A FAST AND SPECTRAL CONVERGENT ALGORITHM FOR RATIONAL-ORDER FRACTIONAL INTEGRAL AND DIFFERENTIAL EQUATIONS*

NICHOLAS HALE† AND SHEEHAN OLVER‡

Abstract. A fast algorithm (linear in the degrees of freedom) for the solution of linear variable-coefficient rational-order fractional integral and differential equations is described. The approach is related to the ultraspherical method for ODEs [S. Olver and A. Townsend, SIAM Rev., 55 (2013), pp. 462–489] and involves constructing two different bases, one for the domain of the operator and one for the range of the operator. The bases are constructed from direct sums of suitably weighted ultraspherical or Jacobi polynomial expansions, for which explicit representations of fractional integrals and derivatives are known, and are carefully chosen so that the resulting operators are banded or almost banded. Geometric convergence is demonstrated for numerous model problems when the variable coefficients and right-hand side are sufficiently smooth.

Key words. fractional derivative, spectral method, ultraspherical polynomials, Jacobi polynomials, Riemann–Liouville, Caputo, Bagley–Torvik

AMS subject classifications. 26A33, 34A08, 65L99

DOI. 10.1137/16M1104901

1. Introduction. Fractional derivatives and fractional differential equations (FDEs) are becoming increasingly prevalent in the mathematical modeling of biological and physical processes [10, 17, 21, 22, 24, 29, 32, 33, 34]. Numerical techniques for computing solutions are typically based on finite differences [9, 23, 42] or finite elements [11, 14, 20], but these usually provide only low-accuracy solutions due to the global nature of fractional derivatives. There have been some recent developments in spectral methods for FDEs [7, 19, 43], but these are only observed to achieve spectral accuracy for special solutions.

This paper concerns the numerical solution of linear equations involving rational-order fractional integrals and derivatives on the interval $[-1, 1]$. For $p, q \in \mathbb{N}$ with $0 < p < q$, the left-sided $p/q$-integral is defined as [31]$^2$

$$
-1^Q_{x} p/q f(x) = \frac{1}{\Gamma(p/q)} \int_{-1}^{x} \frac{f(t)}{(x-t)^{1-p/q}} dt,
$$

and for $m \in \mathbb{N}$, $(m + p/q)$-order derivatives of Riemann–Liouville (RL) and Caputo types are given by

---

*Submitted to the journal’s Methods and Algorithms for Scientific Computing section November 23, 2016; accepted for publication (in revised form) April 23, 2018; published electronically August 9, 2018.

http://www.siam.org/journals/sisc/40-4/M110490.html

†Department of Mathematical Sciences, Stellenbosch University, Stellenbosch, 7602, South Africa (nickhale@sun.ac.za).

‡Department of Mathematics, Imperial College, London, SW7 2AZ, United Kingdom (s.olver@imperial.ac.uk).

$^1$Problems defined on any other bounded interval may be mapped to $[-1, 1]$ by a suitable affine transformation.

$^2$The right-sided rational-integral, $x^Q_{1} p/q$, is similar, but with the limits on the integral changed from $[-1, x]$ to $[x, 1]$ and the bracketed term in the denominator of the integrand negated. Without loss of generality, we focus on the left-sided case.
A FAST ALGORITHM FOR RATIONAL-ORDER FIEs & FDEs

\[ R^{1}_D \frac{d^{m+1}}{dx^{m+1}} \left( -1 Q_{1}^{1-p/q} f(x) \right) \]  
and  
\[ C^{1}_D \frac{d^{m+1}}{dx^{m+1}} \left( -1 Q_{1}^{1-p/q} f(x) \right), \]

respectively. We propose an approach which achieves spectral convergence in linear complexity for a broad class of linear fractional integral equations (FIEs) and FDEs composed of such rational-order operators. We demonstrate the accuracy and flexibility of the method on numerous examples, such as in Figure 5.2, where we solve the generalized second-kind Abel integral equation

\[ u(x) + \lambda \int_{-1}^{x} \frac{u(t)}{(x-t)^{1/3}} = f(x), \quad x \in [-1,1], \]

and in Figure 8.1, where we solve a highly oscillatory fractional Airy equation

\[ i^{3/2} R^{1}_D \frac{d^{3/2}}{dx^{3/2}} u(x) - 10^4 x u(x) = 0, \quad x \in [-1,1], \quad u(-1) = 0, \quad u(1) = 1. \]

The approach is related to the ultraspherical spectral (US) method for ODEs [27] and singular integral equations [35], where the key idea is that the underlying operators are banded when represented by their action on appropriately chosen bases, built out of ultraspherical polynomials. Here, for integral equations, the idea is similar: We exploit the fact that fractional integration is a banded operator between a suitable direct sum space formed of \( q \) weighted Jacobi polynomial bases (which can be related to the “generalized Jacobi functions” of [7] and “polyfractinomials” of [43]), for which an explicit representation of the fractional derivative is available. However, a critical difficulty arises for differential equations: The bases are not compatible in the sense that the weights in the range of the operators differ from those of the domain. To overcome this difficulty we expand the solution as a direct sum of weighted Jacobi polynomials as before but then consider another basis formed as a direct sum of \( q \) different weighted Jacobi polynomial bases to represent the range of the operator (for a total of \( 2q \) bases). \(^3\) If these two direct sum spaces are chosen appropriately, the resulting operators are banded.

There have been two recent additions to the literature which also provide spectral accuracy for FDEs, namely, the works of Zayernouri and Karniadakis [43] and Chen, Shen, and Wang [7]. \(^4\) The foundation of both is the same formula for the fractional integral of weighted Jacobi polynomials (i.e., [2, Theorem 6.72(b)]) which also forms the basis of our own approach (see Theorems 2.1 and 7.2 below). Whereas in this paper we limit our attention to rational-order derivatives, both [7] and [43] deal with arbitrary orders and so are in a sense more general. However, the algorithm proposed by Zayernouri and Karniadakis is collocation based, leading to dense matrices and \( O(N^3) \) complexity. Spectral accuracy is demonstrated for a few select problems, but it is typically subgeometric. Furthermore, the discussion is limited to zero Dirichlet boundary conditions. The algorithm of Chen, Shen, and Wang has linear complexity, but applies only to FDEs of the form \( -1 D^\nu_x u(x) = f(x) \) and \( x D^\nu_x u(x) = f(x) \) (for

\(^3\)When \( q = 2 \), the integral in (1.1) is called the left-sided half-integral, and we shall see below that more elegant formulae can be obtained in this instance using ultraspherical rather than Jacobi polynomials.

\(^4\)There has also been recent work in spectral methods for tempered FDEs (see, for example, [44]), but it is not clear that these approaches provide spectral accuracy in the limit \( \alpha \to 0 \), i.e., the nontempered case.
both RL and Caputo definitions). In this work we shall consider FDEs which are linear combinations of rational-integer order derivatives with more general boundary conditions and demonstrate geometric convergence with linear complexity.

The bulk of this paper is dedicated to introducing the proposed algorithm specifically for the case of half-integral order integrals and derivatives (i.e., $p = 1$ and $q = 2$ in (1.1) and (1.2)), for which the approach and derivation are more intuitive to follow. However, the extension to more general rational-order derivatives and integrals follows readily once the approach is understood for the half-integer order case, and in the penultimate section we describe in some detail how this is achieved and give further examples. As such, the outline of this paper is as follows. In section 2 we introduce some necessary preliminaries regarding ultraspherical polynomials, in particular an explicit formula for their half-integrals and various transformations between different weighted ultraspherical polynomial expansions. In sections 3–5 we use these to derive a fast and geometrically convergent algorithm for a certain class of half-integer order FIEs and FDEs of the RL and the Caputo type, respectively. Section 6 discusses some computational issues relating to these algorithms, such as the efficient computation of the required polynomial coefficients and solution of the linear systems describing the FIEs/FDEs. In section 7 we describe how the ideas of the previous sections may be adapted to consider more general rational-order FIEs, before concluding in section 8 with one final example and some suggestions for future work.

Remark. The experiments in this paper were conducted in MATLAB (code to reproduce all figures is available online at [15]), and a Julia implementation of the algorithm is available in ApproxFun.jl [26].

2. Preliminaries. In this section we consider the required preliminaries needed for working with half-integrals

\[ -\frac{1}{2} Q_{\frac{1}{2}}(x) f(x) = \frac{1}{\sqrt{\pi}} \int_{-1}^{x} \frac{f(t)}{(x-t)\frac{1}{2}} dt \]

and half-integer order derivatives

\[ RL_{x} D_{x}^{m+1/2} f(x) = \frac{d^{m+1}}{dx^{m+1}} \left( -\frac{1}{2} Q_{\frac{1}{2}}(x) f(x) \right) \quad \text{and} \]

\[ C_{x}^{q} D_{x}^{m+1/2} f(x) = -\frac{1}{2} Q_{\frac{1}{2}} \left( \frac{d^{m+1}}{dx^{m+1}} f(x) \right), \quad m \in \mathbb{N}. \]

Our primary tools here are ultraspherical polynomials, specifically Legendre and Chebyshev polynomials, as described below.

Remark. We shall see in section 7 that in the case of general rational-order integrals and derivatives one must instead work with Jacobi polynomials. While it is possible to unify these approaches and use Jacobi polynomials in the half-order case, we find that using the ultraspherical polynomials here leads to cleaner and more elegant formulae and so choose to formulate our algorithm with these instead.

2.1. Ultraspherical polynomials. The ultraspherical (or Gegenbauer) polynomials, $C_{n}^{(\lambda)}(x)$, are orthogonal with respect to the weight function $(1 - x^2)^{\lambda-1/2}$ on the interval $[-1,1]$, where $\lambda > -\frac{1}{2}$ and $\lambda \neq 0$. For any $\lambda > 0$, the degree $n$ ultraspherical polynomial may be defined via the recurrence [12, Eqn. 18.9.1]

\[ C_{n-1}^{(\lambda)}(x) = 0, \quad C_{0}^{(\lambda)}(x) = 1, \quad (n+1)C_{n+1}^{(\lambda)}(x) = 2(n+\lambda)x C_{n}^{(\lambda)}(x) - (n+2\lambda - 1) C_{n-1}^{(\lambda)}(x). \]
The Legendre polynomials, $P_n(x)$, and the second-kind Chebyshev polynomials, $U_n(x)$, are special cases of the ultraspherical polynomials with $\lambda = \frac{1}{2}$ and $\lambda = 1$, respectively. These two will be of particular importance in our algorithms described in sections 3–5 for half-integer order FIEs and FDEs.

For any $x \in \mathbb{C}$, $\lambda > 0$, and $\gamma \in \mathbb{R}$, we define $C^{(\lambda)}(x)$ as the quasimatrix—a “matrix” whose “columns” are functions defined on an interval $[36]$—whose $j$th column is the degree $(j-1)$th ultraspherical polynomial with parameter $\lambda$ weighted by $(1 + x)^\gamma$, i.e.,

$$
C^{(\lambda)}(x) := [(1 + x)^\gamma C_0^{(\lambda)}(x), (1 + x)^\gamma C_1^{(\lambda)}(x), \ldots].
$$

We refer to these as weighted ultraspherical bases and note that the columns of $C^{(\lambda)}(x)$ are related to the “generalized Jacobi functions” of [7] and “polyfractinomials” of [43].

Linear operators which can be applied to one such weighted ultraspherical basis and expanded in another induce infinite-dimensional matrices that can be viewed as acting between different $C^{(\lambda)}$ spaces. For example, given a continuous linear operator $L : X \to Y$ so that $(1 + x)^\lambda C_k^{(\lambda)}(x) \in X$ and $(1 + x)^\gamma c_j^{(\gamma)}(x) \in Y$ with the property

$$
L[(1 + o)^\gamma C_k^{(\lambda)}](x) = \sum_{j=k-\ell}^{k+m} L_{jk}(1 + x)^\gamma c_j^{(\gamma)}(x),
$$

we can associate it with an $m$-banded (i.e., banded with bandwidth $m$) infinite-dimensional matrix

$$
L := \begin{pmatrix}
L_{00} & \cdots & L_{0m} \\
& \ddots & \vdots \\
L_{m0} & \cdots & L_{m1} \\
& \ddots & \ddots \\
& & \ddots \\
L_{m+1,1} & \cdots & \cdots \\
& & \ddots \\
& & & \ddots \\
& & & & \ddots 
\end{pmatrix}.
$$

Since $L$ is banded, multiplication is a well-defined operation on $\mathbb{C}^\infty$, and (2.7) can be viewed as an operator $L : C^{(\lambda)}_\gamma \to C^{(\gamma)}_c$. To relate the operator $L$ and the operator $\mathcal{L}$ we note that, by construction, we have

$$
\mathcal{L}[(1 + o)^\gamma C_k^{(\lambda)}](x) = L C^{(\lambda)}_\gamma(x) \odot \mathcal{L} = C^{(\gamma)}_c(x) L \odot \mathcal{L}.
$$

\footnote{Here and throughout we use $\odot$ to represent the dummy variable in an operator.}
If \( u(x) \in X \), then, assuming that the \( C_\gamma^{(\lambda)}(x) \) are dense in \( X \), there exists \( u \in C_\gamma^{(\lambda)} \) such that \( u(x) = C_\gamma^{(\lambda)}(x)u \). Because \( L \) is continuous, we have
\[
(2.9) \quad Lu = L C_\gamma^{(\lambda)}(x)u = C_\gamma^{(\ell)}(x)Lu,
\]
and therefore applying \( L \) to \( u(x) \) is equivalent to applying \( L \) to \( u \).

The US method [27] for differential equations requires three such banded operators which act on ultraspherical polynomials: conversion, multiplication, and differentiation. We now revisit these operators in the case of weighted ultraspherical polynomials and introduce new operators corresponding to fractional integration and fractional differentiation of half-integer order.

