Research Article

Numerical Solvability and Solution of an Inverse Problem Related to the Gibbs Phenomenon

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We report on the inverse problem for the truncated Fourier series representation of \( f(x) \in BV(-L, L) \) in a form with a quadratic degeneracy, revealing the existence of the Gibbs-Wilbraham phenomenon. A new distribution-theoretic proof is proposed for this phenomenon. The paper studies moreover the iterative numerical solvability and solution of this inverse problem near discontinuities of \( f(x) \).

1. Introduction

This paper reinvestigates the Fourier series representation

\[
g(x) = \frac{a_0}{2} + \sum_{m=1}^{\infty} a_m \cos \frac{\pi}{L} x + b_m \sin \frac{\pi}{L} x,
\]

of a piecewise continuous 2L-periodic signal \( f(x) \in BV(-L, L) \), with

\[
a_0 = \frac{1}{L} \int_{-L}^{L} f(x) \, dx,
\]

\[
a_m = \frac{1}{L} \int_{-L}^{L} f(x) \cos \frac{\pi}{L} x \, dx,
\]

\[
b_m = \frac{1}{L} \int_{-L}^{L} f(x) \sin \frac{\pi}{L} x \, dx.
\]

The Gibbs phenomenon (see, e.g., [3]) is a statement of the fact that the infinite series \( g(x) \) tends to overshoot the positive corner of a discontinuity of \( f(x) \) by \( \sim 9\% \) of the jump size and to undershoot the negative corner by the same amount. An overshoot/undershoot effect that is accompanied with spurious oscillations “ringing” near the discontinuity (as described in the appendix) when the series in (1) is truncated at \( m = N \). However, according to a theorem by Fejer [1], the infinite series \( g(x) \) of a \( BV(-L, L) \) function converges to \( f(x) \) at each point \( x \) of continuity of \( f(x) \).

This shortcoming in the infinite Fourier series representation of piecewise continuous \( f(x) \) was first observed by H. Wilbraham in 1848 and then analyzed in detail [4] by J. W. Gibbs in 1898. The main reasons of the Gibbs-Wilbraham effect are that (i) not all frequencies (only integer ones) are employed in (1), (ii) \( a_m \) and \( b_m \) happen to decay slowly with increasing \( m \), and (iii) the global nature of the approximation of \( f(x) \): the expansion coefficients are obtained, via (2), by integration over the entire period, including the points of discontinuity.

What is unpleasant though, with all of this, is that the Gibbs-Wilbraham effect is generic and is present for any periodic signal \( f(x) \in BV(-L, L) \) with isolated discontinuities. The presence of this effect can in fact lead to quite negative consequences when single infinite Fourier series, multiple infinite Fourier series, or even infinite wavelet series are employed to approximate signals of various dimensions, in many fields such as radio engineering and signal transmission.

The Gibbs-Wilbraham effect can nevertheless have both positive and negative consequences in different applications. The negative consequences call for Gibbs effect reduction, and this can in principle be achieved with the use [4] of a variety of filters. This effect can also be reduced theoretically and for all purposes. In addition to classical mathematical filters, recently (in 2011) Rim and Yun [5] defined a kind of spectral series to filter off completely the Gibbs effect near
a discontinuity. The construction of this series is based on the method of adding the Fourier coefficients of a Heaviside function to the given Fourier partial sums.

This paper is organized as follows. After this introduction, we propose in Section 2 a new distribution-theoretic proof for the Gibbs-Wilbraham effect. In Section 3, we advance the inverse problem of the truncated Fourier series and the new stereographic truncated Fourier series and its relation to resolution of the undershoot/overshoot pair associated to the Gibbs-Wilbraham effect. Here we illustrate how solving the inverse problem for the truncated representation (1) contains a quadratic degeneracy that indicates the existence of the Gibbs phenomenon. Section 4 deals with the numerical iterative solution of the previous inverse problem and its convergence, with a proof and demonstration that stereographic projection does not affect such a numerical solution and its posedness.

2. Distribution-Theoretic Proof for the Gibbs-Wilbraham Effect

Let \( f(x) \in BV(-L, L) \) be a \( 2L \)-periodic piecewise continuously differentiable function on \((-L, L)\), and let \( x_0 \) be a point in \((-L, L)\) at which \( f(x) \) has a discontinuity of the first kind. Consider \( x_i = x_0 - i, x_i^+ = x_i + 0 \) and \( f_i^\pm = f(x_i^\pm) \), to assume, without loss of generality, that \( f_i^+ > f_i^- \) and define the discontinuity jump by

\[
\Delta_i = f_i^+ - f_i^-.
\]

Moreover,

\[
f_c(x) = \begin{cases} 
  f(x); & x \neq x_i \\
  \text{undefined}; & x = x_i.
\end{cases}
\]

Theorem 1. Let a \( 2L \)-periodic \( f(x) \in BV(-L, L) \) be discontinuous at \( x_i \). The Fourier series representation \( g(x) \) of \( f(x) \) converges over \((-L, L)\), as \( N \to \infty \) so as,

\[
g(x) = \begin{cases} 
  f_i^- + \rho \Delta_i; & x = x_i^-\\
  \frac{1}{2}(f_i^- + f_i^+); & x = x_i \\
  f_i^+ + \rho \Delta_i; & x = x_i^+ \\
  f_c(x); & x \neq x_i \text{ or } x_i^+.
\end{cases}
\]

where \( \rho = u(0) \), an uncertainty.

Proof. According to distribution theory [6] when \( g(x) \) coincides a.e. with \( f(x) \) over \((-L, L)\), it may always be differentiated; namely,

\[
g'(x) = f'_c(x) + \Delta_x \delta(x - x_i),
\]

with \( \delta \) as Dirac's delta. Cancellation of this differentiation is performed by sweeping the \( x \)-axis during a compensating integration from \(-L\) to \( L\). Indeed, integration of (6) first from \(-L\) to \( x \) leads to

\[
\int_{-L}^{x} \left[ g'(r) - f'_c(r) \right] dr = \Delta_x \int_{-L}^{x} \delta(r - x_i) dr,
\]

and since \( g(-L) = f_c(-L) \), then

\[
g(x) = f_c(x) + \Delta_x u(x - x_i), \tag{8}
\]

with \( u \) as Heaviside's unit step function.

By a theorem of Fejer [1], \( g(x) \) should converge to \( f_c(x) \) when \( N \to \infty \), \( \forall x \neq x_i \), on \((-L, L)\). Hence the term \( \Delta_x u(x - x_i) \) in (8) can have a compact support only on an infinitesimal interval \( \sigma = \lim_{N \to \infty} \alpha_N = 0 \) to the right of \( x_i \), that is, ending at \( x_i^+ \). Hence

\[
g(x) = f_c(x) + \Delta_x u(x_i^+ - x_i). \tag{9}
\]

Furthermore, despite the fact that \( u(x_i^+ - x_i) = 1, u(x_i^+ - x_i) = u(0) \), which is an uncertainty to be denoted by \( \rho \). Consequently, relation (9) is representable as

\[
g(x) = \begin{cases} 
  f_c(x); & x \neq x_i \\
  f_c(x) + \rho \Delta_i; & x = x_i^+.
\end{cases}
\]

Next complete the sweeping by integrating (6) from \( x \) to \( L \); namely,

\[
\int_{x}^{L} \left[ g'(r) - f'_c(r) \right] dr = \Delta_x \int_{x}^{L} \delta(r - x_i) dr. \tag{11}
\]

Here again since \( g(L) = f_c(L) \), then the left-hand side of (11) becomes \(-g(x) + f_c(x)\). The right-hand side requires standardization by means of the substitution \( r = -y, \) leading to \( dr = -dy \), \([x, L]\) becoming \([-x, -L]\) and

\[
\int_{x}^{L} \delta(r - x_i) dr = -\int_{-x}^{-L} \delta(-y - x_i) dy = -\int_{-x}^{-L} \delta(y + x_i) dy = \int_{-L}^{x} \delta(y + x_i) dy = u(x + x_i).
\]

