## Note

# Nonlinear-Multiple-Function Simultaneous Least Squares Fitting Procedure 

## Introduction

Sometimes physicists working in experimental Physics are faced with the analysis of data coming from different sources, whose interpretation may follow from quite different physical origins, but with the interesting (and desirable) situation that one or more of the physical parameters used in the analysis are common to various of the experiments. In the present work the analysis is restricted to data coming from two different and independent sources, although the procedure may easily be extended to more general cases.

The standard procedure to analyse two sets of data obtained from two different experiments on the same physical system is to analyse each of them separately and then to compare the similarities or the discrepancies between both sets of results. On physical grounds the common parameters must have the same value and in this work a procedure is outlined as to how this aim may be accomplished by combining both sets of data in a simultaneous treatment. Our motivation was to use all possible physical constraints to determine the common parameters.

In the next section the basic theory is outlined, and in the last section an example is given.

## Theory

The general treatment of nonlinear least-squares fit to a given functional form developed by Deming [1] will be used as the basis for the present procedure. Let us consider two sets of data described by

$$
\begin{align*}
& y=f\left(x, P_{1}, P_{2}\right)  \tag{1a}\\
& z=g\left(x, P_{2}, P_{3}\right), \tag{1b}
\end{align*}
$$

where $x$ is the independent variable and $P_{1}, P_{2}, P_{3}$ are the parameters to be determined, and where $P_{2}$ is common to both descriptions. The data are provided in the form

$$
\begin{array}{ll}
X_{i}, Y_{i}, W_{y i}, & 1 \leqslant i \leqslant N \\
X_{j}, Z_{j}, W_{z j}, & 1 \leqslant j \leqslant M \\
244 &
\end{array}
$$

where $Y_{i}$ and $Z_{i}$ are measured values of the properties $y$ and $z$ when the independent variable takes values $X_{i}$ and $X_{j}$; while $W_{y i}^{\prime}$ and $W_{j j}$ are weights assigned to the corresponding data points ( $X_{i}, Y_{t}$ ) and ( $X_{i}, Z_{j}$ ), respectively.

In general, it is expected that,

$$
\begin{array}{ll}
y_{i}=f\left(X_{i}, P_{1}, P_{2}\right) \neq Y_{i}, & 1 \leqslant i \leqslant N, \\
z_{j}=g\left(X_{i}, P_{2}, P_{3}\right) \neq Z_{j}, & 1 \leqslant j \leqslant M, \tag{30}
\end{array}
$$

and therefore $P_{1}, P_{2}$, and $P_{3}$ are determined to minimize $S$, the weighted sum of the squares of the deviations,

$$
\begin{equation*}
S=\sum_{i=1}^{N} W_{y i}\left(Y_{i}-y_{i}\right)^{2}+\sum_{i=1}^{M} W_{i j}\left(Z_{l}-z_{j}\right)^{2} \tag{4}
\end{equation*}
$$

In order to proceed it is necessary to linearize Eqs. (3a) and (3b) in their dependences on the parameters to be determined, and this is accomplished by means of a Taylor power series expansion of the function given by Eqs. (1a) and (1b), on the parameters about the initial values $P_{01}, P_{02}$, and $P_{03}$. If we define $C_{1}, C_{2}$, and $C_{3}$ by

$$
\begin{align*}
& C_{1}=P_{1}-P_{01}  \tag{5a}\\
& C_{2}=P_{2}-P_{02}  \tag{5b}\\
& C_{3}=P_{3}-P_{03} \tag{5c}
\end{align*}
$$

the Gauss normal system of the equations results,

$$
\begin{align*}
& (1,1) C_{1}+(1,2) C_{2}+(1,3) C_{3}=(1,0)  \tag{6a}\\
& (1,2) C_{1}+(2,2) C_{2}+(2,3) C_{3}=(2,0)  \tag{6b}\\
& (1,3) C_{1}+(2,3) C_{2}+(3,3) C_{3}=(3,0) \tag{60}
\end{align*}
$$

