

Master Thesis

# Matrix-free Leja based exponential integrators in Python

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Abstract

## 1 Introduction

Consider the action of the matrix exponential function

$$e^A v, \quad A \in \mathbb{C}^{N \times N}, v \in \mathbb{C}^N.$$

It can be difficult or impossible to compute  $e^A$  in a first step and then the action  $e^A v$  in a separate step. This is especially true in applications where  $N > 10000$  is not uncommon. Furthermore the matrix exponential of a sparse matrix is in general no longer sparse. Therefore it is more feasible to compute the action of the matrix exponential in a single step. This can be done by approximating the matrix exponential with a matrix polynomial  $p_n$  of degree  $n$  in  $A$

$$e^A v \approx p_n(A)v.$$

This approach has many advantages. The cost of the computation of  $p_n(A)v$  mainly depends on the calculation of  $n$  matrix-vector multiplications with  $A$ . Furthermore the explicit knowledge of  $A$  itself is no longer required.  $A$  can be replaced by a linear function, which can be more convenient and saves memory.

## 2 The Leja method

This section serves as an introduction to the Leja method for the exponential function. All proofs can be found in a more general form in [4].

### 2.1 Leja interpolation

Let  $K \subset \mathbb{C}$  be a compact set in the complex plane and  $z_0 \in K$  be arbitrary. The sequence  $(z_k)_{k=0}^\infty$  recursively defined as

$$z_k = \arg \max_{z \in K} \prod_{j=0}^{k-1} |z - z_j|$$

is called a Leja sequence. Due to the maximum principles all elements in the sequence realize their maximum on the border  $\partial K$ . Typically  $z_0$  is also chosen on  $\partial K$ .

For analytical functions  $f: K \rightarrow \mathbb{C}$  the Newton interpolation polynomial  $p_m$  with nodes  $(z_k)_{k=0}^m$  has many beneficial properties.

**Convergence properties:** The sequence  $(p_m)_{m=0}^\infty$  converges maximally to  $f$ : Let  $(p_m^*)_{m=0}^\infty$  be the best uniform approximation polynomials for  $f$  in  $K$ . Then

$$\limsup_{m \rightarrow \infty} \|f - p_m\|_K^{1/m} = \limsup_{m \rightarrow \infty} \|f - p_m^*\|_K^{1/m} \stackrel{K=\mathbb{C}}{=} 0.$$

Furthermore if  $f$  is an entire function, then  $(p_m)_{m=0}^\infty$  converges superlinearly to  $f$ . For entire functions  $f$  the corresponding matrix polynomials achieves similar superlinear convergence

$$\limsup_{m \rightarrow \infty} \|f(A)v - p_m(A)v\|_2^{1/m} = 0,$$

for  $A \in \mathbb{C}^{n \times n}$ ,  $v \in \mathbb{C}^n$ .

**Recursive constructability:** The Newton interpolation polynomial  $p_m$  can be constructed iteratively since the corresponding Leja interpolation points  $(z_k)_{k=0}^m$  are defined recursively. Therefore if the approximation  $p_n \approx f$  is accurate enough for  $n < m$  the interpolation can be stopped early to reduce the cost of the interpolation. Note that this is not possible with Chebyshev nodes.

**Leja sequence can be stored:** For a given  $K$  the Leja interpolation nodes only need to be computed once and for all. These values can be stored a priori and loaded once they are needed for the interpolation. If  $f$  is fixed the same is true for the corresponding divided differences.

In summary the Leja points offer convergence properties similar to Chebyshev nodes for interpolation, while having computational advantages. All results hold true for the corresponding matrix interpolation polynomials.