It is obvious then that (11) is reducible to

\[
g(x) = f_c(x) - \Delta_x u(-x + x_i). \tag{13}
\]

Repeated application of the same arguments employed in the derivation of (9)-(10) to (13) allows for writing it as

\[
g(x) = f_c(x) - \Delta_x u(x_i^+ + x_i), \tag{14}
\]

and subsequently in the form

\[
g(x) = \begin{cases} 
  f_c(x); & x \neq x_i \\
  f_c(x) - \rho \Delta_i; & x = x_i^-.
\end{cases}
\]

Taking into consideration that \( f_c(x_i^+) = f_i^+ \) and \( f_c(x_i^-) = f_i^- \) in (10) and (15), respectively, then combining these equations with Dirichlet's theorem [1] leads to the required result. \( \square \)
It should be pointed out that in the previous proof relations (10) and (15) say nothing specifically about the behavior of \( g(x) \) at \( x^+_m \) and \( x^-_m \), respectively, while asserting the existence of a spiky behavior of \( g(x) \) individually at \( x^+_m \) then at \( x^-_m \). Moreover, the uncertainty \( \rho \) had been resolved in the past (by Wilbraham; see also the Appendix) computationally to satisfy \( \rho = 0.09 \) and long before the advance of the theory of distributions (generalized functions) by S. L. Sobolev in 1936. The number 0.09 for \( \rho \) remains until now, however, a mysterious theoretical puzzle that calls for a rigorous distribution-theoretic justification.

Clearly, in the neighborhood of \( x_m \), \( g(x) \) can only be a rudimentary approximation to \( f(x) \), even when \( N \rightarrow \infty \), with an added spike pair

\[
Q_s(x) = \begin{cases}
+\rho \Delta, & x = x^+_m \\
-\rho \Delta, & x = x^-_m \\
0, & x \notin \{x_m\}
\end{cases} \tag{16}
\]

\( Q_s(x) \) to \( f^s \) at \( x^+_m \). Relation (16) represents the Gibbs-Wilbraham effect which, due to the fact that \( |x^+_m - x^-_m| = 0 \), is not practically computable. This effect degenerates, however, when \( N \) is finite, to an undershoot-overshoot pair \( Q_{\text{u.o}}(x) \) on an interval \( I_{\text{u.o}} = [x_m - \sigma_N, x_m + \sigma_N] \), split by \( \Delta \), at \( x_m \). In this case, it is always possible to compute

\[
g(x) = \begin{cases}
\sim f(x) + Q_{\text{u.o}}(x); & x \in I_{\text{u.o}}, \quad N < \infty \\
\sim f(x); & x \notin I_{\text{u.o}}. \tag{17}
\end{cases}
\]

Furthermore, since \( \lim_{N \to \infty} Q_{\text{u.o}}(x) = Q_s(x) \), then these computations with (17) over \( I_{\text{u.o}} \) can reveal only a feature \( Q_{\text{u.o}}(x) \), associated with the Gibbs-Wilbraham effect \( Q_s(x) \) but not the effect itself.

3. The Inverse Problem for Truncated Fourier Series

Substitution of any value of \( x \in [-L, L] \) in (1) yields a number \( c = g(x) \), which may differ [2] from \( f(x) \) by \( Q_{\text{u.o}}(x) \) if \( x \) is in the neighborhood of a discontinuity of \( f(x) \). Obviously \( c \) satisfies, the nonlinear in \( x \), trigonometric series equation

\[
\sum_{m=1}^{N} \frac{G(y, x, m)}{L} = \sum_{m=1}^{N} a_m \cos \frac{m\pi}{L} x + b_m \sin \frac{m\pi}{L} x \quad \Rightarrow \quad \sum_{m=1}^{N} a_m \cos \frac{m\pi}{L} x + b_m \sin \frac{m\pi}{L} x = c - \frac{a_0}{2}. \tag{18}
\]

which is, however, linear in the data \( y \) = \( \{a_m, b_m\}_{m=1}^{\infty} \).

**Definition 2** (the inverse problem for the truncated Fourier series representation). Given a number \( c = g(x) \) satisfying (18), what is the corresponding \( x \)?

This inverse problem happens to be a discretized version of the similar deconvolution problem [4] of \( x \) from a given number \( q \) in a singular first kind integral equation of the form

\[
\int_{0}^{\infty} K(y, x, \tau) \, d\tau = q, \tag{19}
\]

in which \( y \) represents the data. It is well known, additionally (see, e.g., [7, 8]) that this deconvolution problem can be computationally ill-posed if an arbitrarily small deviation of the \( y \) data may cause an arbitrarily large deviation in the solution \( x \). This situation may further prevail when the operator \( \int_{0}^{\infty} K(y, x, \tau) \, d\tau \) does not continuously depend on \( y \).

Moreover, the inverse problem of the previous definition has a unique solution that satisfies (19) and is expected to accept a double root when the horizontal line \( h(x) = c > f(x) \) intersects with a sharp peak (overshoot) of a truncated \( g(x) \). We shall call this feature a duplicated (quadratic) degeneracy of the solution for this inverse problem.

The existence of the Fourier series truncation peak (and the corresponding two roots for \( x \)) is not directly evident from the structure of \( G(y, x, m) \), and despite its linearity in the data \( y = \{a_m, b_m\}_{m=1}^{\infty} \).

3.1. The Tangent Half-Angle Stereographic Projection. The tangent half-angle substitution (the Euler-Weierstrass sub-stitution [9]) \( \tan(\theta/2) \) is widely used in integral calculus for finding antiderivatives of rational expressions of circular function pairs (cos \( \theta \), sin \( \theta \)). Geometrically the construction goes like this: draw the unit circle, and let \( P \) be the point \((-1, 0)\). A line through \( P \) (except the vertical line) passes through a point \((\cos(\theta), \sin(\theta)) \) on the circle and crosses the \( y \)-axis at some point \( t = \tan(\theta/2) \). Obviously \( t \) determines the slope of this line. Furthermore, each of the lines (except the vertical line) intersects the unit circle in exactly two points, one of which is \( P \). This determines a function from points on the unit circle to points on a straight line. The circular functions determine therefore a map from points \((\cos(\theta), \sin(\theta)) \) on the unit circle to slopes \( t \). In fact this is a stereographic projection [9] that expresses these pairs in terms of one variable: \( \tan(\theta/2) = t \).

Numerical algorithms employing these trigonometric pairs like Fourier series (1) could in principle be influenced by such a projection. In this context, the tangent half-range transformation relates the angle \( \theta_m = m\pi/L \) to the slope \( t_m = \tan(\theta_m/2) \) of a line intersecting the unit circle centered around \( (0, 0) \).

\[
\frac{1 - t_m^2}{1 + t_m^2} = \frac{2\theta_m}{1 + t_m^2} = \cos(\theta_m) \sin(\theta_m) \tag{20}
\]

which goes only once around the circle as \( t_m \) goes from \(-\infty \) to \( \infty \) and never reaches the point \((-1, 0)\), which is approached as a limit when \( t_m \to \pm \infty \).

Clearly the stereographic projection \( t_m = \tan(\theta_m/2) \) is defined on the entire unit circle except at the projection point \( P : (-1, 0) \). Where it is defined, this mapping is smooth and bijective [9]. It is conformal, meaning that it preserves angles. It is however not isometric; that is, it does not preserve distances as

\[
d\theta_m = \frac{2dt_m}{1 + t_m^2}. \tag{21}
\]
The rest of this section illustrates how stereographic projection of Fourier series reveals a quadratic degeneracy that indicates the existence of the Gibbs-Wilbraham effect. It also addresses the question of solving the inverse problem for the truncated Fourier series representations. The impact of stereographic projection on the ill-posedness of the numerical solution of this inverse problem is moreover a main question that has been addressed in this paper.

