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Preface

To be filled in!

Acknowledgment

To be filled in.

Summary and Conclusions

To be filled in!

Contents

Chapter 1

Introduction

Skrive litt om problemet, layouten og slikt.

The equation

$$\dot{y}(t) = Ay(t) + F(t)$$
$$y(0) = y_0$$

(1.1)

often makes an appearance when solving partial differential equations with numerical methods. The author has earlier observed how the heat equation, discretized with finite difference methods to be on the form of equation (1.1) can be solved with the use of the Krylov projection method(KPM) !!!!Cite rapporten min!!!!. This note will continue on the same track, with more focus on the wave equation, and energy preservation. It will also feature a comparison between Symplectic Lanzcos method(SLM) !!!Cite!!! and KPM. SLM is a projection technique that only works on Hamiltonian matrices. Due to this, SLM (claim to) preserve energy better than KPM.

Chapter 2

Explonation

There will here be a short explanation of all solvers, constants, outdata and expressions used in this text. Matlab notation is used where applicable. Fell free to go straight to next chapter, and use this as a reference later in the note.

!!!!!!!!!!!!!!!!Skriv dette sammen til en fin tabell!!!!!!!!!!!!!!!!

m	number of points in each spacial direction
n	Restart variable
k	number of points in time
restart	a boolean value. If restart == 1, Arnoldi or symplectinc lanzcos method will restart
convergence criterion	if restart == 1, restarting will commence until the change in the solution is less than the convergence criteria

Table 2.1: Some constants

iterations	number of restarts performed by Arnoldi or symplectic lanzcos method.
Time	time elapsed when solving the problem.
Error	difference between true solution, and estimated.
Energy error	Difference between the largest and smallest energy during the simulated time.
Error2	Difference in error between orthogonalised solution, and the non-orthogonalised solution.
Energy2	Difference in energy between orthogonalised solution, and the non-orthogonalised solution.

Table 2.2: Outdata

The table below shows how the matrix *A* was made. All test problems where implemented to

`wave`	A = 1/hs^2*gallery('poisson', m-2);
	A = [0,I;-A,0];
`semirandom`	A = rand(2*(m-2)^2);
	A = 0.5*[0,I;-I,0]*(A+A'+m^2*I)
	Keeps the matrix as long as m is the same,
	but creates a new when not.

Table 2.3: All solvers solve problems of the following form: $\dot{y} = Ay + F$, $y(0) = 0$, (F may depend on t, if need be, but not if the energy is to be kept constant). The matrix A is defined above for each equation. I is the identity matrix of size 2*(m-2)^2.

satisfy the condition $u(t,0,y) = u(t,1,y) = u(t,x,0) = u(t,x,1) = 0$. For `semirandom` all non-edge points where chosen randomly, and kept with the same conditions as for the A.

Three different integrators where used:

Trapezoidal rule
Forward Euler
Implicit midpoint rule

Table 2.4: list of Integrators

All problems on the form $\dot{y}(t) = Ay(t) \quad y(0) = y_0$ was transformed to the form $\dot{u}(t) = Au(t) + F \quad u(0) = 0$, with $u = y - y_0$ and $F = Ay_0$.

First we define the implicit midpoint rule:

$$y_{i+1} = y_i + hf\left(t_i + \frac{h}{2}, \frac{1}{2}(y_i + y_{i+1})\right),$$

forward Euler

$$y_{i+1} = y_i + hf(t_i, y_i)$$

and the trapezoidal rule

$$y_{i+1} = y_i + \frac{h}{2}(f(t_i, y_i) + f(t_{i+1}, y_{i+1})).$$

We are solving the system

$$\dot{y}(t) = f(t) = Ay(t) + F(t)$$

A problem with the midpoint rule is also that $F(t)$ is only known in certain points. This can

Implicit midpint rule	$U_i = (I - Ah_t)\backslash(U_{i-2} + h_t AU_{i-2} + 2h_t F_{i-1})$
Forward Euler	$U_i = U_{i-1} + h_t(AU_{i-1} + F_{i-1})$
trapezoidal rule	$U_i = (I - Ah_t/2)\backslash(U_{i-1} + h_t/2 AU_{i-1} + h_t/2(F_i + F_{i-1}))$
The exponential	$U_i = \exp(At_i)U_0$

Table 2.5: Iteration schemes

be worked around in two ways, either by approximating $F(t_n + h/2) = 1/2(F(t_n) + F(t_{n+1}))$, or doubling the step length. The iteration scheme is given in the table below.

