i(k_3 - \delta \delta_0) \cdot x - i\alpha t],
\]

\[
E^{(s)}(x; t) = \int \frac{d^2 q_3}{(2\pi)^2} \left\{ \frac{\varepsilon \omega}{\varepsilon_0} [\hat{q}_3 \alpha_0(q_3) + \hat{s}_3 q_3] A_p(q_3) + (\hat{s}_3 \times \hat{q}_3) \right. \times A_s(q_3) \exp[i(k_3 - \delta \delta_0) \cdot x - i\alpha t].
\]

The subscripts \( p \) and \( s \) denote the \( p \)-polarized (TM) and \( s \)-polarized (TE) components of each of these fields, respectively.
The function $\alpha_0(q_1)$ in Eqs. (7) is defined as

$$\alpha_0(q_1) = \left[\frac{\omega}{c} - q_1^2\right]^{1/2}, \quad \text{Re} \alpha_0(q_1) > 0, \quad \text{Im} \alpha_0(q_1) > 0.$$  (8)

Maxwell's equations imply linear relations between $A_\alpha(q_1)$ and $R_\beta(q_1)$, which we write in the form ($\alpha = p, s, \beta = p, s$)

$$A_\alpha(q_1) = \sum_\beta R_{\alpha\beta}(q_1)\!|k_1\rangle R_\beta(k_1).$$  (9)

The scattering amplitudes $\{R_{\alpha\beta}(q_1\!|k_1\rangle)\}$ play a significant role in the present theory because the mean differential reflection coefficient is given in terms of them. The differential reflection coefficient $|\delta R_{\alpha\beta}(q_1\!|k_1\rangle)/\partial \Omega_\alpha|$ is defined such that $|\delta R_{\alpha\beta}(q_1\!|k_1\rangle)/\partial \Omega_\alpha|d\Omega_\alpha$ is the fraction of the total time-averaged flux in an incident field of $\beta$ polarization the projection of whose wave vector on the mean scattering plane is $k_1$, that is scattered into a field of $\alpha$ polarization, the projection of whose wave vector on the mean scattering plane is $q_1$, within an element of solid angle $d\Omega_\alpha$ about the scattering direction defined by the polar and azimuthal angles ($\theta_\alpha, \phi_\alpha$). It is given by [6,7]

$$\frac{\partial R_{\alpha\beta}(q_1\!|k_1\rangle)}{\partial \Omega_\alpha} = \frac{\omega}{2\pi c} \cos^2 \theta_\alpha \cos \theta_0 |R_{\alpha\beta}(q_1\!|k_1\rangle)|^2,$$  (10)

with (see Fig. 1)

$$k_1 = -\frac{\omega}{c} \sin \theta_0 (\cos \phi_0, \sin \phi_0, 0), \quad q_1 = -\frac{\omega}{c} \sin \theta_\alpha (\cos \phi_\alpha, \sin \phi_\alpha, 0),$$  (11a)

where ($\theta_0, \phi_0$) and ($\theta_\alpha, \phi_\alpha$) are the polar and azimuthal angles of incidence and scattering, respectively. $S$ is the area of the plane $x_3 = 0$ covered by the rough surface. As we are dealing with scattering from a randomly rough surface, it is the average of this function over the ensemble of realizations of the surface profile function that we have to calculate. The contribution to this average from the light scattered incoherently is

$$\left\langle \frac{\partial R_{\alpha\beta}(q_1\!|k_1\rangle)}{\partial \Omega_\alpha} \right\rangle_{\text{incoh}} = \frac{\omega}{S} \left(\frac{\omega}{2\pi c}\right)^2 \cos^2 \theta_\alpha \cos \theta_0 |S_{\alpha\beta}(q_1\!|k_1\rangle)|^2,$$  (12)

Closely related to the matrix of scattering amplitudes $R(q_1\!|k_1\rangle)$ is the scattering matrix $S(q_1\!|k_1\rangle)$ whose elements $\{S_{\alpha\beta}(q_1\!|k_1\rangle)\}$ are given by

$$S_{\alpha\beta}(q_1\!|k_1\rangle) = \alpha_0^{1/2}(q_1) \frac{R_{\alpha\beta}(q_1\!|k_1\rangle)}{\alpha_0^{1/2}(k_1)}.$$  (13)

These elements satisfy the reciprocity relations [8]

$$S_{pp}(q_1\!|k_1\rangle) = S_{pp}(-k_1\!| -q_1\rangle),$$  (14a)

$$S_{ps}(q_1\!|k_1\rangle) = S_{ps}(-k_1\!| -q_1\rangle),$$  (14b)

$$S_{ps}(q_1\!|k_1\rangle) = -S_{sp}(-k_1\!| -q_1\rangle).$$  (14c)

which serve as a check on the correctness of their derivation. In terms of the elements of the scattering matrix, Eq. (10) takes the form

$$\left\langle \frac{\partial R_{\alpha\beta}(q_1\!|k_1\rangle)}{\partial \Omega_\alpha} \right\rangle_{\text{incoh}} = \frac{\omega}{S} \left(\frac{\omega}{2\pi c}\right)^2 \cos \theta_\alpha |S_{\alpha\beta}(q_1\!|k_1\rangle)|^2$$

$$- |S_{\alpha\beta}(-k_1\!| -q_1\rangle)|^2.$$  (15)

This is the definition we will work with.

In the Appendix it is shown that the $ss$ element of the expression given by Eq. (15) obtained on the basis of second-order phase perturbation theory can be written as

$$\left\langle \frac{\partial R_{ss}(q_1\!|k_1\rangle)}{\partial \Omega_\alpha} \right\rangle_{\text{incoh}}$$

$$= \frac{(\varepsilon - 1)^2}{(2\pi)^2} \left(\frac{\omega}{c}\right)^6 \cos \theta_\alpha \left[\frac{\varepsilon \alpha_0(q_1)\alpha_0(k_1)\hat{q}_1 \cdot \hat{k}_1^2}{d_s(q_1) d_s(k_1)^2}\right]^n \exp[-2M(q_1\!|k_1\rangle)]$$

$$\times \sum_{n=1}^{\infty} \frac{4 \delta^2 \alpha_0(q_1) \alpha_0(k_1) \hat{q}_1 \cdot \hat{k}_1^2}{n!}$$

$$\times \int d^2u_1 W^\alpha(u_1) \exp[-i(q_1 - k_1) \cdot u_1].$$  (16)

In writing this expression we have introduced the functions

$$d_p(q_1) = \varepsilon \alpha_0(q_1) + \alpha(q_1),$$  (17a)

$$d_s(q_1) = \alpha_0(q_1) + \alpha(q_1),$$  (17b)

where

$$\alpha(q_1) = \left[\varepsilon \left(\frac{\omega}{c}\right)^2 - q_1^2\right]^{1/2} \quad \text{Re} \alpha(q_1) > 0, \quad \text{Im} \alpha(q_1) > 0.$$  (18)

