# Optical potentials from the scattering cross section by inversion

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The experimental elastic scattering cross section is inverted into an optical potential using a rational parametrization of the scattering function. A regularized error analysis is carried out. The method is successfully applied to the system  $\alpha^{-40}$ Ca at 104 MeV.

## I. INTRODUCTION

Empirical optical potentials are usually inferred from experimental elastic differential cross sections by search on the parameters of an assumed parametrized potential shape. The traditional search procedure based on the Woods-Saxon shape has in the recent past been extended or modified with the aim of making it more "model independent." We mention here the (Woods-Saxon)<sup> $\nu$ </sup> fits,<sup>1</sup> the Fourier-Bessel expansion,<sup>2</sup> the orthogonal polynomial expansion,<sup>3</sup> the sum-of-Gaussians method,<sup>4</sup> and the cubic spline fits.<sup>5-7</sup>

While it is impossible to determine the potential deductively from the cross section with the help of some analytic algorithm, this can in principle be done if the input is not the cross section but the scattering function (or its logarithm, the phase shift). The latter is the inverse scattering problem at fixed energy: calculating the potential from the phase shifts as a function of angular momentum.<sup>8-12</sup>

In a recently proposed inversion method at fixed energy<sup>13-15</sup> the scattering function is interpolated by a rational function of angular momentum, whose poles and zeros determine uniquely the corresponding potential. This method has been applied in nuclear and atomic scattering.<sup>16-19</sup> Inasmuch as the rational interpolation of the scattering function between the integer angular momenta can be regarded as model independent, the same can then be said of the potential derived from this interpolation. A general discussion of potentials connected with simple classes of scattering functions is given in Refs. 20 and 21.

The input for the inverse scattering problem, the scattering function  $S_l$  at integer angular momenta  $l=0,1,2,\ldots$ , would have to be obtained from the experimental elastic cross section by a phase shift analysis (as in Ref. 22). For high energies and heavy particles, for example in heavy-ion scattering, a phase shift analysis which treats each phase shift independently, becomes forbiddingly difficult owing to the large number of phase shifts  $\delta_l$  involved (one for each integer angular momentum l which contributes to the scattering). It appears much simpler to skip fitting the phase shifts: that is, to fit directly the small number of parameters **a** (poles and zeros) in the rational scattering function  $S(l, \mathbf{a})$ , to the experimental cross section.

This "rational scattering function analysis" of the cross section is not only more practical than the traditional phase shift analysis, it has the further great advantage of yielding immediately the associated scattering potential via the algorithm of Ref. 15. Although a search on the parameters **a** has to be made—this is no "deductive" procedure— and although a certain functional dependence of the phase shift  $\delta_l$  on l is implied (though a rather flexible one), this procedure may perhaps be called a model independent inversion of the elastic cross section into the associated potential.

In the inversion the spread of the experimental data as well as the inaccuracies in the fitting procedure will induce errors in the potential.<sup>23,24</sup> Unless only a very few parameters **a** are used in the fit, the inverse problem will become an ill-posed problem, leading to large, unphysical errors in the potential. This can be remedied by a statistical regularization procedure.<sup>25,26</sup>

In the present paper we formulate such an inversion scheme for the cross section and apply it to a realistic case. In the next section, the inversion algorithm for a rational scattering function is briefly recalled, and the fitting of the parameters of this scattering function to the experimental cross section is explained. Section III is devoted to the derivation of the errors or confidence limits of the potential, and to a discussion of the regularization of the inversion procedure. In Sec. IV the method is applied to the scattering of  $\alpha$  particles on <sup>40</sup>Ca at 104 MeV, and the results are compared with those of Ref. 27. Section V contains a summary.

## **II. INVERSION METHOD**

The method<sup>13-15</sup> makes use of the fact that certain simple classes of scattering functions  $S_l = S(l; \mathbf{a})$  charac-

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terized by a finite number of parameters  $\mathbf{a} = \{a_n\}$  are associated with certain easily calculable, analytic classes of local potentials. The determination of the potential ("inversion") then consists in determining the parameters **a** from the input scattering information [the scattering function  $S_i$  at  $i=0,1,2,\ldots$  in the conventional inverse scattering problem or, as in the present work, the cross section  $\sigma(\theta)$  at a finite number of angles  $\theta = \theta_i$ ] by a fitting procedure. Thereafter, the potential is obtained automatically, as a by-product as it were. The search in the fit is on the parameters of the scattering function, not of the potential.

