Integral transform methods: a critical review of various kernels

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Abstract

Some general remarks about integral transform approaches to response functions are made. Their advantage for calculating cross sections at energies in the continuum is stressed. In particular we discuss the class of kernels that allow calculations of the transform by matrix diagonalization. A particular set of such kernels, namely the wavelets, is tested in a model study.
I. INTRODUCTION

Integral transform methods are a very useful tool in all fields of physics, in experiment as well as in theory. In fact one often encounters situations where an observable is not directly accessible, while this is the case for one of its integral transforms. If one denotes as \( f(x) \) the function representing an observable of interest, its integral transform is defined as

\[
g(\sigma_1, \sigma_2, ...) = \int dx f(x) K(x, \sigma_1, \sigma_2, ...).
\]

Depending on the kernel \( K \) it may happen that, while one is not able to access \( f(x) \), there is some reason why it is much easier to have access to \( g(\sigma_1, \sigma_2, ...) \). For example in an experiment the window of a detector may represent the kernel \( K \), since the observer does not have a direct access to the observable itself, but to the result of its folding with that particular window. If one wants to know the original observable one has to unfold the data, namely one has to invert the transform.

In theoretical physics there are many examples of integral transform applications, ranging from QCD, to nuclear physics, to condensed matter physics. In most cases \( f(x) \) is a quantum mechanical quantity involving continuum states. It is well known that such states are the most difficult objects to access theoretically. For example in non relativistic quantum many-body systems, knowing a continuum state implies to solve the many-body body scattering problem, which has a viable solution in a very limited number of cases. Therefore scattering cross sections or spectra involving more than three or four particles are often inaccessible in a direct way, namely by first calculating the wave functions entering the dynamical matrix elements between initial and final states. However, for some particular kernels, the calculation of an integral transform of such cross sections/spectral functions is possible. One then remains with the task of extracting as much information as possible about \( f(x) \), namely one is left with the inversion problem.

The inversion of integral transforms, when they are affected by errors of any nature, suffers from the well known ill-posed problem, namely the inversion may generate even very different functions whose transforms lie within the error uncertainty. Various algorithms deal with this problem. Among them the so called regularization methods, as well as methods based on Bayes' theorem, like the Maximum Entropy Method. The success of an inversion algorithm, however, also depends on the form of the kernel. For example, kernels that are representations of the \( \delta \)-function, like in the Lorentz integral transform (LIT) method \[1\],
are very powerful. The reason is that the width of the $\delta$-function representation allows to concentrate as much information as possible in the transform, facilitating in this way the task of recovering that information in the inversion process.

The LIT method \[1\] has been largely and successfully applied to electro-weak interactions with few-nucleon systems. In view of an evolution and an extension of this method, we would like to discuss here which kinds of kernels satisfy the two essential conditions for getting sufficient knowledge on $f(x)$, i.e. i) the calculability of the transform and ii) the stability of its inversion. (Such a discussion applies equally well to any other few/many-body system of different nature).

II. INTEGRAL TRANSFORMS VIA DIAGONALIZATION METHODS

As was already said above using integral transform approaches is particularly advantageous when the calculation of an observable of interest requires the knowledge of continuum states. Since several accurate methods have been devised for bound-states of few/many-body systems (see \[2\] for a review) it is interesting to ask which are the conditions that a kernel has to fulfill in order to allow the transform to be calculated by bound state methods. Among them we will focus in particular on the methods based on the diagonalization of the Hamiltonian represented on finite norm basis functions.

We start identifying $f(x)$ with the excitation spectrum/response function $S(\omega)$ of a system perturbed by an external probe:

$$S(\omega) = \sum_n |< n|\mathcal{O}|0>|^2 \delta(\omega - E_n + E_0) , \tag{2}$$

where $\mathcal{O}$ and $H$ are excitation operator and Hamiltonian of the system, respectively ($H|n> = E_n|n>$). An integral transform of $S(\omega)$ with a kernel $K(\omega, \sigma)$ is ($\sigma$ may indicate one or more parameters)

