

# Shift optimization for matrix exponential computing

Alexandr Katrutsa

Skolkovo Institute of Science and Technology

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## Problem statement

$$y(t) = \exp(-tA)v$$

- ▶ Integration of differential equations

$$\frac{dy}{dt} = -Ay, \quad y(0) = v$$

- ▶ Graph centrality measure
- ▶ Point cloud manifold descriptors

## Shift-and-invert method

- ▶ Krylov methods are based on

$$\mathcal{K}_k(A, v) = \text{span}(v, Av, A^2v, \dots, A^{k-1}v)$$

- ▶ Approximate in the orthogonal Krylov subspace

$$\exp(-tA)v \approx V_m \exp(-tT_m)e_1$$

- ▶ Matrix exponential is mostly determined by the *smallest* eigenpairs
- ▶ Replace  $\mathcal{K}_k(A, v)$  with  $\mathcal{K}_k((I + \gamma A)^{-1}, v)$
- ▶ Shift parameter  $\gamma$  affects convergence

## How to find optimal shift?

- ▶ Change parameter:  $\delta = \gamma/t$
- ▶ Introduce loss function

$$\|r\|_\infty = \|Ay(t) + y'(t)\|_\infty$$

and find  $\delta$  that minimizes it

- ▶ Introduce method to update  $\delta$  for online setting

## Optimization methods

Main assumption: all vectors  $v$  come from the some fixed subset

- ▶ Optimize-and-run method

$$\delta^* = \arg \min_{\delta \in [0.01, 0.1]} \frac{1}{N} \sum_{i=1}^N \|r(\delta \mid A, v^{(i)}, K, \epsilon, t)\|_\infty,$$

- ▶ Incremental method: sequentially update  $\delta$  for every next vector

## Incremental method

- ▶ Compute not only residual, but also derivative estimate
- ▶ Update  $\delta$  with this derivative estimate
- ▶ Take the next vector and use updated  $\delta$

## Application to PDE integration

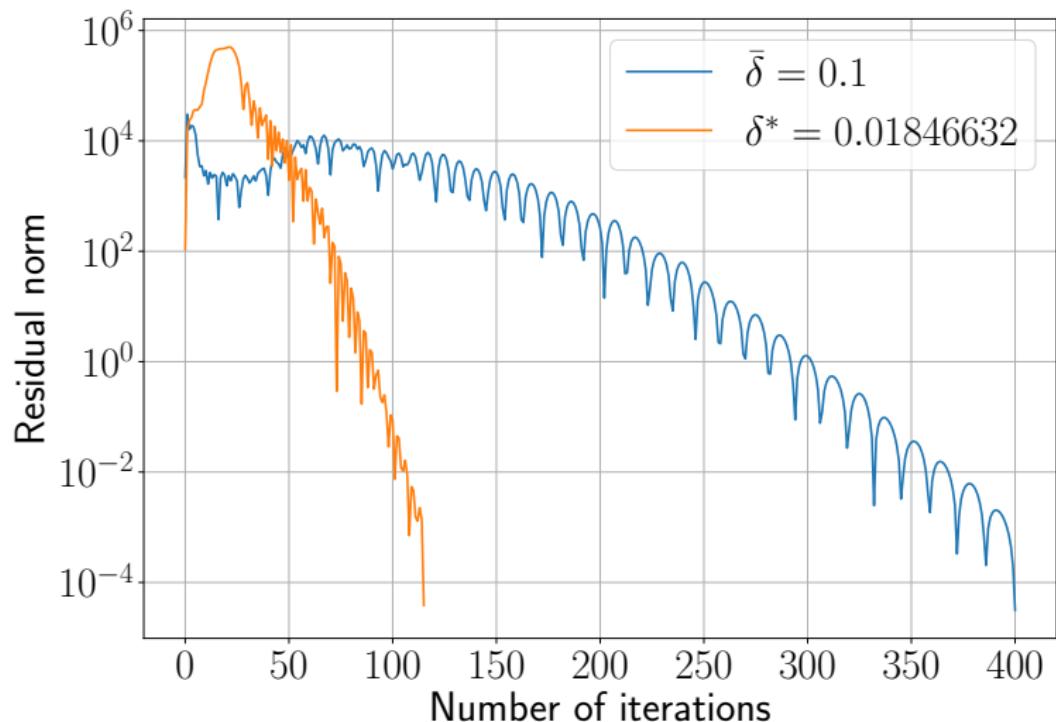
$$\begin{aligned}\frac{\partial u}{\partial t} &= (D_1 u_x)_x + (D_2 u_y)_y, \quad (x, y) \in \Omega = [0, 1] \times [0, 1], \\ u(x, y, 0) &= u_0,\end{aligned}$$

$$D_1 = \begin{cases} 1000, & (x, y) \in [\frac{1}{4}, \frac{3}{4}] \times [\frac{1}{4}, \frac{3}{4}], \\ 0.1, & \text{otherwise,} \end{cases} \quad D_2 = \frac{1}{2} D_1.$$