**Remark.** In sections 3–5 we will seek solutions to FIEs and FDEs formed as linear combinations of Legendre polynomials, \( P_n(x) \), and weighted Chebyshev polynomials of the second kind, \( \sqrt{1 + x}U_n(x) \). Another possibility is to choose a direct sum of Legendre polynomials and weighted Chebyshev polynomials of the first kind, \( T_n(x)/\sqrt{1 + x} \) (for which one can also find explicit and compact formulae for half-integer order integrals and derivatives). We make the decision to use second-kind polynomials for the following reasons: First, \( T_n(x) \) is not an ultraspherical polynomial. In particular, this means that the formulae involving \( T_n(x) \) in the next few sections must be treated separately from \( C_n^{(\lambda)}(x) \), which greatly clutters the exposition. Second, in most applications of interest the solution remains finite, so a basis which remains bounded in the computational interval is preferred. (See also the remark in section 4.1.)

2.2. Conversion operators. We consider two representations of the identity operator, \( I \), which map between different \( C_\gamma^{(\lambda)} \) spaces. First, the relationship [12, Eqn. 18.9.7]
\[
(2.10) \quad C_n^{(\lambda)}(x) = \frac{\lambda}{n + \lambda} \left( C_n^{(\lambda+1)}(x) - C_n^{(\lambda+1)}(x) \right),
\]
induces operators \( S_\lambda : C_\gamma^{(\lambda)} \to C_\gamma^{(\lambda+1)} \) defined by
\[
(2.11) \quad S_\lambda := \begin{pmatrix}
1 & 0 & \frac{-\lambda}{\lambda+2} & \frac{-\lambda}{\lambda+3} & \frac{-\lambda}{\lambda+4} & \cdots \\
\frac{\lambda}{\lambda+2} & 0 & \frac{-\lambda}{\lambda+3} & \frac{-\lambda}{\lambda+4} & \cdots \\
\frac{\lambda}{\lambda+3} & 0 & \frac{-\lambda}{\lambda+4} & \cdots \\
\frac{\lambda}{\lambda+4} & \cdots \\
\vdots & \vdots & \vdots & \ddots & \ddots
\end{pmatrix},
\]
so that if \( u \in C_\gamma^{(\lambda)} \), then \( u(x) = C_\gamma^{(\lambda)}(x)u = C_\gamma^{(\lambda+1)}S_\lambda u \). These are precisely the conversion operators \( S_\lambda \) as described in [27]. Here and in the other operators that follow, when \( S \) is acting on either \( P \) or \( U \), we shall subscript with these rather than the corresponding value of \( \lambda \). That is,
\[
(2.12) \quad S_P := S_{1/2} = \begin{pmatrix}
1 & 0 & \frac{-1}{3} & \frac{-1}{7} & \frac{-1}{11} & \cdots \\
\frac{1}{3} & 0 & \frac{-1}{7} & \frac{-1}{11} & \cdots \\
\frac{1}{7} & 0 & \frac{-1}{11} & \cdots \\
\frac{1}{11} & \cdots \\
\vdots & \vdots & \vdots & \ddots & \ddots
\end{pmatrix} \quad \text{and} \quad S_U := S_1 = \begin{pmatrix}
1 & 0 & \frac{-1}{3} & \frac{-1}{7} & \frac{-1}{11} & \cdots \\
\frac{1}{3} & 0 & \frac{-1}{7} & \frac{-1}{11} & \cdots \\
\frac{1}{7} & 0 & \frac{-1}{11} & \cdots \\
\frac{1}{11} & \cdots \\
\vdots & \vdots & \vdots & \ddots & \ddots
\end{pmatrix}.
\]
A second relationship,

\[(2.13) \quad (1 + x)C_n^{(\lambda)}(x) = \frac{n + 1}{2(n + \lambda)} C_{n+1}^{(\lambda)}(x) + C_n^{(\lambda)}(x) + \frac{n + 2\lambda - 1}{2(n + \lambda)} C_{n-1}^{(\lambda)}(x), \quad \lambda > 0,\]

which can be readily derived from the recurrence relations (2.3), induces operators

\[R_\lambda : C_\gamma^{(\lambda)} \to C_{\gamma-1}^{(\lambda)}, \quad \lambda > 0,\]

so that if \(u \in C_\gamma^{(\lambda)}\), then \(u(x) = C_\gamma^{(\lambda)}(x) u = C_{\gamma-1}^{(\lambda)} R_\lambda u\). Note in particular that \(R_U\) has the simple form

\[R_U := R_1 = \frac{1}{2} \begin{pmatrix}
2 & 1 \\
1 & 2 \\
& \ddots
\end{pmatrix}.\]

### 2.3. Multiplication operators.

As outlined in [35] and [40], polynomial multiplication can be viewed as a banded operator acting on \(C^{(\lambda)}\) spaces. In particular, the basic building block is the Jacobi operator, built out of the three-term recurrence (2.3):

\[(2.16) \quad x C_n^{(\lambda)}(x) = \frac{n + 1}{2(n + \lambda)} C_{n+1}^{(\lambda)}(x) + \frac{n + 2\lambda - 1}{2(n + \lambda)} C_{n-1}^{(\lambda)}(x).\]

In the language of section 2.1, this amounts to choosing \(\mathcal{L} = x\), inducing the operator \(J_\lambda : C_\gamma^{(\lambda)} \to C_\gamma^{(\lambda)}\) defined as

\[J_\lambda := \frac{1}{2} \begin{pmatrix}
0 & \frac{2\lambda}{1+\lambda} \\
\frac{1}{1+\lambda} & 0 \\
\frac{2}{1+\lambda} & 0 & \frac{1+2\lambda}{2+\lambda} \\
& \ddots
\end{pmatrix},\]

so that if \(u \in C_\gamma^{(\lambda)}\), then \(xu(x) = x C_\gamma^{(\lambda)}(x) u = C_\gamma^{(\lambda)} J_\lambda u\). If \(C_n^{(\ell)}(x)\) is another ultraspherical polynomial, then its corresponding three-term recurrence applied to \(J_\lambda\) gives

\[(2.18) \quad C_{n+1}^{(\ell)}(J_\lambda) = 2 \frac{n + \ell}{n + 1} J_\lambda C_n^{(\ell)}(J_\lambda) - \frac{n + 2\ell - 1}{n + 1} C_{n-1}^{(\ell)}(J_\lambda), \quad n \geq 1,\]

with \(C_0^{(\ell)}(J_\lambda) = 0\), \(C_0^{(\ell)}(J_\lambda) = 1\), and the multiplication operator \(\Pi_\lambda[C_\gamma^{(\ell)}] : C_\gamma^{(\lambda)} \to C_\gamma^{(\lambda)}\) may be defined recursively as

\[(2.19) \quad \Pi_\lambda[C_{n+1}^{(\ell)}] = 2 \frac{n + \ell}{n + 1} J_\lambda \Pi_\lambda[C_n^{(\ell)}] - \frac{n + 2\ell - 1}{n + 1} \Pi_\lambda[C_{n-1}^{(\ell)}], \quad n \geq 1,\]
where \( \Pi_\lambda[C_{-1}^{(\ell)}] = 0 \) and \( \Pi_\lambda[C_{0}^{(\ell)}] = I \). Each term in the recursion will increase the bandwidth by 1 (since \( J_\lambda \) has bandwidth 1), so \( \Pi_\lambda[C_{d}^{(\ell)}] \) is banded with bandwidth \( d \). Then, by linearity of \( \Pi_\lambda \) and the orthogonality of ultraspherical polynomials, given any degree \( d \) polynomial \( p \) we may construct

\[
(2.20) \quad \Pi_\lambda \left[ p(x) = \sum_{n=0}^{d} p_n C_{n}^{(\ell)}(x) \right] = \sum_{n=0}^{d} p_n \Pi_\lambda[C_{n}^{(\ell)}],
\]

which also has bandwidth \( d \) and satisfies \( p(x)u(x) = p(x)C_\gamma^{(\lambda)}(x)u = C_\gamma^{(\lambda)}(x)\Pi_\lambda[p(x)]u \) when \( u \in C_\gamma^{(\lambda)} \).

One may take \( \ell = \lambda \), in which case the columns of \( \Pi_\lambda[C_{d}^{(\lambda)}] \) give rise to linearization formulae for products of the form \( C_d^{(\lambda)}(x)C_n^{(\lambda)}(x) \) [12, Eqn. 18.18.22]. Alternatively, one can construct a similar recurrence relationship based on Chebyshev polynomials of the first kind, in which case

\[
(2.21) \quad \Pi_\lambda[T_{n+1}] = 2J_\lambda \Pi_\lambda[C_{n}^{(\ell)}] - \Pi_\lambda[T_{n-1}], \quad n > 1,
\]

with \( \Pi_\lambda[T_0] = 1, \Pi_\lambda[T_1] = x \), and

\[
(2.22) \quad \Pi_\lambda \left[ p(x) = \sum_{n=0}^{d} p_n T_n(x) \right] = \sum_{n=0}^{d} p_n \Pi_\lambda[T_n].
\]

Since Chebyshev coefficients of a polynomial are readily computed by a discrete cosine transform, this can often be more convenient than (2.19). When \( p \) is not a polynomial but a sufficiently differentiable function, we can approximate it to high accuracy by a polynomial. In particular, if \( p \) is analytic in some neighbourhood of \([-1, 1]\), then the polynomial approximation will converge geometrically, and the degree of the approximant (and hence bandwidth of \( \Pi_\lambda \)) will typically be small [39].

### 2.4. Integral operators

The foundation of our approach is the following formula, which shows how the half-integral of certain weighted ultraspherical polynomials may be computed in closed form.

**Theorem 2.1.** For any \( \lambda > 0, n \geq 0 \),

\[
(2.23) \quad -1 \mathcal{Q}_x^{1/2}[(1+\phi)^{\lambda-1/2}C_n^{(\lambda)}](x) = \frac{\Gamma(\lambda + 1/2)}{\Gamma(\lambda)(n + \lambda)}(1 + x)^\lambda \left( C_n^{(\lambda + 1/2)}(x) - C_{n-1}^{(\lambda + 1/2)}(x) \right).
\]

**Proof.** This follows from relating \( C_n^{(\lambda)}(x) \) to the Jacobi polynomial \( P_n^{(\lambda-1/2,\lambda-1/2)}(x) \) and using the closed-form expression for fractional integrals of weighted Jacobi polynomials [2, Theorem 6.72(b)]. Applying the symmetric version of [12, Eqn. 18.9.5] to the right-hand side and converting \( P_n^{(\lambda,\lambda)}(x) \) back to \( C_n^{(\lambda+1/2)}(x) \) yields the required result.

**Corollary 2.2.**

\[
(2.24) \quad -1 \mathcal{Q}_x^{1/2}P_n(x) = \frac{2\sqrt{1+x}}{\sqrt{\pi}(2n+1)}(U_n(x) - U_{n-1}(x))
\]

and

\[
(2.25) \quad -1 \mathcal{Q}_x^{1/2}[(1+\phi)U_n](x) = \frac{\sqrt{\pi}}{2}(P_{n+1}(x) + P_n(x)).
\]
Proof. The first follows immediately from setting $\lambda = 1/2$ in (2.23). For the second, take $\lambda = 1$ in (2.23) and make the observation (see Appendix A) that $n(P_n(x) + P_{n-1}(x)) = (1 + x)(C_n^{(3/2)}(x) - C_{n-1}^{(3/2)}(x))$.  

We may therefore, in the language of section 2.1, consider half-integration as a banded operator between the spaces of Legendre polynomials and weighted Chebyshev polynomials and define the associated banded half-integer order integral operators $Q_{1/2}^P : P \rightarrow U_{1/2}$ and $Q_{1/2}^U : U_{1/2} \rightarrow P$ as

$$Q_{1/2}^P := \frac{2}{\sqrt{\pi}} \begin{pmatrix} 1 & -\frac{1}{3} & \frac{1}{3} & -\frac{1}{5} & \frac{1}{5} & \cdots \end{pmatrix}$$ and $$Q_{1/2}^U := \frac{\sqrt{\pi}}{2} \begin{pmatrix} 1 & 1 & 1 & \cdots \end{pmatrix},$$

respectively. Therefore, letting $u_P \in P$ and $u_{U_{1/2}} \in U_{1/2}$, we have that $-1 Q_{1/2}^P P(x) u_P = U_{1/2}(x) Q_{1/2}^P u_P$ and $-1 Q_{1/2}^U U_{1/2}(x) u_{U_{1/2}} = P(x) Q_{1/2}^U u_{U_{1/2}}$.  