$$\Phi(\sigma) = \int d\omega S(\omega) K(\omega, \sigma) . \tag{3}$$

Substituting $S(\omega)$ in (2), integrating in $d\omega$ and using the Schrödinger equation and the completeness property of the Hamiltonian eigenstates, ($\sum_n |n><n| = I$) the transform can be expressed as

$$\Phi(\sigma) = <0|\mathcal{O}^\dagger K(H - E_0, \sigma) \mathcal{O}|0> . \tag{4}$$
If the kernel is a positive definite function of $\omega$ one can split it into $k\dagger k$, therefore
\[ \Phi(\sigma) = \langle 0|\mathcal{O}\dagger k\dagger(H - E_0, \sigma)k(H - E_0, \sigma)\mathcal{O}|0\rangle > \equiv \langle \tilde{\Psi}|\tilde{\Psi} > . \] (5)

If the transform $\Phi(\sigma)$ exists, namely the integral in (3) is finite and $\mathcal{O}|0\rangle$ has a finite norm (which is the case for all physical excitation operators) one can expand $|\tilde{\Psi} > e < \tilde{\Psi}|$, as well as $\mathcal{O}|0\rangle$ and $< 0|\mathcal{O}\dagger$, on a complete set of finite norm functions ($|l>, |m>, |p>$). Therefore
\[ \Phi(\sigma) = \sum_{l,m,p} < 0|\mathcal{O}\dagger|m><m|k\dagger(H - E_0, \sigma)|l><l|k(H - E_0, \sigma)|p><p|\mathcal{O}|0\rangle > . \] (6)

In this way the kernel is a function of the matrix elements $H_{ml} = < m|H|l >$ or $H_{tp} = < l|H|p >$ of the Hamiltonian represented on those basis functions. After diagonalizing the matrix, $\Phi(\sigma)$ becomes:
\[ \Phi(\sigma) = \sum_{\mu} < 0|\mathcal{O}\dagger|\mu><\mu|k\dagger(\epsilon_{\mu} - E_0, \sigma)|\mu><\mu|k(\epsilon_{\mu} - E_0, \sigma), \sigma|\mu><\mu|\mathcal{O}|0\rangle > . \] (7)

Recombining $k\dagger(\epsilon_{\mu} - E_0, \sigma)k(\epsilon_{\mu} - E_0, \sigma)$ into $K(\epsilon_{\mu} - E_0, \sigma)$ one has
\[ \Phi(\sigma) = \sum_{\mu} K(\epsilon_{\mu} - E_0, \sigma)| < \mu|\mathcal{O}|0\rangle|^2 . \] (8)

Summarizing: it is possible to calculate integral transforms of spectra/response functions, also to continuum, via diagonalization of the Hamiltonian represented on finite norm basis functions, provided that the following conditions are satisfied: i) the integral transform exists, namely $\Phi(\sigma) < \infty$; ii) the 0th moment of $S(\omega)$ exists, namely $\Phi(\sigma) = \int d\omega S(\omega) < \infty \ (\Rightarrow \Phi(\sigma) = < 0|\mathcal{O}\dagger\mathcal{O}|0 > )$; iii) the kernel is a positive definite function of $\omega$ or a linear combination of them.

The convenience of the kernel depends primarily on condition i) (e.g. the moment kernel $K(\sigma, \omega) = \omega^\sigma$, with $\sigma$ integer, may fulfill that condition only for a very limited number of $\sigma$ values), but also by the possibility to invert the transform reliably. In fact there are some kernels for which the inversion of the transform is a seriously ill-posed problem. A well known example is the Laplace kernel $exp[-\sigma\omega]$, widely used in imaginary-time-Monte-Carlo calculations. Another example is $K(\sigma, \omega) = (\omega - \sigma)^{-1}$, suggested in [3] and known as the Stieltjes kernel, which, however, has turned out to be very useful for direct calculations of polarizabilities [4], quantities that also contain the physics of the continuum.
III. A SPECIAL CLASS OF KERNELS: WAVELETS

A special class of kernels, satisfying the iii) condition expressed after Eq. (8), is represented by the wavelets [5]. These kernels combine the advantages of the Fourier analysis with the property of focusing on different ranges of $\omega$, since they are defined on finite intervals. In this respect they present the same advantage as the Lorentzian kernel, whose two parameters rule the position and width of the window-like kernel. Moreover, some of them form a complete set of orthogonal functions, allowing a straightforward and controllable inversion expression for the transform.