Examples of such classes are the *rational scattering* function<sup>13</sup>

$$S_{\rm rat}(l;\mathbf{a}) = S^{(0)}(\lambda) \prod_{n=1}^{N} \frac{\lambda^2 - \beta_n^2}{\lambda^2 - \alpha_n^2} , \qquad (2.1)$$

where  $\lambda = l + \frac{1}{2}$  and  $\mathbf{a} = \{a_n\} = \{\alpha_n, \beta_n\}$ , and  $S^{(0)}(\lambda)$  is the scattering function of a possible reference potential  $V_0(r)$ , e.g., a Coulomb background potential; the *nonrational scattering function*<sup>14</sup>

$$S_{\text{nonrat}}(l;\mathbf{a}) = S^{(0)}(\lambda) \frac{\left| \left| \frac{\sigma_{\beta_{n}}^{(0)} - \sigma_{\alpha_{m}}^{(0)}}{\beta_{n}^{2} - \alpha_{m}^{2}} - \frac{\sigma_{\lambda}^{(0)} - \sigma_{\alpha_{m}}^{(0)}}{\lambda^{2} - \alpha_{m}^{2}} \frac{\sigma_{\beta_{n}}^{(0)}}{\sigma_{\lambda}^{(0)}} \right| \right|}{\left| \left| \frac{\sigma_{\beta_{n}}^{(0)} - \sigma_{\alpha_{m}}^{(0)}}{\beta_{n}^{2} - \alpha_{m}^{2}} - \frac{\sigma_{\lambda}^{(0)} - \sigma_{\alpha_{m}}^{(0)}}{\lambda^{2} - \alpha_{m}^{2}} \right| \right|}$$
(2.2)

with  $n, m = 1, \ldots, N$  and

$$\sigma_{\lambda}^{(0)} = e^{-i\pi(\lambda - 1/2)}S^{(0)}(\lambda)$$

and the "mixed" scattering function,<sup>15</sup> which is a product of expressions (2.1) and (2.2). Since the rational function can only be used when  $\text{Im}\alpha_n^2 > 0$  and  $\text{Im}\beta_n^2 < 0$ ,<sup>14</sup> while the nonrational function is appropriate only if  $|\text{Im}\alpha_n|$ ,  $|\text{Im}\beta_n|$  are reasonably large,<sup>15</sup> the mixed scheme must generally be employed.

The potential associated with these scattering functions is given by

$$V(r) = V_N(r) ; \qquad (2.3)$$

$$V_n(r) = V_{n-1}(r) + V^{(n)}(r); \quad n = 1, \dots, N; \quad (2.4)$$

$$V^{(n)}(r) = \frac{2}{r} (\beta_n^2 - \alpha_n^2) \frac{d}{dr} \left[ \frac{1}{r} \frac{1}{L_{\beta_n}^{(n-1)}(r) - L_{\alpha_n}^{(n-1)}(r)} \right].$$
(2.5)

The  $L_{\lambda}^{(n)\pm}(r)$  are the logarithmic derivatives of the Jost solutions  $f_{\lambda}^{(n)\pm}(r)$  to the potential  $V_n(r)$  in the rational scheme, and of the regular solution  $\phi_{\lambda}^{(n)}(r)$ , in the nonrational scheme.<sup>14,15</sup>

A given set of  $S_l$ , l = 0, 1, ..., is easily interpolated by the rational function (2.1). The parameters  $\mathbf{a} = \{\alpha_n, \beta_n\}$ are determined by setting

$$\prod_{n=1}^{N} (\lambda^2 - \alpha_n^2) = \sum_{n=1}^{N} A_n \lambda^{2n-2} + \lambda^{2N} , \qquad (2.6a)$$

$$\prod_{n=1}^{N} (\lambda^2 - \beta_n^2) = \sum_{n=1}^{N} B_n \lambda^{2n-2} + \lambda^{2N} , \qquad (2.6b)$$

and minimizing

$$\sum_{i=1}^{M} g_{i} \left[ S_{li} \left[ \sum_{n=1}^{N} A_{n} \lambda_{i}^{2n-2} + \lambda_{i}^{2N} \right] - S^{(0)}(\lambda_{i}) \left[ \sum_{n=1}^{N} B_{n} \lambda_{i}^{2n-2} + \lambda_{i}^{2N} \right] \right]^{2}$$
(2.7)

