### Study Guide: Truncation Error Analysis

### Hans Petter Langtangen $^{1,2}$

<sup>1</sup>Center for Biomedical Computing, Simula Research Laboratory <sup>2</sup>Department of Informatics, University of Oslo

### Sep 15, 2013

### Contents

	0.1	Motivation for studying truncation errors		
	0.2	Abstract problem setting		
	0.3	Truncation error for a differential equation problem		
1	Computing truncation errors in finite difference formulas			
	1.1	Example: The backward difference for $u'(t)$		
	1.2	Taylor series		
	1.3	Taylor series inserted in the backward difference approximation . 3		
	1.4	The forward difference for $u'(t)$		
	1.5	The central difference for $u'(t)$ (1)		
	1.6	The central difference for $u'(t)$ (1)		
	1.7	Overview of leading-order error terms in finite difference formulas		
		(1)		
	1.8	Overview of leading-order error terms in finite difference formulas		
		(2)		
	1.9	Overview of leading-order error terms in averages		
	1.10	Software for computing truncation errors		
		Symbolic computing with difference operators		
2	Tru	ncation errors in an ODE for exponential decay		
	2.1	Truncation error of the Forward Euler scheme		
	2.2	Truncation error of the Crank-Nicolson scheme		
	2.3	Test the understanding!		
	2.4	Truncation error of the $\theta$ -rule		
	2.5	Using symbolic software		
	2.6	Empirical verification of the truncation error (1)		
	2.7	Empirical verification of the truncation error $(2)$		
	2.8	Empirical verification of the truncation error in the Forward Euler		
		scheme		
	2.9	Increasing the accuracy by adding correction terms		

	2.10	Lowering the order of the derivative in the correction term	10
	2.11	With a correction term Forward Euler becomes Crank-Nicolson .	10
	2.12	Correction terms in the Crank-Nicolson scheme	11
	2.13	Extension to variable coefficients	12
	2.14	Exact solutions of the finite difference equations	12
	2.15	Computing truncation errors in nonlinear problems (1)	13
	2.16	Computing truncation errors in nonlinear problems (2)	13
3	Trui	acation errors in vibration ODEs	13
	3.1	Linear model without damping	13
	3.2	Computing correction terms	14
	3.3	Model with damping and nonlinearity	14
	3.4	Model with damping and nonlinearity: truncation error analysis	15
	3.5	Extension to quadratic damping	15
	3.6	The truncation error for quadratic damping $(1)$	15
	3.7	The truncation error for quadratic damping $(2)$	16
	3.8	The general model formulated as first-order ODEs	16
	3.9	The forward-backward scheme	16
	3.10	Truncation error analysis	17

### 0.1 Motivation for studying truncation errors

- Definition: The *truncation error* is the discrepancy that arises from performing a finite number of steps to approximate a process with infinitely many steps.
- Widely used: truncation of infinite series, finite precision arithmetic, finite differences, and differential equations.
- Why? The truncation error is an error measure that is easy to compute.

### 0.2 Abstract problem setting

Consider an abstract differential equation

$$\mathcal{L}(u) = 0.$$

Example:  $\mathcal{L}(u) = u'(t) + a(t)u(t) - b(t)$ .

The corresponding discrete equation:

$$\mathcal{L}_{\Delta}(u) = 0.$$

Let now

- $\bullet$  u be the numerical solution of the discrete equations
- $\bullet$   $u_{\rm e}$  the exact solution of the differential equation

Then

$$\mathcal{L}(u_{\mathrm{e}}) = 0,$$

$$\mathcal{L}_{\Delta}(u) = 0$$
.

The numerical solution is computed at mesh points:  $u^n$ ,  $n = 0, ..., N_t$ .

### 0.3 Truncation error for a differential equation problem

- Dream:  $e^n = u_e(t_n) u^n$
- Impossible, except for very simple problems
- Must find other error measures that are easier to calculate
- To what extent does  $u_{\rm e}$  fulfill  $\mathcal{L}_{\Delta}(u_{\rm e}) = 0$ ?
- It does not fit, but we can measure the error  $\mathcal{L}_{\Delta}(u_{\mathrm{e}}) = R$
- R is the truncation error and it is easy to compute

# 1 Computing truncation errors in finite difference formulas

### 1.1 Example: The backward difference for u'(t)

Backward difference approximation to u':

$$[D_t^- u]^n = \frac{u^n - u^{n-1}}{\Delta t} \approx u'(t_n). \tag{1}$$

Define the truncation error of this approximation as

$$R^{n} = u'(t_{n}) - [D_{t}^{-}u]^{n}. {2}$$

The common way of calculating  $\mathbb{R}^n$  is to

- 1. expand u(t) in a Taylor series around the point where the derivative is evaluated, here  $t_n$ ,
- 2. insert this Taylor series in (2), and
- 3. collect terms that cancel and simplify the expression.