If we define $Q_P := Q_{1/2}^P Q_{1/2}^P$ so $Q_P : P \rightarrow P$ is given by

$$Q_P = \begin{pmatrix} 1 & -\frac{1}{3} & 1 & 0 & -\frac{1}{5} & \frac{1}{5} & \cdots \end{pmatrix},$$

we see that this is consistent with the relation

$$-1 Q_{1/2}^P P_n(x) = \int_{-1}^{1} P_n(t) dt = \begin{cases} \frac{1}{2\pi^2} (P_{n+1}(x) - P_{n-1}(x)), & n \geq 1, \\ P_0(x) + P_1(x), & n = 0, \end{cases}$$

for the integral of Legendre polynomials (which can be obtained from [12, Egn. 18.17.1] and [12, Egn. 18.9.6]). We may go further and repeatedly combine the $Q_P$ and $Q_{U_{1/2}}$ operators to define banded operators between the spaces $P$ and $U_{1/2}$ representing integral operators of half-integer order by repeatedly applying the matrices (2.26), i.e., $-1 Q_m^P$ and $-1 Q_m^{U_{1/2}}$. In particular, we have

$$Q_{m+1/2}^P := Q_{1/2}^P \left( Q_{1/2}^U Q_{1/2}^P \right)^m : P \rightarrow U_{1/2}$$

(2.29)

$$Q_{m+1/2}^{U_{1/2}} := Q_{1/2}^U \left( Q_{1/2}^U Q_{1/2}^P \right)^m : U_{1/2} \rightarrow P.$$  

Integral operators of integer order, $Q_m$, acting on these same spaces give rise to $m$-order banded operators $Q_m^P : P \rightarrow P$ and $Q_m^{U_{1/2}} : U_{1/2} \rightarrow U_{1/2}$, which can be constructed likewise by omitting the terms outside the parentheses in (2.29) and (2.30), respectively, or from (2.27).  

\[ \text{Henceforth, we cease (with a few exceptions) to explicitly state such equalities for each operator} \]

\[ \text{we introduce. It should be clear from the context which continuous operator is in question and the} \]

\[ \text{range and domain of the discrete operator from the notation introduced in (2.9).} \]
2.5. Differentiation operators. The final key ingredient for the US method for ODEs is the relationship

\begin{equation}
\frac{d}{dx} C^{(\lambda)}(x) = 2\lambda C^{(\lambda+1)}(x).
\end{equation}

This induces banded derivative operators \( D_\lambda : C^\lambda \to C^{\lambda+1} \) and, more generally, \( D^{(m)}_\lambda : C^\lambda \to C^{\lambda+m} \), defined by

\begin{equation}
D_\lambda := 2\lambda \begin{pmatrix} 0 & 1 & \ddots \\ 1 & & \ddots \\ & \ddots & \ddots \end{pmatrix} \quad \text{and} \quad D^{(m)}_\lambda := 2^m \lambda^{(m)} \begin{pmatrix} \vdots & \ddots & \ddots \\ & 1 & \ddots \\ & & \ddots & 1 \end{pmatrix},
\end{equation}

respectively, where \( \lambda^{(m)} = \lambda(\lambda + 1) \ldots (\lambda + m - 1) \) is the Pochhammer function or “rising factorial.”

We now derive similar such operators for half-integer order derivatives of weighted ultraspherical polynomials. For now we consider only the RL definition, for which we have the following.

**Corollary 2.3.**

\begin{equation}
-1^{1/2} D^{1/2}_x P_n(x) = \frac{1}{\sqrt{x+1}} U_n(x) + U_{n-1}(x)
\end{equation}

and

\begin{equation}
-1^{1/2} D^{1/2}_x \sqrt{x+1} U_n(x) = \frac{\sqrt{x}}{2} \left( C^{(3/2)}(x) + C^{(3/2)}_{n-1}(x) \right).
\end{equation}

**Proof.** The second equation follows immediately from differentiating (2.25) in Corollary 2.2 using (2.31). For the first equation, differentiate the right-hand side of (2.25) via the product rule and make the observation (see Appendix A) that \( 2(1+x) \left( C^{(2)}_{n-1}(x) - C^{(2)}_{n-2}(x) \right) = nU_n(x) + (n+1)U_{n-1}(x) \).

Therefore, similarly to the case of half-integrals above, we may consider half-differentiation as a banded operator acting on \( P \) and \( U^{1/2} \) but now mapping to \( U^{1/2} \) and \( C^{(3/2)} \), respectively. In particular, we have half-derivative operators \( D^{1/2}_P : P \to U^{1/2} \) and \( D^{1/2}_{U^{1/2}} : U^{1/2} \to C^{(3/2)} \), given by the banded infinite-dimensional matrices

\begin{equation}
D^{1/2}_P := \frac{1}{\sqrt{\pi}} \begin{pmatrix} 1 & 1 & \ddots \\ 1 & 1 & \ddots \\ & \ddots & \ddots \end{pmatrix} \quad \text{and} \quad D^{1/2}_{U^{1/2}} := \frac{\sqrt{\pi}}{2} \begin{pmatrix} 1 & 1 & \ddots \\ & 1 & \ddots \\ & & \ddots \end{pmatrix}
\end{equation}

so that \( -1^{1/2} D^{1/2}_P(x) u_P = U^{1/2}(x) D^{1/2}_P u_P \) and \( -1^{1/2} D^{1/2}_{U^{1/2}}(x) u_{U^{1/2}} = C^{(3/2)}(x) D^{1/2}_{U^{1/2}} u_{U^{1/2}} \).

Since the composition of these operators is no longer a mapping between the same space, we cannot construct higher-order derivatives by repeated multiplication as we did higher-order integral operators, i.e., (2.29) and (2.30). However, applying (2.31) \( m \) times to \( P_n(x) \) and (2.34), we may readily write

\begin{equation}
\frac{d^m}{dx^m} P_n(x) = 2^m \left( \frac{1}{2} \right)^{(m)} C^{(m+1/2)}_{n-m}(x) = \frac{2^m \Gamma(m+1/2)}{\sqrt{\pi}} C^{(m+1/2)}_{n-m}(x)
\end{equation}
and

\[
R_L D_x^{m+1/2} \sqrt{1 + x} U_n(x) = 2^m \Gamma(m + 3/2) \left( C_n^{(m+3/2)}(x) + C_n^{(m+3/2)}(x) \right)
\]

and can consider derivative operators \( D^{m}_P : P \to C^{(m+1/2)} \) and \( D^{m+1/2}_U : U_{1/2} \to C^{(m+3/2)} \) defined by

\[
D^{m}_P := \frac{2^m \Gamma(m + 1/2)}{\sqrt{\pi}} \begin{pmatrix} 0 & \cdots & 0 & 1 \\ \vdots & \ddots & \ddots & \ddots \\ 0 & \cdots & 1 & \vdots \\ 1 & \cdots & \ddots & \ddots \\ \end{pmatrix}
\]

and

\[
D^{m+1/2}_U := 2^m \Gamma(m + 3/2) \begin{pmatrix} 0 & \cdots & 0 & 1 & 1 \\ \vdots & \ddots & \ddots & \ddots & \ddots \\ 0 & \cdots & 1 & 1 & \vdots \\ 1 & \cdots & \ddots & \ddots & \ddots \\ \end{pmatrix},
\]

representing \( D^m \) and \( -1 D_x^{m+1/2} \), respectively.

Remark: Observe that \( D^{m}_P \) and \( D^{m+1/2}_U \) are banded with bandwidths \( m \) and \( m+1 \), respectively.

The corresponding operators for \( \frac{d^m}{dx^m} \sqrt{1 + x} U_n(x) \) and \( R_L D_x^{m+1/2} P_n(x) \) are complicated by the \( \sqrt{1 + x} \) weights. Here we must appeal to the product rule for differentiation and derive recursive formulations for \( D^{m}_U \) and \( D^{m+1/2}_P \). We first note the following.

**Lemma 2.4.** For any \( n \geq 0, \lambda > 0, \mu \neq 0 \),

\[
\frac{d}{dx} (1 + x)^\mu C_n^{(\lambda)}(x)
\]

\[
= \lambda (1 + x)^{\mu - 1} \left[ \left( 1 + \frac{\mu - \lambda}{n + \lambda} \right) C_n^{(\lambda+1)}(x) + 2C_{n-1}^{(\lambda+1)}(x) + \left( 1 - \frac{\mu - \lambda}{n + \lambda} \right) C_n^{(\lambda-1)}(x) \right],
\]

where \( C_{-2}^{(\lambda+1)}(x) := 0 \).

**Proof.** Apply the product rule to the left-hand side, then use (2.10), (2.13), and (2.31).

If we define \( D_{\mu,\lambda} : C^{(\lambda)}_\mu \to C^{(\lambda+1)}_{\mu-1} \) as the differentiation operator induced by this relationship, i.e.,

\[
D_{\mu,\lambda} := \lambda \begin{pmatrix} 1 & 2 & 1 \\ 1 & 2 & 1 \\ \vdots & \ddots \end{pmatrix} + \lambda(\mu - \lambda) \begin{pmatrix} \frac{1}{\lambda} & 0 & -\frac{1}{\lambda+2} \\ \frac{1}{\lambda+1} & 0 & -\frac{1}{\lambda+3} \\ \frac{1}{\lambda+2} & 0 & \ddots \end{pmatrix},
\]

(2.40)
we may then define \( D^m_{U_{1/2}} : U_{1/2} \to C^{(m+1)}_{-m+1/2} \) and \( D^{m+1/2}_P : P \to C^{(m+1)}_{-m-1/2} \) as

\[
D^m_{U_{1/2}} := \prod_{k=0}^{m-1} D_{-k+\frac{1}{2},k+1} \quad \text{and} \quad D^{m+1/2}_P := \left( \prod_{k=0}^{m-1} D_{-k-\frac{1}{2},k+1} \right) D^{1/2}_P,
\]

respectively.

**Remark.** Since the \( D_{\mu,\lambda} \) operators each have bandwidth 2 (and recalling that \( D^{1/2}_P \) has bandwidth 1), it is readily verified that \( D^m_{U_{1/2}} \) and \( D^{m+1/2}_P \) will have bandwidths \( 2m \) and \( 2m + 1 \), respectively.

### 2.6. Block operators.

As we shall see in the next section, our approach for solving FIEs and FDEs of half-integer order will be to seek solutions formed as a direct sum of two different weighted ultraspherical polynomials, namely, \( P \oplus U_{1/2} \). Here we introduce some notation to clarify the exposition in the description of the algorithm that follows.

First, suppose \( A \) and \( B \) are two different \( C^{(\lambda)} \) spaces. We define \( [A \oplus B](x) := [A(x), B(x)] \), and if \( u = [a^\top, b^\top]^\top \), where \( a \in A \) and \( b \in B \), then we say \( u \in A \oplus B \) and may write

\[
(2.42) \quad u(x) = [A(x), B(x)]y = \sum_{n=0}^\infty a_n A_n(x) + \sum_{n=0}^\infty b_n B_n(x).
\]

Then for any \( m \in \mathbb{N} \) we define

\[
(2.43) \quad Q^{m/2} := \begin{pmatrix} 0 & Q^{1/2}_{U_{1/2}} \\ Q^{1/2}_P & 0 \end{pmatrix} : P \oplus U_{1/2} \to P \oplus U_{1/2},
\]

\[
(2.44) \quad D^m := \begin{pmatrix} D^m_P & 0 \\ 0 & D^m_{U_{1/2}} \end{pmatrix} : P \oplus U_{1/2} \to C^{(m+1/2)} \oplus C^{(m+1)}_{-m+1/2},
\]

and

\[
(2.45) \quad D^{m+1/2} := \begin{pmatrix} 0 & D^{m+1/2}_U \\ D^{m+1/2}_P & 0 \end{pmatrix} : P \oplus U_{1/2} \to C^{(m+3/2)} \oplus C^{(m+1)}_{-m-1/2},
\]

corresponding to half-integer order integral and derivative operators, respectively. We then have, for example, that if \( u = [\underline{u}_P, \underline{u}_{U_{1/2}}]^\top \), with \( \underline{u}_P \in P \) and \( \underline{u}_{U_{1/2}} \in U_{1/2} \), then

\[
(2.46) \quad Q^{m/2}u(x) = Q^{m/2} [P(x), U_{1/2}(x)]u = [P(x), U_{1/2}(x)]Q^{m/2}u
\]

and

\[
(2.47) \quad D^m u(x) = D^m [P(x), U_{1/2}(x)]u = [C^{(m+1/2)}(x), C^{(m+1)}_{-m+1/2}(x)]D^m u.
\]

For convenience we also introduce the block conversion operators \( E_m : C^{(\ell)}_{\gamma_{01}} \oplus C^{(\ell)}_{\gamma_{02}} \to C^{(\ell)}_{\gamma_{1}} \oplus C^{(\ell)}_{\gamma_{2}} \) and \( E_{m+1/2} : C^{(m+1/2)}_{\gamma_{01}} \oplus C^{(m+1/2)}_{\gamma_{02}} \to C^{(m+1/2)}_{\gamma_{1}} \oplus C^{(m+1/2)}_{\gamma_{2}} \), \( m \in \mathbb{N}^+ \) defined by

\[
(2.48) \quad E_m := \begin{pmatrix} I & 0 \\ 0 & S_m \end{pmatrix} \quad \text{and} \quad E_{m+1/2} := \begin{pmatrix} S_{m+1/2} & 0 \\ 0 & R_{m+1} \end{pmatrix}.
\]
Remark. Note that all of the operators in (2.43)–(2.48) are banded or block-banded. Block-banded matrices become banded when the coefficients are interleaved, which is expanded on below.

As described in section 2.3, polynomial multiplication also results in banded operators. In particular, multiplication by a polynomial \( r(x) \in \mathbb{P}^d \) yields the following \( d \)-banded operator \( \Pi_0[r] : \mathbb{P} \oplus \mathbb{U}_{1/2} \to \mathbb{P} \oplus \mathbb{U}_{1/2} \):

\[
\Pi_0[r] := \begin{pmatrix} \Pi_P[r] & \Pi_U[r] \end{pmatrix}.
\]

(2.49)

More generally, for integer values \( m \), we define multiplication operators

\[
\Pi_m[r] := \begin{pmatrix} \Pi_{m+1/2}[r] \\ \Pi_{m+1}[r] \end{pmatrix}, \quad \Pi_{m+1/2}[r] := \begin{pmatrix} \Pi_{m+3/2}[r] \\ \Pi_{m+1}[r] \end{pmatrix},
\]

(2.50)

which act on the appropriate direct sum spaces.