From now on, we will fix

$$K = [-2, 2], \quad f = e^{\cdot}, \quad z_0 = 2$$

## 2.2 Exploiting the properties of the matrix exponential functions:

## 3 Linear advection diffusion equation

Plot the eigenvalues of the matrix

## 4 Numerical experiments

For the first experiments we will discretize multiple one-dimensional advection-diffusion-reaction equations with hybrid difference schemes.<sup>1</sup> We will always choose an equidistant grid with grid size  $h = \frac{1}{N}$ ,  $N \in \mathbb{N}$  and grid points  $x_i = ih$  for  $i = 0 \dots, N$  on the domain  $\Omega = [0, 1]$ . The resulting ordinary differential equations (ODEs) will be solved with four different integrators. Our goal is to investigate the respective computational costs of these methods while achieving a prescribed relative tolerance `tol`.

<sup>1</sup>Need a source, [https://en.wikipedia.org/wiki/Hybrid\\_difference\\_scheme](https://en.wikipedia.org/wiki/Hybrid_difference_scheme)

**Crank-Nicolson method:** We refer to the Crank-Nicolson method of order 2 as `cn2`. In our implementation of `cn2`, we used the SciPy[7] package `scipy.sparse.linalg.gmres` to solve linear equations. We set the relative tolerance to `tol/s`, where  $s$  is the total number of substeps taken for solving the ODE. This choice guarantees that the sum of errors made by `gmres` is always lower than our specified tolerance `tol`, since we have to solve exactly one linear equation per substep. No preconditioner was used for `gmres`. The Crank-Nicolson method is unconditionally stable and therefore does not have to satisfy the Courant-Friedrichs-Lewy (CFL) conditions imposed by the advective and diffusive part of the differential equations.

**Exponential Rosenbrock-Euler method:** We refer to the Exponential Rosenbrock-Euler method of order 2 as `exprb2`. The approximate the action of the matrix exponential with the Leja method. No hump reduction is used. The maximal interpolation degree is set to 100. Note that the total number of matrix-vector multiplication per time step can still exceed 100 since we have to compute a single matrix norm. This typically happens for  $s = 1$ .

**Explicit midpoint method:** We refer to the explicit midpoint method of order 2 as `rk2`.

**Classical Runge kutta:** We refer to the classical Runge-Kutta method of order 4 as `rk4`.

For our experiments we will often fix one of two different Péclet numbers

$$\text{Pe} = \frac{b}{a}, \quad \text{pe} = \frac{hb}{2a},$$

The Péclet numbers are dimensionless quantities representing the ratio of the advective velocity  $b$  to the diffusive velocity  $a$ . While  $\text{Pe}$  characterizes the original partial differential equation, the grid Péclet number  $\text{pe}$  is the dimensionless quantity for the resulting ODE after discretization. Note that by fixing  $\text{pe}$  for varying grid sizes, we have to change the original partial differential equation. Unless otherwise noted we accomplish that by replacing  $b$  with  $2b$  and  $a$  with  $ah$ .

## 4.1 Experiment 1: Linear advection diffusion equation

Consider the one-dimensional advection-diffusion equation

$$\begin{aligned} \partial_t u &= a \partial_{xx} u + b \partial_x u \quad a, b \geq 0 \\ u_0(t) &= e^{-80 \cdot (t-0.45)^2} \quad t \in [0, 0.1] \end{aligned}$$

with homogeneous Dirichlet boundary conditions on the domain  $\Omega = [0, 1]$ . For a fixed  $N \in \mathbb{N}$  we approximate the diffusive part with second-order central differences on an equidistant grid with grid size  $h = \frac{1}{N}$  and grid points  $x_i = ih$ ,  $i = 0 \dots, N$ .

$$\partial_{xx} u(x_i) = \frac{u(x_{i+1}) - 2u(x_i) + u(x_{i-1}))}{h^2} + \mathcal{O}(h^2)$$

In order to limit numerical instabilities we discretize the advective part with forward differences, similar to the upwind scheme.<sup>2</sup>

$$\partial_x u(x_i) = \frac{u(x_{i+1}) - u(x_i)}{h} + \mathcal{O}(h)$$

The resulting system of ordinary differential equation is given by

$$\partial_t u = Au.$$

Some eigenvalues of  $A$  can have an extremely large negative real part. Therefore, since no explicit Runge-Kutta method is A-stable, this imposes very stringent conditions on the time step size  $\tau$  for rk2 and rk4.<sup>3</sup> We will refer to the Courant-Friedrich-Lewy (CFL) conditions imposed by the advective and diffusive part of  $A$  respectively by  $C_{adv}$  and  $C_{dif}$ .