The function $M(q_1\!|k_1\rangle)$ is given by (see the Appendix)

$$M(q_1\!|k_1\rangle) = 2 \delta^2 \alpha_0^{1/2}(q_1) \alpha_0^{1/2}(k_1)$$

$$\times \int \frac{d^2p_1}{(2\pi)^2} \Re F(q_1\!|p_1\!|k_1\rangle) g(|p_1 - k_1\rangle),$$  (19)

where

$$F(q_1|p_1|k_1) = \text{sgn} \left(\hat{q}_1 \cdot \hat{k}_1\right) \left\{ \frac{1}{2} \alpha(q_1) + \alpha(k_1) \right\} \hat{q}_1 \cdot \hat{k}_1 + (\varepsilon - 1)(\hat{q}_1 \times \hat{p}_1) \frac{\alpha_0(p_1) \alpha_0(p_2)}{d_p(p_1)} (\hat{p}_1 \times \hat{k}_1)$$

$$- (\varepsilon - 1) \left(\frac{\omega}{c}\right)^2 \hat{q}_1 \cdot \hat{p}_1 |p_1\rangle,$$  (20)

with sgn(·) denoting the sign function.
We now turn to an evaluation of the ingredients in Eq. (16). We begin with the expression for \(2M(q_i|k_i)\) given by Eqs. (19)–(20). With the use of Eqs. (5)–(6) we rewrite it in terms of \(W(x_i)\):

\[
2M(q_i|k_i) = \delta^2 \alpha_0^{1/2}(q_i)\alpha_0^{1/2}(k_i) \operatorname{sgn}(\hat{q}_i \cdot \hat{k}_i) \left\{ 2[\alpha(q_i) + \alpha(k_i)](\hat{q}_i \cdot \hat{k}_i) + \frac{(\varepsilon - 1)}{\pi^2} \Re \int_0^\infty d p_i p_i \int_{-\pi}^\pi d \phi_p \int_0^\infty dx_1 x_1 W(x_i) \times \int_{-\pi}^{\pi} d \phi_x \exp[-i p_i x_i \cos(\phi_p - \phi_x)] \times \left[ \frac{\alpha_0(p_i)\alpha(p_i)}{d_p(p_i)} \sin(\phi_p - \phi_x) \sin(\phi_x - \phi_p) \right. \\
\left. - \frac{(\omega/c)^2}{\delta_x(p_i)} \cos(\phi_q - \phi_p) \cos(\phi_q - \phi_x) \right]\}.
\] (21)

where \(\phi_q, \phi_p, \phi_x, \) and \(\phi_s\) are the azimuthal angles of the unit vectors \(\hat{q}_i, \hat{p}_i, \hat{k}_i, \) and \(\hat{s}_i\), respectively, measured from the positive \(x_i\) axis (see Fig. 1). On evaluating the angular integrals this result becomes

\[
2M(q_i|k_i) = 2\delta^2 \alpha_0^{1/2}(q_i)\alpha_0^{1/2}(k_i) \left\{ \alpha(q_i) + \alpha(k_i) - (\varepsilon - 1) \Re \left[ \int_0^\infty d p_i p_i \left( \frac{\alpha_0(p_i)\alpha(p_i)}{d_p(p_i)} + \frac{(\omega/c)^2}{\delta_x(p_i)} \right) \right] \int_0^\infty dx_1 x_1 \\
\times W(x_i) J_0(p_i x_i) J_0(k_i x_i) + \int_0^\infty d p_i p_i \left[ - \frac{\alpha_0(p_i)\alpha(p_i)}{d_p(p_i)} + \frac{(\omega/c)^2}{\delta_x(p_i)} \right] \int_0^\infty dx_1 x_1 W(x_i) J_2(p_i x_i) J_2(k_i x_i) \right\}.
\] (22)

where \(J_n(z)\) is a Bessel function of the first kind and order \(n\), and we have used the relation \(x \operatorname{sgn}(x) = |x|\). Finally, due to the circular symmetry of \(W(u_i)\) we obtain the result

\[
\int d^2 u_i \ W^* (u_i) \exp(-i q_i \cdot u_i) = 2\pi \int_0^{\infty} du_i u_i W^*(u_i) J_0(|q_i| - |k_i| u_i).
\] (23)

In the case of normal incidence \((k_i = 0)\) and in-plane \((\hat{q}_i \parallel \hat{k}_i)\) scattering, Eq. (16) becomes

\[
\left( \frac{\partial R_{ss}(q_i)}{\partial \Omega_s} \right)_{\text{inc oh}} = \frac{(\varepsilon - 1)^2}{(2\pi)^2} \left( \frac{\omega}{c} \right)^6 \frac{\cos \theta_s}{d_s(q_i)d_s(0)^2} \left| \frac{4\delta^2 \alpha_0(q_i)\alpha_0(0)}{n!} \int d^2 u_i \ W^*(u_i) \exp(-i q_i \cdot u_i) \right|^2
\] (24)

and Eq. (22) can be written as

\[
2M(q_i|0) = 2\delta^2 \alpha_0^{1/2}(q_i)\alpha_0^{1/2}(0) \left\{ \alpha(q_i) + \alpha(0) - (\varepsilon - 1) \Re \left[ \int_0^\infty d p_i p_i \left( \frac{\alpha_0(p_i)\alpha(p_i)}{d_p(p_i)} + \frac{(\omega/c)^2}{\delta_x(p_i)} \right) \right] \int_0^\infty dx_1 x_1 W(x_i) J_0(p_i x_i) J_0(k_i x_i) \right\}.
\] (25)

From Eq. (11) it is noted that for normal incidence \(\hat{k}_i = (\cos \phi_0, \sin \phi_0, 0)\), even if \(k_i/k_0\) is not well defined in this case. Moreover, for scattering into directions that are normal to the mean surface we have \(\hat{q}_i = \hat{k}_i\), so for all directions \(q_i\) corresponding to in-plane scattering \(|\hat{q}_i \cdot \hat{k}_i| = 1\). These results were used in arriving at the expressions presented in Eqs. (24) and (25).

**IV. THE INVERSE PROBLEM**

To determine the function \(W(x_i)\) from in-plane scattering data defined as

\[
\left( \frac{\partial R_{ss}(\theta_s)}{\partial \Omega_s} \right)_{\text{inc oh,input}} = \left( \frac{\partial R_{ss}(q_i)}{\partial \Omega_s} \right)_{\text{inc oh}} \left| \frac{\partial R_{ss}(q_i)}{\partial \Omega_s} \right|_{\text{inc oh}} |q_i| k_i = 1
\]

we assume an analytic form for it that contains adjustable parameters. The values of these parameters, together with the rms height \(\delta\), are determined by minimizing a cost function with respect to variations of these parameters. The cost function we use is

\[
\chi^2(\mathcal{P}) = \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} d \theta_s \left[ \left( \frac{\partial R_{ss}(\theta_s)}{\partial \Omega_s} \right)_{\text{inc oh,input}} \right]^2 - \left( \frac{\partial R_{ss}(\theta_s)}{\partial \Omega_s} \right)_{\text{inc oh,calc}}^2
\] (26)

where \(\mathcal{P}\) denotes the set of variational parameters used to characterize \(\langle \partial R_{ss}(\theta_s)/\partial \Omega_s \rangle_{\text{inc oh,calc}}\). The minimization of this function with respect to the elements of \(\mathcal{P}\) was carried out by the use of the routine "lmidif1" contained in the Fortran package MINPACK which is part of the general purpose mathematical library SLATEC [9]. The routine lmidif1 implements a modified version of the Levenberg-Marquardt algorithm [10,11], and it calculates the Jacobian by a forward-difference approximation.

The function \(\langle \partial R_{ss}(\theta_s)/\partial \Omega_s \rangle_{\text{inc oh,input}}\) was obtained from rigorous, nonperturbative, purely numerical solutions [8,12] of the reduced Rayleigh equation for the scattering of polarized light from a two-dimensional randomly rough penetrable dielectric surface [13].
for an ensemble of random surfaces generated [12] on the basis of expressions for \( W(|x|) \) of either the exponential form

\[
W(|x|) = \exp \left( \frac{-x^a}{a^*} \right),
\]

(27a)

or the Gaussian form

\[
W(|x|) = \exp \left\{ - \left( \frac{x^a}{a^*} \right)^2 \right\}.
\]

(27b)

In Eqs. (27), \( a \) denotes the transverse correlation length of the surface roughness.