with respect to the 2N complex parameters  $A_n$  and  $B_n$ , n = 1, ..., N. The number of fitting points is equal to  $M \ge 2N$ , and the  $g_i$  are suitable weights. The minimization of expression (2.7) leads to a system of 2N linear equations with a unique solution for the coefficients  $A_n$ and  $B_n$ . From these the poles  $\alpha_n^2$  and zeros  $\beta_n^2$  are found immediately. With the convention that all  $\text{Im}\alpha_n < 0$  and all  $\text{Im}\beta_n > 0$ , we find that for  $\text{Re}\alpha_n$ ,  $\text{Re}\beta_n < 0$  we have genuine Regge poles and zeros, while for  $\text{Re}\alpha_n$ ,  $\text{Re}\beta_n > 0$ we have false Regge poles and zeros.<sup>14</sup> The former are taken into account in the rational form (2.1), and the latter in the nonrational form (2.2), with the condition that the imaginary parts of the false poles and zeros are sufficiently large ( $\ge 2$ ). The potential is then calculated as explained in Ref. 15.

Thus, the inversion starting from the scattering function  $S_l$  involves a unique "best" interpolation which is readily found by solving a linear system of equations. On the other hand, the inversion starting from the experimental cross section  $\sigma(\theta)$  is not so straightforward. Here we introduce a "rational" cross section interpolation,

$$\sigma(\theta; \mathbf{a}) = |f(\theta; \mathbf{a})|^2, \qquad (2.8)$$

$$f(\theta; \mathbf{a}) = f_{\text{Coul}}(\theta) + (2ik)^{-1} \sum_{l=0}^{\infty} (2l+1) [S_{\text{rat}}(l; \mathbf{a}) - e^{2i\sigma_l}] P_l(\cos\theta), \qquad (2.9)$$

where  $f_{\text{Coul}}(\theta)$  and  $\sigma_l$  are the Coulomb amplitude and phases, respectively. The parameters  $\mathbf{a} = \{a_n\} = \{\alpha_n, \beta_n\}$  are determined by minimizing the sum of least squares

$$\chi^{2}(\boldsymbol{\sigma};\mathbf{a}) \equiv \sum_{i=1}^{M} \left[\sigma_{i} - \sigma(\theta_{i};\mathbf{a})\right]^{2} / (\Delta \sigma_{i})^{2} , \qquad (2.10)$$

where  $\sigma = {\sigma_i} = {\sigma(\theta_i)}$ , and the  $\theta_i$  are the measurement angles;  $\Delta \sigma_i$  is the experimental error of  $\sigma_i$ .

Here the minimization problem does not lead to linear equations, but a nonlinear search on the 4N real parameters **a** must be made instead. This is unavoidable in any scheme: The cross section can in general be analyzed only by a search procedure, be it on the phase shifts, or on the poles or zeros of the rational scattering function. As mentioned in the Introduction, we regard the latter as more practicable than the former, quite apart from its "invertibility" into a potential. The search on the parameters **a** must be started with some starting values  $\mathbf{a}^{(a)}$ . These may be obtained by a rough Woods-Saxon-type fit of the cross section  $\sigma(\theta)$ , whereupon the scattering function  $S_I$  of that Woods-Saxon-type potential is used to calculate the parameters  $\mathbf{a}^{(a)}$  from the linear equation discussed earlier.

#### **III. ERROR AND REGULARIZATION**

Regarding the experimental errors as purely statistical, in particular, assuming that the experimental cross section values  $\sigma_i$  are independent stochastic variables with a normal distribution of variance  $(\Delta \sigma_i)^2$ , then<sup>24,25</sup>

$$P(\boldsymbol{\sigma} \mid \mathbf{a}) = \operatorname{const} \exp[-(\frac{1}{2})\chi^2(\boldsymbol{\sigma};\mathbf{a})]$$
(3.1)

is the conditional probability for  $\sigma_i$  if the "true" cross section is given by the expression (2.8) with some fixed but unknown parameters **a**. One finds estimates  $\hat{\mathbf{a}}$  of these parameters in terms of the experimental quantities  $\sigma_i$ ,  $(\Delta \sigma_i)^2$  by maximizing the probability (3.1), i.e., minimizing the function  $\chi^2(\sigma; \mathbf{a})$  with respect to the variables **a** (as was indeed the recipe of the previous section). The minimum value is then given by

$$\chi^2_{\min}(\boldsymbol{\sigma}) = \chi^2(\boldsymbol{\sigma}; \mathbf{\hat{a}}) ; \qquad (3.2)$$

its expectation value over the distribution  $P(\sigma | \mathbf{a})$  of the variables  $\sigma$ , is<sup>25</sup>

$$\overline{\chi^2_{\min}} = F , \qquad (3.3)$$

where F = M - 4N is equal to the number of degrees of freedom.