$$C_{adv} = \frac{b\tau}{h} \leq 1, \quad C_{dif} = \frac{a\tau}{h^2} \leq \frac{1}{2}$$

In our case the problem is fully linear and therefore `exprb2` simplifies to the computation of the action of the matrix exponential function with the Leja method. We write `expleja` for the single precision Leja method approximation. Note that reference solution was computed with double precision and therefore uses different nodes.

In order to keep the solution from vanishing, we fix  $b = 1$  and only consider coefficients  $a \in [0, 1]$ . The advection-diffusion ratio scaled by the grid size  $h$  is represented by the grid Péclet number

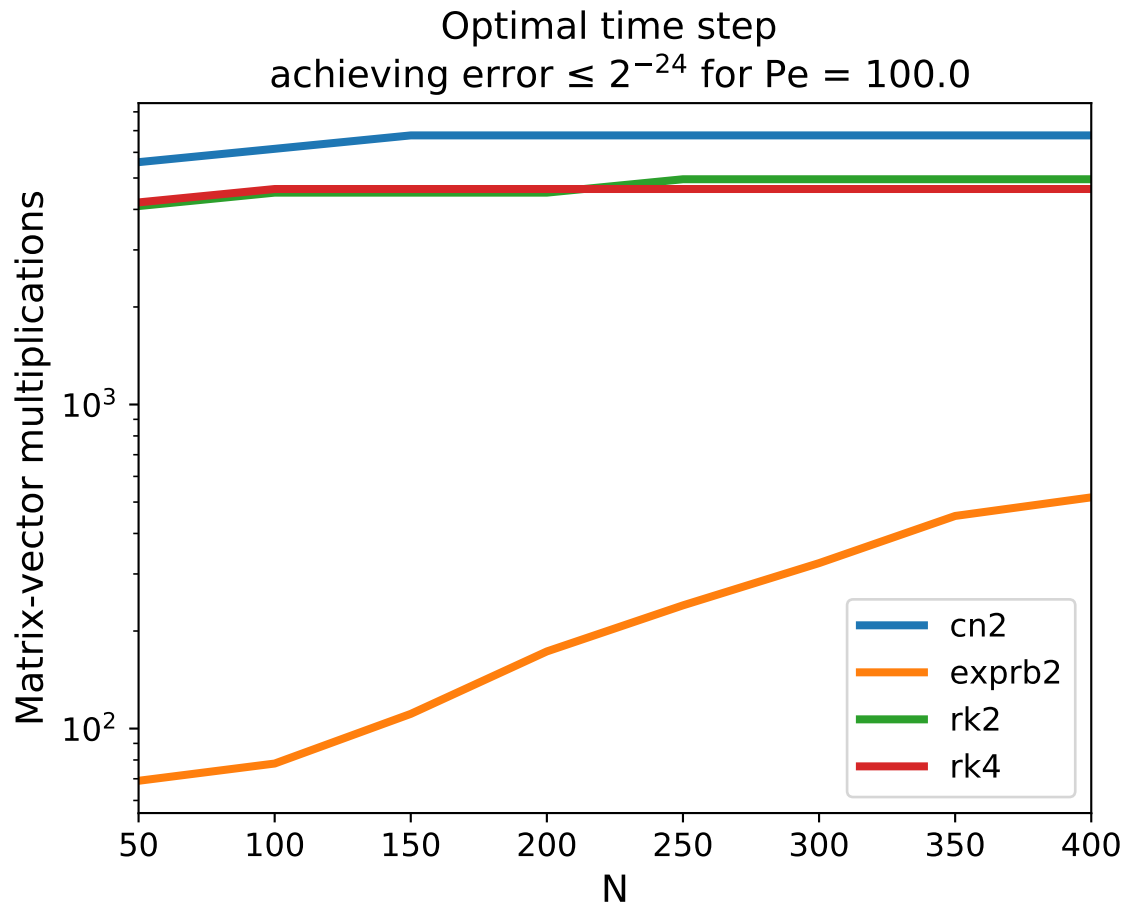
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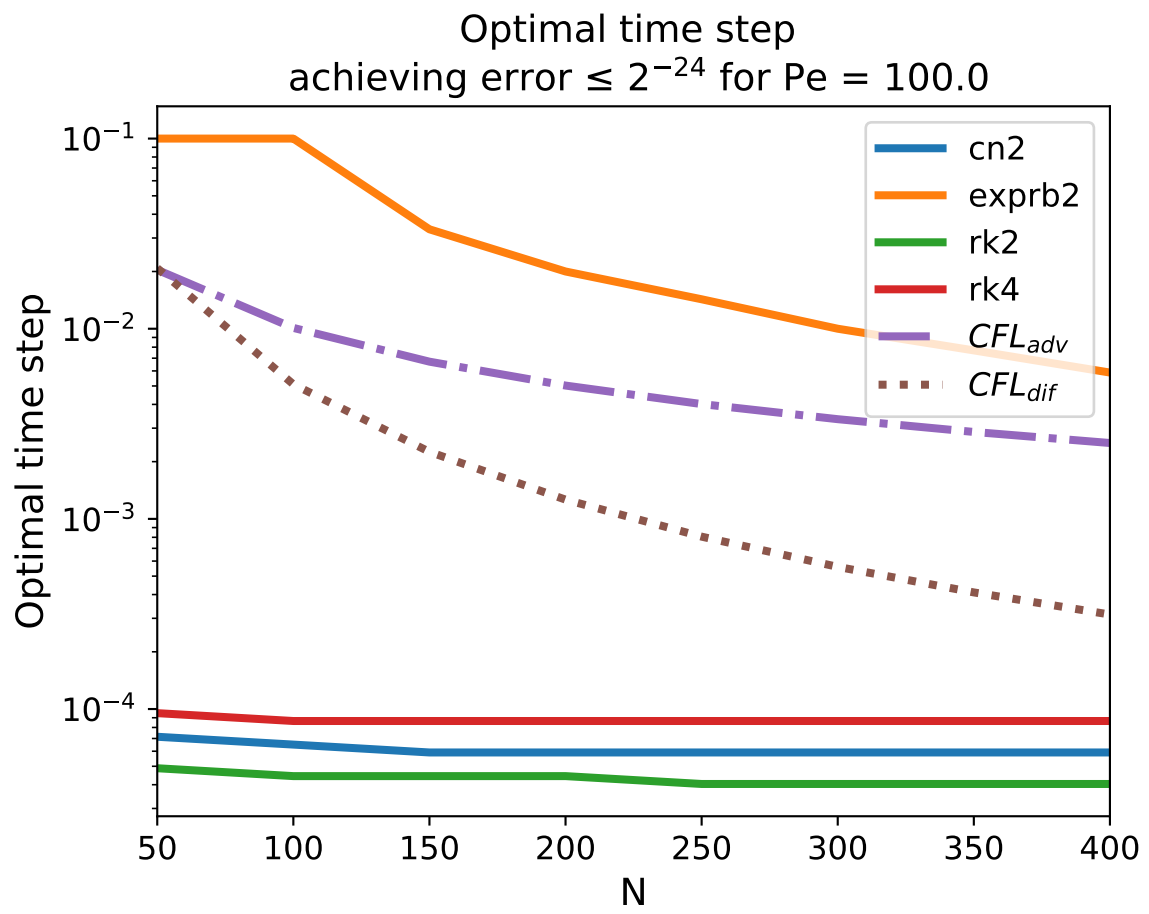
<sup>2</sup>Maybe create a separate section on hybrid difference schemes? There we can also analyze the resulting matrix  $A$  itself and plot the eigenvalues. I need sources for that though.

<sup>3</sup>See section ??