where

$$
\begin{align*}
& (i, j)=\sum_{k=1}^{N} W_{y k} f_{p i}^{k} f_{p i}^{k}+\sum_{n=1}^{M} W_{z n} g_{p i}^{n} g_{p j}^{n} \\
& (i, 0)=\sum_{k=1}^{N} W_{y k} f_{p i}^{k}\left(Y_{k}-y_{k}\right)+\sum_{n=1}^{M} W_{z n} g_{p i}^{n}\left(Z_{n}-z_{n}\right) \tag{7b}
\end{align*}
$$

with

$$
\begin{align*}
& f_{p i}^{k}=\left.\frac{\delta f}{\delta P_{i}}\right|_{P_{1}=P_{10}, P_{2}=P_{20}, x=X_{h}}  \tag{8a}\\
& g_{p i}^{n}=\left.\frac{\delta g}{\delta P_{i}}\right|_{\rho_{2}=P_{20} . P_{3}=P_{30}, x=X_{n}} . \tag{8b}
\end{align*}
$$

The solution of Eqs. (6a)-(6c) gives the corrections to the initial values of the parameters, and from Eqs. (5a)-(5c) we obtain

$$
\begin{align*}
& P_{11}=P_{01}+C_{1}  \tag{9a}\\
& P_{12}=P_{02}+C_{2}  \tag{9b}\\
& P_{13}=P_{03}+C_{3} \tag{9c}
\end{align*}
$$

These values for the parameters are used in turn as initial values and a new iteration is performed to correct the values $P_{11}, P_{12}$, and $P_{13}$ for the parameters given by Eqs. (9a)-(9c). The procedure is iterated until convergence is achieved to the desired degree of approximation, which may be expressed in the form that changes in the parameters and/or in $S$ between two successive iterations that are below a certain arbitrary value.

Once the convergence has been obtained, the values for the parameters providing the best simultaneous fit to Eqs. (1a) and (1b) in the least-squares sense have been determined.

It has been shown by Deming [2] that an estimate of the standard deviation in the $i$ th parameter being fitted is given by

$$
\begin{equation*}
P_{i}=\left[\frac{S}{N+M-n} A_{i i}^{-1}\right]^{1 / 2} \tag{10}
\end{equation*}
$$

where $n$ is the number of parameters being fitted ( 3 in the present case), and $A_{i i}^{-1}$ indicates the element ( $i, i$ ) of the inverse matrix to that of the system of Eqs. (6a)-(6c).

The experimental uncertainties in both dependent variables and in the independent variable are taken into account through the effective weights $[1,3,4]$,

$$
\begin{array}{ll}
W_{y i}=\left[\left(\Delta y_{i}\right)^{2}+\left(f_{x}^{i} \Delta x_{i}\right)^{2}\right]^{-1}, & 1 \leqslant i \leqslant N \\
W_{z i}=\left[\left(\Delta z_{j}\right)^{2} \mid\left(g_{x}^{j} \Delta x_{j}\right)^{2}\right]^{-1}, & 1 \leqslant j \leqslant M \tag{11b}
\end{array}
$$

where $\left(\Delta x_{i}, \Delta y_{i}\right)$ are the experimental uncertainties affecting the data point $\left(X_{i}, Y_{i}\right)$ (similarly for ( $\left.\Delta x_{j}, \Delta z_{j}\right)$ ), and

$$
\begin{align*}
& f_{x}^{i}=\left.\frac{\delta f}{\delta x}\right|_{P_{1}=P_{01}, P_{2}=P_{02}, x=x_{i}}, 1 \leqslant i \leqslant N,  \tag{12a}\\
& g_{x}^{j}=\left.\frac{\delta g}{\delta x}\right|_{P_{2}=P_{02}, P_{3}=P_{03}, x=x_{i}}, 1 \leqslant j \leqslant M . \tag{12b}
\end{align*}
$$

