# A fast method to implement non-local discrete transparent boundary conditions for the Schrödinger equation

Discrete transparent boundary conditions (DTBCs) for the time-dependent Schrödinger equation were introduced in the numerical simulations of whole space problems in order to reduce the computational domain to a finite region. They include a convolution w.r.t. time with a weakly decaying kernel that leads to very costly numerical evaluation for large-time simulations. As a remedy we construct approximate DTBCs with a kernel having the form of a finite sum-of-exponentials, which can be evaluated in an efficient recursion.

#### 1. Introduction

Discrete transparent boundary conditions for the discrete 1D–Schrödinger equation

$$-iR(\psi_{j,n+1}-\psi_{j,n}) = \psi_{j+1,n+1}-2\psi_{j,n+1}+\psi_{j-1,n+1}+\psi_{j+1,n}-2\psi_{j,n}+\psi_{j-1,n}-wV_{j,n+\frac{1}{2}}(\psi_{j,n+1}+\psi_{j,n}), \quad (1)$$

where  $R = 4\Delta x^2/\Delta t$ ,  $w = 2\Delta x^2$ ,  $V_{j,n+\frac{1}{2}} := V(x_j, t_{n+\frac{1}{2}})$ ,  $x_j = j\Delta x$ ,  $j \in \mathbb{Z}$ ; and  $V(x,t) = V_- = const$  for  $x \le 0$ ;  $V(x,t) = V_+ = const$  for  $x \ge X$ ,  $t \ge 0$ ,  $\psi(x,0) = \psi^I(x)$ , with supp  $\psi^I \subset [0,X]$ , were introduced in [1]. The DTBC at e.g. the left boundary point j = 0 reads, cf. Th. 3.8 in [2]:

$$\psi_{1,n} - s_0 \psi_{0,n} = \sum_{k=1}^{n-1} s_{n-k} \psi_{0,k} - \psi_{1,n-1}, \quad n \ge 1.$$
<sup>(2)</sup>

The convolution kernel  $\{s_n\}$  can be obtained by explicitly calculating the inverse Z-transform of the function  $\hat{s}(z) := \frac{z+1}{z}\hat{\ell}(z)$ , where  $\hat{\ell}(z) = 1 - i\zeta \pm \sqrt{-\zeta(\zeta + 2i)}$ ,  $\zeta = \frac{R}{2}\frac{z-1}{z+1} + i\Delta x^2 V_-$  (choose sign such that  $|\hat{\ell}(z)| > 1$ ).

The use of (2) for calculations permits us to avoid any boundary reflections and it renders the fully discrete scheme unconditionally stable, like the Crank-Nicolson scheme (1). However, the linearly in t increasing numerical effort to evaluate the DTBCs can sharply raise the total computational costs. A strategy to overcome this drawback will be the key issue of this paper.

### 2. Approximation of Convolution Coefficients by Sums of Exponentials

We evaluate numerically the several first convolution coefficients  $s_n$  appearing in the DTBC (2):  $s_n \approx s_n^{(N)} = \rho^n N^{-1} \sum_{k=0}^{N-1} \hat{s}(\rho e^{i\varphi_k}) e^{in\varphi_k}$ ,  $n = 0, 1, \ldots, N-1$ . Here  $\varphi_k = 2\pi k/N$ , and  $\rho > 1$  is a regularization parameter.

Our fast method to calculate the discrete convolution in (2) is based on the approximation of the coefficients  $s_n$  by the following ansatz (sum of exponentials):

$$s_n \approx \tilde{s}_n := \begin{cases} s_n, & n = 0, \dots, \nu - 1, \\ \sum_{l=1}^{L} b_l q_l^{-n}, & n = \nu, \nu + 1, \dots, \end{cases}$$
(3)

where  $L, \nu \in \mathbb{N}$  are fixed numbers. In order to find the required  $\{b_l, q_l\}$ , we fix L and  $\nu$  in (3) (e.g.  $\nu = 1$ ), and consider the Padé approximation  $\frac{P_{L-1}(x)}{Q_L(x)}$  for the formal power series:  $f(x) := s_{\nu} + s_{\nu+1}x + s_{\nu+2}x^2 + \dots, \quad |x| \leq 1$ .

Theorem 1. Let the polynomial  $Q_L(x)$  have L simple roots  $q_l$  with  $|q_l| > 1$ , l = 1, ..., L. Then

$$\tilde{s}_n = \sum_{l=1}^{L} b_l q_l^{-n}, \qquad n = \nu, \nu + 1, \dots, \quad where \quad b_l := -\frac{P_{L-1}(q_l)}{Q'_L(q_l)} q_l^{\nu-1} \neq 0, \qquad l = 1, \dots, L.$$

R e m a r k 1. According to the definition of the Padé algorithm, we have  $\tilde{s}_n = s_n$  for  $n = \nu, \nu+1, \ldots, 2L+\nu-1$ . For the remaining  $\tilde{s}_n$  with  $n > 2L + \nu - 1$ , the following estimate is proved:  $|\tilde{s}_n - s_n| = \mathcal{O}(n^{-\frac{3}{2}})$ .

R e m a r k 2. All our practical calculations confirm that the *condition* of Theorem 1 holds for any desired L, although we cannot prove this.

## 3. Fast Evaluation of the Discrete Convolution with an "Exponential" Kernel

Given the approximation (3) of the discrete convolution kernel appearing in the DTBC (2), the convolution

$$C^{(n)}(u) := \sum_{k=1}^{n-\nu} u_k \tilde{s}_{n-k}$$
(4)

of a discrete function  $u_k$ ,  $k = 1, 2, \ldots$ , can be calculated efficiently by *recurrence* formulas, cf. [3]:

Theorem 2. The function  $C^{(n)}(u)$  from (4) for  $n \ge \nu + 1$  is represented by

$$C^{(n)}(u) = \sum_{l=1}^{L} C_{l}^{(n)}(u), \quad \text{where} \quad C_{l}^{(n)}(u) = q_{l}^{-1} C_{l}^{(n-1)}(u) + b_{l} q_{l}^{-\nu} u_{n-\nu} \quad \text{for} \quad n \ge \nu + 1, \quad C_{l}^{(\nu)}(u) \equiv 0.$$

The recursion permits us to drastically reduce the computational effort of evaluating DTBCs for long-time computations  $(n \gg 1)$ :  $\mathcal{O}(L * n)$  instead of  $\mathcal{O}(n^2)$  arithmetic operations.

### 4. Numerical Example

As an example, we consider (1) on  $0 \le x \le 1$  with  $V_- = V_+ = 0$ , and non-zero initial data  $\psi^I$ . The reference solution  $\psi_{ref}$  with  $\Delta x = 1/160$ ,  $\Delta t = 2 \cdot 10^{-5}$  is obtained by using exact DTBCs (2) at the ends x = 0 and x = 1. We vary the parameter L = 20, 30, 40, 50 in (3), find the corresponding approximate DTBCs, and show the error of the approximate solution  $\psi_a$  measured in  $\frac{||\psi_a - \psi_{ref}||_{L_2}(t)}{||\psi^I||_{L_2}}$ . The result up to n = 15000 is shown in the figure.



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### 5. References

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