

A sparse spectral method for Volterra integral equations

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Introduction

Volterra integral equations

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Define the *Volterra integral operator*

$$(\mathcal{V}_K u)(x) := \int_0^{\ell(x)} K(x, y) u(y) dy,$$

where $K(x, y)$ is called the kernel, $u(y)$ is a given function of one variable.
The limits of integration are either

$$\ell(x) = x \quad \text{or} \quad \ell(x) = 1 - x.$$

We introduce a sparse spectral method to find numerical approximations to the solution of Volterra integral equations of the first and second kind, i.e. to find u satisfying

$$\mathcal{V}_K u = g \quad \text{or} \quad (I + \mathcal{V}_K)u = g.$$

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Function approximation with orth. polynomials

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We expand functions using a complete basis of orthogonal polynomials:

$$f(x) = \sum_{n=0}^{\infty} P_n(x) f_n = \mathbf{P}(x)^T \mathbf{f},$$

where

$$\mathbf{P}(x) := \begin{pmatrix} P_0(x) \\ P_1(x) \\ \vdots \end{pmatrix}, \quad \mathbf{f} := \begin{pmatrix} f_0 \\ f_1 \\ \vdots \end{pmatrix}.$$

This works analogously for bivariate orthogonal polynomials

$$f(x, y) = \sum_{n=0}^{\infty} \sum_{k=0}^{\infty} P_{n,k}(x, y) f_{n,k}.$$

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Jacobi operators

Introduction

We can compute $xf(x)$ if $f(x)$ is given in coefficient vector form:

$$\mathbf{P}(x)^T \mathbf{J}^T \mathbf{f} = xf(x).$$

This is efficiently possible because the Jacobi polynomials satisfy a three term recurrence relationship, making \mathbf{J} a tridiagonal operator.

Analogously on the triangle

Introduction

We use the Jacobi polynomials shifted to the $[0, 1]$ interval and denote them by $\tilde{P}^{(\alpha, \beta)}$, which allows us to write the bivariate Jacobi polynomials on the triangle as:

$$P_{k,n}^{(\alpha, \beta, \gamma)}(x, y) = (1 - x)^k \tilde{P}_{n-k}^{(2k + \beta + \gamma + 1, \alpha)}(x) \tilde{P}_k^{(\gamma, \beta)}\left(\frac{y}{1 - x}\right).$$

As in the 1-dimensional case we can define Jacobi operators (now block tridiagonal) J_x and J_y , one for each variable:

$$\mathbf{P}(x, y)^T J_x^T \mathbf{f}_\Delta = xf(x, y),$$

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Spectral method for Volterra integral equations

The Volterra operator on coefficient space

Spectral method for Volterra integral equations

We build the operator

$$\int_0^{1-x} f(y) dy = \mathbf{P}(x)^T W_Q Q_y E_y \mathbf{f}_{[0,1]}$$

from two parts:

1 Q_y is the integral operator

$$\mathbf{P}(x)^T W_Q Q_y \mathbf{f}_\Delta = \int_{y=0}^{1-x} f(x, y) dy,$$

2 E_y extends a one-dimensional function on $[0, 1]$ to the triangle:

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The operators

Spectral method for Volterra integral equations

Using properties of the Jacobi polynomials one can derive

$$Q_y = \begin{pmatrix} 1 & & & \\ & 1 & 0 & & \\ & & \ddots & \ddots & \ddots \end{pmatrix}, \quad E_y = \begin{pmatrix} \times & & & \\ & \times & & \\ & & \ddots & \\ & & & \ddots \end{pmatrix}$$

where $E_y(n)_j = \frac{(-1)^{j+n}(2j-1)}{n}$. This also means that

$$(Q_y E_y)(n) = D_y(n) = \frac{(-1)^{n+1}}{n}.$$

Dealing with kernels

Spectral method for Volterra integral equations

Assuming a monomial expansion for the kernel¹

$$K(x, y) = \sum_{n=0}^{\infty} \sum_{j=0}^n k_{nj} x^{n-j} y^j,$$

the primary part of the Volterra integration operator is

$$\begin{aligned} Q_y K(J_x^\top, J_y^\top) E_y &= Q_y \left(\sum_{n=0}^{\infty} \sum_{j=0}^n k_{nj} (J_x^\top)^{n-j} (J_y^\top)^j \right) E_y \\ &= \sum_{n=0}^{\infty} \sum_{j=0}^n k_{nj} (J^\top)^{n-j} Q_y E_y (J^\top)^j. \end{aligned}$$

where we made use of

$$\begin{aligned} Q_y J_x^\top f_\Delta &= J^\top Q_y f_\Delta, \\ J_y^\top E_y f_{[0,1]} &= E_y J^\top f_{[0,1]}. \end{aligned}$$

¹We actually use a modified variation of Clenshaw's algorithm on the triangle due to see S. Olver, A. Townsend and G. Vasil (2019).

