



Exponential Integrators using Matrix Functions: Krylov Subspace Methods and Chebyshev Expansion approximations

The HPC Approach

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Marlon Brenes

brenesnm@tcd.ie

<https://www.tcd.ie/Physics/research/groups/qusys/people/navarro/>



Outline

- Exponential Integrators
- Brief Introduction: Chebyshev expansion for matrix functions
- Brief Introduction: Krylov subspace techniques
- HPC Approach:
 - Relevance and importance of an HPC approach
 - Parallelisation strategy
- Outlook



Exponential Integrators

Matrix functions

The problem

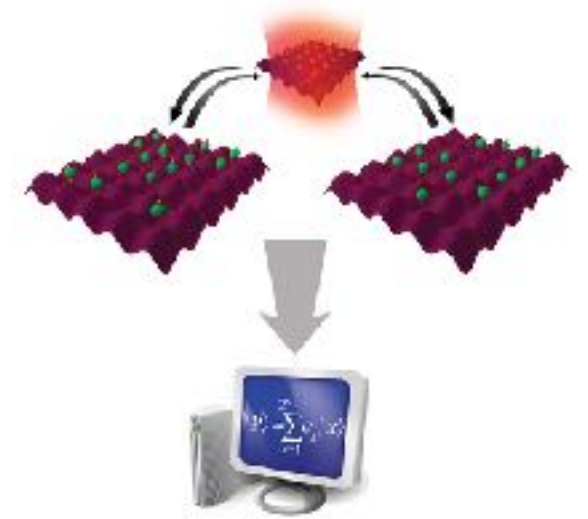
- Consider a problem of the type

$$\frac{d\mathbf{w}(t)}{dt} = \mathbf{A}\mathbf{w}(t), \quad t \in [0, T]$$

$$\mathbf{w}(0) = \mathbf{v}, \text{ initial condition}$$

- Its analytic solution is $\mathbf{w}(t) = e^{t\mathbf{A}}\mathbf{v}$
- Things to consider:
 - \mathbf{A} is a matrix
 - The exponential of the matrix is not really required, but merely it's **action** on the vector \mathbf{v}

Exponential Integrators



- Mathematical models of many physical, biological and economic processes systems of linear, constant-coefficient ordinary differential equations
- Growth of microorganisms, population, decay of radiation, control engineering, signal processing...
- More advanced: MHD (magnetohydrodynamics), quantum many-body problems, reaction-advection-diffusion equations...

Numerical approach

- The idea to use exponential functions for the matrix is **not new**
 - Was considered impractical...
- The development of Krylov subspace techniques to the **action** of the matrix exponential substantially changed the landscape
- Different types of solution evaluation for matrix exponentials:
 - ODE methods: numerical integration
 - Polynomial methods
 - Matrix decomposition methods

Numerical approach

- We're interested in the case where \mathbf{A} is **large** and **sparse**
- A *sole* implementation may not be reliable for *all* types of problems
- Chebyshev expansion approach
- The technique of Krylov subspaces has been proven to be very efficient for **many** classes of problems
- Convergence is faster than applying the solution to linear systems in both techniques



Take-home message #1:
There's a big number of problems in science and engineering that can be tackled using exponential integrators and matrix functions



Polynomial approximation: Chebyshev expansion

A brief introduction

Definition

- We intend to employ a polynomial expansion as an approximation
- Let us start with a definition of the matrix exponential by convergent power series:

$$e^{t\mathbf{A}} = \mathbf{I} + t\mathbf{A} + \frac{t^2 \mathbf{A}^2}{2!} + \dots$$

- An effective computation of the **action** of this operator on a vector is the main topic of this talk

Chebyshev expansion

- We intend to employ a **good** converging polynomial expansion
- Explicit computation of e^{A^t} **has** to be avoided
- **Key component:** Efficient matrix-vector product operations
- **Upside:** Efficient parallelisation and vectorisation, extremely simple approach
- **Downside:** Requires computation of two eigenvalues, not so versatile

Chebyshev polynomials

- The polynomials are defined by a three-term recursion relationship in the interval $x \in [-1, 1]$:

$$T_0(x) = 1$$

$$T_1(x) = x$$

$$T_{n+1}(x) = 2xT_n(x) - T_{n-1}(x)$$

- $T_n(x)$ constitutes an orthogonal basis, therefore one can write:

$$f(x) = \sum_{n=0}^{\infty} b_n T_n(x) \approx \sum_{n=0}^N b_n T_n(x)$$

- with $b_n = \frac{2 - \delta_n}{\pi} \int_{-1}^1 \frac{f(y) T_n(y)}{\sqrt{1 - y^2}} dy$

Bessel coefficients for the particular case of the exponential function

Chebyshev polynomials

- We're interested in applying the approximation to the action of the operator on a vector
- First step: Find the extremal eigenvalues of \mathbf{A} , more on this later

$$\lambda_{min} \quad \lambda_{max}$$

- Second step: Rescale operator such that it's spectrum is bounded by $[-1, 1]$

$$\mathbf{A}' = 2 \frac{\mathbf{A} - \lambda_{min} \mathbf{I}}{\lambda_{max} - \lambda_{min}} - \mathbf{I}$$

- Third step: Use the Chebyshev recursion relation

$$f(t\mathbf{A}')\mathbf{v} = e^{t\mathbf{A}'}\mathbf{v} \approx \sum_{n=0}^N b_n T_n(t\mathbf{A}')\mathbf{v}$$