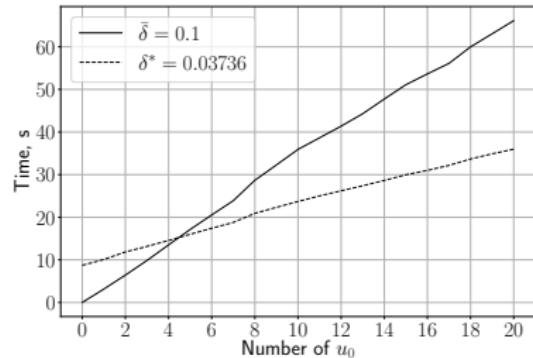
- ▶  $u_0 \sim \mathcal{N}(\mu, \Sigma)$
- ▶  $\mu$  – random vector from  $\Omega$
- ▶  $\Sigma = 0.05I$
- ▶ Second order finite difference discretization

## Convergence speed up

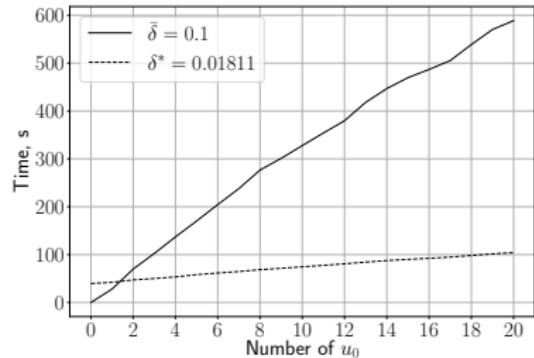
- ▶  $n = 300$ , matrix size is  $90000 \times 90000$
- ▶  $t = 2 \cdot 10^{-4}$



# Total running time comparison



(a)  $n = 200, t = 10^{-4}$

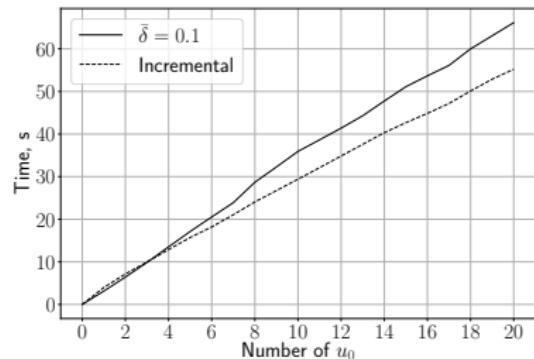


(b)  $n = 200, t = 4 \cdot 10^{-4}$

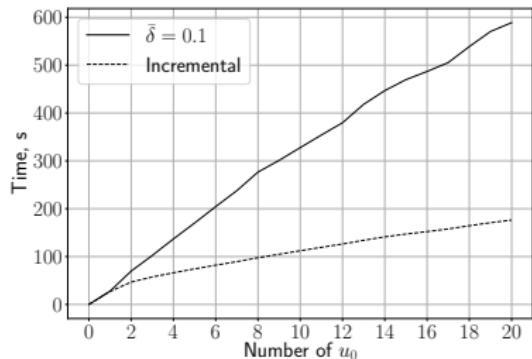
## Comparison average number of iterations

	$\bar{\delta} = 0.1$	$\delta^*$
$n = 200, t = 10^{-4}$	88.4	48.5
$n = 200, t = 4 \cdot 10^{-4}$	292.05	93.45
$n = 300, t = 10^{-4}$	237.75	100.6
$n = 300, t = 2 \cdot 10^{-4}$	517.7	150.75

## Incremental method



(a)  $n = 200, t = 10^{-4}$



(b)  $n = 200, t = 4 \cdot 10^{-4}$

## Summary

- ▶ Rational Krylov approximation of matrix exponential
- ▶ Residual norm as loss function to evaluate quality of shift
- ▶ Incremental method for online setting