Multiplication by square root-weighted polynomials, \( \sqrt{1 + x}s(x) \), \( s \in \mathbb{P}^d \), is complicated by the need to convert between \( C^{(\lambda)} \) and \( C^{(\lambda+1/2)} \) bases. For example, in the case of \( \sqrt{1 + x}s(x) \) multiplying a vector in \( \mathbb{P} \oplus \mathbb{U}_{1/2} \) we require the upper triangular conversion or “connection” operators \( \hat{M} : \mathbb{P} \to \mathbb{U} \) and \( \hat{L} : \mathbb{U} \to \mathbb{P} \) [1, 38] so that

\[
\Pi_0[0,s] := \begin{pmatrix} \Pi_U[s] \hat{M} \\ \Pi_U[(1 + \diamond)s] \end{pmatrix}.
\]

(2.51)

More generally, multiplication by \( r(x) + \sqrt{1 + x}s(x) \) yields the operator \( \Pi_0[r,s] : \mathbb{P} \oplus \mathbb{U}_{1/2} \to \mathbb{P} \oplus \mathbb{U}_{1/2} \):

\[
\Pi_0[r,s] := \begin{pmatrix} \Pi_P[r] \\ \Pi_U[s] \hat{M} \\ \Pi_U[(1 + \diamond)s] \end{pmatrix}.
\]

(2.52)

Remark. \( \Pi_0[0,s] \) and \( \Pi_0[r,s] \) are neither banded nor block-banded; however, they are (upon reordering) lower-banded. We give an example with such a weighted nonconstant coefficient below but will otherwise limit our attention to the case when the nonconstant coefficients are smooth (i.e., well approximated by an unweighted polynomial).

3. Half-integer order integral equations. We now use the operators described above to derive an algorithm for integral equations of half-integer order.

3.1. Half-integral equations. We first consider Abel-like integral equations of the form

\[
\sigma u(x) + \int_{-\lambda}^{\lambda} Q_x^{1/2} u(x) = e(x) + \sqrt{1 + x}f(x), \quad x \in [-1,1],
\]

(3.1)

where \( e(x) \) and \( f(x) \) are smooth (typically analytic in some neighborhood of \( [-1,1] \)) and \( \sigma > 0 \).

\[\text{Remark.} \quad \text{We use} \ \hat{L} \ \text{and} \ \hat{M} \ \text{here, as} \ L \ \text{and} \ M \ \text{are typically used to denote the conversion operators} \ \hat{M} : \mathbb{P} \to \mathbb{T} \ \text{and} \ \hat{L} : \mathbb{T} \to \mathbb{P}, \ \text{respectively, where} \ T \ \text{is the quasimatrix whose columns are formed of Chebyshev polynomials of the first kind,} \ T_n(x).\]
Motivated by the block operators in section 2.6, we make the ansatz that the solution \( u(x) \) may be expressed as a linear combination of Legendre polynomials and weighted Chebyshev polynomials:

\[
u(x) = \sum_{n=0}^{\infty} a_n P_n(x) + \sqrt{1 + x} \sum_{n=0}^{\infty} b_n U_n(x).
\]

(3.2)

Assuming the coefficients \( a_0, a_1, \ldots \) and \( b_0, b_1, \ldots \) satisfy the conditions (2.5), we may, in the language of section 2, write

\[
u(x) = [P(x), U_{1/2}(x)] \begin{pmatrix} a \\ b \end{pmatrix}.
\]

(3.3)

Applying the block half-integral operator (2.43) with \( m = 0 \), we have that

\[
-1 Q^{1/2} u(x) = [P(x), U_{1/2}(x)] Q^{1/2} \begin{pmatrix} a \\ b \end{pmatrix}
\]

(3.4)

and hence

\[
su(x) + -1 Q^{1/2} u(x) = [P(x), U_{1/2}(x)] (\sigma I + Q^{1/2}) \begin{pmatrix} a \\ b \end{pmatrix}.
\]

(3.5)

Letting

\[
e(x) = \sum_{n=0}^{\infty} e_n P_n(x) \quad \text{and} \quad f(x) = \sum_{n=0}^{\infty} f_n U_n(x)
\]

(3.6)

or, equivalently,

\[
e(x) + \sqrt{1 + x} f(x) = [P(x), U_{1/2}(x)] \begin{pmatrix} e \\ f \end{pmatrix}
\]

(3.7)

and equating coefficients, we arrive at the (infinite-dimensional) linear system of equations

\[
(\sigma I + Q^{1/2}) \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} \sigma I & Q^{1/2} \\ Q^{1/2} & \sigma I \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} e \\ f \end{pmatrix}.
\]

(3.8)

Note that both the diagonal and the off-diagonal blocks of the operator in (3.8) are banded, and by interleaving the coefficients in \( a \) and \( b \) (i.e., \( (a^T, b^T) \mapsto (a_0, b_0, a_1, b_1, \ldots)^T \)) we arrive at a banded operator (in this case, tridiagonal). Taking a finite section approximation (i.e., truncating each of the summations in (3.2) and (3.6) and hence the block operators in (3.8)) at a suitable length \( N \), we arrive at a \( 2N \times 2N \) tridiagonal matrix system, which can be solved directly in \( O(N) \) floating point operations for the approximate coefficients \( a \) and \( b \) of \( u(x) \) in (3.2). Alternatively, one can use the adaptive QR approach described in [27] and solve the infinite-dimensional (3.8) system to a required accuracy without a priori truncation (see section 6.2). Convergence and stability are discussed in more detail in Appendix B.

**Example 1.** We consider the second-kind Abel integral equation

\[
u(x) + -1 Q^{1/2} u(x) = 1
\]

(3.9)

Formally this direct sum-space defines a frame [8]. We discuss the consequences of this in section 6.1 and Appendix B.1.
with solution [30, section 11.4-1]

\begin{equation}
    u(x) = e^{1+x} \text{erfc}(\sqrt{1+x}),
\end{equation}

where \text{erfc} is the complimentary error function. Note that the solution \( u(x) \) takes the form \( p(x) + \sqrt{1+x}q(x) \), where \( p(x) \) and \( q(x) \) are functions analytic in some neighborhood of \([-1,1] \), and so any attempt to approximate \( u(x) \) by a polynomial \( u(x) \approx p_N(x) \) or a weighted polynomial \( u(x) \approx \sqrt{1+x}q_N(x) \) will achieve only algebraic convergence as the degree of the polynomial is increased. However, using the direct sum basis \( P \oplus U_{1/2} \) we are able to approximate such a function and hence solve the FDE (3.9) with spectral accuracy.

To solve the FDE (3.9) we form the linear system (3.8) with \( \sigma = 1, \xi = [1, 0, \ldots]^T \), and \( f = [0, 0, \ldots]^T \); truncate at some length \( N \); and solve with \( \backslash \) in MATLAB. The results are shown in Figure 3.1. The first image shows a plot of the computed solution with \( N = 20 \). The \textit{spy} plot in the second image verifies that, upon reordering, the associated linear system (3.8) is tridiagonal. The third image shows two measures of the error in the approximation as \( N \) varies. Solid line: Infinity norm error of solution approximated on a 100-point equally spaced grid. Dashed line: 2-norm difference between the coefficients of the approximated solution when truncating at sizes \( N \) and \([1.1N]\). In both cases, geometric convergence is observed.

\begin{figure}[h]
  \centering
  \includegraphics[width=\textwidth]{figure3.png}
  \caption{(a) Approximate solution to (3.9). (b) MATLAB \textit{spy} plot showing that truncated linear system (3.8) is tridiagonal. (c) Two measures of the error in the approximation as \( N \) varies. Solid line: Infinity norm error of solution approximated on a 100-point equally spaced grid. Dashed line: 2-norm difference between the coefficients of the approximated solution when truncating at sizes \( N \) and \([1.1N]\). In both cases, geometric convergence is observed.}
\end{figure}

3.2. Half-order integral equations with nonconstant coefficients. In much the same way as in the US method for ODEs [27], the approach outlined above is readily extended to FDEs with nonconstant coefficients. In particular, to solve problems of the form

\begin{equation}
    u(x) + r(x) -1 Q^{1/2}_2 u(x) = e(x) + \sqrt{1+x}f(x),
\end{equation}

we can appeal to the discussion in section 2.3 and construct multiplication operators \( \Pi_P[r] \) and \( \Pi_U[r] \), which act on Legendre and second-kind Chebyshev series,
respectively. If \( r(x) \) is a polynomial of degree \( d \), then these two operators will have bandwidth \( d \). If \( r(x) \) is not a polynomial but is sufficiently smooth (for example, analytic), then the coefficients in its ultraspherical polynomial expansion will decay rapidly, and we may consider \( \Pi_P[r] \) and \( \Pi_U[r] \) as banded for practical purposes. The resulting linear system then takes the form

\[
(I + \Pi_0[r]Q^{1/2})(\begin{pmatrix} a \\ b \end{pmatrix}) = \begin{pmatrix} e \\ f \end{pmatrix},
\]

where \( \Pi_0[r] \) is defined in section 2.6.

We may also consider problems of the form

\[
u(x) + r_1(x) - i \sqrt{1 + x}^2 [r_2 u](x) = e(x) + \sqrt{1 + x} f(x),
\]

in which case the linear system becomes

\[
(I + \Pi_0[r_1]Q^{1/2}\Pi_0[r_2])(\begin{pmatrix} a \\ b \end{pmatrix}) = \begin{pmatrix} e \\ f \end{pmatrix}.
\]

Similarly, problems of the form

\[
u(x) + (r(x) + \sqrt{1 + x} s(x)) - i Q_x^{1/2} u(x) = e(x) + \sqrt{1 + x} f(x)
\]

and

\[
u(x) + -i Q_x^{1/2}(r + \sqrt{1 + x} s) u(x) = e(x) + \sqrt{1 + x} f(x)
\]

may also be solved with linear systems

\[
(I + \Pi_0[r, s]Q^{1/2})(\begin{pmatrix} a \\ b \end{pmatrix}) = \begin{pmatrix} e \\ f \end{pmatrix}
\]

and

\[
(I + Q^{1/2}\Pi_0[r, s])(\begin{pmatrix} a \\ b \end{pmatrix}) = \begin{pmatrix} e \\ f \end{pmatrix},
\]

respectively. However, these systems are no longer banded, and we will lose the linear complexity of our algorithm. Upon reordering they are dense above the diagonal and banded below, so Gaussian elimination will have \( O(N^2) \) complexity (see Example 3 below).

**Example 2.** We adapt the example of section 1 so that

\[
u(x) + e^{-(1+x)/2} - i Q_x^{1/2}[e^{(1+x)/2} u](x) = e^{-(1+x)/2}
\]

with solution

\[
u(x) = e^{(1+x)/2} \text{erfc}(\sqrt{1 + x}).
\]

Again taking \( u(x) \) as in (3.2), the resulting linear system defining the coefficients \( a \) and \( b \) is of the form

\[
(I + \Pi[e^{-(1+x)/2}Q^{1/2}\Pi_0[e^{(1+x)/2}])(\begin{pmatrix} a \\ b \end{pmatrix}) = \begin{pmatrix} e \\ 0 \end{pmatrix},
\]
A FAST ALGORITHM FOR RATIONAL-ORDER FIEs & FDEs

Fig. 3.2. (a) Solution to (3.19). (b) MATLAB spy plot of (3.21) showing the banded structure. (c) Solid line: Infinity norm error of solution approximated on a 100-point equally spaced grid. Dashed line: 2-norm difference between the coefficients of the approximated solution when truncating at sizes $N$ and $\lceil 1.1N \rceil$. Again, geometric convergence is observed.

where $e$ are the Legendre coefficients of the function $e^{-(1+x)/2}$. (See section 6.1 for discussion on how these are computed.) As in the previous example, we truncate each of the expansions at a suitable length $N$ and solve the resulting finite-dimensional banded matrix problem using \ in MATLAB.

The results are shown in Figure 3.2. The left panel shows the approximate solution computed with $N = 20$. The center panel shows a spy plot of the discretized, truncated, and reordered operator (3.21). The nonconstant coefficients in (3.19) mean the resulting matrix is no longer tridiagonal; however, it is banded independently of $N$ and can be solved by \ in linear time as $N \to \infty$. The precise bandwidth depends on the number of Chebyshev coefficients required to approximate the nonconstant coefficients, in this case $e^{\pm(1+x)/2}$, to machine-precision accuracy. Here the number of coefficients required is around 14, and the resulting matrix has a bandwidth of approximately 43. The final panel shows the accuracy of the computed solution as $N$ increases, using the same two forms of the error estimate as described in Example 1. Geometric convergence is again observed, but here both measures of the error plateau due to rounding error in the computation of the Chebyshev coefficients of the functions $e^{\pm(1+x)/2}$.

Example 3. Here we solve

$$u(x) - \text{erfc}\sqrt{1 + x} - 1 = 1.$$  \hspace{1cm} (3.22)

The linear system satisfied by the coefficients $a$ and $b$ is then of the form

$$
\begin{pmatrix}
I + \Pi_0
-1, \frac{\text{erf}(\sqrt{1 + x})}{\sqrt{1 + x}}
\end{pmatrix}Q^{1/2}
\begin{pmatrix}
a \\
b
\end{pmatrix}
= 
\begin{pmatrix}
1 \\
0
\end{pmatrix}.
$$  \hspace{1cm} (3.23)

We may truncate and solve (3.23), and the results of such are shown in Figure 3.3. Here, in the spy plot in the center panel we see that, as expected, the required change of bases $P \leftrightarrow U_{1/2}$ are no longer banded, and hence neither is the reordered version of (3.23). However, the reordered matrix is lower-banded, and MATLAB’s \ will require $O(N^2)$ operations to solve such systems directly via Gaussian elimination.