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The method for integral equations

Spectral method for Volterra integral equations

Equations of first kind, $V_K u = g$, turn into

$$\tilde{\mathbf{P}}^{(1,0)}(x)^T Q_y K(J_x^T, J_y^T) E_y \mathbf{u} = \tilde{\mathbf{P}}^{(1,0)}(x)^T \mathbf{q},$$

where now \mathbf{q} is the coefficient vector of $q(x) = \frac{g(x)}{1-x}$.

Equations of second kind, $(I + V_K)u = g$, turn into

$$\tilde{\mathbf{P}}^{(1,0)}(x)^T \left(\mathbb{1} - (\mathbb{1} - J^T) Q_y K(J_x^T, J_y^T) E_y \right) \mathbf{u} = \tilde{\mathbf{P}}^{(1,0)}(x)^T \mathbf{g}.$$

Quick notes on convergence

Sketch for second kind

Quick notes on convergence

$$\begin{array}{ccc} L^2(0,1) & \xrightarrow{\mathcal{V}_K} & L^2(0,1) \\ \mathcal{E} \downarrow & & \uparrow \mathcal{E}^{-1} \\ \ell^2 & \xrightarrow{\mathbf{V}_K} & \ell^2 \end{array}$$

For Volterra integral equations of second kind the operator to be inverted is of the form $(\mathbb{1} + \mathbf{V}_K)$ and \mathbf{V}_K compact.

Sketch for first kind (I)

Quick notes on convergence

Since V_K compact from ℓ^2 to ℓ^2 we instead consider $V_K : \ell^2 \rightarrow \ell_1^2$ where ℓ_λ^2 denotes the Banach space with norm

$$\|\mathbf{u}\|_{\ell_\lambda^2} = \sqrt{\sum_{n=0}^{\infty} ((1+n)^\lambda |u_n|)^2} < \infty.$$

Then the operator can be brought into the form

$$V_K = D(T_f + C),$$

where T is a symmetric Toeplitz operator with real entries and with symbol f , C is compact and D is diagonal, bounded and invertible.

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Sketch for first kind (II)

Quick notes on convergence

The symbol of the Toeplitz operator part is uniquely determined by the coefficients of the kernel to be

$$f(z) = \sum_{n=0}^M \sum_{j=0}^n k_{nj} \cos^{2n} \left(\frac{\theta}{2} \right) \quad \text{where} \quad z = e^{i\theta}.$$

The resulting condition for convergence of the method is found to be:

$$\forall x \in [0, 1] : K(x, x) \neq 0.$$

Implementation in Julia under ApproxFun.jl framework

Three examples

Implementation in Julia under ApproxFun.jl framework

We seek numerical solutions u_1 , u_2 and u_3 to the following three Volterra integral equations of second kind.

$$\text{Let } C(x) = \frac{e^{-10\pi x}(1+20\pi)-2+\cos(10\pi x)+\sin(10\pi x)}{20\pi}.$$

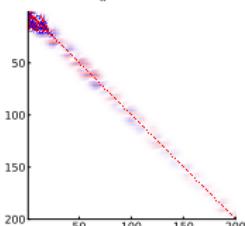
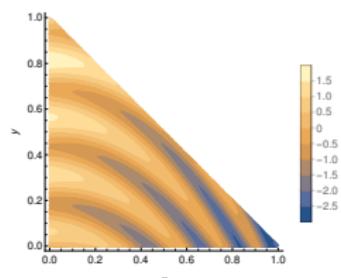
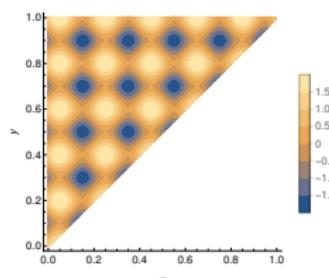
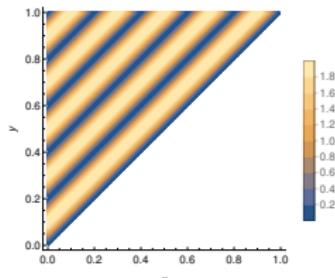
$$u_1(x) = C(x) + \int_0^x (1 - \cos(10\pi x - 10\pi y)) u_1(y) dy \quad (1)$$

$$u_2(x) = \frac{e^{\frac{x}{2}}}{\pi} + \int_0^x (\sin(10\pi x) + \cos(10\pi y)) u_2(y) dy \quad (2)$$