If F is constant, then trapezoidal rule is equivalent to the midpoint rule. Since we are mainly interested in the case where the energy is constant, this becomes a good approximation. The problem arises when using the restart with a projection method, because in this case F is never constant.

compareintegrateenergy.jpg

The figure below shows the difference between the different time-integration methods

As we can see from the figure above the trapezoidal rule and midpoint rule is about equally good at approximating the error, while forward Euler is just terrible.
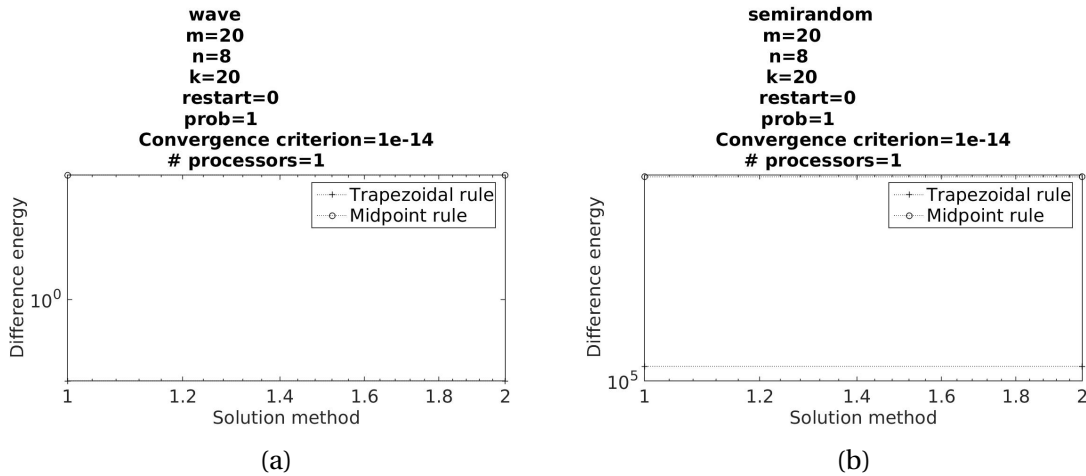

Figure 2.1: Figure of the difference in energy as a function of number of points in time. In the figure on the left, forward Euler is included, but because of it's poor convergence it is dropped from the rest of the figures in this chapter. The figure to the left shows that there is minimal difference between the trapezoidal rule and the midpoint rule.
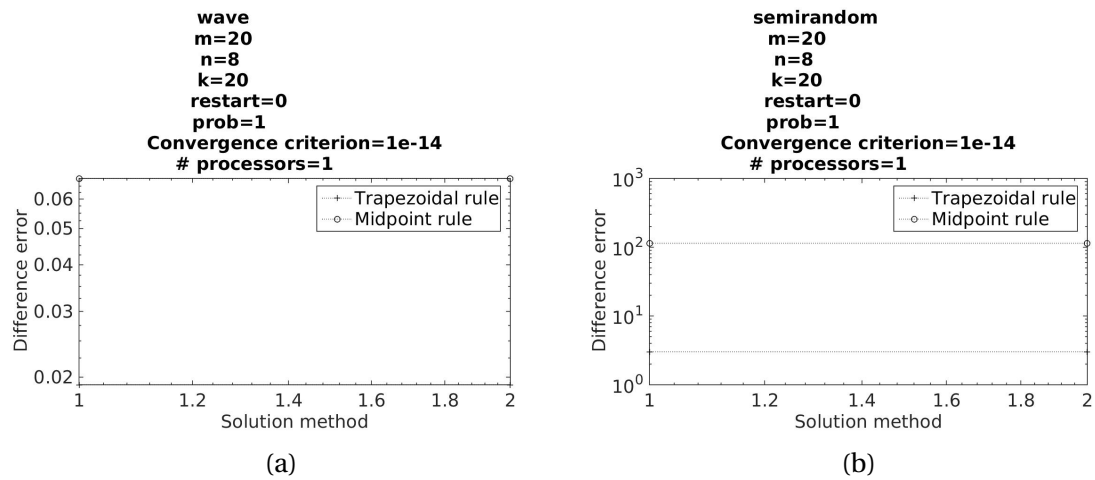
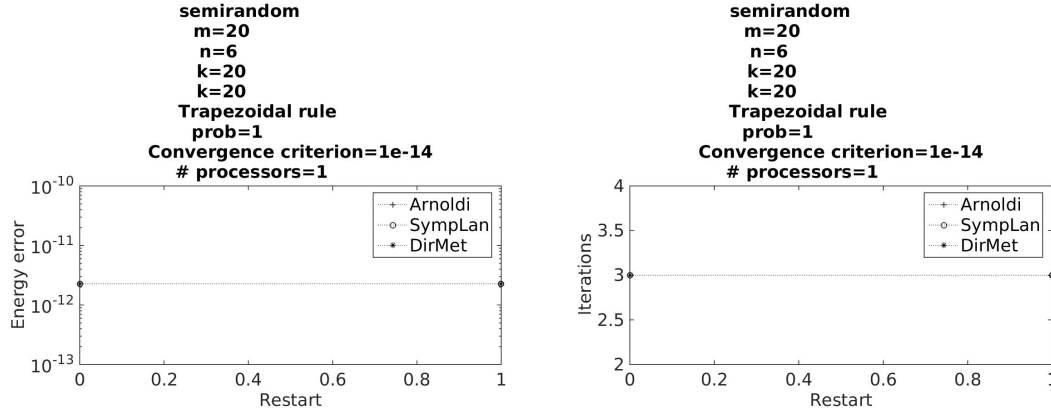**wave**
m=20
n=8
k=20
restart=0
prob=1
Convergence criterion=1e-14
 # processors=1