In practice, the parametrized cross section (2.8) will not, in general, represent the true cross section exactly (although it might get quite close with sufficiently many parameters a). Then, Eq. (3.3) is to be replaced by

$$\chi^2_{\min} \geq F$$
 . (3.4)

Thus,

$$f = f(\boldsymbol{\sigma}; \hat{\mathbf{a}}) \equiv \chi^2(\boldsymbol{\sigma}; \hat{\mathbf{a}}) / F$$
(3.5)

is an estimate of the goodness of fit, the true fit giving f=1 on the average.

We now regard expression (3.1) as a distribution in the parameters **a** for given experimental values  $\sigma$ , and define the average over a function  $F(\mathbf{a})$ ,

$$\langle F(\mathbf{a}) \rangle \equiv \int F(\mathbf{a}) P(\boldsymbol{\sigma} \mid \mathbf{a}) d\mathbf{a} / \int P(\boldsymbol{\sigma} \mid \mathbf{a}) d\mathbf{a}$$
. (3.6)

Then, the covariance of the parameters **a** defines the error matrix<sup>25</sup>  $(\Delta a_n = a_n - a_n)$ 

$$\epsilon_{nm} \equiv \langle \Delta a_n \Delta a_m \rangle = \alpha_{nm}^{-1} f . \tag{3.7}$$

Here,

$$\alpha_{nm} = \frac{1}{2} \frac{\partial^2 \chi^2}{\partial a_n \partial a_m} \bigg|_{\hat{\mathbf{a}}}$$
$$= \sum_{i=1}^{M} \frac{\partial \sigma(\theta_i; \mathbf{a})}{\partial a_n} \frac{\partial \sigma(\theta_i; \mathbf{a})}{\partial a_m} \bigg|_{\hat{\mathbf{a}}} \frac{1}{(\Delta \sigma_i)^2}$$
(3.8)

is the curvature, or normal matrix, and the parabolic approximation is assumed valid near the minimum of  $\chi^2$ .

This yields for the covariance of the potential  $V(r) = V(r; \hat{a})$ ,

$$\langle \Delta V(r) \Delta V(r') \rangle = \sum_{n,m=1}^{4N} \frac{\partial V(r)}{\partial a_n} \frac{\partial V(r')}{\partial a_m} \bigg|_{\hat{a}} \epsilon_{nm}$$
 (3.9)

Thus, we have for the error

$$[\overline{\Delta V(r)}]^2 \equiv \langle [\Delta V(r)]^2 \rangle = \sum_{n,m=1}^{4N} \frac{\partial V(r)}{\partial a_n} \frac{\partial V(r)}{\partial a_m} \bigg|_{\widehat{a}} \epsilon_{nm} .$$
(3.10)

If the potential is complex, Eq. (3.9) is to be generalized to a covariance for the real and imaginary parts, and Eq. (3.10) holds for  $\operatorname{Re} V(r)$  and  $\operatorname{Im} V(r)$  separately. We remark that one could analogously define the covariance for the scattering function  $S(\lambda)$ , and make an error analysis on the phase shifts in terms of our rational parametrization.

The fit to the experimental cross section will generally be the better (i.e., f will be the closer to unity) the larger the number 4N of parameters. But as N increases, the normal matrix  $\alpha_{nm}$  becomes ill conditioned,<sup>26,29</sup> and the errors  $\overline{\Delta V(r)}$  will become very large. A family of highly fluctuating, unphysical potentials will be found which, however, all give rise to fits of the cross section  $\sigma(\theta)$ within the experimental errors.