The function \( \langle \partial R_{ss} (\theta_s) / \partial \Omega_1 \rangle \text{incoh,calc} \) was obtained by evaluating the expression for it obtained using phase perturbation theory [Eq. (16)] for the trial function assumed to represent \( W(|x|) \). Several forms for this trial function were used in our calculations. In the first set of forms we assumed an exponential or Gaussian trial function, that is,

\[
W(|x|) = \exp \left( \frac{-x^a}{a^*} \right),
\]

(28a)

or

\[
W(|x|) = \exp \left\{ - \left( \frac{x^a}{a^*} \right)^2 \right\}.
\]

(28b)

In this case the variational parameters of the reconstruction are \( \delta^*, a^*, \gamma^* \), and potentially \( \epsilon^* \).

V. RESULTS

We will now illustrate the inversion method developed here by applying it to the reconstruction of \( W(|x|) \), first by the use of one of the trial functions (28) and then by the use of the more general trial function (29).

A. Exponentially correlated surface roughness

For the first scattering system we consider, it is assumed that the surface-height autocorrelation function \( W(|x|) \) is exponential, Eq. (27a), and characterized by a transverse correlation length \( a = 158.20 \) nm and an rms height of the surface \( \delta = 9.50 \) nm. The medium above the surface is vacuum and the dielectric constant of the substrate is \( \epsilon = 2.64 \) (photoresist).

The wavelength (in vacuum) of the \( s \)-polarized incident light is \( \lambda = 632.8 \) nm. For this geometry and by the method of Ref. [12], we calculated the mean differential reflection coefficients by averaging the results from 5000 realizations of the surface profile function. For normal incidence, the in-plane, \( s \)-to-\( s \) co-polarized incoherent component of the mean differential reflection coefficient (DRC) obtained in this way is presented as a function of the scattering angle \( \theta_s \) by open circles in Fig. 2(a) [the same data set also appears in Figs. 3(a)–5(a)]. These data constitute the input function \( \langle R_{ss} (\theta_s) / \partial \Omega_1 \rangle \text{incoh,input} \) for our first set of reconstruction examples.

As our first example of reconstruction based on this data set, we assume that the trial function \( W(|x|) \) has the exponential form given by Eq. (28a). The set of variational parameters is therefore \( \mathcal{P} = \{ \delta^*, a^* \} \). The use of a mean differential reflection coefficient generated by the use of a known \( W(|x|) \)
in our inversion approach enables us to assess the quality of the reconstructions we obtain. By starting the minimization procedure with the values \( \delta^* = 2.00 \) nm and \( a^* = 75.00 \) nm, the values of these parameters that minimize the cost function \( \chi^2(\mathcal{P}) \), Eq. (26), were found to be \( \delta^* = 9.519 \) nm and \( a^* = 158.565 \) nm, to be compared with the values \( \delta = 9.50 \) nm and \( a = 158.20 \) nm used to generate the input data. In the minimization procedure we assumed that both \( \delta^* \) and \( a^* \) were restricted to positive values. The inversion is quite accurate. The function \( \frac{\partial R_{ss}(\theta_s)}{\partial \Omega_{\|}} \) calculated with the reconstructed values of \( \delta^* \) and \( a^* \) by means of second-order phase perturbation theory is plotted as the solid curve in Fig. 2(a) while the reconstructed correlation function \( W(\lambda s) \) is plotted as the solid curve in Fig. 2(b). The reconstructed \( W(\lambda s) \) is nearly superimposed on the input \( W(\lambda s) \) [open symbols Fig. 2(b)]. The shaded region in Fig. 2(b), and in the subsequent plots of \( W(\lambda s) \), represents the magnitude of the difference between the input and reconstructed values of this function. This difference is seen to be very small.

In the preceding example it was assumed that the dielectric constant of the scattering medium was known. For our second example we take the input data from our first example, given by the open circles in Fig. 2(a), but now assume that together with the roughness parameters the dielectric constant of the substrate is unknown. Therefore the variational parameter set is now \( \mathcal{P} = \{ \delta^*,a^*,\varepsilon^* \} \). The results of this inversion are shown in Fig. 3, and it is seen that also in this case a rather good reconstruction is obtained. By starting the minimization procedure with the values \( \delta^* = 2.00 \) nm, \( a^* = 75.00 \) nm, and \( \varepsilon^* = 2.00 \), the values of these parameters that minimize the cost function \( \chi^2(\mathcal{P}) \) were determined to be \( \delta^* = 9.272 \) nm, \( a^* = 158.042 \) nm, and \( \varepsilon^* = 2.718 \). These are to be compared with the input values \( \delta = 9.50 \) nm, \( a = 158.20 \) nm, and \( \varepsilon = 2.64 \). The function \( \frac{\partial R_{ss}(\theta_s)}{\partial \Omega_{\perp}} \) calculated with the reconstructed values of \( \delta^*,a^*,\varepsilon^* \) by means of second-order phase perturbation theory is plotted as the solid curve in Fig. 3(a), while the reconstructed correlation function \( W(\lambda s) \) is plotted as the solid curve in Fig. 3(b). From a comparison of the results presented in Figs. 2 and 3 it is seen that the addition of a single variational parameter changes the reconstruction of \( W(\lambda s) \) only marginally.

A more stringent test of our inversion scheme is obtained when the trial function for \( W(\lambda s) \) has a functional form that differs from the form assumed in generating the input data that the reconstruction is based on. As our third example, we therefore present results of our calculations when the trial \( W(\lambda s) \) is assumed to have the stretched exponential form given by Eq. (29). The set of variational parameters is now \( \mathcal{P} = \{ \delta^*,a^*,\gamma^* \} \). By starting the minimization procedure with the values \( \delta^* = 2.00 \) nm, \( a^* = 75.00 \) nm, and \( \gamma^* = 2.00 \), the values of these parameters that minimize the cost function were found to be \( \delta^* = 9.425 \) nm, \( a^* = 161.717 \) nm, and \( \gamma^* = 1.012 \). These values are fairly close to the input values \( \delta = 9.50 \) nm, \( a = 158.20 \) nm, and \( \gamma = 1.0 \) used in obtaining the simulation data. However, the importance of this example is to show that our minimization procedure is in fact able to distinguish a Gaussian form for the correlation function from an exponential form. In Fig. 4(a) we plot the function \( \frac{\partial R_{ss}(\theta_s)}{\partial \Omega_{\perp}} \) calculated with means of the second-order phase perturbation theory for the reconstructed values of \( \delta^*,a^*,\gamma^* \) (solid curve), together with a plot of the input function (open circles). The agreement between these two results is quite good. In Fig. 4(b) we present plots of the input (open circles) and reconstructed (solid curve) correlation functions \( W(\lambda s) \). The latter curve very nearly coincides with the former curve.