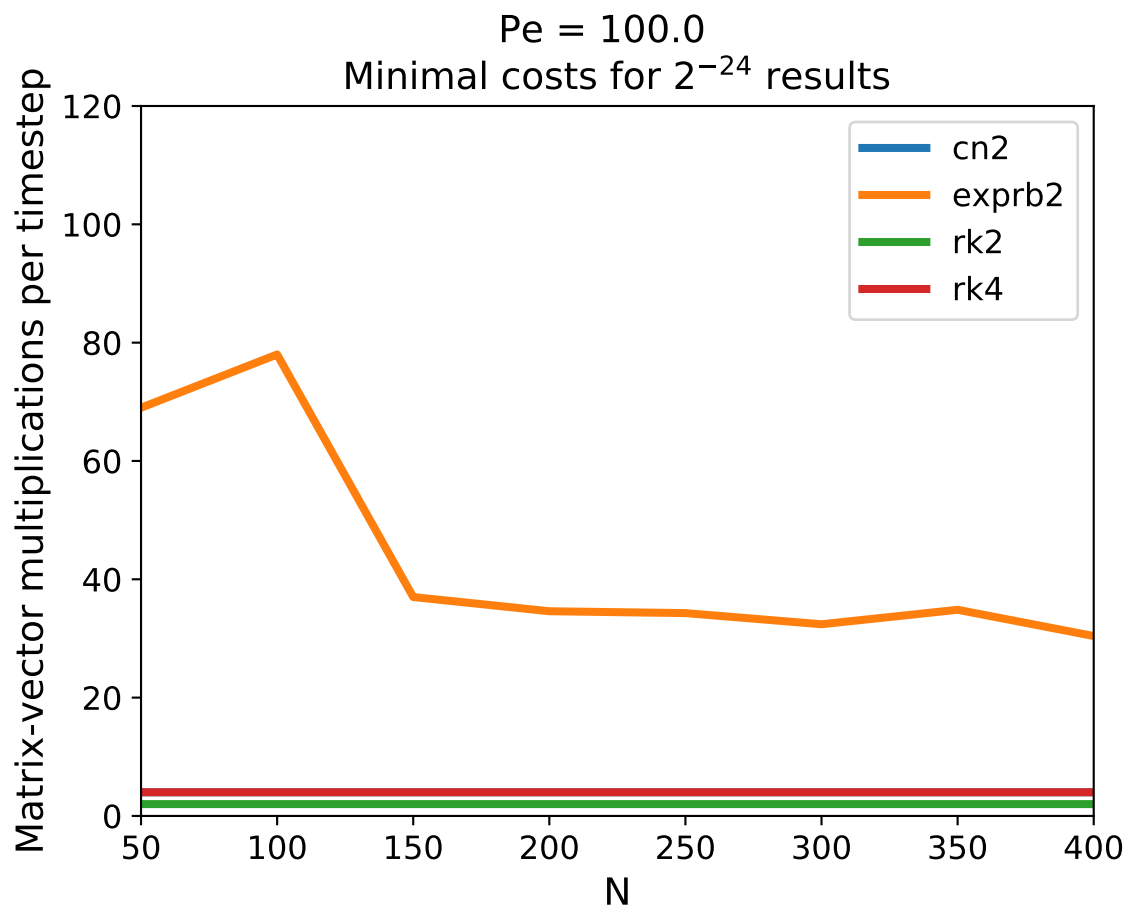
## 5 Appendix

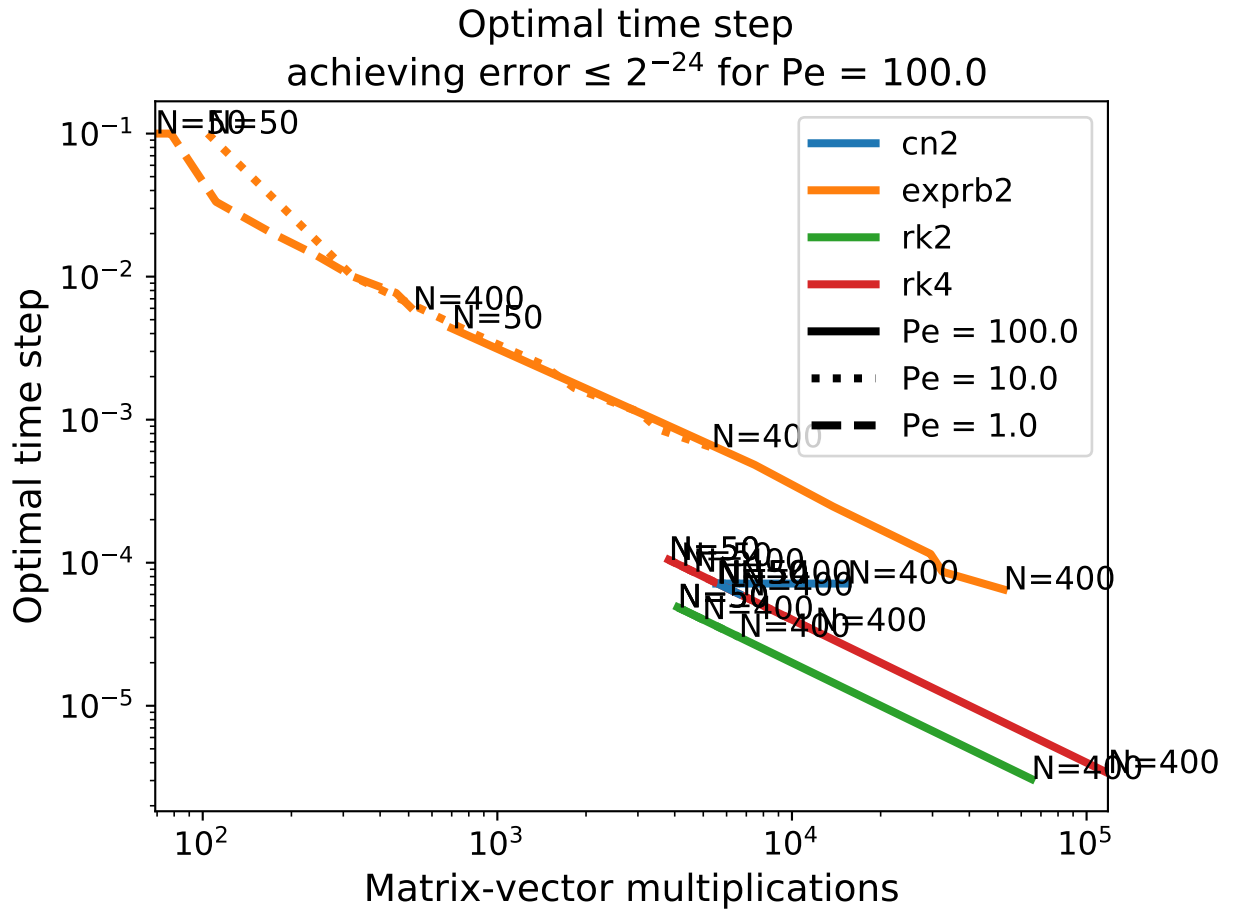
### 5.1 Experiment 1











## 5.2 Experiment 1.5

In the matrix-free case the linear operator  $A$  is not explicitly given. In order to compute the matrix norm  $\|A\|_2$  we use power iterations to estimate the absolutely largest eigenvalue of  $A$ . A priori it is not clear how many power iterations  $it$  are necessary for a good approximation.