Our analysis starts however by illustrating how the tangent half-angle stereographic projection can provide a new insight into the behavior of truncated Fourier series near a discontinuity of \( f(x) \).

**Lemma 3.** The solution \( x \) of the inverse problem for the truncated Fourier series representation satisfies

\[
\sum_{m=1}^{N} H(y, x, m) = \sum_{m=1}^{N} \left( a_m^2 + b_m^2 \right) - \left( a_m \tan m(\pi/2L)x - b_m \right)^2 \left( a_m^2 \tan^2 m(\pi/2L)x + 1 \right) = c - a_0/2.
\]

Proof. By studying solvability for \( x \) of (18) Let \( t_m = \tan m(\pi/2L)x \), and then \( \cos m(\pi/2L)x = (1 - t_m^2)/(1 + t_m^2) \), and \( \sin m(\pi/2L)x = 2t_m/(1 + t_m^2) \). Further consideration of these facts in (18) leads to the tangent half-angle stereographically transformed, \( \mathbb{G}[y] = 1 \), form

\[
- \sum_{m=1}^{N} \mathbb{G}[G(y, x, m)] = \sum_{m=1}^{N} \left( a_m^2 t_m^2 - b_m t_m - a_m \right) \left( t_m^2 + 1 \right) = a_0/2 - c.
\]

The harmonic trinomial in the left-hand side of (23) has two harmonic zeros, \( v_m^\pm \), for each \( x \); namely,

\[
\begin{align*}
\xi_m^\pm &= \frac{2L}{m\pi} \tan^{-1} t_m^k, \\
t_m^k &= \frac{b_m}{a_m} \pm \sigma_m, \\
\sigma_m &= \sqrt{a_m^2 + b_m^2}/a_m.
\end{align*}
\]

It is worth noting at this point that these \( v_m^\pm \) zeros should, by no means, be confused with the possible roots of the \( x \) variable. Substitution then of (24) in (23) leads to

\[
\sum_{m=1}^{N} a_m \left( t_m - t_m^k \right) \left( t_m - t_m^k \right) \left( t_m^2 + 1 \right) = a_0/2 - c.
\]

which is the same as

\[
\sum_{m=1}^{N} a_m \left( t_m - b_m/a_m - \sigma_m \right) \left( t_m - b_m/a_m + \sigma_m \right) \left( t_m^2 + 1 \right) = a_0/2 - c.
\]

Equation (26) is easily transformable to the required result (23); here the proof completes.

3.2. The Stereographic Fourier Series. To further analyze the variability of stereographically projected Fourier series, we make the substitution

\[
\theta_m = \theta_m(x) = m\pi/L x,
\]

which is the same as

\[
\sum_{m=1}^{N} a_m \left( t_m - b_m/a_m - \sigma_m \right) \left( t_m - b_m/a_m + \sigma_m \right) \left( t_m^2 + 1 \right) = a_0/2 - c.
\]
in $G(y, x, m)$ to rewrite it as

$$G(y, x, m) = a_m \cos \theta_m(x) + b_m \sin \theta_m(x).$$  \hfill (34)

The tangent half-angle (stereographic) transformation $t_m = \tan(\theta_m/2)$ or $\theta_m = 2 \tan^{-1} t_m$ is conceived as

$$\theta_m(x) = 2 \tan^{-1} t_m(x) = \Phi[t_m(x)],$$  \hfill (35)

and subsequently

$$G = G[\theta_m(x)] = G[\Phi[t_m(x)]] = H[t_m(x)].$$  \hfill (36)

It is clear then that this stereographic projection effectively leads to insertion of an additional bracket into the composition of $G$ as a function of $x$. This paves the way to the definition

$$\mathfrak{S}[G[\theta_m(x)]] = G[\Phi[t_m(x)]] = H[t_m(x)] = H(y, x, m).$$  \hfill (37)

Proposition 4. The following:

$$\mathfrak{S} \left\{ \frac{d}{dx} \sum_{m=1}^{N} G[\theta_m(x)] \right\} = \frac{d}{dx} \mathfrak{S} \left\{ \sum_{m=1}^{N} G[\theta_m(x)] \right\}.$$  \hfill (38)

always holds.

Proof. Since $N$ is finite, then the proof of $\mathfrak{S}[(d/dx)G] = (d/dx)\mathfrak{S}[G] = (d/dx)H$ implies the correctness of this proposition. So let us differentiate $G(x) = G[\theta_m(x)]$ with respect to $x$ by applying the chain rule; namely,

$$\frac{dG}{dx} = \frac{dG}{d\Phi} \frac{d\Phi}{dt_m} \frac{dt_m}{dx} = \frac{dG}{d\theta_m} \frac{d\theta_m}{dt_m} \frac{dt_m}{dx},$$  \hfill (39)

which is the same as

$$\frac{dG}{dx} = \left[-a_m \sin \theta_m + b_m \cos \theta_m\right] \frac{2}{(1 + t_m^2)} \frac{\theta_m}{2} \frac{\sec^2 \left(\frac{\theta_m}{2}\right)}{dx}.$$

Take then the $\mathfrak{S}$ transformation of the previous relation to arrive, after some algebra, at

$$\mathfrak{S} \left\{ \frac{d}{dx} G \right\} = -m \frac{\pi}{L} \frac{\left(b_m t_m^2 + 2a_m t_m - b_m\right)}{(1 + t_m^2)}.$$  \hfill (41)

Invoke then

$$H = H(y, x, m) = \frac{(a_m + 2b_m t_m - a_m t_m^2)}{(1 + t_m^2)},$$  \hfill (42)

to differentiate it with respect to $x$ as

$$\frac{d}{dx} H = \frac{d}{dx} \mathfrak{S}[G] = -m \frac{\pi}{L} \frac{\left(b_m t_m^2 + 2a_m t_m - b_m\right)}{(1 + t_m^2)},$$  \hfill (43)

which is the required result.

Let us rewrite (1) in the form

$$\varphi(x) = g(x) - c = 0$$  \hfill (44)

then try to differentiate it term-by-term as

$$\varphi'(x) = \frac{d}{dx} g(x) = \frac{d}{dx} \sum_{m=1}^{\infty} G(y, x, m)$$

$$= - \sum_{m=1}^{\infty} m \frac{\pi}{L} \left(-b_m \cos \frac{\pi}{L} x + a_m \sin \frac{\pi}{L} x\right)$$  \hfill (45)

$$= - \sum_{m=1}^{\infty} m \frac{\pi}{L} G^*(y, x, m),$$

where $G^*(y, x, m)$ is the harmonic conjugate of $G(y, x, m)$. The trigonometric series $g^*(x) = \sum_{m=1}^{\infty} G^*(y, x, m)$ is called the trigonometric series conjugate (see, e.g., [10]) to $g(x)$.

In addition to the fact that the convergence of $g^*(x)$ can sometimes become worse than convergence of $g(x)$ (see, e.g., [10]) the presence of the $m(\pi/L)$ factor in $(d/dx)g(x)$ can aggravate the convergence problem of (45) when the coefficients $a_m$ and $b_m$ do not decay fast enough.