In our final example assuming an exponentially correlated surface, we again use the stretched exponential trial function for \( W(\lambda s) \), but now also assume that the dielectric constant of the scattering medium is unknown. The set of variational parameters is now \( \mathcal{P} = \{ \delta^*,a^*,\varepsilon^*,\gamma^* \} \). We start the minimization of the cost function \( \chi^2(\mathcal{P}) \) with the values \( \delta^* = 2.00 \) nm, \( a^* = 75.00 \) nm, \( \varepsilon^* = 2.00 \), and \( \gamma^* = 2.00 \). The values of these parameters that minimize the cost function are found to be \( \delta^* = 9.774 \) nm, \( a^* = 166.709 \) nm, \( \varepsilon^* = 2.507 \), and \( \gamma^* = 1.027 \). The proximity of these values to the input values, that
This figure is the same as Fig. 2 except that now the trial function for $\delta^\star$ assumed in the reconstructions were during the different reconstruction scenarios based on in-plane, correlated surface:

$$\delta = a \delta^\star$$

for the purpose of inversion, is characterized by a Gaussian function assumed in the earlier set of examples. Compared to correlation function instead of the exponential correlation function of the form (28a) was used, while in the last scenario based on in-plane, correlated surfaces using various variational parameter sets, $P$, that are subsets of $\{\delta^\star, a^\star, \epsilon^\star, \gamma^\star\}$. In such cases, the starting values assumed in the minimization will be $[2 \text{ nm}, 75 \text{ nm}, 2, 1]$, respectively, if nothing is said to indicate otherwise.

FIG. 4. Reconstruction of the rms roughness $\delta^\star$, the transverse correlation length $a^\star$, and the exponent $\gamma^\star$ from in-plane scattering data. This figure is the same as Fig. 2 except that now the trial function for $W(|x|)$ has the stretched exponential form given by Eq. (29). The reconstructed surface-roughness parameters are found to have the values $\delta^\star = 9.425 \text{ nm}, a^\star = 161.717 \text{ nm}, \text{ and } \gamma^\star = 1.012$.

are the same as those used previously, is poorer than for the first three examples, but the reconstructed values are still quite satisfactory. The reconstructed function $\left\langle \partial R_{ss}(\theta_s)/\partial \Omega_1 \right\rangle_{\text{incoh,calc}}$ [Fig. 5(a)] and the reconstructed correlation function $W(|x|)$ [Fig. 5(b)] calculated with these reconstructed values are still in good quantitative agreement with the corresponding input functions.

The parameters for the scattering system obtained in the different reconstruction scenarios detailed in this subsection are summarized in Table I.

| TABLE I. Summary of the scattering system parameters obtained during the different reconstruction scenarios based on in-plane, s-t-s co-polarized scattering data corresponding to an exponentially correlated surface: $\delta^\star$, $a^\star$, $\epsilon^\star$, and $\gamma^\star$. |  |
|---|---|---|---|---|
| $\delta^\star$ (nm) | $a^\star$ (nm) | $\epsilon^\star$ | $\gamma^\star$ | Comments |
| 9.519 | 158.565 | – | – | Fig. 2 |
| 9.272 | 158.042 | 2.718 | – | Fig. 3 |
| 9.425 | 161.717 | – | 1.012 | Fig. 4 |
| 9.774 | 166.709 | 2.507 | 1.027 | Fig. 5 |

The second scattering system, from which we will use data for the purpose of inversion, is characterized by a Gaussian correlation function instead of the exponential correlation function assumed in the earlier set of examples. Compared to the previous scattering system, in addition to the different form of $W(|x|)$, the only parameters that have changed are the rms roughness and the dielectric constant of the substrate; they now have the values $\delta = 15.82 \text{ nm}$ (an increase of more than 65% compared to its previous value) and $\epsilon = 2.6896$, respectively. Except for the angles of incidence, all other parameters characterizing the scattering system remained unchanged, i.e., $a = 158.20 \text{ nm}$ and $\lambda = 632.8 \text{ nm}$.

For these parameters, a computer simulation approach [12] was used to generate scattering data that were obtained by averaging the results from 24000 surface realizations. Results obtained this way are presented as open symbols in Fig. 6(a) for the polar angle of incidence $\theta_0 = 50.2^\circ$. It is these data we will base our inversion on in this subsection; that is, here this data set represents $\left\langle \partial R_{ss}/\partial \Omega_1 \right\rangle_{\text{incoh,input}}$.

Motivated by the reconstruction done in Sec. VA using scattering data obtained for the exponentially correlated surface, we will now perform similar inversions for Gaussian-correlated surfaces using various variational parameter sets, $P$, that are subsets of $\{\delta^\star, a^\star, \epsilon^\star, \gamma^\star\}$. In such cases, the starting values assumed in the minimization will be $[2 \text{ nm}, 75 \text{ nm}, 2, 1]$, respectively, if nothing is said to indicate otherwise.

Figure 6 provides the results of the reconstruction for $P = \{\delta^\star, a^\star\}$ under the assumption that the trial function used for $W(|x|)$ is of the Gaussian form, Eq. (28b). In this way, the reconstruction procedure resulted in the numerical values $\delta^\star = 15.873 \text{ nm}$ and $a^\star = 158.000 \text{ nm}$. These values agree rather well with the values assumed in generating the scattering data used in the inversion. Moreover, the input and reconstructed correlation functions, as well as the absolute difference between them, are depicted in Fig. 6(b); the solid red line in Fig. 6(a) represents the mean DRC predicted by the inversion.

We now continue by reconstructing the same variational parameter sets, $P$, as were used in Sec. VA for the exponential surface roughness; the only difference now is that we will assume the form of the trial function (28b) where we previously used Eq. (28a). A summary of the reconstructed parameters is presented in Table II. Moreover, in Fig. 7 the input and
FIG. 5. Reconstruction of the rms roughness $\delta^\ast$, the transverse correlation length $a^\ast$, the exponent $\gamma^\ast$, and the dielectric constant of the substrate $\varepsilon^\ast$ from in-plane scattering data. This figure is the same as Fig. 2 except that now the trial function for $W(|x|)$ has the stretched exponential form given by Eq. (29), and the dielectric constant of the scattering medium is assumed to be unknown. The reconstructed surface-roughness parameters are found to have the values $\delta^\ast = 9.774 \text{ nm}$, $a^\ast = 166.709 \text{ nm}$, $\gamma^\ast = 1.027$, and the reconstructed value of the dielectric constant is $\varepsilon^\ast = 2.507$.

different reconstructed correlation functions obtained in this way are presented, together with comparisons of them. The mean DRCs that result from these reconstructions are visually indistinguishable from those of Fig. 6(a), and such plots have therefore not been presented.

From Figs. 6–7 and Table II we see that the reconstructed results for the Gaussian-correlated surface roughness are in general good, at least for the angle of incidence $\theta_0 = 50.2^\circ$ assumed here. The quality of the results obtained is on a par with the results obtained previously for the exponentially correlated surface roughness, even though in the Gaussian case the rms roughness is significantly larger and the angle of incidence is nonzero.

It should be mentioned that we have also performed reconstructions based on input data corresponding to other polar angles of incidence than $\theta_0 = 50.2^\circ$, and it was found that the values of the reconstructed parameters essentially remained unchanged; if there were any changes at all, the quality of the reconstruction seemed to improve for smaller polar angles of incidence. Assuming different initial values of the parameters of the set $\mathcal{P}$ seemed not to affect the reconstruction. Hence, our results seem to indicate that reconstruction based on different input data, at least for the scattering data that we considered, influences the numerical values of the reconstructed parameters of the surface-height correlation function only to a small extent. Furthermore, the reconstruction seems to be reliable for a wide range of angles of incidence.