## Example

An example, in which the parameters to be fitted are involved in a strong nonlinear form in Eqs. (1a)-(1b), is that provided by

$$
\begin{align*}
\Delta v_{Q}(T)= & v_{Q}(0)-v_{Q}(T)=\frac{3 T^{4}}{T_{\mathrm{D}}^{3}} \int_{0}^{T_{D} \cdot T} \frac{x^{3}}{\left(e^{x}-1\right)} d x \\
& +\frac{500}{\exp \left(h v_{\mathrm{E}} / k T\right)-1}  \tag{13a}\\
C(T)= & \frac{9 R T^{3}}{T_{\mathrm{D}}^{3}} \int_{0}^{T_{D} T} \frac{x^{4} e^{x}}{\left(e^{x}-1\right)^{2}} d x \\
& +\frac{3 R\left(h v_{\mathrm{E}} / k T\right)^{2} \exp \left(h v_{\mathrm{E}} / k T\right)}{\left[\exp \left(h v_{\mathrm{E}} / k T\right)-1\right]^{2}} \tag{13b}
\end{align*}
$$

where $T$ is the temperature; $h, k$, and $R$ are Planck's, Boltzmann's, and the gas constants, respectively. The parameters to be determined are $T_{\mathrm{D}}$ and $v_{E}$, which are the Debye temperature and the Einstein frequency, respectively, characteristics of the crystalline solid under investigation. Equation (13a) describes a typical temperature dependence of the NQR (nuclear quadrupole resonance) transition frequency measured in $\mathrm{kHz}[5,6]$; while Eq. (13b) describes a typical specific heat expression for six normal modes: 3 acoustics and 3 optic [7].

The data to be fitted was generated by means of Eqs. (13a)-(13b), using $T_{\mathrm{D}}=100 \mathrm{~K}$ and $v_{\mathrm{E}}=100 \mathrm{~cm}^{-1}$, and the value were dispersed in a Gaussianrandom way with standard deviations of 2 kHz and $2 \%$ for $A r_{Q}(T)$ and $C(T)$, respectively. Table I shows one particular set of data and the values produced with the obtained parameters: $T_{\mathrm{D}}=(98.233 \pm 0.0057) \mathrm{K}$ and $v_{\mathrm{E}}=(100.18 \pm 0.12) \mathrm{cm}^{-1}$.

It is interesting to mention that in the case in which the dependence of Eqs. (la)-(1b) on the parameters is a linear one, the first iteration produces the final values. On the other hand, if the parameters are involved in a nonlinear form, several iterations are needed before convergence is achieved. Table II shows the values for the parameters, obtained in the example, after each iteration, starting from different initial guessed values.

The program is written in Basic for the Commodore PET 2001 computer, and a listing is available upon request.

TABLE I
Data Used and Values Determined for the NQR Frequency and the Specific Heat

| $\begin{gathered} T \\ (K) \end{gathered}$ | $\begin{aligned} & \Delta v_{Q \exp } \\ & (\mathrm{kHz}) \end{aligned}$ | $\Delta v_{Q \text { calc }}$ <br> ( kHz ) | $\begin{gathered} C_{\text {exp }} \\ (J / \operatorname{mol} K) \end{gathered}$ | $\begin{gathered} C_{\text {cale }} \\ (J / \mathrm{mol} K) \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: |
| 5 |  |  | 0.2418 | 0.2562 |
| 10 | 0.60 | 0.20 |  |  |
| 15 |  |  | 5.6877 | 5.6781 |
| 20 | 4.30 | 2.81 |  |  |
| 25 |  |  | 15.1800 | 15.4268 |
| 30 | 13.70 | 11.65 |  |  |
| 35 |  |  | 23.4448 | 24.4483 |
| 40 | 26.10 | 28.43 |  |  |
| 45 |  |  | 31.3605 | 31.1789 |
| 50 | 53.50 | 52.04 |  |  |
| 55 |  |  | 34.9928 | 35.8719 |
| 60 | 79.00 | 80.71 |  |  |
| 65 |  |  | 37.7319 | 39.1282 |
| 70 | 111.30 | 113.00 |  |  |
| 75 |  |  | 40.7179 | 41.4272 |
| 80 | 145.60 | 147.89 |  |  |
| 85 |  |  | 43.9971 | 43.0894 |
| 90 | 186.90 | 184.68 |  |  |
| 95 |  |  | 45.7958 | 44.3206 |
| 100 | 224.50 | 222.88 |  |  |
| 105 |  |  | 44.7777 | 45.2534 |
| 110 | 263.90 | 262.17 |  |  |
| 115 |  |  | 44.6634 | 45.9748 |
| 120 | 299.60 | 302.29 |  |  |
| 125 |  |  | 47.5434 | 46.5428 |
| 130 | 347.30 | 343.07 |  |  |
| 135 |  |  | 47.7316 | 46.9973 |
| 140 | 384.20 | 384.37 |  |  |
| 145 |  |  | 46.3083 | 47.3662 |
| 150 | 428.60 | 426.11 |  |  |
| 155 |  |  | 46.3255 | 47.6695 |
| 160 | 471.40 | 468.20 |  |  |
| 165 |  |  | 47.7183 | 47.9217 |
| 170 | 513.20 | 510.59 |  |  |
| 175 |  |  | 47.3500 | 48.1336 |
| 180 | 549.70 | 553.22 |  |  |
| 185 |  |  | 48.7830 | 48.3132 |
| 190 | 594.40 | 596.07 |  |  |
| 195 |  |  | 48.2661 | 48.4667 |
| 200 | 636.90 | 639.09 |  |  |