Next we may stereographically transform $(\partial/\partial x)g(x)$, using the notation $\mathfrak{S}[(d/dx)g(x)] = [(d/dx)\mathfrak{S}g(x)]$, where

$$\left[ \frac{d}{dx} \mathfrak{S}g(x) \right] = - \sum_{m=1}^{\infty} m \frac{\pi}{L} G^*(y, x, m),$$  \hfill (46)

Similarly, for the stereographic Fourier series representation

$$\mathfrak{S}g(x) = \sum_{m=0}^{\infty} H(y, x, m),$$  \hfill (47)

in which

$$H(y, x, m) = \frac{(a_m + 2b_m t_m - a_m t_m^2)}{(1 + t_m^2)},$$  \hfill (48)

we may approximate (44) by

$$\psi(x) = \mathfrak{S}g(x) - c = 0.$$  \hfill (49)

Differentiate then (49) term-by-term as

$$\psi'(x) = \frac{d}{dx} \mathfrak{S}g(x) = \frac{d}{dx} \sum_{m=1}^{\infty} H(y, x, m)$$

$$= - \sum_{m=1}^{\infty} m \frac{\pi}{L} H^*(y, x, m)$$  \hfill (50)

$$= - \sum_{m=1}^{\infty} m \frac{\pi}{L} \left(b_m t_m^2 + 2a_m t_m - b_m\right).$$

\hfill \Box
Note here that
\[ H^* (\gamma, x, m) = \frac{b_m t_m^2 + 2a_m t_m - b_m}{(1 + t_m^2)} \]
\[ = \frac{b_m \tan m(\pi/2L)x - a_m}{b_m (\tan^2 m(\pi/2L) + 1)} \]
is the harmonic conjugate of \( H(x, y, m) \) and that
\[ H^* (x, y, m) = \tilde{G}^* (y, x, m). \tag{52} \]

This rather interesting fact leads to a remark that follows.

Remark 5. Away from points of discontinuity of \( f(x) \), the stereographically transformed differentiated Fourier series \([d/dx]g(x)]\), which defines \( \tilde{G}^* (y, x, m) \), is the same as the derivative \( d/dx)g(x) \), which defines \( H^*(x, y, m) \), of the stereographic Fourier series \( g(x) \). Moreover, even at points of discontinuity of \( f(x) \), \([d/dx]g(x)]\) is \( (d/dx)g(x) \).

Theorem 6 (see [1]). Let \( f(x) \) be (i) continuous on \([-L, L]\) such that (ii) \( f(-L) = f(L) \), and let (iii) \( f'(x) \) be piecewise continuous over \([-L, L] \); then the corresponding Fourier series \( g(x) \) of (1) or (47) is differentiable at each point where (iv) \( f''(x) \) exists.

The previous remark can incidentally have useful applications when differentiating certain slowly converging Fourier series, as illustrated by the following example.

Example 7. Consider the \( 2\pi \)-periodic odd signal \( f(x) \), defined over one period, by \( f(x) = x \). Its Fourier series is known to be
\[ g(x) = 2 \sum_{m=1}^{\infty} \frac{(-1)^{m+1}}{m} \sin mx, \tag{53} \]
that is, with \( a_m = 0, \forall m, b_m = 2((-1)^{m+1}/m) \), and \( m(\pi/L) = m \).

Differentiating term-by-term, we obtain
\[ g'(x) = 2 \sum_{m=1}^{\infty} (-1)^{m+1} \cos mx, \tag{54} \]
which is not uniformly convergent because \( f(x) \) is not only discontinuous at \( x = \pm \pi \) but \( f(-\pi) \neq f(\pi) \) and \( f''(\pm \pi) \) does not exist. In particular, at \( x = 0 \), the sum of the series \( g'(0) = 2 \sum_{m=1}^{\infty} (-1)^{(m+1)} \) oscillates between 0 and 1, that is, divergent. It is also well known, nevertheless, that the series \( \sum_{m=1}^{\infty} (-1)^{m+1} \) is \( C - 1 \) summable \([1, 2]\) to 1/2 and that turns out to yield the right answer \( g'(0) = f'(0) = 1 \). Moreover, at \( x = \pi \), the sum of the series \( g'(\pi) = 2 \sum_{m=1}^{\infty} (-1)^{2m+1} \to -\infty \), that is, divergent.

Alternatively, by means of (47)
\[ \tilde{g}(x) = \sum_{m=1}^{\infty} b_m \frac{2t_m}{1 + t_m^2} = 2 \sum_{m=1}^{\infty} \frac{(-1)^{m+1}}{m} \sin mx \tag{55} \]
and by means of (46)
\[ \tilde{g}'(x) = \left[ \frac{d}{dx} \tilde{g}(x) \right] = -\sum_{m=1}^{\infty} m\tilde{G}^* (y, x, m) \]
\[ = -\sum_{m=1}^{\infty} mb_m \frac{t_m^2 - 1}{1 + t_m^2} \tag{56} \]
\[ = 2 \sum_{m=1}^{\infty} (-1)^{m+1} \left( 1 - \tan^2 m(1/2)x \right) \]
\[ = 2 \sum_{m=1}^{\infty} (-1)^{m+1} \cos mx. \]

Here, at \( x = \pi \), the sum of the series \( \tilde{g}'(\pi) = 2 \sum_{m=1}^{\infty} (-1)^{2m+1} \to -\infty \) is also divergent.

Note in general that, for even signals, that is, when \( b_m = 0, \forall m \), it follows from (22) that
\[ H(y, x, m)|_{b_m=0} = a_m \frac{1 - t_m^2}{1 + t_m^2} = a_m \cos m\frac{\pi}{L}x \]
\[ = G(y, x, m)|_{b_m=0}. \tag{57} \]

Moreover, for odd signals, that is, when \( a_m = 0, \forall m \), the situation is not of the same clarity.

Indeed, according to (22),
\[ H(y, x, m)|_{a_m=0} = \frac{b_m}{0 + t_m^2}, \tag{58} \]
which is an uncertainty. Luckily, however, this uncertainty is not an essential one and can straightforwardly be resolved by resorting to the stereographic form (23) of (22), which yields the correct result
\[ H(y, x, m)|_{a_m=0} = b_m \frac{2t_m}{1 + t_m^2} = b_m \sin m\frac{\pi}{L}x \tag{59} \]

Also when \( b_m = 0, \forall m \),
\[ H^*(y, x, m)|_{b_m=0} = \frac{a_m}{0 + t_m^2}, \tag{60} \]
and this is resolvable in a similar fashion.

Both \( G(y, x, m) \), of (2), and \( H(y, x, m) \), of (22), equivalent forms are nonlinear trigonometric functions of the \( x \) solution. However, while \( G(y, x, m) \) is linear in \( y = [a_m, b_m]_{m=1}^{\infty} \), the stereographic \( H(y, x, m) \) is nonlinear in this \( y \). This can raise a question on a possibility for \( (\partial/\partial x)g(x) \neq (\partial/\partial x)\tilde{g}(x) \) at singular points of \( f(x) \). A question that will be answered negatively in the next section. Furthermore, since the data \( y = [a_m, b_m]_{m=1}^{\infty} \) is discrete, then the operators \( \sum_{m=1}^{\infty} G(y, x, m) \) and \( \sum_{m=1}^{\infty} H(y, x, m) \) do not depend continuously on \( y \), and this is the main reason for ill-posedness of the present inverse problem.
In conclusion, the quadratic degeneracy of the solution of the present inverse problem, explicit when using $H(y, x, m)$, is a nonlinear indicator of the existence of the peak $Q_{\gamma y x}(x)$ pertaining to the Gibbs phenomenon. However, the numerical ill-posedness, associated with the relative values of a trial iterative variable in the $y$ set, happens to remarkably remain invariant under the tangent half-angle stereographic projection.