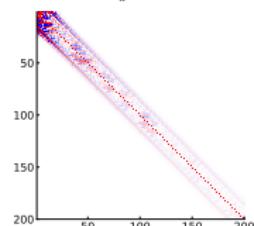
$$u_3(x) = e^{x^2-2x} + \int_0^{1-x} (-2x + y + \sin(25x^2 + 8\pi y)) u_3(y) dy \quad (3)$$

A look at the kernels and operators

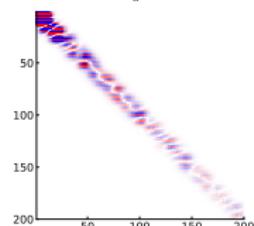
Implementation in Julia under ApproxFun.jl framework



(a) $K_1(x, y)$



(b) $K_2(x, y)$

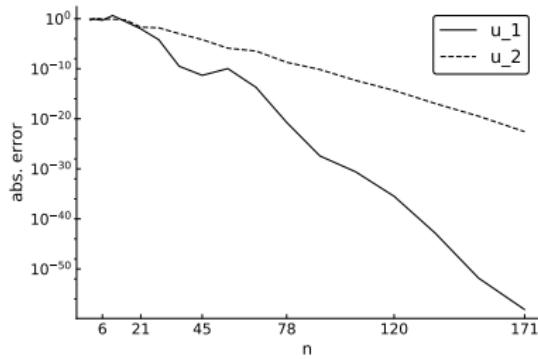


(c) $K_3(x, y)$

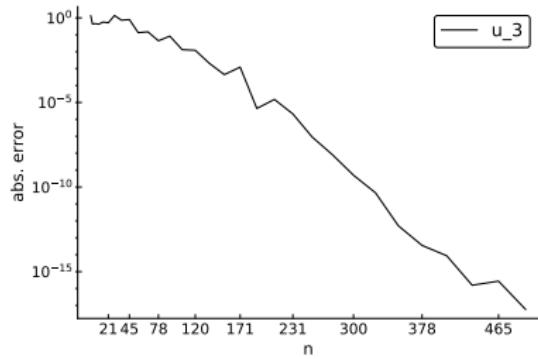
Figure: Kernel contour plots and operator spy plots.

Convergence of numerical experiments

Implementation in Julia under ApproxFun.jl framework



(a)



(b)

Figure: Absolute errors for equations (1–3). $u_1(x)$ is compared to the analytic solution, $u_2(x)$ and $u_3(x)$ are compared to a solution computed with $n = 5050$.

Discussion

- The Volterra integral operator is banded on an appropriate basis of orthogonal polynomials.
- This can be used in a highly efficient sparse spectral method for Volterra integrals and integral equations.
- The method is not restricted to convolution kernels.
- We have a working implementation of this method under ApproxFun.jl framework.
- As it is the method only works for linear Volterra integral equations but an extension to non-linear cases is conceivable - we are working on it.

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Clenshaw's algorithm

The algorithm makes use of the polynomial basis' recurrence relationships to reduce function evaluation to the solution of a triangular linear system:

$$Q_y K(J_x, J_y) E_y = (\mathbf{e}_0 \otimes \mathbb{1}) \mathcal{L}_V^{-T} (\mathbf{K}_{\Delta} \otimes Q_y E_y),$$

with

$$\mathcal{L}_V = \begin{pmatrix} (\mathbb{1}_1 \otimes \mathbb{1}) & & \\ (A_0^x \otimes \mathbb{1}) - (\mathbb{1}_1 \otimes J_{\diamond}) & (B_0^x \otimes \mathbb{1}) & \\ (A_0^y \otimes \mathbb{1}) - (\mathbb{1}_1 \otimes \diamond J) & (B_0^y \otimes \mathbb{1}) & \\ (C_0^x \otimes \mathbb{1}) & (A_1^x \otimes \mathbb{1}) - (\mathbb{1}_2 \otimes J_{\diamond}) & (B_1^x \otimes \mathbb{1}) \\ (C_0^y \otimes \mathbb{1}) & (A_1^y \otimes \mathbb{1}) - (\mathbb{1}_2 \otimes \diamond J) & (B_1^y \otimes \mathbb{1}) \\ & \ddots & \ddots & \ddots \end{pmatrix}.$$

After appropriate preconditioning this system can be solved via backward substitution, see S. Olver, A. Townsend and G. Vasil (2019).