TABLE II
Values for $T_{D}, v_{\epsilon}$, and $S$ Obtained at Each Iteration for Different Initial Values.

| Iteration | $\begin{gathered} T_{\mathrm{D}} \\ (\mathrm{~K}) \end{gathered}$ | $\begin{gathered} v_{E} \\ \left(\mathrm{~cm}^{-1}\right) \end{gathered}$ | $S$ |
| :---: | :---: | :---: | :---: |
| 0 | $+100.0000$ | $+100.0000$ | 170.980 |
| 1 | + 98.1990 | $+100.1820$ | 39.492 |
| 2 | + 98.2326 | +100.1802 | 39.442 |
| 0 | $+200.0000$ | $+50.0000$ | 700864.220 |
| 1 | -265.4045 | + 77.5762 | 417452.502 |
| 2 | +94.3701 | + 86.4021 | 21390.004 |
| 3 | +97.3702 | + 98.2079 | 404.167 |
| 4 | +98.2131 | $+100.1394$ | 39.596 |
| 5 | + 98.2327 | $+100.1802$ | 39.442 |
| 0 | $+50.0000$ | $+150.0000$ | 329326789 |
| 1 | + 82.4187 | + 73.6757 | 135841.401 |
| 2 | + 92.9680 | + 93.0386 | 6715.532 |
| 3 | + 97.7857 | + 99.6598 | 73.119 |
| 4 | +98.2293 | +100.1775 | 39.443 |
| 5 | + 98.2327 | $+100.1802$ | 39.442 |
| 0 | $+50.0000$ | $+50.0000$ | 1453485.580 |
| 1 | + 75.6493 | + 75.1321 | 149833.034 |
| 2 | + 91.5756 | +93.8813 | 6686.251 |
| 3 | +97.6540 | +99.7895 | 70.473 |
| 4 | +98.2279 | +100.1789 | 39.443 |
| 5 | + 98.2327 | $+100.1802$ | 39.442 |
| 0 | $+150.0000$ | $+150.0000$ | 135509.306 |
| 1 | + 47.4357 | + 77.9977 | 444805.004 |
| 2 | + 79.9207 | + 96.6547 | 25228109 |
| 3 | + 95.1202 | $+100.2570$ | 503.244 |
| 4 | +98.1241 | $+100.1879$ | 39.962 |
| 5 | +98.2324 | $+100.1803$ | 39.442 |
| 6 | +98.2327 | $+100.1802$ | 39.442 |

## Acknowledgments

Partial financial support provided by Consejo Nacional de Investigaciones Cientificas y Témicas (CONICET) and by Consejo de Investigaciones Cientificas y Tecnológicas de la Provincia de Córdoba (CONICOR), both from Argentina, are gratefully acknowledged.

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Received: November 5, 1986; revised March 4, 1987.

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