Following the same arguments of Wilbraham, we may cancel the artificially induced differentiation by integrating (50) to represent the stereographic Fourier series $\tilde{g}(x)$ in integral form as

$$
\tilde{g}(x) = \frac{a_0}{2} - \sum_{m=1}^{\infty} \int_{0}^{x} m \frac{b_m (\tan^2 m (\pi/2L) \xi + 1)}{(\tan^2 m (\pi/2L) \xi + 1)} d\xi
$$

or

$$
\tilde{g}(x) = \frac{a_0}{2} - \sum_{m=1}^{\infty} \int_{0}^{x} m \frac{b_m (\tan^2 m (\pi/2L) \xi + 1)}{(\tan^2 m (\pi/2L) \xi + 1)} d\xi
$$

4. Numerics of Solving the Inverse Problem

The solution of the present inverse problem is in fact a problem of finding the roots of (44) when $g(x)$ is truncated at $m = N$. For that reason, we shall focus on such roots in the neighborhood of a step-like discontinuity of $f(x)$ at $x = x^*$, say, where $f(x^* - e) < f(x^* + e)$, $0 < e \ll 1$. The associated with the Gibbs phenomenon overshoot of $g(x)$ is $a_0/2 + \sum_{m=1}^{N} G(y, x, m)$ or $a_0/2 + \sum_{m=1}^{N} H(y, x, m)$ over $f(x^* + e)$ and underfoot of $g(x)$ below $f(x^* - e)$. The intersection of a horizontal line $h(x) = c \geq f(x^* + e)$, corresponding to $g'(x) = 0$, can be at two points $\alpha_1$ and $\alpha_2$ on both sides of the peak location $\alpha_0$, representing two distinct roots of $\phi(x) = 0$. It can also be at one point $\alpha_0$, corresponding to the maximum of the truncated $g(x)$; that is, a double root of $g(x) = 0$.

Obviously, $\alpha_1 < \alpha_1 < \alpha_0 < \alpha_2$, where $\alpha_1$ and $\alpha_2$ are possible points of inflection, defined by

$$
\phi''(\alpha_1) = \phi''(\alpha_1) = \phi''(\alpha_2) = \phi''(\alpha_2) = 0.
$$

The existence of both $\alpha_1$ and $\alpha_2$ or at least one of them, namely, $\alpha_2$, is however not mandatory. Moreover, for the $\alpha_1$, $\alpha_0$, and $\alpha_2$ points we expect to have

$$
g'(\alpha_1) = \phi'(\alpha_1) > 0,
$$

$$
\phi''(\alpha_1) > 0,
$$

$$
\phi'''(\alpha_1) < 0
$$

Equation (44) can be solved for $x : \phi(x) = 0$ by a Newton-Raphson process

$$
x_{n+1} = x_n - \frac{\phi(x_n)}{\phi'(x_n)},
$$

which is repeated until a sufficiently accurate value is reached. This process, for each series truncation number $N$, rewrites as

$$
x_{n+1} = x_n + \frac{\sum_{m=1}^{N} G(y, x_n, m) + a_0/2 - c}{\sum_{m=1}^{N} m (\pi/L) G^*(y, x_n, m)},
$$

or in the stereographic form

$$
x_{n+1} = x_n + \frac{\sum_{m=1}^{N} H(y, x_n, m) + a_0/2 - c}{\sum_{m=1}^{N} m (\pi/L) H^*(y, x_n, m)},
$$

4.1. Computation of the Peak Location $\alpha_0$. Being a maximization problem for $\phi(x) = 0$, the determination of $\alpha_0$ by the Newton-Raphson process [11] becomes

$$
x_{n+1} = x_n - \frac{\phi'(x_n)}{\phi''(x_n)}
$$

or

$$
x_{n+1} = x_n - \frac{\psi'(x_n)}{\psi''(x_n)},
$$

in stereographic form.
Striaghtforwardly, this process can be expressed as
\[ x_{n+1} = x_n - \frac{\sum_{m=1}^{N} mG^*(y, x_n, m)}{\sum_{m=1}^{N} m^2 (\pi/L) G(y, x_n, m)} \]  
(71)
or in the stereographic form
\[ x_{n+1} = x_n - \left[ \frac{\sum_{m=1}^{N} m (\pi/L) H^*(y, x_n, m)}{\sum_{m=1}^{N} m^2 (\pi/L^2) ((a_m t_m^2 - 2b_m t_m - a_m) / (1 + t_m^2))} \right] \]
(72)
which is the same as
\[ x_{n+1} = x_n + \left[ \frac{\sum_{m=1}^{N} m \left( (b_m t_m^2 + 2a_m t_m - b_n) / (1 + t_m^2) \right)}{\sum_{m=1}^{N} m^2 (\pi/L) \left( (a_m t_m^2 - 2b_m t_m - a_m) / (1 + t_m^2) \right)} \right] \]
(73)
It is remarkable how this algorithm is distinctively independent of \( c \).

**Proposition 8.** The ill-posedness of the inverse problem of the truncated Fourier series is invariant under stereographic projection.

**Proof.** The tangent half-angle stereographic projection, though not isometric (distance preserving), is nevertheless variably preserving. Indeed, the main ingredients of this inverse problem are, namely, \( y = [a_m, b_m]_{m=1}^{N}, c, \) and \( x_0, \) with \((d/dx)G[\theta_m(x_0)] \). Under this projection \( \theta_m(x_0) \) transforms to \( t_m(x_0) = \tan[\theta_m(x_0)/2], \) and \( G[\theta_m(x_0)] \) transforms to \( H[t_m(x_0)]. \) But according to Proposition 4, \((d/dx)H[t_m(x_0)] = (d/dx)G[\theta_m(x_0)] \). Hence the variation of \( G(y, x_0, m) \) with \( x_0 \) in the Fourier series inverse problem and the variation of \( H(y, x_0, m) \) with \( x_0 \) in the stereographic form are identical.

Equivalently, this makes the ill-posedness of this inverse problem invariant under stereographic projection. \( \square \)

Expected manifestations of the correctness of this proposition are as follows: (i) the results of computations by the (71) and (73) algorithms for \( \alpha \) with the same \( x_0 \) must be identical and (ii) the results of computations by the (65) and (67) algorithms for \( \alpha_1 \) (or \( \alpha_2 \)) with the same \( x_0 \) and \( c \) must also be identical.

### 4.2. Convergence Analysis.

Let us focus first on the convergence of the iterative process (65) or (67). In many situations where term-by-term differentiability does not hold, the Newton-Raphson method can always be replaced by some quasi-Newton method [12, 13], where \( \phi' \) (\( x_0 \)) or \( \psi' \) (\( x_0 \)) in (65) or (67) is replaced by a suitable approximation, in particular, in the chord method \( \phi' \) (\( x_0 \)) is replaced by \( \phi' \) (\( x_0 \)) for all iterations.

If \( \epsilon_n = x_n - x_0, \epsilon_{n+1} = \alpha - x_{n+1}, \) and \( \beta_n \) is some point between \( x_0, x_n \), it can be easily proved (see [11] or [12]) that
\[ \epsilon_{n+1} = \frac{\phi''(\beta_n)}{2 \phi'(x_n)} \epsilon_n^2 \]
(74)
Taking absolute value of both sides gives
\[ |\epsilon_{n+1}| = \frac{\phi''(\beta_n)}{2 \phi'(x_n)} \epsilon_n^2 \]
(75)
implies quadratic convergence; that is, \( \epsilon_n \) is squared (the number of correct digits roughly doubles) at each step. The same arguments apply of course to the \( \psi(x) = 0 \) equation.

In actual fact for a distinct \( \alpha \) the convergence of \( \{x_n\} \) in (65)–(67) is at least quadratic in the neighborhood of this \( \alpha \). The process may face a difficulty when \( \phi'(x_n) = -\sum_{m=1}^{N} m(\pi/L)G^*(y, x_n, m) \) or \( \psi'(x_n) = -\sum_{m=1}^{N} m(\pi/L)H^*(y, x_n, m) \) diverges. This is however a special but not a general problem, and (75) always holds if the following conditions are satisfied:

(i) \( \phi'(x) \neq 0, \forall x \in I, \) where \( I = [\alpha - \eta, \alpha + \eta] \) for some \( \eta \geq |\alpha - x_0| \).

(ii) \( \psi'(x) \) is finite \( \forall x \in I. \)

(iii) \( x_0 \) is sufficiently close to \( \alpha \).

The term sufficiently close in this context means the following:

(a) Taylor approximation is accurate enough such that we can ignore higher order terms.