In this case we do not know an explicit form for the solution, and so the right panel shows only the accuracy estimate based upon comparison of successive approximations. We again observe geometric convergence in the number of degrees of freedom until convergence plateaus at around the level of machine precision.
3.3. Higher-order integral equations. The approach outlined in the previous few examples extends readily to higher-order integral equations of half-integer order. The general form of the problem we consider is

\begin{equation}
\mathcal{L}u(x) = e(x) + \sqrt{1 + xf(x)},
\end{equation}

where

\begin{equation}
\mathcal{L}u(x) = \alpha^0(x)u(x) + \sum_{k=1}^{2m} \alpha^k(x) -1Q_x^{k/2}[\beta^k]u(x),
\end{equation}

\begin{equation}
\alpha^k(x) = p^k(x) + \sqrt{1+xf^k(x)}, \quad \beta^k(x) = r^k(x) + \sqrt{1+xs^k(x)}, \quad k = 0, 1, \ldots, 2m,
\end{equation}

and all the functions $e(x), f(x), p^k(x), q^k(x), r^k(x), s^k(x), k = 0, 1, \ldots, 2m$ are assumed analytic in some neighborhood of $[-1, 1]$. If we continue to take (3.2) as our ansatz, then we arrive at the infinite-dimensional linear system

\begin{equation}
\begin{pmatrix}
P_0^0 \begin{bmatrix} p^0 \end{bmatrix}, q^0 \end{bmatrix} + \sum_{k=1}^{2m} P_0^0 \begin{bmatrix} p^k \end{bmatrix}, q^k \end{bmatrix} Q_x^{k/2}P_0^0 \begin{bmatrix} r^k \end{bmatrix}, s^k \end{bmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} e \\ f \end{pmatrix},
\end{equation}

where $e$ and $f$ are as in (3.6).

As before, the precise form of this operator will depend on both $m$ and the number of Chebyshev coefficients required to represent the functions $p^k(x), q^k(x), r^k(x), s^k(x)$. However, if the $q^k(x)$ and $s^k(x)$ are all identically zero, then (after reordering) the operator will remain banded independently of $N$. Otherwise, it will be lower-bounded, as in Example 3.

Example 4. For simplicity, we choose a constant coefficient problem so that the banded structure of the resulting operator is readily observed. In particular, we solve

\begin{equation}
u(x) - 1Q_x^{1/2}u(x) + -1Q_x^1u(x) - -1Q_x^{3/2}u(x) + -1Q_x^2u(x) = 1,
\end{equation}

which may be expressed as

\begin{equation}
\begin{pmatrix} I - Q^{1/2} + Q^1 - Q^{3/2} + Q^2 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}.
\end{equation}
A FAST ALGORITHM FOR RATIONAL-ORDER FIEs & FDEs

Fig. 3.4. (a) Approximate solution to (3.28). (b) MATLAB spy plot of (3.29) showing the banded structure of the linear system. (c) 2-norm difference between the coefficients of the approximated solution when truncating at sizes \(N\) and \([1.1N]\). Even for higher-order problems, geometric convergence is obtained. As in Example 1, since there are no constant coefficients whose Chebyshev coefficients must be computed, this measure of the error continues to converge below machine precision.

As in the previous examples, we truncate this operator at a given size \(N\) and solve the resulting finite-dimensional problem. The results are shown in Figure 3.4 with the first and second panels showing the approximated solution and spy plot of (3.29) when \(N = 20\), respectively. The third panel shows the convergence of the solution. As in Example 3, we do not know the true solution, so we estimate the error by the 2-norm difference between the coefficients of the approximated solution when truncating at sizes \(N\) and \([1.1N]\). We again observe geometric convergence, and as in Example 1, the error continues to decrease even below the level of machine precision for this constant coefficient problem.

4. FDEs: Riemann–Liouville definition. Our approach here for FDEs of the RL type will be similar to the FIEs above. The main difference will stem from the fact that the operators \(D^m\) and \(D^{m+1/2}\) defined in section 2.6 no longer map to the same direct sum spaces and we must make use of the block-banded conversion operators \(E_m\) and \(E_{m+1/2}\) (analogous to how the conversion operators \(S_{\lambda}\) are used in [27]).

4.1. Differential equation of order 1/2. Consider the FDE

\[
(4.1) \quad u(x) + \frac{RL}{-1} D_x^{1/2} u(x) = e(x) + \frac{1}{\sqrt{1+x}} f(x), \quad x \in [-1, 1], \quad u(-1) < \infty,
\]

(sometimes called a “fractional relaxation equation”) and make the same ansatz as before that

\[
(4.2) \quad u(x) = \sum_{n=0}^{\infty} a_n P_n(x) + \sqrt{1+x} \sum_{n=0}^{\infty} b_n U_n(x) = [P(x), U_{1/2}(x)] \left( \begin{array}{c} a \\ b \end{array} \right).
\]

From section 2.6 we have that

\[
(4.3) \quad \frac{RL}{-1} D_x^{1/2} u(x) = \left[ C^{(3/2)}(x), U_{-1/2}(x) \right] D^{1/2} \left( \begin{array}{c} a \\ b \end{array} \right),
\]

where \(D^{1/2}\) is defined in (2.44). As mentioned above and unlike in the case of the integral equations, the range of \(\frac{RL}{-1} D_x^{1/2}\) is not the same as that of \(u(x)\). However,
we can find a banded transform from \( P \oplus U_{1/2} \) to \( C^{(3/2)}(x) \oplus U_{-1/2} \) using \( E_{1/2} \) so that
\[
(4.4) \quad u(x) = \left[ C^{(3/2)}(x), U_{-1/2}(x) \right] E_{1/2} \begin{pmatrix} \frac{a}{b} \end{pmatrix}.
\]
This time letting
\[
(4.5) \quad e(x) + \frac{1}{\sqrt{1+x}} f(x) = \left[ C^{(3/2)}(x), U_{-1/2}(x) \right] \begin{pmatrix} \frac{e}{f} \end{pmatrix}
\]
and equating coefficients leads to the linear system of equations
\[
(4.6) \quad \left( E_{1/2} + D^{1/2} \right) \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} S_P \\ D_{1/2}^{1/2} \\ D_{1/2}^{1/2} R_U \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} e \\ f \end{pmatrix}.
\]
Again, each block of the operators in (4.6) are banded, and by interleaving the coefficients so that \( (a^T, b^T) \mapsto (a_0, b_0, a_1, b_1, \ldots)^T \) we can convert the above to a banded system.

**Remark.** One can show that the null space of the operator on the left-hand side of (4.1) acting on functions in \( L_1 \) is
\[
(4.7) \quad v(x) = \frac{E_{1/2,1/2}(-\sqrt{1+x})}{\sqrt{1+x}},
\]
where \( E_{1/2,1/2} \) is the Mittag–Leffler function [12, Eqn. 10.46.3] [6, p. 13], which is unbounded at \( x = -1 \). Since our trial space (4.2) contains only bounded functions on \([-1, 1]\), we need not enforce a boundary condition explicitly in this case. We discuss boundary constraints in more detail momentarily. To allow solutions which are unbounded at the left end of the domain, one possibility is to use instead the ansatz \( u(x) = P(x)a + T_{-1/2}(x)b \), where \( T(x) \) is the quasimatrix whose columns are formed of Chebyshev polynomials of the first kind, \( T_n(x) \), \( n = 0, 1, \ldots \). One can derive similar banded operators to all those introduced in section 2, but we omit the details (which are complicated by the fact that \( T_n(x) \) is not an ultraspherical polynomial and therefore many of the formulae in section 2 differ subtly).

**Example 5.** We consider a modification of the second-kind Abel integral equation in Example 1, namely,
\[
(4.8) \quad u(x) + \int_{-1}^x D_x^{1/2} u(x) = \frac{1}{\sqrt{\pi} \sqrt{1+x}}, \quad u(-1) < \infty,
\]
with solution
\[
(4.9) \quad u(x) = e^{1+x} \text{erfc}(\sqrt{1+x}).
\]
To solve we choose an \( N \) and form the system (4.6) with \( e = 0 \) and \( f = [1/\sqrt{\pi}, 0, 0, \ldots]^T \). Results are shown in Figure 4.1. As usual, the first two panels show a plot of the solution and a spy plot of the discretized operator when \( N = 20 \). The spy plot verifies that the matrix is banded (here with bandwidth 4) and hence that the system (4.6) can be solved in linear time with MATLAB’s \( \setminus \). The final panel shows both the infinity norm error (approximated on a 100-point equally spaced grid) of computed solution compared to the exact solution (4.9) (solid line) and the 2-norm difference between the coefficients of the approximated solution when truncating at sizes \( N \) and \( \lceil 1.1N \rceil \) (dashed line). Again we observe geometric convergence.
4.2. Differential equation of order 1 and 1/2. Here we consider

\begin{equation}
u(x) + RL_{-1} D^{1/2}_x u(x) + u'(x) = e(x) + \frac{1}{\sqrt{1+x}} f(x), \quad x \in [-1, 1], \end{equation}

along with a suitable initial or boundary condition or some other functional constraint (see below). Now

\begin{equation}u'(x) = \left[ C^{(3/2)}(x), C^{(2)}_{-1/2}(x) \right] D \left( \frac{a}{b} \right), \end{equation}

where $D$ is defined in (2.44). We must modify the spaces of both $u$ and $D^{1/2}u$ accordingly and so arrive at

\begin{equation}\left( E_1 E_{1/2} + E_1 D^{1/2} + D \right) \left( \frac{a}{b} \right) = \left( \frac{e}{f} \right), \end{equation}

where

\begin{equation}\begin{aligned}e(x) &= \sum_{n=0}^{\infty} e_n C^{(3/2)}_n(x), \\f(x) &= \sum_{n=0}^{\infty} f_n C^{(2)}_n(x). \end{aligned}\end{equation}

Again, by interleaving the coefficients, we can make the above operator banded.

4.2.1. Boundary conditions. In this case, the kernel of the operator in (4.10) is smooth, and we must enforce a boundary condition to ensure that the linear system (4.12) is invertible. The topic of boundary conditions in FDEs is complicated, and it is beyond the scope of this paper to give a full treatment here. Here we simply show how certain boundary conditions/side constraints can be applied to linear systems such as (4.12) and leave it to the reader to determine how many and what form of constraints are applicable to their FDE.

For example, consider the functional constraint $B^x \colon u(x) = c$. Given scalar $x \in [-1, 1]$ we can construct this functional acting on a basis in $C^{(\lambda)}_{\gamma}$ as a row vector by defining $B^x_{\lambda, \gamma} : C^{(\lambda)}_{\gamma} \to \mathbb{C}$ as $B^x_{\lambda, \gamma} := C^{(\lambda)}_{\gamma}(x)$. In particular, $x = -1$ corresponds to a boundary/initial condition on the left and $x = +1$ to a boundary condition on the right. Some useful cases are
NICHOLAS HALE AND SHEEHAN OLVER

Fig. 4.2. (a) Approximate solution to (4.16). (b) MATLAB spy plot of (4.15) showing the almost-banded structure. (c) 2-norm difference between the coefficients of the approximated solution when truncating at sizes $N$ and $\lceil 1.1N \rceil$ showing geometric convergence.

\[
B_P^{-1} = [1, -1, 1, -1, \ldots], \quad B_{U_{1/2}}^{-1} = [0, 0, \ldots],
\]
\[
B_{P}^{+1} = [1, 1, \ldots], \quad \text{and} \quad B_{U_{1/2}}^{+1} = \sqrt{2}[1, 2, 3, 4, \ldots].
\]

Combining such operators to act on our direct sum expansion of the solution $u(x)$, we have, for example, $B^{-1} : P \oplus U_{1/2} \to \mathbb{C}$ given by $B^{-1} = [B_P^{-1}, B_{U_{1/2}}^{-1}]$, and our system (4.12) augmented with the boundary condition $u(-1) = c$ becomes

\[
\begin{pmatrix}
B^{-1} \\
E_1 E_{1/2} + E_1 D^{1/2} + D
\end{pmatrix}
\begin{pmatrix}
\frac{a}{b} \\
\frac{c}{f}
\end{pmatrix}
= \begin{pmatrix}
c \\
\ell
\end{pmatrix}.
\]

Upon the usual reordering of the coefficients, this becomes an almost-banded infinite matrix—that is, banded apart from a finite number of dense rows—and when truncated to a $(2N + 1) \times (2N + 1)$ finite matrix is solvable in $O(N)$ operations using either a Schur complement factorization about the $(1, 1)$ entry, the Woodbury matrix identity, or the adaptive QR method described in [27]. See section 6.2 for more details.

Example 6. Consider the case of (4.10) where the right-hand side is zero and $u(-1) = 1$,

\[
u(x) + RL_{D}^{1/2}u(x) + u'(x) = 0, \quad x \in [-1, 1],
\]

which amounts to taking $c = 1$ and $\ell = f = 0$ in (4.15). The computed solution is depicted in the left panel of Figure 4.2. The middle panel verifies the almost-banded nature of the operator (4.15), and the right panel demonstrates geometric convergence.

4.3. Nonconstant coefficients. Nonconstant coefficients can be dealt with in a similar way as described for FIEs in section 3.2. We omit the details.

4.4. Higher order. Similarly to the case of integral equations, the approach outlined above can be extended to higher-order derivatives. Consider the general $m$th half-integer order FDE

\[
\mathcal{L}u(x) = \alpha^{[0]}(x)u(x) + \sum_{k=1}^{2m} \alpha^{[k]}(x) - P_{x}^{k/2}[\beta^{[k]}u](x),
\]
where the nonconstant coefficients $\alpha^{[k]}(x)$ and $\beta^{[k]}(x)$ are analytic in some neighborhood of $[-1, 1]$. If we continue to take as our anzatz solution the function

$$u(x) = \begin{bmatrix} P(x), U_{1/2}(x) \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix},$$

then we have $L : P \oplus U_{1/2} \to C(m+1/2) \oplus C_{-m+1/2}$ given by

$$L : P \oplus U_{1/2} \to \begin{bmatrix} C^{m+1/2} \oplus C_{-m+1} \end{bmatrix},$$

where we have defined $\beta^{[0]}(x) = 1$ for the sake of brevity.

**Example 7.** Consider the classical Bagley–Torvik equation [4, 5]

$$u''(x) + RL^{-1}D^{1/2}u(x) + u(x) = 0, \quad x \in [-1, 1],$$

but here treated as a boundary value problem with

$$u(-1) = 1 \quad \text{and} \quad u(1) = 0.$$ 

Following the approach outlined above, we arrive at the infinite-dimensional linear system

$$\begin{bmatrix} B^{-1} \\ D^2 + E_2E_{1/2}E_1(D^{1/2} + E_{1/2}) \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix},$$

which can be solved in the same manner as before. The solution is depicted in Figure 4.3.