C. Sensitivity to noise

Any experimental data set will contain some level of noise. Therefore, it is imperative to have reliable inversion
TABLE II. Summary of the values reconstructed from in-plane scattering data obtained for the polar angle of incidence \( \theta_0 = 50.2^\circ \) and corresponding to a Gaussian-correlated surface characterized by \( \delta = 15.82 \) nm and \( a = 158.20 \) nm. The dielectric constant of the substrate was \( \varepsilon = 2.6896 \). Note that a Gaussian correlation function corresponds to an exponent \( \gamma = 2.00 \) for the stretched exponential. The symbol “—” indicates that the corresponding parameter was not reconstructed and instead had the value assumed in numerically generating the input data. When inverting for any of the parameters in the set \( \{ \delta^*, a^*, \varepsilon^*, \gamma^* \} \) the initial values used were \( \{2 \text{ nm}, 75 \text{ nm}, 2, 1\} \), respectively.

| \( \delta^* \) (nm) | \( a^* \) (nm) | \( \varepsilon^* \) | \( \gamma^* \) | Comments |
|--------------------|----------------|-------------------|--------------|---------|
| 15.922             | 157.928        | –                 | –            | Fig. 6  |
| 16.161             | 157.785        | 2.645             | –            | Fig. 7(a) |
| 16.170             | 154.592        | –                 | 1.929        | Fig. 7(b) |
| 16.180             | 157.148        | 2.651             | 1.986        | Fig. 7(c) |

approaches that can be applied successfully to data sets containing noise. Until now, we have used simulated data as the basis for the reconstruction, and the only source of uncertainty (or noise) in such simulations results is the finite number of surface realizations used to obtain them. However, since a sufficiently high number of realizations has been used in generating such data, the uncertainty has been modest. To start investigating the sensitivity of the reconstructed parameters to noise, we will, for reasons of comparison, base it on the data set used in Sec. VB [open symbols in Fig. 6(a)] and add noise to it. Due to the way these simulation results were generated [12], only 36 points exist in this data set. However, experimental angular resolved scattering data sets typically will have significantly better angular resolution (and, thus, more points). Therefore, to better mimic the more relevant experimental situation, we have interpolated by splines these simulation results [solid symbols in Fig. 8(a)] to an angular resolution of \( \Delta \theta = 1^\circ \) for \( \theta_2 \) in the interval from \(-90^\circ \) to \(90^\circ \). Then, to these (locally smooth) interpolated data, we have added multiplicative Gaussian white noise of a standard deviation of 5% and zero mean [gray erratic signal oscillating around zero in Fig. 8(a) resulting in the open blue data in Fig. 8(a)]. It is this latter data set that will be used as the noisy input signal for the reconstructions to be performed below. In passing, it is noted that additive noise (of constant amplitude) mainly will affect the tails of the in-plane angular intensity distributions of the scattered light due to their typical convex shape. In this study we have chosen to focus on multiplicative noise since it affects the whole intensity distribution and therefore represents a greater challenge for the reconstruction approach.

Results for the reconstructed parameters based on this Gaussian white noisy data set, performed in a fashion identical to what was done in the preceding subsection, are listed in Table III. From this table, we observe that the results obtained are rather good, even for the significant noise level assumed. Moreover, one finds that the quality of the reconstruction is not dramatically degraded compared to what was previously obtained using the non-noisy data set (see Table II). Not surprisingly, the poorest results of the inversion are obtained for the variational parameter set \( \mathcal{P} \) of cardinality 4, and this

FIG. 7. Correlation functions obtained by reconstructing several variational parameter sets \( \mathcal{P} \) for the Gaussian surface roughness parameters defined in Fig. 6: (a) \( \mathcal{P} = \{ \delta^*, a^*, \varepsilon^* \} \), (b) \( \mathcal{P} = \{ \delta^*, a^*, \gamma^* \} \), and (c) \( \mathcal{P} = \{ \delta^*, a^*, \varepsilon^*, \gamma^* \} \). The numerical values of the reconstructed parameters are listed in Table II. The shaded areas represent the absolute error between the input and reconstructed correlation functions.

“worst case” is presented as solid red lines in Fig. 8. The results presented in Table III and Fig. 8, which for the non-noisy case should be compared to Table II and Fig. 7(c), support the view that the reconstruction approach presented in this paper is able to produce reliable results also when the input data are noisy.

\[ \text{TABLE II. Summary of the values reconstructed from in-plane scattering data obtained for the polar angle of incidence } \theta_0 = 50.2^\circ \text{ and corresponding to a Gaussian-correlated surface characterized by } \delta = 15.82 \text{ nm and } a = 158.20 \text{ nm. The dielectric constant of the substrate was } \varepsilon = 2.6896. \text{ Note that a Gaussian correlation function corresponds to an exponent } \gamma = 2.00 \text{ for the stretched exponential. The symbol “—” indicates that the corresponding parameter was not reconstructed and instead had the value assumed in numerically generating the input data. When inverting for any of the parameters in the set } \{ \delta^*, a^*, \varepsilon^*, \gamma^* \} \text{ the initial values used were } \{2 \text{ nm}, 75 \text{ nm}, 2, 1\}, \text{ respectively.} \]

| \( \delta^* \) (nm) | \( a^* \) (nm) | \( \varepsilon^* \) | \( \gamma^* \) | Comments |
|-------------------|----------------|-------------------|--------------|---------|
| 15.922            | 157.928        | –                 | –            | Fig. 6  |
| 16.161            | 157.785        | 2.645             | –            | Fig. 7(a) |
| 16.170            | 154.592        | –                 | 1.929        | Fig. 7(b) |
| 16.180            | 157.148        | 2.651             | 1.986        | Fig. 7(c) |

FIG. 7. Correlation functions obtained by reconstructing several variational parameter sets \( \mathcal{P} \) for the Gaussian surface roughness parameters defined in Fig. 6: (a) \( \mathcal{P} = \{ \delta^*, a^*, \varepsilon^* \} \), (b) \( \mathcal{P} = \{ \delta^*, a^*, \gamma^* \} \), and (c) \( \mathcal{P} = \{ \delta^*, a^*, \varepsilon^*, \gamma^* \} \). The numerical values of the reconstructed parameters are listed in Table II. The shaded areas represent the absolute error between the input and reconstructed correlation functions.

“worst case” is presented as solid red lines in Fig. 8. The results presented in Table III and Fig. 8, which for the non-noisy case should be compared to Table II and Fig. 7(c), support the view that the reconstruction approach presented in this paper is able to produce reliable results also when the input data are noisy.
For instance, when reconstructing 4 parameters, the relative errors in the reconstructed correlation lengths are about 3.5% and −0.7% for the noisy and non-noisy cases, respectively, which is not dramatic given the level of noise that was added to the input data.

It should be mentioned that we found that if the parameters were estimated using a noisy version of the original scattering data instead of the interpolated data, as done above, the results using the trial function (28b) essentially remained unchanged. However, for the stretched exponential form (29), the results were more sensitive to the initial values used in the minimization, resulting in less robust results than those obtained by the use of the former data set. We will see in the next subsection that this can also be the case when using multiple angles of incidence.

