On the Numerical Determination of Relaxation and Retardation Spectra for Linearly Viscoelastic Materials

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Synopsis

Knowledge of the relaxation spectrum is important because (1) it provides an intrinsic characterization of the mechanical properties for linearly viscoelastic materials and (2) it offers a rational way to derive the coefficients for a Prony or Dirichlet series representation of the relaxation modulus of importance to some engineering analyses. A numerical solution based on Simpson quadrature leads to an unstable solution in the sense that a decrease in integration intervals produces a progressively worse solution which oscillates between positive and negative values. This difficulty may be overcome by requiring that the curvature of the relaxation spectrum with respect to the relaxation times be minimized. The method is tested on the modified power law and good agreement with the exact and numerically determined relaxation spectrum is obtained. However, when the same method is used to determine the retardation spectrum, only the unstable solution is obtained, although the form of the integral equation is the same. This different behavior is attributed to the difference in the characteristics of the relaxation and retardation spectral functions.

Introduction

From the standpoint of polymer physics or polymer chemistry, the relaxation spectrum provides an intrinsic material characterization not modulated by particular laboratory test variations. Furthermore, it may be desirable to relate molecular parameters such as chain composition and conformation to the spectrum rather than a particular response curve.

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An additional need for knowledge of spectral functions may arise in
some engineering problems. While it is not always necessary from an
engineering standpoint to determine the relaxation or creep spectrum
for stress analysis problems, it is at times convenient to characterize
viscoelastic material functions by a finite Prony or Dirichlet series
\[ R(t) = R_\infty + \sum_{n=1}^{N} r_n e^{-t/r_n} \]  
where \( R \) is the response to a step stress or strain and \( R_\infty \) is the re-
sponse at infinite time. Schapery\(^1\) proposed a collocation method
based on minimization of the square error for the determination of the
coefficients \( r_n \) when the characteristic times \( r_n \) are chosen arbitrarily
throughout the transition region of the material. It turns out that
when the number of terms in the series (1) is increased in order to
improve the smoothness of the representation some of the coefficients
\( r_n \) are negative. The dynamic properties derived from such a series
could be physically unrealistic. In such a case, knowledge of the
spectral functions would provide a rational means to derive from it
positive coefficients by a straightforward integration.

Mathematical Preliminaries

Because it is not possible to devise a test for the determination of
the spectra, they must be calculated from material responses mea-
sured in particular test histories. In general, the material response
\( R(t) \) can be represented in the form
\[ R(x) = \int_{0}^{\infty} H(\tau)B(x,\tau)(d\tau/\tau) \]  
where \( H(\tau) \) is the spectral function and \( B(x,\tau) \) is a function of relaxa-
tion time \( \tau \) and time measure \( x \) (time or frequency). Roesler et al.\(^2-4\)
have treated the solution of eq. (2) in terms of Fourier series synthesis
with special application to dynamic viscoelastic responses. Twomey\(^5\)
and Philips\(^6\) have considered the numerical solution of eq. (2) for
the Kernel \( B(x,\tau) = e^{-x/\tau} \). In this case the integral is of the Laplace
inverse of an experimentally determined function. The special
difficulty that arises in this computation is that the solution of the
integral equation is not stable under reasonable perturbations.\(^7\) This
instability manifests itself in progressively worse solutions as one attempts to calculate the function at more and more points in the sense that the solution oscillates with ever wider excursions about the—presumably smooth—exact solution.

In order to arrive at a non-oscillating or metastable solution, it is possible⁵ to solve the integral equation subject to the smoothness constraint that the curvature of the solution with respect to its variable be minimized. To demonstrate this let us write eq. (2) in the form

\[ R(z) = \ln 10 \int_{-\infty}^{\infty} H(\tau(y)) B[x, \tau(y)] \, dy \]  

by virtue of the transformation \( y = \log \tau \). Assuming that \( H(\tau) \) contributes to the integral (3) over a finite domain of \( \tau \),* one may use Simpson quadrature to write eq. (3) as

\[ \sum_{i=1}^{N} W_i K_{ij} H_i = R_j + \epsilon_j \]  

where \( j = \text{integer such that } 1 \leq j < N \), \( R_j = R(z_j) \), \( 1 \leq j < N \); \( K_{ij} = \ln 10 B(x_j, \tau_j) \); \( W_i = \text{Simpson quadrature coefficients } \{1/3, 4/3, 3/3, \ldots, 4/3, 1/3\} \).