(b) \( 1/2)|\phi''(x_n)|/|\phi'(x_n)| < K|\phi''(\alpha)|/|\phi'(\alpha)| \) for some \( K < \infty. \)

(c) \( K|\phi''(\alpha)|/|\phi'(\alpha)| \epsilon_n < 1 \) for \( n \in \mathbb{N}. \)

Consequently, (75) can be expressed in the following way:
\[ |\epsilon_{n+1}| = M \epsilon_n^2, \]
(76)
where
\[ M = \sup_{x \in I} \left| \frac{\phi''(x)}{2 \phi'(x)} \right|. \]
(77)
The initial point \( x_0 \) has to be chosen such that conditions (i) through (iii) are satisfied, where the third condition requires that \( M|\epsilon_0| < 1. \) Moreover, with increasing \( N \) the distance \( |\alpha_1 - \alpha_2| \) is known to decrease in a way making the satisfaction of these conditions very difficult and actually impossible as \( N \to \infty. \) Here it should be underlined that the iterative method (65)–(67) may fail to converge to the \( \alpha \) root in the following situations:

(i) If a point like \( \alpha_0 \) is included in \( I, \) the method will terminate due to division by zero.

(ii) When the initial estimate \( x_0 \) is poor, the pertaining wrong \( I \) can contribute to nonconvergence of the algorithm.

As for convergence of the Newton-Raphson process towards \( x_0, \) let \( \epsilon_n = x_0 - x_n, \epsilon_{n+1} = \alpha_0 - x_{n+1}, \) and \( \theta_n \) is some point between \( x_n \) and \( \alpha_0, \) to arrive at
\[ \epsilon_{n+1} = -\frac{\phi'''(\theta_n)}{2 \phi''(x_n)} \epsilon_n^2 \]
(78)
\[ |\epsilon_{n+1}| = \left| \frac{\phi'''(\theta_n)}{2 \phi''(x_n)} \right| \epsilon_n^2. \]
(79)
The convergence of \( \{x_n\} \) in (71)–(73) is at least quadratic in the neighborhood of this \( \alpha_0 \). The process may face a difficulty, however, when \( \phi''(x_n) = -\sum_{m=1}^\infty m^2(\pi/L)^2 \cos (\pi m x_n) \) or \( \psi''(x_n) \) diverges. Moreover (79) always holds if the following conditions are satisfied:

(i) \( \phi''(x) \neq 0, \forall x \in I \), where \( I = [\alpha_0 - \eta, \alpha_0 + \eta] \) for some \( \eta \geq |(\alpha_0 - x_0)| \).

(ii) \( \phi''(x) \) is finite \( \forall x \in I \).

(iii) \( x_0 \) is sufficiently close to \( \alpha_0 \).

Sufficient closeness in this context means that

(a) Taylor approximation is accurate enough such that we can ignore higher order terms;

(b) \( (1/2)|\phi''''(x_n)/\phi''(x_n)| < K|\phi''''(\alpha_0)/\phi''(\alpha_0)| \) for some \( K < \infty \);

(c) \( K|\phi''''(\alpha_0)/\phi''(\alpha_0)|e_n < 1 \) for \( n \in \mathbb{N} \).

Consequently,

\[ |e_{n+1}| = Me_n^2, \tag{80} \]

where

\[ M = \sup_{x \in I} \frac{1}{2} \frac{\phi''''(x)}{\phi''(x)}. \tag{81} \]

4.3. Applications. The standard iterative algorithms, (71) or (73), for finding the location of overshoots in \( Q_{x_0, \sigma_0}(x) \) of the Gibbs effect employ the fact that the first derivative of a peak, in \( g(x) \) or \( \tilde{g}(x) \), respectively, has a downward-going-zero-crossing at the peak maximum. But the presence of many neighboring smaller peaks in the ringing close to the overshoot will cause many undesirable zero-crossings to be invoked if the initial trial root \( x_0 \) is not correctly chosen. This fact makes the numerical problem of peak finding in the Gibbs effect of pathological difficulty. A difficulty that aggravates with the increase in the number \( N \) of truncated terms in the employed Fourier series. Indeed, with the increase of \( N \) the number of undesirable zero-crossings increases in the neighborhood of \( \alpha_0 \), as illustrated in Section 1. This can increasingly lead to oscillatory and even unstable values in sequence \( \{x_n\} \), the more the distance \( |x_0 - \alpha_0| \) contains unwanted zero-crossings.

These problems have been resolved in the two examples that follow, which simultaneously illustrate the applicability of Proposition 8.

Example 9. Consider the 2\( \pi \)-periodic odd signal \( f(x) \) of Example A.1, defined over one period, by

\[ f(x) = \begin{cases} -1, & -\pi < x < 0 \\ 1, & 0 \leq x < \pi \end{cases}, \tag{82} \]

which has, for each truncation number \( N \), the truncated Fourier series approximations

\[ g(x) = \frac{4}{\pi} \sum_{m=1}^N \frac{1}{(2m-1)} \sin (2m-1)x, \]

\[ \tilde{g}(x) = \frac{8}{\pi} \sum_{m=1}^N \frac{b_m t_m}{(1 + t_m^2)} = \frac{8}{\pi} \sum_{m=1}^N \frac{1}{(2m-1)(1 + t_m^2)} \tag{83} \]

\[ = \frac{8}{\pi} \sum_{m=1}^N \frac{1}{(2m-1)} \left[ 1 + \left( \frac{\tan ((2m-1)/2)x}{\tan ((2m-1)/2)x} \right)^2 \right]. \]

This signal has a jump discontinuity at \( x = 0 \) of magnitude 2. We shall focus our computations on the Gibbs phenomenon overshoot near this discontinuity.

Here is a printout list of results for the parameters of the posing inverse problem, via computations by both the direct and stereographic algorithms.

It is remarkable how results by both algorithms are practically identical and share the same following features that are observed by results of the direct algorithm, discussed below.

Case I \((N = 40)\). The computations of \( \alpha_0 \) for the previous example, using (71), with \( x_0 = 0.02 \) lead to \( x_1 = 0.039109 \); this varies quickly with \( n \) (only beyond the 4th decimal) to stabilize towards \( x_4 = x_5 = \cdots = 0.039299 = \alpha_0 \). Here using (1) gives \( g(\alpha_0) = g(0.039299) = 1.179030 \), which is quite close to the theoretically expected value, by (A.7), of \( \sim 1.18 \).

Then using (65) for \( c = 1.09 \) and \( x_0 = 0.000381 \) leads to \( x_1 = 0.021697 \); this varies quickly with \( n \) (only beyond the 3rd decimal) to stabilize quickly towards \( x_5 = x_6 = \cdots = 0.028368 = \alpha_1 \). Also using (65) for \( c = 1.09 \) and \( x_0 = 0.04 \) leads to \( x_1 = 0.067699 \); this varies quickly with \( n \) (beyond the 2nd decimal) to stabilize quickly towards \( x_5 = x_6 = \cdots = 0.052873 = \alpha_2 \).

For both \( \alpha_1 \) and \( \alpha_2 \), using (1) yields \( g(\alpha_1) = g(0.028368) = g(\alpha_2) = g(0.052873) = 1.09 = c \), as expected.

Case II \((N = 100)\). Computations of \( \alpha_0 \), using (71), with \( x_0 = 0.02 \) lead to \( x_1 = 0.014221 \); this varies quickly with \( n \) (only beyond the 4th decimal) to stabilize towards \( x_3 = x_4 = x_5 = \cdots = 0.015708 = \alpha_0 \). Here using (1) gives \( g(\alpha_0) = g(0.015708) = 1.178990 \). This result appears to be slightly excessive, perhaps due to increased rounding off errors with increasing \( N \) to 100.

Then using (65) for \( c = 1.09 \) and \( x_0 = 0.015237 \) leads after a negative value to \( x_1 = 0.014462 \); this varies with \( n \) (beyond the 3rd decimal) to stabilize slowly towards \( x_5 = x_6 = \cdots = 0.0113348 = \alpha_1 \). The used value of \( x_0 \) is apparently not appropriate, and this illustrates the generic ill-posedness of this inverse problem.