### TABLE III. Same as Table II, but now the data set used in the inversion consisted of an interpolated version of the data set used to produce the results of Table II, with added multiplicative Gaussian white noise of a standard deviation of 5% (and mean zero). In Fig. 8, the noisy input data set is depicted as blue open symbols, while the noise appears in gray.

| $\delta^\star$ (nm) | $a^\star$ (nm) | $\varepsilon^\star$ | $\gamma^\star$ | Comments |
|---------------------|----------------|---------------------|----------------|----------|
| 15.955              | 157.705        | –                   | –              | –        |
| 16.151              | 157.571        | 2.653               | –              | –        |
| 15.946              | 157.824        | –                   | 2.003          | –        |
| 15.941              | 163.661        | 2.606               | 2.102          | Fig. 8   |

### D. Inversion of data obtained from multiple angles of incidence

When the experimental setup is prepared to measure the in-plane scattering of light for one angle of incidence, it is relatively straightforward to perform additional measurements for other angles of incidence. Therefore, it is of interest to study how the reconstructed parameters will depend on using multiple angles of incidence, and thus several data sets, in the inversion. In order to include, say, $N$ data sets into the reconstruction, the cost function used in the minimization is generalized to

$$\chi^2(\mathcal{P}) = \sum_{n=1}^{N} \chi_n^2(\mathcal{P}),$$

where $\chi_n^2(\mathcal{P})$ is defined by Eq. (26) and corresponds to the contribution to the total cost function $\chi^2(\mathcal{P})$ from a single angle of incidence $\theta_0^{(n)} \in \{0^{\circ}, 1.6^{\circ}, 25.3^{\circ}, 50.2^{\circ}\}$ [12]. A series of joint inversions were then performed based on the three resulting data sets seen as open symbols in Fig. 9(a). The parameters obtained by such an approach are presented in Table IV and the resulting mean DRCs obtained when reconstructing $\delta^\star$, $a^\star$, and $\varepsilon^\star$ are presented as solid lines in Fig. 9(a). A comparison of the results presented in Tables II and IV reveals that including additional data sets into the inversion scheme, at least for the data sets we used in obtaining these results, did not change the estimates of the parameters in any significant way. If there is any change, it may seem as if a multangle reconstruction may slightly improve the results when using the trial function.
(28b), while it becomes slightly worse for the trial function (29). However, before one can draw firm conclusions on this issue, proper estimates of the error bars associated with each of these parameters will have to be obtained, something that is outside the scope of the present work.

If now noise is added to these data sets after first interpolating them to a higher angular resolution, as was done above for the single data set corresponding to $\theta_0 = 50.2^\circ$, then on performing multiangle reconstructions based on the resulting data, the values presented in Table V are obtained [see also Fig. 9(b)]. The first thing to observe from Table V is that the multiangle inversions based on the stretched exponential trial function (29) produce rather inaccurate results compared to inversions based on only one of these data sets (see Table III). Potentially one could first estimate the exponent of the stretched exponential from a single data set, since we have found that it does not matter which of the data sets we use, and then use this value for $\gamma^*$ as a fixed parameter in a multiangle inversion. However, we will not consider this situation farther here, since we feel that it is probably more fruitful to consider alternative forms of the trial functions.

On the other hand, when using the Gaussian trial function (28b) in the minimization, a comparison of the results presented in Tables III and V shows that the multiangle reconstructions are producing more accurate results than those obtained when the inversion is based on the single angle scattering data set corresponding to $\theta_0 = 50.2^\circ$. We have also found this result to be true when any of the other two data sets were used in the single-angle inversion. This is an important result, since it may indicate that including several experimental data sets into the inversion may improve the estimates of the parameters. The results from Table V also hint at the importance of the choice taken for the trial function, since $a^*$, $\gamma^*$, and $\varepsilon^*$ are used in the single-angle and multi-angle non-noisy case, respectively, to obtain the minimum in a cost function, $\chi (P)$, that has a deep and well-defined minimum, in contrast to many local minima of comparable depths.

It is also interesting to note that the results obtained for the multiangle noisy case (Table V) seem to be more accurate than those for the corresponding multiangle non-noisy case (Table IV). However, here it is important to recall that significantly more points are used in the inversion in the former than in the latter case, and we speculate that this could be the main reason for the improvement.

### E. Computational cost of the inversion scheme

In principle, an inversion scheme, similar to the one proposed in this work, could be based on one of the rigorous simulation approaches that recently have become available to calculate light scattering from two-dimensional

| $\delta^*$ (nm) | $a^*$ (nm) | $\varepsilon^*$ | $\gamma^*$ | Comments |
|----------------|------------|-----------------|------------|----------|
| 15.920         | 157.649    | –               | –          | –        |
| 16.074         | 157.929    | 2.658           | –          | Fig. 9(a) |
| 16.308         | 152.620    | –               | 1.896      | –        |
| 16.208         | 155.652    | 2.665           | 1.953      | –        |

| $\delta^*$ (nm) | $a^*$ (nm) | $\varepsilon^*$ | $\gamma^*$ | Comments |
|----------------|------------|-----------------|------------|----------|
| 15.843         | 158.694    | –               | –          | –        |
| 15.822         | 158.863    | 2.669           | –          | Fig. 9(b) |
| 16.866         | 145.591    | –               | 1.749      | –        |
| 17.050         | 140.664    | 2.730           | 1.675      | –        |

The solid lines, independent of color, represent the reconstructed mean DRCs. The parameter values obtained from the reconstructions are listed in Tables IV and V.

FIG. 9. The same as (a) Fig. 6(a) and (b) Fig. 8(a) but for the polar angles of incidence $\theta_0 = 1.6^\circ$, 25.3°, and 50.2° and multiangle reconstruction of the data sets corresponding to these angles of incidence assuming the variational parameter set $P = \{\delta^*, a^*, \varepsilon^*, \gamma^*\}$. The solid lines, independent of color, represent the reconstructed mean DRCs. The parameter values obtained from the reconstructions are listed in Tables IV and V.
randomly rough surfaces [6,7,14]. However, for such an inversion scheme to be practically relevant, it has to be numerically efficient, since during the inversion process, the forward-scattering problem has to be solved for a large set of parameters. From this perspective, the rigorous numerical simulation approaches mentioned previously fall short since they typically require days of computer time, or more, to run for just one set of parameters.

In contrast, the phase perturbation theoretical approach to the forward scattering problem that we base our inversion scheme on is computationally efficient. For instance, for an angular resolution of 1°, it takes less than 2 s to obtain the mean DRC in the plane of incidence for one set of surface parameters (and one angle of incidence). Furthermore, the CPU times it took to perform the four inversions whose results are given in Table II were 25 s and 140 s when reconstructing 2 and 4 parameters, respectively (initial values as in Sec. VB). The remaining two inversions required CPU times somewhere in between the two previously given CPU times. All reported computer CPU times are based on the use of a single computer core on an Intel i7 960 CPU operating at 3.20 GHz.

Even if such CPU times do depend strongly on the angular resolution of the data set used in the inversion, the amount of noise that it contains, and the initial values from which the reconstruction is started, these numbers for the computational cost do illustrate that the proposed inversion method is rather quick to perform; this, together with its robustness, should make it useful in practical situations.

VI. DISCUSSION AND CONCLUSIONS

In this paper we have shown that second-order phase perturbation theory can be used to model the incoherent scattering of light by a two-dimensional randomly rough dielectric surface and to calculate the mean differential reflection coefficient for such a surface. As a result it has been expected that it should be an effective tool for the inversion of light-scattering data to obtain statistical properties of a random surface on which the mean differential reflection coefficient depends. This expectation has been borne out for weakly rough two-dimensional randomly rough dielectric surfaces. Together with several parametrized forms for the normalized-surface-height autocorrelation function $W(|\mathbf{x}|)$, namely an exponential, a Gaussian, and a stretched exponential, and the minimization of a cost function with respect to the parameters defining these forms, phase perturbation theory has been used in this paper to determine $W(|\mathbf{x}|)$, the rms height of the surface, the transverse correlation length of the surface roughness, the dielectric constant of the scattering medium when it was not known in advance, as well as the exponent of the stretched exponential. The function $W(|\mathbf{x}|)$ has been reconstructed quite accurately. The agreement between the reconstructed values of $\delta$, $\alpha$, and $\epsilon$ and the input values of these parameters is gratifyingly satisfactory.