In principle, one would like to solve eq. (4) with errors \( \epsilon_j = 0 \). One finds, however, that the inverse \( [A]^{-1} \) of the matrix

\[ [A] = A_{ij} = W_i K_{ij} \]  

becomes more ill-conditioned as its size \( N \) increases if \( B(x, \tau) \) has exponential character. The ill condition manifests itself in that its elements alternate in size and become very large, leading, in the inversion process, to small differences of large numbers. In terms of the solution \( H_i \) this means that small changes in the vector components \( R_j \) may result in large changes in the resultants \( H_i \), by virtue of the relation

\[ H_i = A_{ij}^{-1} R_j \]  

To improve this situation we define the square relative error

\[ \epsilon = \sum_{j=1}^{N} (\epsilon_j/R_j)^2 \]  

* Practically speaking, it has been found that approximately two decades to either side of the transition region are sufficient.
and the second-order difference

\[ C = \sum_{i=1}^{N} (H_{i+1} - 2H_i + H_{i-1})^2 \]  

(8)

as a measure of the curvature of \( H(\tau) \) with respect to \( \tau \). It will be recalled that \( \epsilon = \epsilon(H_i) \) is related to the \( H_i \) by eq. (4). It would be desirable to minimize the square error \( \epsilon \) with respect to the numbers \( H_i \). If the curvature of the function \( H(\tau) \) were known, this would lead to the well-known variational problem with a constraint which can be treated by the method of Lagrange multipliers. Since the curvature is not known, one might still expect to obtain satisfactory results by minimizing the linear combination

\[ F(H_i) = \epsilon(H_i) + \lambda C(H_i) \]

(9)

and treat \( \lambda \) as an unknown constant to be chosen more or less arbitrarily rather than through prescription of the curvature constraint. In effect one minimizes thus a linear combination of the square error and the local curvature. * Under consideration of eq. (4) and the definition (5), the minimization of eq. (9) namely

\[ \frac{\partial F(H_i)}{\partial H_i} = 0 \quad j = 1, 2, 3, 4, \ldots N \]

(10)

leads, upon using eqs. (4), (3), (7), and (8), to the set of linear equations

\[ \sum_{i=1}^{N} \left\{ \sum_{j=1}^{N} \left( A_{jk}A_{ik}/R_j^2 \right) + \lambda S_{ik} \right\} H_i = \sum_{j=1}^{N} A_{jk}/R_j \]

(11)

where \( S_{ki} \) represents the symmetric banded matrix

\[
\begin{array}{cccccccc}
1 & -2 & 1 & 0 & 0 & 0 & . & 0 \\
-2 & 5 & -4 & 1 & 0 & 0 & . & 0 \\
1 & -4 & 6 & -4 & 1 & 0 & . & 0 \\
0 & 1 & -4 & 6 & -4 & 1 & . & 0 \\
\vdots & & & & & & & \vdots \\
0 & 0 & 0 & 1 & 4 & 6 & 4 & 1 \\
0 & 0 & 0 & 0 & 1 & -4 & 5 & -2 \\
0 & 0 & 0 & 0 & 0 & 1 & -2 & 1 \\
\end{array}
\]

* Some further improvements may be obtained if for smooth functions \( H(\tau) \) the curvature is treated less locally by writing

\[ C = \sum_{i=1}^{N} (H_{i+1} - 2H_i + H_{i-1})^2 \]
This may be written more compactly in matrix form as

\[ |M| |H| = |V| \]  

with

\[ M_{ik} = A_{ik} \sum_{j=1}^{N} \left( A_{jk}/R_j \right) + \lambda S_{ik} \]

\[ V_k = \sum_{j=1}^{N} A_{jk}/R_j \]

which equation has the formal solution

\[ |H| = |M|^{-1} |V|. \]  

**Sample Calculations for the Relaxation Modulus**

In order to test the solution method, it is necessary to obtain an exact solution of a typical response function for comparison. For this purpose we have chosen the function

\[ R(t) = \frac{(E_g - E_e)}{1 + t/\tau} = \int_0^\infty H(\tau) e^{-t/\tau} d\tau \]  

which equation has as the solution the modified power law for the relaxation spectrum

\[ H(\tau) = \frac{E_g - E_e}{\Gamma(n)} \left( \frac{\tau}{\tau_0} \right)^n e^{-\tau_0/\tau} \]  

with \( E_g \) and \( E_e \) as the glassy and equilibrium modulus, respectively.