In order to get familiar with these kernels, and in view of the possibility to use them in physical cases, we have chosen to investigate the performances of a particularly simple class of them, i.e. the Haar wavelets. They have been the first kind of wavelets to be introduced in 1909 by Alfred Haar [6] as an example of a numerable orthonormal system of $L_2$ functions. They are generated by the mother wavelet defined by:

$$\psi(z) = \begin{cases} 
1 & \text{if } 0 < z < \frac{1}{2} \\
-1 & \text{if } \frac{1}{2} < z < 1 \\
0 & \text{elsewhere}
\end{cases} \quad (9)$$

Therefore, denoting with $\Omega$ the energy range of interest, the wavelet kernel is given by

$$K_W(\omega, k, j) = 2^j \left[ \Theta \left( \omega - \frac{k}{2^j} \Omega \right) + \Theta \left( \omega - \frac{k + 1}{2^j} \Omega \right) - 2 \Theta \left( \omega - \frac{2k + 1}{2^{j+1}} \Omega \right) \right], \quad (10)$$

where $\Theta$ is the Heaviside function and $k$ and $j$ are integer numbers determining the position and width of the wavelet, respectively ($j = -\infty, \ldots, -1, 0, 1, \ldots, \infty$ and $k = 0, 1, 2 \ldots k_{\max}$ with $k_{\max} = 2^j - 1$). One can easily verify that the Haar wavelets fulfill the zero-mean condition and are orthonormal. Some of them are plotted in Fig. [1]. With such a kernel $\Phi(k, j)$ becomes a function of integer variables, and because of the orthonormality relation of the wavelets, the inversion is simply obtained by

$$S(\omega) = \sum_{j} \sum_{k=0}^{k_{\max}} \Phi(k, j) K_W(\omega, k, j). \quad (11)$$

In the following we show the quality of the reconstruction of a model $S(\omega)$, starting from the knowledge of its wavelet transform. For the test we have chosen an $S(\omega)$ resembling a physical spectrum characterized by a narrow resonance and a larger peak, a situation which
FIG. 1: Left: $K_W(\omega, k, j)$ at fixed $k$ for various $j$. Right: $K_W(\omega, k, j)$ at fixed $j$ for various $k$

is often problematic for common inversion methods. It is instructive to see the quality of the inversion at different values of $j_{\text{max}}$ in Fig. 2a and 2b. The values of the index $j$, in fact, represent the various resolution powers of the wavelets. Therefore fixing $j_{\text{max}}$ corresponds to fix the maximum resolution power of the wavelet transform. One can notice that a maximum resolution slightly smaller than the width of the resonance allows a rather good description of both the resonance and the larger peak (of course a much courser resolution is enough to reproduce the larger peak only).

We have also studied the propagation of both systematic and random errors in the inversion procedure. Eq. (11), a consequence of the orthonormality of the wavelets, ensures that large instabilities cannot be generated. The steps in the reconstructed $S(\omega)$ are the consequence of using discrete wavelets. We intend to use also continuous wavelets, even if they have the drawback to be in general non-orthonormal. More important will be the investigation of the realistic case, when $\Phi(k, j)$ is calculated by diagonalization. As Eq. (8) shows, depending on the basis functions used to represent $H$ (see discussion in [6]) and on the matrix size, $\Phi(k, j)$ contains the information on the dynamics only via a discrete set of eigenvalues $\epsilon_\mu$. Their density in the range $\Omega$ determines the maximum resolution one can use. In this sense, one has again, as for the Lorentz kernel, a controlled resolution prediction of $S(\omega)$, namely one will be able to predict only structures comparable or larger than the average distance among the various $\epsilon_\mu$. The ideal case will be when this coincides with the experimental resolution.
FIG. 2: Model $S(\omega)$ (black curve) and the result of the inversion of the wavelet transform (red curve) for $j_{\text{max}} = 4$ (a) and $j_{\text{max}} = 5$ (b) in the sum of Eq. (11).

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