By keeping the number of parameters low, the class of potentials can be kept "smooth," at the possible expense of the goodness of fit. This has been called "parametric regularization."<sup>20</sup> A more satisfying method is the statistical regularization scheme,<sup>26</sup> which has been introduced in nuclear scattering by Krappe and Rossner.<sup>29</sup> (Another method of regularization is that of Ref. 30.) As applied to the present situation, the method stipulates that the searched-for potential be as close to a pregiven a priori potential "as the experimental data allow." Since the potential is determined by the parameters a, we consider an equivalent set of a priori parameters  $\mathbf{a}^{(a)}$ , which, as the notation indicates, are conveniently identified with the starting values of the parameter search. These a priori parameters determined a normal a priori distribution with assumed variance  $1/\gamma$ :<sup>26</sup>

$$P_{a \text{ priori}}(\mathbf{a}) = \operatorname{const} \exp\left[-(\gamma/2) \sum_{n=1}^{4N} (a_n - a_n^{(a)})^2\right].$$
 (3.11)

Together with the experimental information described by the distribution  $P(\sigma | \mathbf{a})$  of Eq. (3.1) we have the *a posteriori* distribution (Bayes formula)

$$P_{a \text{ post}}(\mathbf{a} \mid \boldsymbol{\sigma}) = \frac{P_{a \text{ priori}}(\mathbf{a})P(\boldsymbol{\sigma} \mid \mathbf{a})}{\int P_{a \text{ priori}}(\mathbf{a})P(\boldsymbol{\sigma} \mid \mathbf{a})d\mathbf{a}} .$$
 (3.12)

This conditional probability in the parameters  $\mathbf{a}$ , for given values  $\boldsymbol{\sigma}$ , is to be maximized to yield estimates for the *a* posteriori parameters  $\mathbf{a}$ . That is, one must determine the minimum of the expression

$$\widetilde{\chi}^{2}(\boldsymbol{\sigma};\mathbf{a}) \equiv \gamma \sum_{n=1}^{4N} (a_{n} - a_{n}^{(a)})^{2} + \chi^{2}(\boldsymbol{\sigma};\mathbf{a})$$
(3.13)

with respect to the variables  $a_n$ , which results in the *a* posteriori estimate  $\hat{\mathbf{a}}(\gamma) = \{a_n(\gamma)\}$ .

For  $\gamma = 0$  we obtain the previous goodness-of-fit value f of Eq. (3.3) without a priori information; as  $\gamma$  increases, the value of

$$f(\gamma) \equiv \chi^2(\boldsymbol{\sigma}; \hat{\mathbf{a}}(\gamma)) / F \tag{3.14}$$

increases as well, i.e., the fit gets worse. However, the regularized error matrix<sup>26,29</sup>

$$\epsilon_{nm}(\gamma) = [\beta(\gamma)^{-1}]_{nm} f(\gamma) , \qquad (3.15)$$

where

$$\beta_{nm}(\gamma) = \alpha_{nm}(\gamma) + \gamma \delta_{nm} \tag{3.16}$$

and  $\alpha_{nm}(\gamma)$  is  $\alpha_{nm}$  of Eq. (3.8) evaluated at  $\hat{\mathbf{a}}(\gamma)$ , has reduced, stabilized matrix elements, and the errors of the potential cease to be fluctuating between large limits.

As to the value of  $\gamma$ , i.e., the degree to which the a priori information is taken into account, one may require that it be chosen such that the value of the function  $\chi^2_{\min}(\gamma) \equiv \chi^2(\sigma; \hat{a}(\gamma))$  associated with the minimum value of  $\tilde{\chi}^2$ ,  $\tilde{\chi}^2_{\min} = \tilde{\chi}^2(\sigma; \hat{\mathbf{a}}(\gamma))$ , be equal to the unregularized average value of  $\chi^2$  (Philipps condition):<sup>31</sup>

$$\chi^2_{\min}(\gamma) = \langle \chi^2 \rangle . \tag{3.17}$$

Thus, one allows the *a posteriori* "minimum"  $\chi^2(\gamma)$  value to be as large as the unregularized, "unbiased" average  $\chi^2$ value  $\langle \chi^2 \rangle > \chi^2_{\min}$ .

Alternatively, one may require that the a posteriori average of  $\chi^2$  be equal to the unregularized average of  $\chi^2$ (Turchin condition):<sup>26</sup>

$$\langle \chi^2 \rangle_{a \text{ post}} = \langle \chi^2 \rangle , \qquad (3.18)$$

where

$$\langle \chi^2 \rangle_{a \text{ post}} \equiv \int \chi^2(\sigma; \mathbf{a}) P_{a \text{ post}}(\mathbf{a} \mid \sigma) d\mathbf{a}$$
 (3.19)

This last condition gives less weight to the a priori information than the former.