In Figure 1 we show a comparison of results using eq. (14) as the test data at \( N = 43 \) points to calculate \( H(\tau) \) at \( N = 39 \) points, for different values of the smoothing factor \( \lambda \). If the curvature constraint is eliminated \( (\lambda = 0) \) so that the procedure reduces to one of minimizing the square error without any curvature constraint, the solution oscillates considerably while even a small value of \( \lambda \) improves the solution considerably.*

Another factor which influences the solution is the range of relaxation times. This range must be large enough to include the range over which the spectrum is essentially nonzero, yet one may not be too liberal in this choice, particularly at small values for \( \tau \). The

* The values of \( \lambda \) are so small here because all the data is normalized by the relaxation modulus values \( R_i = R(\tau_i) \).
Fig. 1. Comparison of relaxation spectrum for different smoothness constrains $\lambda$. 
reason for this limitation is the behavior of the inverse of the matrix 
[M], eq. (13). In the region of small \( \tau \) the function \( R(t) \) has a plateau 
and the problem becomes one of determining locally the solution for \( H(\tau) \) which renders \( R(t) \) nearly a constant. This solution is very 
sensitive to round-off errors in \([M]^{-1}\) and the accuracy of the pre-
scribed values of \( R(t) \). Indeed, in the case of experimentally deter-
dined functions \( R(t) \) it may be necessary to locally smooth the data 
numerically.

Figure 2 gives a comparison of the numerical solution (\( \lambda = 10^{-6} \)) 
with the exact one along with two approximations of the Widder-
Post formula. Note that the approximate solutions are close to the 
exact one to the right of the maximum where the numerical solution 
is locally less adequate though close on the average. The latter 
compares very well with the exact solution near the peak, whereas 
the approximate solutions achieve peak values considerably lower 
than either of the other solutions. It appears thus that a combina-
tion of the two solution methods might give good results. A calcu-
lation of the error as presented in Figure 3 shows further that the 
umerical solution has a nearly vanishing mean error while the two 
approximations have small but constant errors.

Application to Creep Compliance

The creep compliance may be written in spectral form as

\[
D_{\text{erp}}(t) = D_\varepsilon + \int_0^\infty L(\tau)(1 - e^{-i\tau}) \, d\tau/\tau
\]  

or alternately

\[
R'(t) = D_\varepsilon - D_{\text{erp}}(t) = \int_0^\infty L(\tau)e^{-i\tau} \, d\tau/\tau
\]  

with \( D_\varepsilon = D_\varepsilon + \int_0^\infty L(\tau) \,(d\tau/\tau) \) as the long-time elastic compliance 
and \( D_\varepsilon \) the glassy compliance. Equation (17) is of the same form 
as eq. (14) and could therefore, in principle, be solved in the same 
way as before. As a first test case we have chosen the mathematical 
relation

\[
D_{\text{erp}}(t) = 1/E_{\text{rel}}(t)
\]
which, physically speaking, is only an approximation, $E_{rel}(t)$ being given by the power law (14)

$$E_{rel}(t) = E_x + R(t).$$

The solution to eq. (17) is shown as a strongly unstable, oscillating function in Figure 4; no variation of the smoothness coefficient $\lambda$
Fig. 3. Error in relaxation spectrum as determined numerically and by the Widder-Post method.

changed the character of the solution. A second attempt was made using the measured properties for the equivoluminal composition of Solithane 113° with the same result of an unstable solution. Inasmuch as the integral equations (14) and (17) are the same and the function $R(t)$ decreases in both cases monotonically with time in a very similar manner, the distinction between solvable and unsolvable equations may be rather fine. It is generally accepted that relaxation spectra increase rapidly with $\tau$ and decrease slowly after the peak is passed whereas the converse is true for the retardation spectrum. From the standpoint of the cited experiments with the numerical inversion scheme presented here, it appears that the determination
of the relaxation spectra is feasible and is not feasible if the spectrum has the characteristics of the retardation spectrum.

**Conclusion**

We have attempted to automate the determination of spectral distribution functions from experimental data using relatively simple numerical techniques. Although the straightforward numerical solution of the singular integral equations always leads to unstable solutions the introduction of a smoothness constraint leads to acceptable solutions if the spectra are of the relaxation type but not if they are of the retardation type. While the quality of the numerical solution may not be much better than that of standard approximate methods, it should be borne in mind that the current method lends itself, where applicable, to routine calculation and circumvents to a large degree the guesswork and subjective interpretation associated with other approximate methods.

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References


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