This agreement remains very good when significant multiplicative Gaussian white noise is included in the input data. When simulated data for multiple angles of incidence are used in the inversion scheme, it is found that in the absence of noise in the input data the quality of the reconstructions is slightly poorer than when only a single angle of incidence is used. However, when noise is introduced into the input data the use of results obtained from multiple angles of incidence yields slightly better reconstructions than when data from only a single angle of incidence are used. The reasons for this behavior of the reconstructions is not understood, and deserve further study.

An investigation of the computational cost of the inversion approach developed here shows that it is quite computationally efficient compared to the several orders of magnitude higher computational cost of carrying out the inversions by the use of scattering data obtained by rigorous simulations. The CPU times required for carrying out an inversion calculation using our approach for a given angular resolution, one set of surface parameters, and a single angle of incidence, namely seconds, are short enough that this approach can be useful in practical situations.

The inversion approach developed here needs to be explored to determine ranges of roughness, wavelength, and dielectric parameters for which it gives reliable results. Error estimates for the reconstructed values of the roughness and material parameters sought should be obtained. The use of more flexible trial functions for $W(|\mathbf{x}|)$ in reconstructions should be explored, as well as the use of more than one wavelength. Reflectivity data can serve as the basis of an inversion scheme based on phase perturbation theory, and its utility for this purpose should be examined. This issue will be explored in subsequent work.

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APPENDIX: DERIVATION OF EQUATION (16)

Even if we in this work are concerned with a scattering geometry where the substrate is a dielectric, this assumption will not be made in this appendix. Instead, the dielectric function of the substrate will here be assumed complex, so that the substrate can be either a dielectric or a metal. This generalization is done in order to facilitate future reference to the results of this appendix, and because the expressions can be derived simultaneously for a dielectric or metallic substrate with little extra effort.

The starting point for our derivation of Eq. (16) is Eqs. (12), (15), and (16)–(18) of Ref. [15], and the definition of the scattering matrix $S(\mathbf{q}|\mathbf{k})$ in terms of the matrix of the scattering amplitudes $R(\mathbf{q}|\mathbf{k})$, Eq. (13). From these results we obtain for the $ss$ element of the scattering matrix the expansion

$$
S_{ss}(\mathbf{q}|\mathbf{k}) = S_{ss}^{(0)}(\mathbf{q}|\mathbf{k}) - i S_{ss}^{(1)}(\mathbf{q}|\mathbf{k}) - \frac{1}{2} S_{ss}^{(2)}(\mathbf{q}|\mathbf{k}) + \cdots,
$$

(A1)
where the superscript denotes the order of the corresponding term in the surface profile function $\xi(x)$. The coefficient $S_{ss}^{(0)}(q_||k_||)$ is given by

$$
S_{ss}^{(0)}(q_||k_||) = (2\pi)^2 \delta(q_|| - k_||) \frac{\alpha_0(k_||) - \alpha(k_||)}{\alpha_0(k_||) + \alpha(k_||)} \\
= (2\pi)^2 \delta(q_|| - k_||) \left( \frac{\alpha_0(q_||) - \alpha(q_||)}{\alpha_0(q_||) + \alpha(q_||)} \right)^{1/2} \left( \frac{\alpha_0(k_||) - \alpha(k_||)}{\alpha_0(k_||) + \alpha(k_||)} \right)^{1/2} \\
= (2\pi)^2 \delta(q_|| - k_||) \left[ \frac{\alpha_0^2(q_||) - \alpha^2(q_||)}{[\alpha_0(q_||) + \alpha(q_||)][\alpha_0(k_||) + \alpha(k_||)]} \right]^{1/2} \\
= (2\pi)^2 \delta(q_|| - k_||) \left( 1 - \varepsilon (\omega/c)^2 \right) \\
$$

where the functions $\alpha_0(q_||)$ and $\alpha(q_||)$ are defined by Eqs. (8) and (18), respectively. The coefficient $S_{ss}^{(1)}(q_||k_||)$ is found to be

$$
S_{ss}^{(1)}(q_||k_||) = 2(1 - \varepsilon) \left( \frac{\omega}{c} \right)^2 \frac{\alpha_0^2(q_||) - \alpha^2(q_||)}{[\alpha_0(q_||) + \alpha(q_||)][\alpha_0(k_||) + \alpha(k_||)]} (\hat{q}_|| \cdot \hat{k}_||) \hat{\xi}(q_|| - k_||). \\
$$

while the coefficient $S_{ss}^{(2)}(q_||k_||)$ is given by

$$
S_{ss}^{(2)}(q_||k_||) = -\frac{4\alpha_0^2(q_||) - \alpha^2(q_||)}{\alpha_0(q_||) + \alpha(q_||)[\alpha_0(k_||) + \alpha(k_||)]} \left\{ \frac{1}{2} \left( \frac{\omega}{c} \right)^2 \left[ \alpha(q_||) + \alpha(k_||) \right] \hat{\xi}(q_|| - k_||) \\
+ \frac{1}{2} \left( \frac{\omega}{c} \right)^2 \int \frac{d^2 p_||}{(2\pi)^2} \hat{\xi}(q_|| - p_||)(\hat{q}_|| \cdot \hat{p}_||) \alpha(p_||)(\hat{p}_|| \times \hat{k}_||) \hat{\xi}(p_|| - k_||) \right\} \\
+ \frac{4\alpha_0^2(q_||) - \alpha^2(q_||)}{\alpha_0(q_||) + \alpha(q_||)[\alpha_0(k_||) + \alpha(k_||)]} \int \frac{d^2 p_||}{(2\pi)^2} \hat{\xi}(q_|| - p_||) \\
\times \left\{ \frac{1 - \varepsilon}{\varepsilon} (\hat{q}_|| \times \hat{p}_||) (\hat{q}_|| \times \hat{k}_||) \alpha_0^2(p_||) + \alpha(p_||) \right\} \left[ \hat{\xi}(p_|| - k_||) \\
+ \left( \frac{\omega}{c} \right)^2 \frac{\alpha_0(q_||) - \alpha(q_||)}{\alpha_0(q_||) + \alpha(q_||)} \right] (\hat{q}_|| \cdot \hat{p}_||)(\hat{q}_|| \cdot \hat{k}_||) \hat{\xi}(p_|| - k_||) \\
\right\} \\
= -4(1 - \varepsilon) \left( \frac{\omega}{c} \right)^2 \frac{\alpha_0^2(q_||) - \alpha^2(q_||)}{[\alpha_0(q_||) + \alpha(q_||)][\alpha_0(k_||) + \alpha(k_||)]} \\
\times \int \frac{d^2 p_||}{(2\pi)^2} \hat{\xi}(q_|| - p_||) \left\{ \frac{1}{2} \left[ \alpha(q_||) + \alpha(k_||) \right] (\hat{q}_|| \cdot \hat{k}_||) + \frac{1 - \varepsilon}{\varepsilon} (\hat{q}_|| \times \hat{k}_||) \alpha(p_||)(\hat{p}_|| \times \hat{k}_||) \\
- \left( \frac{\omega}{c} \right)^2 \frac{\alpha_0(q_||) - \alpha(q_||)}{\alpha_0(q_||) + \alpha(q_||)} \right\} \left[ \hat{\xi}(p_|| - k_||) \\
\right\} \\
= -4(1 - \varepsilon) \left( \frac{\omega}{c} \right)^2 \frac{\alpha_0^2(q_||) - \alpha^2(q_||)}{[\alpha_0(q_||) + \alpha(q_||)][\alpha_0(k_||) + \alpha(k_||)]} \\
\times \int \frac{d^2 p_||}{(2\pi)^2} \hat{\xi}(q_|| - p_||) \left\{ \frac{1}{2} \left[ \alpha(q_||) + \alpha(k_||) \right] (\hat{q}_|| \cdot \hat{k}_||) + \frac{1 - \varepsilon}{\varepsilon} (\hat{q}_|| \times \hat{k}_||) \alpha(p_||)(\hat{p}_|| \times \hat{k}_||) \\
- \left( \frac{\omega}{c} \right)^2 \frac{\alpha_0(q_||) - \alpha(q_||)}{\alpha_0(q_||) + \alpha(q_||)} \right\} \left[ \hat{\xi}(p_|| - k_||) \\
\right\}. \\
$$