The perturbations of \( \alpha_2 \), with \( x_0 = 0.017000 \), lead to \( x_1 = 0.020434 \); this varies with \( n \) (beyond the 3rd decimal) to stabilize quickly towards \( x_4 = x_5 = x_6 = \cdots = 0.021149 = \alpha_2 \).
For both $\alpha_1$ and $\alpha_2$, using (1), yields $g(\alpha_1) = g(0.0113348) = g(0.021149) = 1.09 = c$, as expected.

Case III ($N = 150$). The computations of $\alpha_0$, using (71), with $x_0 = 0.005$ lead to $x_1 = 0.010595$; this varies insignificantly with $n$ (only beyond the 4th decimal) to stabilize towards $x_7 = x_8 = x_9 = \cdots = 0.010472 = \alpha_0$. Here using (1) gives $g(\alpha_0) = g(0.010472) = 1.178980$, which does not significantly differ from $g(\alpha_0)$ when $N = 100$.

Then using (65) for $c = 1.09$ and $x_0 = 0.010158$ leads after a negative value to $x_4 = 0.009643$; and this varies (beyond the 3-d decimal) to stabilize quickly towards $x_7 = x_8 = x_9 = \cdots = 0.007565 = \alpha_1$.

The pertaining computations of $\alpha_2$, with $x_0 = 0.010800$, lead to $x_7 = 0.025998$; this varies (beyond the 2nd decimal) to stabilize towards $x_{14} = x_{15} = x_{16} = x_{17} = x_{18} = \cdots = 0.014099 = \alpha_2$.

For both $\alpha_1$ and $\alpha_2$, using (1) yields $g(\alpha_1) = g(007565) = g(\alpha_0) = g(0.014099) = 1.09 = c$.

During the previous computations, summarized in Table 1, it has been observed that varying $x_0$ in the present Newton-Raphson process, for all $N$, can lead to strong oscillations and even completely unstable values for $\alpha_0$, $\alpha_1$, and $\alpha_2$ of the Fourier series inverse problem, which though ill-posed is practically regularizable. Apart from this drawback, the (71) with (65) iterative algorithms (i) converge quickly to the theoretically expected values and are (ii) of demonstrated stability of the iterations, for properly chosen $x_0$ (which act as effective regularizing parameters). Moreover, computations by the stereographic algorithm, invoking (73) and (67), share the same magnitude and convergence rates with results by the direct algorithm. Finally, it is demonstrated that the present signal exhibits an overshoot, when $N = 150$, of ~9% of a jump discontinuity, of magnitude 2, that is, located at a distance of $\sim 0.0105$.

To illustrate robustness of both numerical algorithms and their possible effective regularizability we shall study also the numerical inverse problem for the Fourier series representation of an arbitrary (nonsymmetric) periodic signal.

Example I0. Consider the $2\pi$-periodic signal $f(x)$, defined over one period, by

$$f(x) = \begin{cases} -\frac{x}{\pi}, & -\pi < x < 0 \\ 1, & 0 \leq x < \pi. \end{cases}$$

It has, for each truncation number $N$, the truncated Fourier series approximations

$$g(x) = \frac{1}{4} + \sum_{m=1}^{N} \frac{\lfloor (-1)^{m-1} \rfloor}{\pi^2 m^2} \cos mx + \frac{2 - \lfloor (-1)^{m} \rfloor}{\pi m} \sin mx.$$ 

$$\overline{g}(x) = \frac{1}{4}$$

$$+ \sum_{m=1}^{N} \left\{ \frac{\lfloor (-1)^{m-1} \rfloor / \pi^2 m^2 + 2 \lfloor (-1)^{m} \rfloor / \pi m \tan ((2m-1)/2)x - \lfloor (-1)^{m-1} \rfloor / \pi^2 m^2 \tan ((2m-1)/2)x^2} {1 + [\tan ((2m-1)/2)x]^2} \right\}. \quad (85)$$

Case III ($N = 150$). The computations of $\alpha_0$, with $x_0 = 0.007$, lead also to an identical sequence $\{x_n\}$ that converges by both algorithms after the $x_2$ iteration to 0.020915 with $g(\alpha_0) = \overline{g}(\alpha_0) = 1.17891$.

Therefore the present signal exhibits an overshoot when $N = 150$, of ~9% of a jump discontinuity of magnitude 2, that is, located at a distance of $\sim 0.0209$.

Comparison of the computed Gibbs overshoot parameters of the two considered signals with jump discontinuities at the same $x = 0$ of the same magnitude illustrates that the relative size of the overshoot depends essentially exclusively on the magnitude of the stimulating jump discontinuity. For instance, based on the present numerical results, one can suggest here that nonsymmetric signals have equal overshoots with symmetric signals. Their respective $\alpha_0$’s are however different. Indeed the nonsymmetric signal has a significantly larger $\alpha_0$ (0.0209) than the $\alpha_0$ (0.0105) of the symmetric signal.
Table 1: Results of computations for Example 9.

| N  | α_1  | α_0  | α_2  |
|----|------|------|------|
| 40 | 0.028368 | 0.039299 | 0.052873 |
| 100 | 0.011335 | 0.015705 | 0.021149 |
| 150 | 0.007565 | 0.010472 | 0.014099 |

Table 2: Results of computations for Example 10.

| N  | α_1  | α_0  | α_2  |
|----|------|------|------|
| 40 | 0.0564765 | 0.0781391 | 0.105155 |
| 100 | 0.0226546 | 0.0313516 | 0.0422024 |
| 150 | 0.0151124 | 0.0209153 | 0.028156 |

Table 3: Summary of computations of all overshoot parameters in Example 10.

| N  | x_0  | x_1  | α_0  | g(α_0) |
|----|------|------|------|--------|
| 100 | 0.030 | x_3  | 0.0313516 | 1.1787 |
| 200 | 0.015 | x_3  | 0.0157691 | 1.1789 |
| 300 | 0.008 | x_3  | 0.0104648 | 1.1789 |

| N  | x_0  | x_1  | α_1  | c |
|----|------|------|------|--|
| 100 | 0.025 | x_4  | 0.0222027 | 1.08 |
| 200 | 0.010 | x_3  | 0.011117  | 1.08 |
| 300 | 0.008 | x_3  | 0.00772095 | 1.10 |

| N  | x_0  | x_1  | α_2  | c |
|----|------|------|------|--|
| 100 | 0.35  | x_4  | 0.0429439 | 1.08 |
| 200 | 0.018 | x_3  | 0.0214960 | 1.08 |
| 300 | 0.012 | x_3  | 0.0133360 | 1.08 |

A further analysis has also been performed of Example 10 with N = 100, 200, 300, to compute the corresponding α_0 with α_1, α_2 that correspond simultaneously to c = 1.08, 1.10, 1.12 as illustrated in Figure 1. Results of these computations are summarized in Table 3 and then used for the partial Q_0,α_0(x) plots, of Figure 2, that interpolate α_1, α_2 (for c = 1.08, 1.10, 1.12) with α_0, for each N = 100, 200, 300. In this table, x_n stands for the first convergence step of the |x_n| sequence of Newton Raphson iterations.

These results turn out to be extremely stable against variations of the trial initial root x_0. Indeed, in the computationally most difficult situation, namely, when N = 300, a change of x_0 from 0.008 to 0.020 leads to a variation from N = 100, 200, 300.