**Remark.** If we instead consider the FDE $u''(x) + RL^{-1}D^{3/2}u(x) + u(x) = 0$, then we simply change the final block-row of the system (4.22) to $D^2 + E_2E_{3/2}E_1(D^{3/2} + E_{1/2})$. Similarly, we could incorporate a Neumann or fractional Neumann boundary condition at, say, the right boundary by changing the $B^{+1}$ row to the appropriate functional row.
5. FDEs: Caputo definition. FDEs with the Caputo definition of the fractional derivative can be readily solved by combining our approach for FIEs described in section 3 with an integral reformulation of the problem. In particular, setting \( v(x) = u^{[m]}(x) \) and therefore \( u(x) = Q^{[m]} v(x) + p(x) \), \( p(x) \in P^{[m]-1} \), it follows from the definition of the Caputo derivative that an \( m \)th-order FDE in \( u(x) \) becomes an \( m \)th-order FIE in \( v(x) \) with \( [m] \) additional boundary constraints to determine the coefficients of the polynomial \( p(x) = c_0 + c_1 P_1(x) + \ldots + c_{[m]-1} P^{[m]-1}(x) \). We proceed by example.

Example 8. Consider the Caputo fractional relaxation equation

\[
(5.1) \quad u(x) + -\frac{C}{1} D_x^{1/2} u(x) = 0, \quad u(-1) = 1,
\]

which has the solution

\[
(5.2) \quad u(x) = e^{1+x} \text{erfc}(\sqrt{1+x}).
\]

Letting \( v = u' \) we have \( u = Q v + c_0 \), and (5.1) becomes

\[
(5.3) \quad Q v(x) + -1 Q_x^{1/2} v(x) + c_0 = 0
\]

\[
(5.4) \quad Q v(-1) + c_0 = 1.
\]

In operator form, we may write this as the infinite-dimensional system

\[
(5.5) \quad \begin{pmatrix}
1 & B^{-1} Q \\
\hat{e}_1 & Q + Q^{1/2}
\end{pmatrix} \begin{pmatrix}
c_0 \\
\hat{a} / \hat{b}
\end{pmatrix} = \begin{pmatrix}
1 \\
0
\end{pmatrix},
\]

where

\[
(5.6) \quad v(x) = \sum_{n=0}^{\infty} \hat{a}_n P_n(x) + \sqrt{1+x} \hat{b}_n U_n(x) = [P(x), U_{1/2}(x)] \begin{pmatrix}
\hat{a} \\
\hat{b}
\end{pmatrix}
\]

and \( B^{-1} = [B^{-1}_P, B^{-1}_{U_{1/2}}] \). After truncating and solving this system for the approximate coefficients of \( v(x) \), we can recover those of \( u(x) \) via

\[
(5.7) \quad \begin{pmatrix}
\hat{a} \\
\hat{b}
\end{pmatrix} = Q \begin{pmatrix}
\hat{a} \\
\hat{b}
\end{pmatrix} + \begin{pmatrix}
c_0 \\
0
\end{pmatrix}
\]

so that, as usual, \( u(x) = [P(x), U_{1/2}(x)] [\hat{a} \ T, \hat{b} \ T] \ T \). Figure 5.1 shows the results. As in the case of RL FDEs we see that the resulting discretised system is almost banded and that the approximation converges geometrically in the number of degrees of freedom.

Example 9. Consider the Bagley–Torvik equation from Example 7 but now using the Caputo definition of the half-derivative:

\[
(5.8) \quad u''(x) + -\frac{C}{1} D_x^{1/2} u(x) + u(x) = 0, \quad x \in [-1, 1],
\]

\[
(5.9) \quad u(-1) = 1, \quad u(1) = 0.
\]

This time letting \( v = u'' \) we have \( u = Q^2 v + c_0 P_0(x) + c_1 P_1(x) \) and

\[
(5.10) \quad v(x) + -1 Q_x^{3/2} v(x) + c_1 Q_x^{1/2} P'_1(x) + Q^2 v(x) + c_0 P_0(x) + c_1 P_1(x) = 0
\]

\[
(5.11) \quad Q^2 v(-1) + c_0 + c_1 P_1(-1) = 1
\]

\[
(5.12) \quad Q^2 v(1) + c_0 + c_1 P_1(1) = 0.
\]
A FAST ALGORITHM FOR RATIONAL-ORDER FIEs & FDEs

A2479

Fig. 5.1. (a) Solution to the FDE (5.1). (b) MATLAB `spy` plot of (5.5) showing the almost-banded structure. (c) Solid line: Infinity norm error of solution approximated on a 100-point equally spaced grid. Dashed line: 2-norm difference between the coefficients of the approximated solution when truncating at sizes \( N \) and \([1.1N]\). As in the case of FIEs and RL-type FDEs of the previous two sections, geometric convergence is observed.

Fig. 5.2. (a) Approximate solution to the Bagley–Torvik equation (5.8). (b) MATLAB `spy` plot of (5.13) showing the almost-banded structure. (c) 2-norm difference between the coefficients of the approximated solution when truncating at sizes \( N \) and \([1.1N]\). Compare with the solution on the RL Bagley–Torvik equation in Figure 4.3.

We may write this as

\[
\begin{pmatrix}
1 & -1 \\
0 & 1
\end{pmatrix}
\begin{pmatrix}
\xi_1 \\
\xi_0
\end{pmatrix}
+ \begin{pmatrix}
B^{-1}Q^2 \\
B^+Q^2
\end{pmatrix}
\begin{pmatrix}
\xi_1 \\
\xi_0
\end{pmatrix}
= \begin{pmatrix}
1 \\
0
\end{pmatrix},
\]

where \( \xi(x) = \sum_{n=0}^\infty a_n P_n(x) + \sqrt{1-x^2} b_n U_n(x) \) and we have used the fact that \( P_1(\pm 1) = \mp 1 \). Once we have solved this system for the approximate coefficients of \( \xi \), we can recover those of \( u \) via

\[
\begin{pmatrix}
a \\
b
\end{pmatrix} = \begin{pmatrix}
a \\
b
\end{pmatrix} + \begin{pmatrix}
c_0 \\
c_1 \\
0
\end{pmatrix},
\]

where \( \bar{0} \) is a vector of zeros of length \( 2N - 2 \). Figure 5.2 shows the results. Compare with the RL version in Example 7.

6. Computational issues. In this section we outline some practical considerations required to perform computations.
6.1. Representing the right-hand side. An essential part of this approach is representing the right-hand side in the direct sum basis $P \oplus U_{1/2}$ and its higher-order cousins involving higher-order ultraspherical polynomials. A substantial issue is that given a general right-hand side $g(x)$, the decomposition as, for example,

$$g(x) = \sum_{n=0}^{\infty} c_n P_n(x) + \sqrt{1+x} \sum_{n=0}^{\infty} f_n U_n(x),$$

is not unique: $P(x)$ and $U_{1/2}(x)$ form a frame [8]. In the context of this numerical approach, uniqueness is not critical, as any expansion of this form is suitable provided we can approximate $g(x)$ well by taking finite number of terms.

We will assume we are given $e(x)$ and $f(x)$ that can be evaluated pointwise so that

$$g(x) = e(x) + \sqrt{1+x}f(x).$$

In this case, we can calculate the number of Chebyshev coefficients of $e$ and/or $f$ to within a required tolerance using an adaptive algorithm [3, 13]. The algorithm is based on the discrete cosine transform and hence takes $O(d \log d)$ operations to compute $d$ coefficients. Calculating coefficients in a basis $C^{(\lambda)}$, $\lambda \in \mathbb{N}^+$ proceeds in $O(\lambda d)$ operations by applying the conversion operators (2.11). Calculating Legendre coefficients from $d$ Chebyshev coefficients can be accomplished in $O(d \log^2 d)$ operations using recently developed fast transforms [16, 38], and from these coefficients in a basis $C^{(\lambda+1/2)}$, $\lambda \in \mathbb{N}^+$ can again be calculated in $O(\lambda d)$ operations by the conversion operators (2.11). The coefficients in nonconstant coefficient problems can be computed in an analogous manner.

Remark. In typical applications $d$ (the number of coefficients required to represent the right-hand side or nonconstant terms) is much smaller than $N$ (the discretization size of the system), and the claim that the proposed method is linear in the degrees of freedom is justified. The exception is when the linear problem arises from the linearization of a nonlinear problem. In this case the number of polynomial terms required to approximate the nonconstant coefficients will be the same as the for the solution (i.e., $d \approx N$). The linear systems resulting from discretization are then dense and require $O(N^3)$ operations to solve via Gaussian elimination. A spectrally accurate algorithm with linear complexity is still an open problem even in the case of ODEs.

6.2. Solving the linear systems. We have described an approach to reduce FDEs and FIEs to banded or almost-banded infinite-dimensional linear systems. A natural approach to approximating the solutions to the resulting equations is the finite section method: truncate the infinite-dimensional systems to $2N \times 2N$ finite-dimensional linear systems. This is an effective and easy to implement approach that achieves $O(N)$ complexity using standard LAPACK routines in the banded case or using the Woodbury formula in the almost-banded case.

Alternatively, one can solve using the adaptive QR method [27], which can be thought of as performing linear algebra directly on the infinite-dimensional linear system [28]. In this case, the number of coefficients needed to represent the solution

---

10 The case where we may only sample $g(x)$ is beyond the scope of this paper, though solving a least squares system with more points than coefficients can perform well in practice. Another situation that arises in practical settings is where $g(x)$ is specified by a formula such as $\exp(x) \sqrt{1+x} \cos x + \exp((1 + x)/2) \text{erfc}(\sqrt{1+x})$. The approach taken by ApproxFun is to overload each operation to automatically determine an appropriate decomposition.
within a specified tolerance of the error in residual is determined adaptively while preserving the linear complexity. A benefit of this approach, in addition to the adaptivity, is that it is not prone to the discretization introducing ill-posed equations. Left and right half-integral and half-derivative operators are implemented in the ApproxFun.jl package [26] for Julia, which uses the adaptive QR method.

6.3. Evaluating the result. The outputs of the algorithm we have described in the proceeding sections are coefficients of Legendre and weighted Chebyshev expansion (3.2) of the solution. Typically one is more interested in function values of the solution, but precisely what values are required depends entirely on the application. If only a few functions values are required, then the simplest approach is to use Clenshaw’s algorithm. This is the approach we have taken in the results above. If the solution is required at many points, then the fast transforms mentioned in section 6.1 can again be utilized to do this efficiently.

7. Rational-order equations. Here we consider the extension to problems involving rational-order integrals and derivatives. The general principle is the same as that which we have seen previously for half-integer orders, but an immediate consequence of moving to the rational-order case is that ultraspherical discretizations are no longer sufficient. A rational-order derivative of an ultraspherical polynomial does not typically have a short-term expansion in terms of other ultraspherical polynomials, so instead we must consider weighted Jacobi polynomials,

\[ P_\gamma^{(\alpha,\beta)}(x) := (1 + x)^\gamma [P_0^{\alpha,\beta}(x), P_1^{\alpha,\beta}(x), \ldots], \]

and their associated space of coefficients, \( P_\gamma^{(\alpha,\beta)} \). In the case of half-integer order FIEs and FDEs we required a direct sum space formed of two ultraspherical bases (i.e., Chebyshev and Legendre). Here, for a rational-order integral or derivative of order \( p/q \), we require a direct sum space formed of \( q \) such weighted Jacobi polynomials.

**Definition 7.1.** We denote by \( P_{[q]} \) the direct sum space formed of weighted Jacobi bases of the form \( P_{k/q}^{(1-k/q,k/q)} \) for \( k = 0, \ldots, q-1 \), i.e.,

\[ P_{[q]} := \bigoplus_{k=0}^{q-1} P_{k/q}^{(1-k/q,k/q)} = P_{(1,0)}^{(1)} \oplus P_{1/q}^{(1-1/q,1/q)} \oplus \ldots \oplus P_{1-2/q}^{(2/q,1-2/q)}(x) \oplus P_{1-1/q}^{(1/q,1-1/q)}(x), \]

and by \( P_{[q]}(x) \) the quasimatrix\(^{11}\)

\[ P_{[q]}(x) := \left[ P_{0}^{(1)}(x), P_{1/q}^{(1-1/q,1/q)}(x), \ldots, P_{1-2/q}^{(2/q,1-2/q)}(x), P_{1-1/q}^{(1/q,1-1/q)}(x) \right]. \]

If \( u^{[k]} \in P_{k/q}^{(1-k/q,k/q)} \) for \( k = 0, \ldots, q-1 \), then we say \( u \in P_{[q]} \) and may write

\[ u(x) = P_{[q]}(x)u = \sum_{k=0}^{q-1} (1 + x)^{k/q} \sum_{n=0}^{\infty} u_n^{[k]} P_n^{(1-k/q,k/q)}(x), \quad \text{where} \quad u = \begin{pmatrix} u_0^{[0]} \\ u_1^{[1]} \\ \vdots \\ u_{q-1}^{[q-1]} \end{pmatrix}, \]

\(^{11}\)Observe that each “column” in (7.3) is itself a quasimatrix!
We begin with rational-order integrals of order \( p/q \), where \( p, q \in \mathbb{N}^+ \). For brevity we focus only on constant coefficient problems, but the ideas of section 3.2 are readily applicable.

**7.1. Rational-order integral equations.** The foundation of our approach is the following formula, similar to that of Theorem 2.1, but here showing how the fractional integral of weighted Jacobi polynomials may be computed in closed form.