Péclet: 100.0, sf: 1.1

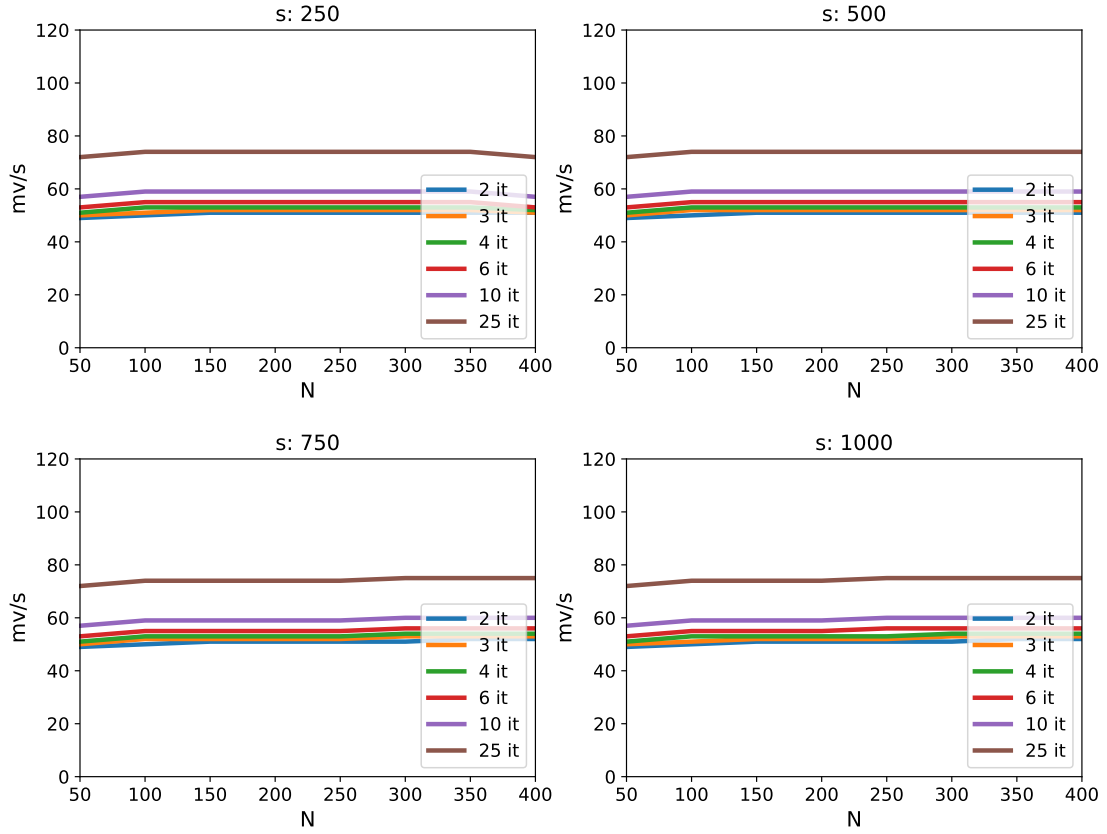


Figure 1: Space dimension  $N$  vs costs  $mv$  per timestep  $s$  for the exponential Rosenbrock method `exprb2`. Results are only shown if they achieve single precision.

## References

- [1] M. Caliari, A. Ostermann. Implementation of exponential Rosenbrock-type integrators, *Applied Numerical Mathematics* 59 (2009), 568-581.
- [2] A. Al-Mohy, N. Higham. Computing the action of the matrix exponential, with an application to exponential integrators, *SIAM Journal on Scientific Computing* 33 (2011), 488-511.
- [3] L. Reichel. Newton interpolation at Leja points, *BIT Numerical Mathematics* 30 (1990), 332-346.
- [4] M. Caliari, M. Vianello, L. Bergamaschi. Interpolating discrete advection-diffusion propagators at Leja sequences, *Journal of Computational and Applied Mathematics* 172 (2004), 79-99.
- [5] M. Caliari, P. Kandolf, A. Ostermann, S. Rainer. The Leja method revisited: backward error analysis for the matrix exponential, *SIAM Journal on Scientific Computation*, Accepted for publication (2016). arXiv:1506.08665.
- [6] Python Software Foundation. Python Language Reference, version 2.7. Available at <http://www.python.org>. Manual at <https://docs.python.org/2/>. [Online; accessed 2015-12-14]
- [7] E. Jones, E. Oliphant, P. Peterson, SciPy: Open Source Scientific Tools for Python, Available at <http://www.scipy.org/>. [Online; accessed 2015-12-14]