In obtaining this expression we have used the result that

$$
\hat{\xi}^{(2)}(q_|| - k_||) = \int \frac{d^2 p_||}{(2\pi)^2} \hat{\xi}(q_|| - p_||) \hat{\xi}(p_|| - k_||). \\
$$
When we substitute the results given by Eqs. (A2) into Eq. (A1) we find that through terms of second order in the surface profile function, \( S_{ss}(\mathbf{q}_l|\mathbf{k}_i) \) takes the form

\[
S_{ss}(\mathbf{q}_l|\mathbf{k}_i) = \text{sgn} (\mathbf{q}_l \cdot \mathbf{k}_i) \frac{(1 - \varepsilon (\omega/c)^2}{d_s(q_l) d_s(k_i)} \int d^2 x_i \exp [-i(\mathbf{q}_l - \mathbf{k}_i) \cdot \mathbf{x}_i] \left\{ 1 - 2i\alpha_0^{1/2}(q_l)\alpha_0^{1/2}(k_i)\text{sgn}(\mathbf{q}_l \cdot \mathbf{k}_i) \right\},
\]

where the expression inside the curly brackets in Eq. (A5) is a continuous function of the lateral scattering wave vector \( q \).

We next express Eq. (A4) in the form of a Fourier integral:

\[
S_{ss}(\mathbf{q}_l|\mathbf{k}_i) = \text{sgn} (\mathbf{q}_l \cdot \mathbf{k}_i) \int d^2 x_i \exp [-i(\mathbf{q}_l - \mathbf{k}_i) \cdot \mathbf{x}_i] \left\{ 1 - 2i\alpha_0^{1/2}(q_l)\alpha_0^{1/2}(k_i)\text{sgn}(\mathbf{q}_l \cdot \mathbf{k}_i) \right\},
\]

where

\[
F(\mathbf{q}_l|\mathbf{p}_l|\mathbf{k}_i) = \text{sgn} (\mathbf{q}_l \cdot \mathbf{k}_i) \int d^2 x_i \exp [-i(\mathbf{q}_l - \mathbf{k}_i) \cdot \mathbf{x}_i] \left\{ 1 - 2i\alpha_0^{1/2}(q_l)\alpha_0^{1/2}(k_i)\text{sgn}(\mathbf{q}_l \cdot \mathbf{k}_i) \right\},
\]

and the functions \( d_s(p_l) \) and \( d_s(k_i) \) are defined in Eq. (17). One notes from Eq. (A6) that the expression inside the curly brackets in Eq. (A5) is a continuous function of the lateral scattering wave vector \( \mathbf{q}_l \) (and in particular at \( \mathbf{q}_l = \mathbf{0} \)).

From Eq. (A5) we find that

\[
\langle S_{ss}(\mathbf{q}_l|\mathbf{k}_i) \rangle = \text{sgn} (\mathbf{q}_l \cdot \mathbf{k}_i) \int d^2 x_i \exp [-i(\mathbf{q}_l - \mathbf{k}_i) \cdot \mathbf{x}_i] \left\{ 1 - 2i\alpha_0^{1/2}(q_l)\alpha_0^{1/2}(k_i)\text{sgn}(\mathbf{q}_l \cdot \mathbf{k}_i) \right\},
\]

which follows that

\[
|\langle S_{ss}(\mathbf{q}_l|\mathbf{k}_i) \rangle |^2 = \left( \frac{1 - \varepsilon (\omega/c)^2}{d_s(q_l) d_s(k_i)} \int d^2 x_i \exp [-2M(\mathbf{q}_l|\mathbf{k}_i)] \int d^2 x_i \exp [-i(\mathbf{q}_l - \mathbf{k}_i) \cdot (\mathbf{x}_i - \mathbf{x}_i')].
\]

where

\[
2M(\mathbf{q}_l|\mathbf{k}_i) = 4\varepsilon^2 \alpha_0^{1/2}(q_l)\alpha_0^{1/2}(k_i) \text{Re} \int d^2 p_l \int d^2 x_i \exp [-i(\mathbf{p}_l - \mathbf{k}_i) \cdot (\mathbf{x}_i - \mathbf{x}_i')].
\]

We next find that

\[
|\langle S_{ss}(\mathbf{q}_l|\mathbf{k}_i) \rangle |^2 = \left( \frac{1 - \varepsilon (\omega/c)^2}{d_s(q_l) d_s(k_i)} \int d^2 x_i \int d^2 x_i' \exp [-i(\mathbf{q}_l - \mathbf{k}_i) \cdot (\mathbf{x}_i - \mathbf{x}_i')] \left\{ 1 - 2i\alpha_0^{1/2}(q_l)\alpha_0^{1/2}(k_i)\text{sgn}(\mathbf{q}_l \cdot \mathbf{k}_i)\text{sgn}(\mathbf{x}_i - \mathbf{x}_i') \right\},
\]

(10)

This expression for \( S_{ss}(\mathbf{q}_l|\mathbf{k}_i) \) is manifestly reciprocal; i.e., it satisfies Eq. (14b). Moreover, for reasons of later convenience, in writing Eq. (A4) we have factored out a phase \( \text{sgn}(\mathbf{q}_l \cdot \mathbf{k}_i) \), where \( \text{sgn}(-) \) denotes the sign function defined by \( x = \text{sgn}(x)|x| \).
From this result we obtain

\[
\|S_{x}(q||k)\|^2 = \left|\frac{(1 - \varepsilon)(\omega/c)^2}{d_s(q||d_s(k||)}\right|^2 \int d^2x_\parallel \int d^2x_\parallel \exp[-i(q||k||) \cdot (x_\parallel - x_\parallel)] \\
\times \exp[-2M(q||k||)] \int d^2x_\parallel \int d^2x_\parallel \exp[-i(q||k||) \cdot (x_\parallel - x_\parallel)] \\
\times \exp[4\varepsilon^2\alpha_0(q||\alpha_0(k||)(q||k||)^2W(|x_\parallel - x_\parallel|)].
\] (A11)

Thus, we have finally

\[
\|S_{xx}(q||k)\|^2 - \|S_{xx}(q||k)\|^2 = S\left|\frac{(1 - \varepsilon)(\omega/c)^2}{d_s(q||d_s(k||)}\right|^2 \exp[-2M(q||k||)] \int d^2u_\parallel \exp[-i(q||k||) \cdot u_\parallel] \\
\times \exp[4\varepsilon^2\alpha_0(q||\alpha_0(k||)(q||k||)^2W(|u_\parallel|)] - 1.
\] (A12)

The substitution of this result into Eq. (15) yields Eq. (16).

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