(a)

semirandom
m=20
n=8
k=20
restart=0
prob=1
Convergence criterion=1e-14
 # processors=1

(b)

Figure 2.2

Chapter 3

Symplectic Lanzcos method preserves the energy without restart

We here want to show that if we use symplectic Lanczos method, the restart will not change the energy.

semirandom
m=20
n=6
k=20
k=20
Trapezoidal rule
prob=1
Convergence criterion=1e-14
processors=1

semirandom
m=20
n=6
k=20
k=20
Trapezoidal rule
prob=1
Convergence criterion=1e-14
processors=1

The figures above implies that restarting the symplectic Lanzcos method does not change the energy. But for Arnoldi it changes quite a bit. To be completely sure that this is what happens, we are going to do everything a bit more precise.

The residual energy of the symplectic Lanczos method is

$$\frac{1}{2}e_r^\top J A e_r + e_r^\top J r_{n+1} e_{2n}^\top z$$

with $e_r = y - y_n$ (analytical solution minus approximated solution) and r_{n+1} is given by the symplectic lanzcos method.

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Chapter 4

Some results and explanation

We will now show some numerical experiments using a randomly generated Hamiltonian matrix(`semirandom`).

Chapter 5

Compare time integration methods

This section will be concerned with comparing the integrators. How well they estimate the error and energy will be the primary concern.

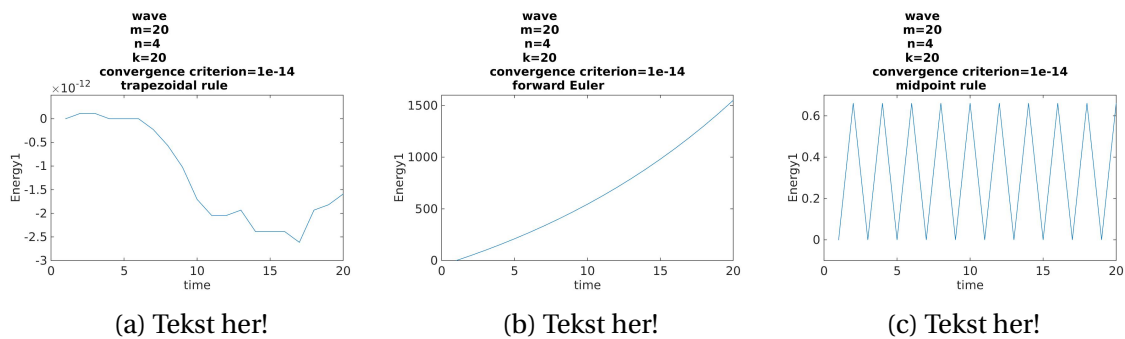(a) Tekst her!   (b) Tekst her!   (c) Tekst her!

Figure 5.1: Figure of the difference in energy at a point in time. It is clear that only trapezoidal rule gives a suitable approximation of the error. Forward euler has an ever increasing energy, while the midpoint rule gives periodic energy.

After looking at figure 5.1 it is easy to conclude that trapezoidal rule outperforms the other two methods.