Now<sup>26</sup>

$$\langle \chi^2 \rangle = \chi^2_{\min} + 4N ,$$

$$\langle \chi^2 \rangle_{a \text{ post}} = \chi^2_{\min}(\gamma) + 4N_{\text{eff}}(\gamma) ,$$

$$(3.20)$$

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where

$$4N_{\rm eff}(\gamma) = \sum_{n,m}^{4N} \epsilon_{nm}(\gamma) \alpha_{nm}(\gamma) \big|_{\hat{a}(\gamma)}$$
(3.21)

is the effective number of parameters  $[4N_{\text{eff}}(0)=4N]$ . Therefore, condition (3.17) implies (Philipps)

$$\chi^2_{\min}(\gamma) = \chi^2_{\min} + 4N$$
, (3.22)

while condition (3.18) means (Turchin)

$$\chi^{2}_{\min}(\gamma) + 4N_{\text{eff}}(\gamma) = \chi^{2}_{\min} + 4N$$
 (3.23)

In either case, one starts with a large value of  $\gamma$ , when the left-hand sides are larger than the right-hand sides, and reduces it until Eqs. (3.22) and (3.23) are satisfied.

#### IV. APPLICATION TO $\alpha$ -<sup>40</sup>Ca AT 104 MeV

## A. Data and a priori potential

For a demonstration of the method, we "invert" the elastic cross section data of Gils et al.<sup>27</sup> for  $\alpha$  on <sup>40</sup>Ca at 104 MeV, since these authors appear to provide the best available data which, moreover, they have fitted by a potential using the Fourier-Bessel method. These data are shown in Figs. 4 and 5. Gils *et al.*<sup>27</sup> have made a "rough" fit to these with a Woods-Saxon potential with parameters V=151.9 MeV,  $r_{01}=1.407$  fm,  $a_1=1.248$  fm (real form factor squared), and W=20.3 MeV,  $r_{02}=1.607$  fm,  $a_2=0.672$  fm,  $R_i=r_{0i}A^{1/3}$ , plus a double-folding Coulomb potential<sup>27</sup> (this is fairly close to a chargedsphere potential). This rough fit achieves a goodness of fit value f = 3.3

The scattering function calculated with this potential can be very accurately fitted (at M=80 angular momenta  $l_i$ ) by a rational scattering function of the form (2.1) with

$$S^{(0)}(\lambda) = e^{i\eta \ln(\lambda^2 + \lambda_c^2)}$$
(4.1)

corresponding to a Coulomb reference function<sup>15</sup> with  $\eta = 40 \ e^2 / \hbar v = 1.2354, \ \lambda_c = 8\eta$ , and the 4N = 32 real parameters  $\mathbf{a}^{(a)} = \{\alpha_n^{(a)}, \beta_n^{(a)}\}$  given in Table I. (We remark that for convenience two zeros  $\beta_n$  are taken into account in the nonrational form although they should be counted as rational according to Sec. II.)

-9.61602 + i4.31102

	n	$\alpha_n$	$\beta_n$
	1	15.83920 - i 25.21440	24.64080 + i11.58200
Nonrational	2	24.449 70 - <i>i</i> 11.318 20	-18.03700 + i 18.37020
	3	20.10140 - i6.58927	-4.11930 + i 5.64255
	4	- 8.34948- <i>i</i> 7.86954	-21.31370 + i0.42403
	5	-23.03760 - i5.09580	-19.20990 + i1.35463
Rational	6	-10.95910 - i4.89835	-16.781200+i2.22191
	7	-18.684 50 - <i>i</i> 7.709 72	-13.71500 + i 3.17024

-14.862 30 - *i* 6.346 00

TABLE I. A priori parameters (" $\gamma \rightarrow \infty$ ").

The minimum value found,  $\tilde{\chi}_{\min}^2(\gamma) = \tilde{\chi}^2(\sigma; \hat{\mathbf{a}}(\gamma))$ , is plotted as a function of  $\gamma$  in Fig. 1(a); the corresponding functions  $\chi_{\min}^2(\gamma) = \chi^2(\sigma; \hat{\mathbf{a}}(\gamma)), \langle \chi^2 \rangle_{a \text{ post}}$ , and  $4N_{\text{eff}}(\gamma)$  are displayed in Figs. 1(b)–(d), respectively.