**Theorem 7.2 ([2, Theorem 6.72(b)]).** For any \( 0 \leq \mu < 1 \), \( \alpha, \beta \geq 0 \), \(-1 < x < 1\), and \( n \geq 0 \),

\[
(7.5) \quad -1 Q_x^\mu [(1 + x)^\beta P_n^{(\alpha, \beta)}(x)] = \frac{\Gamma(\beta + n + 1)}{\Gamma(\beta + \mu + n + 1)} (1 + x)^{\beta + \mu} P_n^{(\alpha - \mu, \beta + \mu)}(x).
\]

We define the infinite-dimensional matrix \( Q^\mu_n : \mathbf{P}_\beta^{(\alpha, \beta)} \to \mathbf{P}_\beta^{(\alpha - \mu, \beta + \mu)} \) induced by this relationship, so that if \( u \in \mathbf{P}_\beta^{(\alpha, \beta)} \), then \(-1 Q_x^\mu \mathbf{P}_\beta^{(\alpha, \beta)}(x)u = \mathbf{P}_\beta^{(\alpha - \mu, \beta + \mu)}(x)Q^\mu_n u \).

We also consider two conversion operators, \( S_{\alpha, \beta} : \mathbf{P}_\gamma^{(\alpha, \beta)} \to \mathbf{P}_\gamma^{(\alpha + 1, \beta)} \) and \( R_{\alpha, \beta} : \mathbf{P}_\gamma^{(\alpha, \beta + 1)} \to \mathbf{P}_\gamma^{(\alpha, \beta)} \) (akin to (2.11) and (2.13)), induced by [12, Eqn. 18.9.5]

\[
(7.6) \quad (2n + \alpha + \beta + 1)P_n^{(\alpha, \beta)}(x) = (n + \alpha + \beta + 1)P_n^{(\alpha + 1, \beta)}(x) - (n + \beta)P_n^{(\alpha + 1, \beta - 1)}(x),
\]

and [12, Eqn. 18.9.6]

\[
(7.7) \quad (n + \frac{1}{2}\alpha + \frac{1}{2}\beta + 1)(1 + x)P_n^{(\alpha, \beta + 1)}(x) = (n + 1)P_n^{(\alpha, \beta)}(x) + (n + \beta)P_n^{(\alpha, \beta - 1)}(x),
\]

respectively, so that if \( u(x) = u \in \mathbf{P}_\beta^{(\alpha, \beta)} \), then \( \mathbf{P}_\gamma^{(\alpha, \beta)}(x)u = \mathbf{P}_\gamma^{(\alpha + 1, \beta)}(x)S_{\alpha, \beta}u = \mathbf{P}_\gamma^{(\alpha, \beta - 1)}(x)R_{\alpha, \beta}u \). Combining \( Q^\mu_n, S_{\alpha, \beta} \), and \( R_{\alpha, \beta} \), we construct a \((1/q)\)th-order integral operator on \( \mathbf{P}_{[q]} \) as follows.

**Theorem 7.3.** Consider any \( q \in \mathbb{N}^+ \). If \( u \in \mathbf{P}_{[q]} \) so that \( u(x) = \mathbf{P}_{[q]}(x)u \), then the operator

\[
Q_{[q]}^{1/q} := \begin{pmatrix}
Q_{0/0}^{1/q} & \cdots & 0 \\
0 & Q_{1/0}^{1/q} & \cdots \\
\vdots & \vdots & \ddots & \vdots \\
0 & \cdots & 0 & Q_{1/q}^{1/q}
\end{pmatrix}
\]

satisfies

\[
-1 Q_{[q]}^{1/q} \mathbf{P}_{[q]}(x)u = \mathbf{P}_{[q]}(x)Q_{[q]}^{1/q} u
\]

**Proof.** We have from Theorem 7.2 that for \( k = 0, \ldots, q - 2 \),

\[
-1 Q_x^{1/q} \mathbf{P}_{k/q}^{(1-k/q, k/q)}(x)u[k] = \mathbf{P}_{(k+1)/q}^{(1-(k+1)/q, (k+1)/q)}(x)Q_{k/q}^{1/q}u[k]
\]

(7.10)

and for the final block, from the definitions of \( R_{0,0} \) and \( S_{0,0} \), that

\[
-1 Q_x^{1/q} \mathbf{P}_{1/q}^{(1/q, 1-1/q)}(x)u[q-1] = \mathbf{P}_{1/q}^{(0,1)}(x)Q_{1/q}^{1/q}u[q-1]
\]

(7.11)

\[
= \mathbf{P}_{0}^{(1,0)}(x)S_{0,0}R_{0,0}Q_{1/q}^{1/q}u[q-1].
\]

\( \square \)
Corollary 7.4. For any \( p, q \in \mathbb{N}^+ \) the operator
\[
Q_{[q]}^{p/q} := \left[ Q_{[q]}^{1/q} \right]^p
\]
is block-banded and satisfies
\[
-1 Q_{[q]}^{p/q} P_{[q]}(x) u = P_{[q]}(x) Q_{[q]}^{p/q} u.
\]

Proof. Equation (7.13) follows from \( p \) applications of \( Q_{[q]}^{1/q} \) on \( P_{[q]} \). That \( Q_{[q]}^{p/q} \) is block-banded follows from the fact that each of the blocks is formed by a product of banded matrices.

Remark. It is possible to construct an equivalent representation of the operator \( Q_{[q]}^{p/q} \) directly (rather than by repeated applications/multiplication of \( Q_{[q]}^{1/q} \)) by using a block matrix similar to that of (7.8) but containing entries of the form \( Q_{[q]}^{p/q} k/q \) and other suitable \( R- \) and \( S- \) type conversion matrices. However, while this may have some performance benefits, for clarity of exposition and convenience implementation we give preference to the construction as given in Corollary 7.4.

To solve an integral equation with terms of the form \( -1 Q_{[q]}^{p/q} x u(x) \), one then makes an ansatz that the solution \( u(x) \) may therefore be written as in (7.4), i.e.,
\[
[ u_0^0, u_0^1, u_0^2, \ldots, u_0^{q-1}, u_1^0, u_1^1, u_1^2, \ldots, u_1^{q-1}, u_2^0, \ldots ]
\]
the operator becomes banded with bandwidth \( \mathcal{O}(q) \). If each of the infinite sums in (7.4) is truncated at \( N \) terms, then the resulting linear system can be solved in \( \mathcal{O}(qN) \) operations.

Example 10. We demonstrate our method on the generalized second-kind Abel integral equation:
\[
u(x) + -1 Q_{[q]}^{p/q} u(x) = 1.
\]
Unfortunately, except for the special case of \( p/q = 1/2 \) considered in Example 1, there is no closed-form solution for (7.15) in general. However, if \( 0 < p/q < 1 \), there is a convergent series solution \([30, \text{section 2.1–7}]\)
\[
u(x) = 1 + \sum_{\ell=1}^{\infty} (-1)^{\ell} \frac{(1 + x)^{(lp/q)}}{\Gamma((lp/q + 1))}, \quad x \in [-1, 1].
\]
In particular, we take \( p = 2 \) and \( q = 3 \), so that our basis consists of weighted Jacobi polynomials of the form \( P_n^{(1,0)}(x) \), \((1 + x)^{1/3} P_n^{(2/3,1/3)}(x)\), and \((1 + x)^{2/3} P_n^{(1/3,2/3)}(x)\), and the infinite-dimensional linear system we must solve is
\[
(I + Q_{[3]}^{2/3}) \begin{pmatrix} u_0^0 \\ u_0^1 \\ u_0^2 \end{pmatrix} = \begin{pmatrix} \xi_0 \\ 0 \\ 0 \end{pmatrix},
\]
Fig. 7.1. (a) Approximate solution to (7.15) when \( p/q = 2/3 \). (b) MATLAB \texttt{spy} plot of the reordered linear system. As in the case of half-integer order FIEs, a banded operator is obtained. (c) (solid) Infinity norm error of solution (approximated on a 100-point equally spaced grid) as compared to the series solution (7.16). (dashed) Magnitude of the coefficients in the weighted Jacobi polynomial expansion of the solution, \( u[0], u[1], u[2] \). As before, geometric convergence is observed.

where \( e_0 = [1, 0, 0, \ldots]^\top \) and \( 0 = [0, 0, 0, \ldots]^\top \). Since \( I \) is diagonal and \( Q^{2/3}_2 \) has banded blocks, reordering the coefficients as described above results in a banded linear system, as shown in the middle panel of Figure 7.1, which can be solved as described in section 6.2. The resulting Jacobi polynomial coefficients of the approximate solution evaluated using Clenshaw’s scheme (or a more efficient approach, such as [38]) to obtain the solution in the left panel of Figure 7.1. The final panel of Figure 7.1 shows the error in the obtained solution as \( n \) is increased and also the magnitude of the coefficients in the solution for the case \( N = 20 \).

7.2. Rational-order FDEs.

7.2.1. Caputo-type derivatives. Similar to the way described in section 5 for half-integer order FDEs, Caputo FDEs of rational order can be reformulated as rational-order integral equations, which can be solved as described in the previous section. We omit the details.

7.2.2. RL-type derivatives. Here we may make use of the following.

**Theorem 7.5.** For any \( 0 \leq \mu < 1, \alpha, \beta \geq 0, \) and \( n \geq 0, \)

\[
(7.18) \quad RL_{-1}D_x^\mu[(1 + x)^\beta P_n^{(\alpha, \beta)}(x)] = \frac{\Gamma(\beta + n + 1)}{\Gamma(\beta - \mu + n + 1)}(1 + x)^{\beta - \mu}P_n^{(\alpha + \mu, \beta - \mu)}(x).
\]

**Proof.** This follows from the fundamental theorem of calculus applied to (7.5).

Similarly to before, we denote by \( D^\mu_\beta \) the infinite-dimensional operator induced by this relationship so that if \( u \in P^{(\alpha, \beta)}_\beta \), then

\[
(7.19) \quad RL_{-1}D_x^\mu P^{(\alpha, \beta)}_\beta(x)u = P^{(\alpha + \mu, \beta - \mu)}_\beta(x)D^\mu_\beta u.
\]

The difficulty here, as in the half-integer case of section 4, is that one cannot construct a banded block operator from such operators which maps \( P^{[q]}_\beta \) to itself. One must use conversion matrices similar to \( E_m \) and \( E_{m+1/2} \) as described in section 4. An additional problem is that when \( p \geq q, \) (7.19) naively applied to \( P^{[q]}_\beta \) will result in Jacobi polynomials with negative integer parameters, which are not classically
A fast Algorithm for rational-order FIEs & FDEs

These difficulties are not insurmountable, and one can extend the approach we consider in this paper to such problems; however, in the interest of brevity, we omit the details for a later publication.

8. Conclusion. By writing the solution in an appropriately constructed basis (in particular, a direct sum of Legendre, \( P_n(x) \), and weighted Chebyshev polynomials of the first kind, \( \sqrt{1+x} U_n(x) \)) we have successfully solved a broad class of half-integer order FIEs and FDEs with spectral accuracy in linear complexity. Some analysis of the half-integral equation described in section 3.1 can be found in section B.1. We also described how the approach can be extended to arbitrary rational-order FIEs and FDEs by using appropriate weighted Jacobi polynomial bases. For the rational-order case we demonstrated that the linear complexity and geometric convergence were maintained, but the implied constant in the former is proportional to the denominator, \( q \), in the rational degree of the problem.

The main objective of this paper was to introduce the algorithm and demonstrate its applicability, and several examples of both constant and nonconstant coefficient linear problems were presented. There are several opportunities for future extensions. Nonlinear problems (through linearization and Newton’s method), time-dependent problems (through method of lines), and partial fractional differential equations on rectangular domains (using ideas related to [37]) should be relatively straightforward, and we hope to solve problems of this type in a future publication. The questions of stability raised in Appendix B also require further investigation. An open problem is adapting the approach to problems involving the two-sided fractional derivative, used to define the fractional Laplacian. The known formula for the fractional (or even half-) integral of Jacobi polynomials does not allow for weighting at both the left and the right end of the domain simultaneously, which would be required to capture the singular behavior of two-sided derivatives.

Example 11. We close with one final example which demonstrates both the high accuracy and the linear complexity of the approach described in this paper when applied to a more challenging problem than those shown in the previous few sections. In particular, let’s consider fractional Airy equations of the form

\[(8.1)\quad \varepsilon i^{3/2} \int_{-1}^{x} D_x^{3/2} u(x) - xu(x) = 0, \quad x \in [-1, 1], \quad u(-1) = 0, \ u(1) = 1,\]

with \( \varepsilon > 0 \). Although complex-valued, this nonconstant coefficient FDE is of the form discussed in section 4, and we may solve accordingly using the algorithm described. The first and second panels of Figure 8.1 show the real and imaginary parts of the solution for \( \varepsilon = 10^{-4} \), and we see behavior qualitatively similar to that of the well-known classical Airy equation. Experimentally we find that an accuracy of \( 10^{-10} \) requires around 750 degrees of freedom (i.e., \( N \approx 375 \)), and forming and solving the almost-banded linear system representing the fractional differential operator and boundary conditions takes under a tenth of a second on a 2014 Desktop PC using the MATLAB implementation [15]. The third panel shows the computational times to form and solve the systems when the number degrees of freedom is artificially increased (as would be required for smaller values of \( \varepsilon \)). Using the Woodbury formula

\[2^{12}\]Li and Xu [18] have recently constructed a definition of negative parameter Jacobi polynomials in terms of orthogonality with respect to a Sobolev inner product, avoiding many of the pitfalls that arise from analytically continuing the classical Jacobi polynomials to negative integers. Using these polynomials may allow for reliable generalization of the results to negative parameters.
to solve the almost-banded linear system, we see that linear complexity is obtained. Finally, we note that this RL FDE can be readily solved using ApproxFun [26] with just a few commands:

```plaintext
using ApproxFun
S = Legendre() ⊕ JacobiWeight(0.5, 0, Ultraspherical(1))
D1S = LeftDerivative(S, 1.5)
x = Fun()
u = [Dirichlet(); 0.0001*im^1.5*D1S - x] \ [[0, 1], 0]
```

Appendix A. Miscellaneous proofs. The following results are required in the proofs of Corollaries 2.2 and 2.3 in sections 2.4 and 2.5, respectively.