- The recursion can be truncated up to a desired tolerance

Chebyshev recursion relation

- The recursion relation

$$f(t\mathbf{A}')\mathbf{v} = e^{t\mathbf{A}'}\mathbf{v} \approx \sum_{n=0}^N b_n T_n(t\mathbf{A}')\mathbf{v}$$

- Goes as follows:

$$\phi_0 = \mathbf{v}$$

$$\phi_1 = t\mathbf{A}'\mathbf{v}$$

$$\phi_{n+1} = 2t\mathbf{A}'\phi_n - \phi_{n-1}$$

- Then:

$$f(t\mathbf{A}')\mathbf{v} \approx \sum_{n=0}^N b_n \phi_n$$

- Until desired tolerance

Chebyshev polynomials

- Why do we choose the Chebyshev polynomials as basis set? Why not another polynomial set?
- Because of the asymptotic property of the Bessel function!
 - When the order n of the polynomial becomes larger than the argument, the function decays **exponentially** fast
 - This means that in order to obtain a good approximation, an exponentially decreasing amount of terms are required in the expansion as a function of the argument (related to t , λ_{min} and λ_{max})



Take-home message #2:

The Chebyshev expansion approach provides a numerically stable and scalable approach at the cost of some restrictions of the problem



Krylov subspace techniques to evaluate the solution

A brief introduction

Krylov subspace techniques

- We intend to employ a combination of a Krylov subspace technique and other known methods for matrix exponential
- Explicit computation of e^{A^t} **has** to be avoided
- **Key component:** Efficient matrix-vector product operations
- **Upside:** Extremely versatile
- **Downside:** Storage of the subspace for large problems, “time scales”

Main idea

- Building a Krylov subspace of dimension m

$$\mathcal{K}_m = \text{span}\{v, \mathbf{A}v, \mathbf{A}^2v, \dots, \mathbf{A}^{m-1}v\}$$

- The idea is to approximate the solution to the problem by an element of \mathcal{K}_m
- In order to manipulate the subspace, it's convenient to generate an orthonormal basis

$$V_m = [v_1, v_2, \dots, v_m] \quad v_1 = v / \|v\|_2$$

- This can be achieved with the **Arnoldi algorithm**

Algorithm: Arnoldi

1. *Initialize*: Compute $v_1 = v / \|v\|_2$.
2. *Iterate*: Do $j = 1, 2, \dots, m$
 - (a) Compute $w := Av_j$
 - (b) Do $i = 1, 2, \dots, j$
 - i. Compute $h_{i,j} := (w, v_i)$
 - ii. Compute $w := w - h_{i,j}v_i$
 - (c) Compute $h_{j+1,j} := \|w\|_2$ and $v_{j+1} := w / h_{j+1,j}$.

- Step 2-b is a modified Gram-Schmidt process.
- **Lanczos** can be applied for the case of symmetric matrices

Krylov subspace techniques

- The Arnoldi procedure produces a basis V_m of \mathcal{K}_m and an upper Hesseberg matrix \mathbf{H}_m of dimension $m \times m$ with coefficients h_{ij}
- We start by the relation given by

$$\mathbf{A}V_m = V_m\mathbf{H}_m + h_{m+1,m}v_{m+1}e_m^T$$

- Where v_{m+1} satisfies $V_m^T v_{m+1} = 0$ and $e_m \in I_m$
- From which we obtain

$$\mathbf{H}_m = V_m^T \mathbf{A} V_m$$

- Therefore, \mathbf{H}_m is the projection of the linear transformation \mathbf{A} onto the Krylov subspace \mathcal{K}_m with respect to the basis V_m

Krylov subspace techniques

- Given that \mathbf{H}_m is a projection of the linear operator, an approximation can be made such that

$$e^{t\mathbf{A}}v \approx ||v||_2 V_m e^{t\mathbf{H}_m} e_1$$

- The approximation is **exact** when the dimension of the Krylov subspace is equal to the dimension of the linear transformation
- Error

$$||e^{t\mathbf{A}}v - ||v||_2 V_m e^{t\mathbf{H}_m} e_1|| \leq 2||v||_2 \frac{(t||A||_2)^m e^{t||A||_2}}{m!}$$

Krylov subspace techniques

- With this approach:
 - A **large sparse** matrix problem is approximated by a **small dense** matrix problem
- There are several methods to evaluate the small dense matrix exponential
 - Series methods, ODE methods, diagonalisation, matrix decomposition methods...
 - Padè methods
- This method can be used to compute λ_{min} and λ_{max} for the Chebyshev approach, very effective



Take-home message #3:

Krylov subspace methods to evaluate the solution provides a more versatile and less restricted approach, at the expense of higher computational cost and memory consumption



HPC Approach

Relevance and importance

- A platform for numerical calculations
 - Important for current research
- Undertake simulations efficiently
- Quantum physics, CFD, finite-element methods...
- Establish an instance where HPC approach has been used recently in scientific research

Parallelisation strategy

$$y = Ax$$

- The linear operator is **sparse** (low density) and **huge** (big dimension, i.e, large amount of degrees of freedom)
- Approach:
 - **Distribute the operator among processing elements**



Take-home message #4:

Your numerical evaluation of the exponential integrator using matrix functions is only going to be **as good as** your implementation of the sparse matrix-vector product

Outlook

- Two different approximation methods to target the solutions to matrix exponentials, which can be interpreted analytically as solutions to a particular set of differential equations
- We will discuss the practical approach using techniques, practices and libraries from the HPC perspective next session...



Thank you!