It is seen that, unfortunately, the search program does

not always yield a unique minimum; rather, in various re-

gions of  $\gamma$ , different families of solutions yield the lowest

minimum. For the fulfillment of the Philipps condition

(3.17) [cf. Fig. 1(b)] we choose family 2 because family 3

### B. Search for the optimal parameters

Starting with the parameters  $\mathbf{a}^{(a)}$ , with  $\gamma$  fixed at various values between 0 and 10.0, a search is made on the parameters **a** to find the minimum of the function  $\tilde{\chi}^{2}(\sigma;\mathbf{a})$  of Eq. (3.13), using M=140 scattering angles  $\theta_{i}$ . This search is carried out with the help of the program INBAFI,<sup>32</sup> which is based on the Marquardt algorithm.<sup>33</sup> In the search the signs of the real and imaginary parts of the parameters were kept fixed at their *a priori* values.



FIG. 1. (a) The minimum value  $\tilde{\chi}_{\min}^2(\gamma)$  determined by search on the parameters **a** at fixed  $\gamma$ . Three different families are found by varying the starting values of **a** for different  $\gamma$ . (b) The function  $\chi_{\min}^2(\gamma)$  associated with the minimum value  $\tilde{\chi}_{\min}^2(\gamma)$ . The intersection with  $\langle \chi^2 \rangle = 177.7$  yields the "Philipps value"  $\gamma = 1.45$ . (c) The *a posteriori* average  $\langle \chi^2 \rangle_{a \text{ post}}$  associated with the minimum value  $\tilde{\chi}_{\gamma}^2(\gamma)$ . The intersection with  $\langle \chi^2 \rangle = 177.7$  yields the "Turchin value"  $\gamma = 0.195$ . (d) The effective number of parameters  $N_{\text{eff}}(\gamma)$  associated with the minimum value  $\tilde{\chi}_{\gamma}^2(\gamma)$ .

	п	$\alpha_n$	$\beta_n$	
	1	15.668 60 - <i>i</i> 25.595 90	27.32180 + i 10.47610	
Nonrational	2	25.05960 - i11.24840	-18.28120+i18.37290	
	3	19.863 10 - <i>i</i> 6.126 95	-3.31099+i5.34660	
Rational	4	-9.07368-i4.36567	-20.60050+i1.43151	
	5	-25.83360 - i7.48736	-20.29170 + i2.68098	
	6	-19.12520 - i4.02579	-15.52830+i0.59620	
	7	-16.918 40 - <i>i</i> 7.557 36	-12.20950+i2.87075	
	8	-14.48340 - i 5.21621	- 8.25091 + <i>i</i> 4.03462	

TABLE II. Optimal parameters for condition (3.22) ( $\gamma = 1.45$ ).

minimum of  $\tilde{\chi}^2$  related to family 1. This results in the value  $\gamma = 1.45$ . For the fulfillment of the Turchin condition (3.18) [cf. Fig. 1(c)] only family 1 can be used, yielding the value  $\gamma = 0.195$ . The parameters  $\hat{a}(\gamma)$  for these two solutions are given in Tables II and III.

### C. The potentials

The potentials (2.3) calculated from the two sets of parameters are shown in Figs. 2 and 3, together with the errors  $\overline{\Delta V(r)}$  calculated by formula (3.10); the error matrix  $\epsilon_{nm}$  is here of course calculated with the help of Eq. (3.15). According to Fig. 9.2b of Ref. 24, the resulting error bands, or confidence intervals, correspond to a probability content of about 55% (number of parameters: 4N=32). It is seen that the final potentials are significantly shifted away from their *a priori* shape.

The Philipps condition (3.17) yields a smoother potential, with smaller confidence intervals, than the Turchin condition (3.18). The price for this is a slightly worse fit (larger value of f). The fits to the cross section resulting from an inversion using the Philipps and Turchin conditions are shown in Figs. 4 and 5, respectively. To the eye they appear as equally perfect, although f=1.675 for the Philipps condition and f=1.417 for the Turchin condition. If no *a priori* information is used at all ( $\gamma=0$ ) a fit with f=1.35 is obtained, which indicates that the rational scattering function with 4N=32 real parameters is only just adequate statistically to describe the experimental cross section (F = 108).<sup>25</sup> The corresponding unregularized potential, however, has errors which generally attain values of to 100% of the potential itself.