**Lemma A.1.** For any \( n > 0 \) and \( \lambda > 0 \), the ultraspherical polynomials \( C_n^{(\lambda)}(x) \) satisfy the relationship

\[
2\lambda(1 + x)(C_n^{(\lambda+1)}(x) - C_n^{(\lambda+1)}(x)) = ((n + 1)C_n^{(\lambda)}(x) + (n + 2\lambda)C_n^{(\lambda)}(x)).
\]

**Proof.** Applying (2.13) to \( C_n^{(\lambda+1)}(x) \) and \( C_n^{(\lambda+1)}(x) \) gives, upon rearrangement,

\[
(1 + x)\left(C_n^{(\lambda+1)}(x) - C_n^{(\lambda+1)}(x)\right)
= \frac{1}{2n + 2\lambda + 1} \left(C_{n+1}^{(\lambda+1)}(x) - C_{n-1}^{(\lambda+1)}(x)\right) + \frac{1}{2n + 2\lambda} \left(C_n^{(\lambda+1)}(x) - C_{n-2}^{(\lambda+1)}(x)\right).
\]

Applying (2.10) to each of the bracketed terms on the right-hand side and canceling common terms gives the required result. \( \Box \)

**Corollary A.2.** The Legendre polynomials, \( P_n(x) \), and the Chebyshev polynomials, \( U_n(x) \), satisfy

\[
(n + 1)P_n(x) + P_{n-1}(x) = (1 + x)\left(C_{n-1}^{(3/2)}(x) - C_{n-2}^{(3/2)}(x)\right)
\]

and

\[
U_n(x) + (n + 1)U_{n-1}(x) = 2(1 + x)\left(C_{n-1}^{(2)}(x) - C_{n-2}^{(2)}(x)\right).
\]
Proof. Take \( n \to n - 1 \) with \( \lambda = \frac{1}{2} \) and \( \lambda = 1 \) in (A.1), respectively.

Appendix B. Convergence and stability results.

B.1. Convergence. Note that the decompositions of the right-hand side and solution of (3.1) in the forms (3.2) and (3.6) are not unique, so the well-posedness of (3.8) is not immediate. However, the Schur complement of the (1,1) block of (3.8) yields

\[
(Q_P - \sigma^2 I) \bar{a} = Q_U^{1/2} f - \sigma \bar{g}, \\
\sigma \bar{b} = f - Q_P^{1/2} \bar{a},
\]

where \( Q_P = Q_U^{1/2} Q_P^{1/2} \) is the indefinite integral operator acting on the Legendre basis (recall (2.27)). The fact that \( Q_P \) is banded along with the decaying properties of its entries leads to a proof of convergence whenever the original equation (3.1) is solvable in \( L^2[-1,1] \).

Definition B.1. Define the Banach space \( \ell_2^k \) with norm

\[
\| f \|_{\ell_2^k}^2 = \sum_{k=0}^{\infty} (k + 1)^{2\lambda} f_k^2.
\]

Lemma B.2. Let \( \Psi := \text{diag}(\sqrt{2}, \sqrt{3}, \sqrt{5}, \sqrt{7}, \ldots) \). If \( \sigma^2 \) is an \( \ell^2 \) eigenvalue of \( \tilde{Q}_P := \Psi Q_P \Psi^{-1} \), then \( \sigma \) (as well as \( -\sigma \)) is an \( L^2[-1,1] \) eigenvalue of \( \tilde{Q}_x^{1/2} \).

Proof. Note that \( \| P_n \| = \sqrt{\frac{2}{2n + 1}} \); hence, conjugating by \( \Psi \) recasts the operator to acting on expansions in the orthonormalized Legendre polynomials \( \tilde{P}_n(x) := P_n(x) \sqrt{\frac{2n + 1}{2}} \). The assumption on \( \sigma^2 \) being an \( \ell^2 \) eigenvalue enforces that any eigenvector \( \bar{a} \) of \( \tilde{Q}_P \) corresponds to the normalized Legendre coefficients of a function \( a(x) \) in \( L^2[-1,1] \) with norm \( \| \bar{a} \|_{\ell^2} \).

The entries of \( \tilde{Q}_P \) decay like \( 1/k \); see (2.27), which implies that \( \tilde{Q}_P : \ell_2^k \to \ell_2^{k+1} \). It follows immediately that \( \bar{a} \in \ell_2^k \) for all \( \lambda; \bar{a} \in \ell_2^2 \) implies that \( \bar{a} = \sigma^{-2} \tilde{Q}_P \bar{a} \in \ell_2^{\lambda+1} \).

In particular, \( \bar{a} \in \ell^1 \). We can bound

\[
\| \sqrt{1 + xU_k} \| = \sqrt{\int_{-1}^{1} (1 + x) \frac{\sin^2(k + 1) \cos^{-1} x}{\sin^2 x} dx} \\
= \sqrt{\int_{0}^{\pi} (1 + \cos \theta) \frac{\sin^2(k + 1) \theta}{\sin \theta} d\theta} \leq 2\pi(k + 1)
\]

since Lagrange's trigonometric identities ensure that

\[
\left| \frac{\sin(k + 1) \theta}{\sin \theta} \right| \leq k + 1.
\]

Thus, the \( O(1/\sqrt{k}) \) decay in \( \tilde{Q}_P^{1/2} \Psi^{-1} \) cancels the \( O(\sqrt{k}) \) growth from \( \| \sqrt{1 + xU_k} \| \), and we have

\[
\| (\sqrt{1 + xU_0(x)}, \sqrt{1 + xU_1(x)}, \ldots) \tilde{Q}_P^{1/2} \Psi^{-1} \bar{a} \| \leq C \| \bar{a} \|_{\ell^2} < \infty.
\]

That is, the entries of \( \bar{b} = \sigma^{-1} \tilde{Q}_P^{1/2} \Psi^{-1} \bar{a} \) correspond to the second-kind Chebyshev coefficients of a function \( b(x) \) such that \( \sqrt{1 + x} b(x) \) is in \( L^2[-1,1] \). We therefore have an \( L^2[-1,1] \) eigenvector \( a(x) + \sqrt{1 + x} b(x) \), satisfying
\[ Q_{x_x}^{1/2}(a(x) + \sqrt{1 + xb(x)}) = Q_{x_x}^{1/2}(P(x)\Psi^{-1}a + U_{1/2}(x)b) \]
\[ = U_{1/2}(x)Q_P^{1/2}\Psi^{-1}a + P(x)Q_{U}^{1/2}b \]
\[ = \sigma U_{1/2}(x)b + \sigma P(x)\Psi^{-1}a \]
\[ = \sigma(a(x) + \sqrt{1 + xb(x)}). \]

**Lemma B.3.** If \( \sigma I + Q \) is invertible in \( L^2[-1,1] \), then \( I + QP \) is invertible in \( L^2 \) for all \( \lambda \) and in \( L^1 \). If \( \epsilon, f \in L^1 \), and \( a = (\sigma I + QP)^{-1}(Q_{U}^{1/2}f - \sigma \epsilon) \), then \( u(x) = a(x) + \sqrt{1 + xb(x)} \) satisfies
\[
(\sigma I + Q^{1/2})u(x) = e(x) + \sqrt{1 + x}f(x)
\]
for \( e(x) = P(x)\epsilon, f(x) = U_{1/2}(x)\epsilon, a(x) = P(x)a, \) and \( b(x) = \sigma^{-1}(f(x) - U_{1/2}(x)Q_{U}^{1/2}a) \).

**Proof.** The decay in the entries of \( P \) and bandedness imply that \( \|P_NQ_P - Q_P\|_{L^2} \to 0 \): \( P \) is compact in \( L^2 \) (and, by a similar argument, in \( L^1 \)). Compactness guarantees that the operator only has discrete eigenvalues. However, the previous lemma ensures that if \( \sigma I + Q^{1/2} \) is invertible in \( L^2[-1,1] \), then \( \sigma \epsilon \) is not an \( L^2 \) eigenvalue of \( Q_P \), and hence \( \sigma^2 I + Q_P \) is invertible in \( L^2 \). But any \( L^2 \) eigenvector is an eigenvector in \( L^2 \) for all \( \lambda \geq 0 \) (and in \( L^1 \)), as \( Q_P \) induces additional decay and, trivially, any \( L^2 \) eigenvector is automatically an \( L^2 \) eigenvector. Thus, we know that \( \sigma \) is also not an \( L^2 \) (or \( L^1 \)) eigenvalue, and the operator is invertible.

Therefore, if \( \epsilon, f \in L^1 \), then \( a \in L^1 \); hence, by the logic of the previous lemma, \( a(x) + \sqrt{1 + xb(x)} \in L^2[-1,1] \). We have thus constructed the unique \( L^2[-1,1] \) solution of \( (\sigma I + Q^{1/2})u(x) = e(x) + \sqrt{1 + x}f(x) \).

We now consider the finite section approximation of (3.8); i.e., we define the projection operator \( P_N : L^2 \to \mathbb{R}^N \) and consider the \( 2N \times 2N \) finite section approximation

(B.6) \[
\begin{pmatrix}
\sigma I_N \\
P_NQ_P^{1/2}P_N^T \\
\sigma I_N \\
\end{pmatrix}
\begin{pmatrix}
a_N \\
b_N \\
\end{pmatrix}
= \begin{pmatrix}
P_N\epsilon \\
P_Nf \\
\end{pmatrix}.
\]

This leads to an approximation:
\[
a(x) \approx a_N(x) = P(x)a \]
\[
b(x) \approx b_N(x) = \sigma^{-1}U(x)(P_Nf - Q_P^{1/2}a) \]
\[
u(x) \approx u_N(x) = a_N(x) + \sqrt{1 + xb_N(x)}.
\]

**Theorem B.4.** If \( \sigma I + Q^{1/2} \) is invertible in \( L^2[-1,1] \) and \( \epsilon, f \) are in \( L^1 \), then the finite section approximation to \( (B.1) \) converges to the true solution of (3.1) in \( L^2[-1,1] \).

**Proof.** Note that, because \( Q_P^{1/2} \) is upper triangular and \( Q_U^{1/2} \) is lower triangular, we have
\[
P_NQ_U^{1/2}P_N^TP_NQ_P^{1/2}P_N^T = P_NQ_P^{1/2}Q_U^{1/2}P_N^T = P_NQ_P^TP_N^T.
\]
A FAST ALGORITHM FOR RATIONAL-ORDER FIEs & FDEs

Fig. B.1. Left: Pseudospectra (computing using EigTool [41]) of the operator $Q_{1/2}$ truncated to a $200 \times 200$ matrix. Right: The same for (3.8) with $\sigma = 0$.

It follows that $a_N$ is also a solution to the $n \times n$ finite section of (B.1):

$$P_N(Q - \sigma^2 I)P_N^T a_N = P_N(Q_{1/2} f - \sigma e).$$

If the condition of this theorem holds, then by the previous lemma, $\sigma^2$ is not an eigenvalue of $Q_P$. $Q_P$ is a compact operator on $\ell^1$; therefore, the finite section approximation $a_N$ converges to $g$ in an $\ell^1$ sense (this follows from a Neumann series argument; see e.g., [27, Theorem 4.5]). This implies convergence of $a_N(x)$ to $a(x)$ in $L^2[-1, 1]$ and convergence of $b_N(x)$ to $b(x)$ in $L^2[-1, 1]$; hence, $u_N(x)$ converges to $u(x)$ in $L^2[-1, 1]$.

**Corollary B.5.** If $e, f \in \ell^2$, then the finite section approximation converges in $\ell^2_\lambda$. If this condition holds for all $\lambda$, then $u_N$ converges in $L^2[-1, 1]$ at a spectrally fast rate. Similarly, if $e, f$ decay exponentially, then $u_N$ converges in $L^2[-1, 1]$ exponentially fast.

**Proof.** The first statement follows from the operator being a compact perturbation of the identity in all $\ell^2_\lambda$ spaces; hence, the same argument as Theorem B.4 applies. The second statement follows from relating convergence in $\ell^2_\lambda$ to fast convergence in $\ell^1$. The exponentially fast convergence follows similarly by adapting the results to the exponentially weighted norm $\sqrt{\sum_{k=0}^\infty |R^k f_k|^2}$.

**B.2. Stability.** Unfortunately, solvability of the resulting equation is not the only issue: We must also consider conditioning. Now, $v(x) = e^{x/\sigma^2}$ is the solution to $Q u(x) - \sigma^2 u(x) = e^{-1/\sigma^2}$; hence, for $\sigma \ll 1$, $v(x)$ is approximately in the kernel of $Q u(x) - \sigma^2 I$. Therefore, we should expect the solution of the above system and hence the system (3.8) to be ill-conditioned when $\sigma \ll 1$. Indeed, the pseudospectral plot of the two (truncated) linear systems in Figure B.1 confirms this.

Decreasing $\sigma$ in this way is equivalent to a change of variables from $[-1, 1]$ to a longer “time” domain (if we consider the independent variable as time). In particular, let $y = \frac{1}{\sigma^2} x + c$ and $u(x) = v(y)$; then, substituting to (2.1), we find

$$-1 Q_{1/2} x u(x) = \sigma - \alpha Q_{1/2} y v(y).$$

Investigating the singular values of the operator suggests that as $\sigma \to 0$, it is only a single singular value that decays to zero and that it might be possible to regularize the problem. However, this is beyond the scope of the current paper, and we avoid this limiting case for now.

**Acknowledgments.** We thank Daniel Hauer (U. Sydney) for discussions related to convergence in higher-order norms, Marcus Webb (K.U. Leuven) for discussions on FDEs, and Alex Townsend (Cornell) for some useful suggestions.
REFERENCES