5. Conclusions

This paper demonstrates that the quadratic degeneracy of the solution of the present inverse problem, explicit when using H(γ, x, m), is a nonlinear indicator of the existence of Gibbs’ phenomenon. It provides also a distribution-theoretic proof for the existence of this phenomenon. The reported analysis of numerical solvability of this problem illustrates the following:

(i) When \[ \sum_{m=1}^{\infty} m(\pi/L)G^*(\gamma, x, m) = \sum_{m=1}^{\infty} m(\pi/L)H^*(\gamma, x, m) \] converge continuously in x and are
nonzero at \( \alpha_1 \) (or \( \alpha_2 \)), there exists a neighborhood (defined by \( I \)) of \( \alpha_1 \) such that, for all starting values \( x_0 \) in the neighborhood, \( \{x_n\} \) will converge to \( \alpha_1 \) (or \( \alpha_2 \)).

(ii) When \( \sum_{m=1}^{\infty} m(\pi/L)G^*(y,x,m) = \sum_{m=1}^{\infty} m(\pi/L)H^*(y,x,m) \) converge continuously in \( y \) and are nonzero at \( \alpha_1 \) and \( \alpha_2 \) (but not at \( \alpha_0 \)) and if \( g(x) \) has a second derivative, which might be zero, at \( \alpha_1 \) and \( \alpha_2 \) (and at \( \alpha_0 \)) then the convergence of \( \{x_n\} \) is quadratic or faster. If \( g''(x) \neq 0 \) at \( \alpha_1 \) and \( \alpha_2 \), then the convergence of \( \{x_n\} \) is merely quadratic.

(iii) The convergence of the Newton-Raphson iterative process to \( \alpha_0 \) can also be quadratic, but restrictions on the derivatives of \( g(x) \) have to be observed up to the third order. Otherwise, the convergence of \( \{x_n\} \) to \( \alpha_0 \) can slow down to become only linear [12] with a rate \( \log^2 \).

(iv) The reported computations clearly illustrate that \( \alpha_0 \to 0 \) and \( \{\alpha_1 - \alpha_2\} \to 0 \) when \( N \to \infty \); that is, the infinite Fourier series will sum to \( f(x) \) in the neighborhood of a discontinuity, except at the discontinuity itself.

**Appendix**

To illustrate the Gibbs-Wilbraham effect we shall use here the same arguments employed by Wilbraham in 1848 in analyzing a similar square wave.

**Example A.1.** Let \( f(x) \) be a square wave periodic signal, which is defined over one period by

\[
 f(x) = \begin{cases} 
 -1, & -\pi < x \leq 0 \\
 1, & 0 < x < \pi.
\end{cases}
\]  

(A.1)

Its Fourier series representation is

\[
g(x) = \frac{4}{\pi} \sum_{m=1}^{\infty} \frac{1}{2m-1} \sin(2m-1)x.
\]  

(A.2)

This defines the truncated series

\[
 S_N(x) = \frac{4}{\pi} \sum_{m=1}^{N} \frac{1}{2m-1} \sin(2m-1)x,
\]  

(A.3)

which may be differentiated as

\[
 \frac{d}{dx} S_N(x) = \frac{4}{\pi} \sum_{m=1}^{N} \cos(2m-1)x.
\]  

(A.4)

Since

\[
 2 \sum_{m=1}^{N} \cos(2m-1)x = \frac{\sin 2Nx}{\sin x},
\]  

(A.5)

then

\[
 \frac{d}{dx} S_N(x) = \frac{2}{\pi} \frac{\sin 2Nx}{\sin x}.
\]  

(A.6)

It is possible to cancel the previous artificially induced differentiation by integrating the last result as

\[
 S_N(x) = \int_0^x \frac{d}{dx} S_N(\xi) \, d\xi = \frac{2}{\pi} \int_0^x \frac{\sin 2N\xi}{\sin \xi} \, d\xi,
\]  

(A.7)

This integral starts at zero when \( x = 0 \) in agreement with (A.2) and increases until \( 2N\xi = \pi \). Then \( \xi = \pi/2N \) defines a turning point \( x_1 \). The integral starts to decrease, at which point the numerator \( \sin 2N\xi \) goes negative. For large \( N \) the denominator \( \sin \xi \) remains positive.

The form (A.7), distinctively from (A.3), reveals that \( S_N(x) \) has turning points at the zeros of \( \sin 2N\xi \). These occur when \( 2N\xi = r\pi; r = 1, 2, 3, 4, 5, \ldots \), that is, at \( \xi = r(\pi/2N) \). Clearly then \( x_1 \sim \xi_1 = \pi/2N \) are turning points for \( S_N(x) \).

In particular \( x_1 \sim \xi_1 = \pi/2N \) defines the overshoot peak location \( \alpha_0 \) and \( S_N(x_1) = (2/\pi) \pi/2N \sin 2N\xi/\sin \xi \) is its magnitude. To evaluate \( S_N(x_1) \) numerically, let us change the variable of integration; namely, \( 2N\xi = w \), or \( \xi = w/2N \), to rewrite the previous integral as \( S_N(x_1) = (2/\pi) \int_0^w (\sin w/\sin(w/2N)) \, dw/w2N \). Clearly when \( N \to \infty \), \( \sin w/\sin(w/2N) \to 2 \), and

\[
 \lim_{N \to \infty} S_N(x_1) = \frac{2}{\pi} \int_0^\pi \frac{\sin w}{w} \, dw = \frac{2}{\pi} \sin(\pi) = 1.18,
\]  

(A.8)

correct to two decimal points.

As for \( x_2, x_3, x_4, \ldots \), \( x_n \), \( \lim_{N \to \infty} [S_N(x) - f(x)] \) decreases as we move away from the discontinuity, but it should be remarked however that the location of the overshoot \( \alpha_0 = x_1 = \pi/2N \) moves towards the discontinuity according to

\[
 \lim_{N \to \infty} \alpha_0 = \lim_{N \to \infty} \pi/2N = 0.
\]  

(A.9)

**Conflict of Interests**

The author declares that there is no conflict of interests regarding the publication of this paper.

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**References**

[1] P. L. Butzer and R. J. Nessel, *Fourier Analysis and Approximation, Volume I: One-Dimensional Theory*, Birkhäuser, 1971.
[2] R. E. Edwards, *Fourier Series: A Modern Introduction*, Holt, Rinehart & Winston, 1967.
[3] W. J. Thompson, “Fourier series and the Gibbs phenomenon,” *American Journal of Physics*, vol. 60, no. 5, pp. 425–429, 1992.
[4] J. Jerri, *The Gibbs Phenomenon in Fourier Analysis, Splines & Wavelet Approximations*, Kluwer Academic Publishers, Boston, Mass, USA, 1990.
[5] K. S. Rim and B. I. Yun, “Gibbs phenomenon removal by adding Heaviside functions,” *Advances in Computational Mathematics*, vol. 38, no. 4, pp. 683–699, 2013.

[6] A. H. Zemanian, *Distribution Theory and Transform Analysis*, McGraw-Hill, 1965.

[7] G. R. Ayers and J. C. Dainty, “Iterative blind deconvolution method and its applications,” *Optics Letters*, vol. 13, no. 7, pp. 547–549, 1988.

[8] N. H. S. Haidar, “A nonlinear method for a severely ill-posed generalized moment problem,” *Journal of Inverse & Ill-Posed Problems*, vol. 5, no. 4, pp. 323–335, 1997.

[9] G. L. Bradley and K. J. Smith, *Calculus*, Prentice Hall, 1995.

[10] P. Öffner, T. Sonar, and M. Wirz, “Detecting strength and location of jump discontinuities in numerical data,” *Applied Mathematics*, vol. 4, no. 12, pp. 1–14, 2013.

[11] K. E. Atkinson, *An Introduction to Numerical Analysis*, John Wiley & Sons, New York, NY, USA, 1989.

[12] E. Suli and D. F. Mayers, *An Introduction to Numerical Analysis*, Cambridge University Press, Cambridge, UK, 2003.

[13] P. Wang, “A third-order family of Newton-like iteration methods for solving nonlinear equations,” *Journal of Numerical Mathematics and Stochastics*, vol. 3, no. 1, pp. 13–19, 2011.