It is seen from Fig. 1(d) that the effective number of real parameters for the regularized fits is  $4N_{\rm eff} \approx 19$  for  $\gamma = 1.45$  and  $4N_{\rm eff} \approx 24$  for  $\gamma = 0.195$ . That is, instead of the 32 real parameters introduced in the rational interpolation, one effectively needs only about 20 independent parameters in order to obtain acceptable fits to the cross section in the light of the assumed *a priori* information.

The regularized inversion of the high-quality  $\alpha$ -<sup>40</sup>Ca scattering cross section data<sup>27</sup> yields smooth potentials with fairly narrow confidence limits. The two potentials obtained by the two different methods of choosing an optimal value of  $\gamma$  are practically the same in the (physically important) surface region, but they differ in the interior. Owing to the absorption in this region, the actual value of the potential is not very important there, however. Nevertheless, the fact that the two potentials differ from one another outside their confidence limits would have to be attributed to a significant difference between the Philipps and Turchin conditions together with a certain (physically irrelevant) rigidity of the rational scattering function analysis (a certain "model dependence" showing up in unimportant parts of the potential, but not in the cross section). The choice of the a priori potential may of course also prejudice certain local  $\tilde{\chi}^2$  minima. In the physically relevant surface region all these ambiguities are kept to a minimum.

	n	$\alpha_n$	$\beta_n$
	1	14.183 80 - <i>i</i> 24.358 90	30.12000 + i6.93593
Nonrational	2	24.580 90 - <i>i</i> 12.936 70	-19.37310+i19.13310
	3	19.237 90 — <i>i</i> 6.852 95	-4.25310+i5.78143
Rational	4	-9.92588-i6.19839	-19.79550+i0.00162
	5	-29.10120 - i 6.14675	-21.72710+i1.76678
	6	-21.93900 - i 3.66432	-15.69840 + i 3.11650
	7	-17.97760-i4.30960	-13.55640 + i0.00112
	8	-12.28170 - i 5.27397	- 0.001 67 + <i>i</i> 3.998 07

TABLE III. Optimal parameters for condition (3.23) ( $\gamma = 0.195$ ).



FIG. 2. The optical potential -V - iW with confidence limits corresponding to probability content 55% for  $\gamma = 1.45$ . Also shown is the *a priori* potential  $-V^{(a)} - iW^{(a)}$ .

Gils et al.<sup>27</sup> have also used their data to derive a potential by the Fourier-Bessel method. They keep the imaginary part of the potential fixed at the a priori Woods-Saxon shape, searching only on the Fourier-Bessel coefficients of the real part. Their fit results in a slightly larger f value than ours, f=2.2, and in a smoother real potential



FIG. 3. Same as Fig. 2 for  $\gamma = 0.195$ .



FIG. 4. Experimental elastic cross section and the rational fit for  $\gamma = 1.45$ .

with narrower confidence limits. In the physically important surface region, both the real and imaginary parts of their potential are deeper than ours by more than their error limits. The small errors in their potential are apparently connected with the "parametric regularization" through the use of a low number (9-12) of free parameters in the Fourier-Bessel method. The discrepancies between the Fourier-Bessel potential and the ones obtained in the present work evidently stem from this difference in the number of free parameters.

#### **V. CONCLUSION**

From the present analysis it emerges that the inversion scheme based on a rational scattering function analysis of the experimental elastic cross section leads to reasonable results. This inversion method is not an analytic pro-



FIG. 5. Same as Fig. 4 for  $\gamma = 0.195$ .

cedure, since it involves a parameter search in going from the cross section to the scattering function. But cross sections can in general be analyzed only by some search procedure; instead of the traditional phase shift search analysis we propose here the somewhat more restricted rational analysis, which involves less parameters, yet is apparently still flexible enough to represent realistic cross sections. The rational analysis automatically yields the corresponding potential, including confidence limits. The *a priori* parameters used in the present work are admittedly introduced *ad hoc* on the basis of reasonableness. A more model independent starting set would indeed be desirable. Further research should also deal with the problem of the occurrence of local minima. The method should facilitate the investigation of the sensitivity of the potential to the cross section in various angular regions.

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