### 1.2 Taylor series

General Taylor series expansion from calculus:

$$f(x+h) = \sum_{i=0}^{\infty} \frac{1}{i!} \frac{d^i f}{dx^i}(x) h^i.$$

Here: expand  $u^{n-1}$  around  $t_n$ :

$$u(t_{n-1}) = u(t - \Delta t) = \sum_{i=0}^{\infty} \frac{1}{i!} \frac{d^{i}u}{dt^{i}} (t_{n}) (-\Delta t)^{i}$$
$$= u(t_{n}) - u'(t_{n}) \Delta t + \frac{1}{2} u''(t_{n}) \Delta t^{2} + \mathcal{O}(\Delta t^{3}),$$

- $\mathcal{O}(\Delta t^3)$ : power-series in  $\Delta t$  where the lowest power is  $\Delta t^3$
- Small  $\Delta t$ :  $\Delta t \gg \Delta t^3 \gg \Delta t^4$

# 1.3 Taylor series inserted in the backward difference approximation

$$\begin{split} u'(t_n) - [D_t^- u]^n &= u'(t_n) - \frac{u(t_n) - u(t_{n-1})}{\Delta t} \\ &= u'(t_n) - \frac{u(t_n) - (u(t_n) - u'(t_n)\Delta t + \frac{1}{2}u''(t_n)\Delta t^2 + \mathcal{O}(\Delta t^3))}{\Delta t} \\ &= \frac{1}{2}u''(t_n)\Delta t + \mathcal{O}(\Delta t^2)) \end{split}$$

Result:

$$R^{n} = \frac{1}{2}u''(t_{n})\Delta t + \mathcal{O}(\Delta t^{2}).$$
(3)

The difference approximation is of first order in  $\Delta t$ . It is exact for linear  $u_e$ .

### 1.4 The forward difference for u'(t)

Forward difference:

$$u'(t_n) \approx [D_t^+ u]^n = \frac{u^{n+1} - u^n}{\Delta t}.$$

Define the truncation error:

$$u'(t_n) = [D_t^+ u]^n + R^n$$
.

Expand  $u^{n+1}$  in a Taylor series around  $t_n$ ,

$$u(t_{n+1}) = u(t_n) + u'(t_n)\Delta t + \frac{1}{2}u''(t_n)\Delta t^2 + \mathcal{O}(\Delta t^3).$$

We get

$$R = -\frac{1}{2}u''(t_n)\Delta t + \mathcal{O}(\Delta t^2).$$

### 1.5 The central difference for u'(t) (1)

For the central difference approximation,

$$u'(t_n) \approx [D_t u]^n, \quad [D_t u]^n = \frac{u^{n+\frac{1}{2}} - u^{n-\frac{1}{2}}}{\Delta t},$$

we write

$$u'(t_n) - [D_t u]^n = R^n,$$

and expand  $u(t_{n+\frac{1}{2}})$  and  $u(t_{n-1/2})$  in Taylor series around the point  $t_n$  where the derivative is evaluated:

$$\begin{split} u(t_{n+\frac{1}{2}}) = & u(t_n) + u'(t_n) \frac{1}{2} \Delta t + \frac{1}{2} u''(t_n) (\frac{1}{2} \Delta t)^2 + \\ & \frac{1}{6} u'''(t_n) (\frac{1}{2} \Delta t)^3 + \frac{1}{24} u''''(t_n) (\frac{1}{2} \Delta t)^4 + \mathcal{O}(\Delta t^5) \\ u(t_{n-1/2}) = & u(t_n) - u'(t_n) \frac{1}{2} \Delta t + \frac{1}{2} u''(t_n) (\frac{1}{2} \Delta t)^2 - \\ & \frac{1}{6} u'''(t_n) (\frac{1}{2} \Delta t)^3 + \frac{1}{24} u''''(t_n) (\frac{1}{2} \Delta t)^4 + \mathcal{O}(\Delta t^5) \;. \end{split}$$

### 1.6 The central difference for u'(t) (1)

$$u(t_{n+\frac{1}{2}}) - u(t_{n-1/2}) = u'(t_n)\Delta t + \frac{1}{24}u'''(t_n)\Delta t^3 + \mathcal{O}(\Delta t^5).$$

Collecting terms in  $[u' = D_t u + R]^n$  we find

$$R = -\frac{1}{24}u'''(t_n)\Delta t^2 + \mathcal{O}(\Delta t^4),\tag{4}$$

Note:

- Second-order scheme since the leading term is  $\Delta t^2$
- Only even powers of  $\Delta t$

# 1.7 Overview of leading-order error terms in finite difference formulas (1)

$$[D_{t}u]^{n} = \frac{u^{n+\frac{1}{2}} - u^{n-\frac{1}{2}}}{\Delta t} = u'(t_{n}) + R^{n},$$

$$R^{n} = \frac{1}{24}u'''(t_{n})\Delta t^{2} + \mathcal{O}(\Delta t^{4})$$

$$[D_{2t}u]^{n} = \frac{u^{n+1} - u^{n-1}}{2\Delta t} = u'(t_{n}) + R^{n},$$

$$R^{n} = \frac{1}{6}u'''(t_{n})\Delta t^{2} + \mathcal{O}(\Delta t^{4})$$

$$[D_{t}^{-}u]^{n} = \frac{u^{n} - u^{n-1}}{\Delta t} = u'(t_{n}) + R^{n},$$

$$R^{n} = \frac{1}{2}u''(t_{n})\Delta t + \mathcal{O}(\Delta t^{2})$$

$$[D_{t}^{+}u]^{n} = \frac{u^{n+1} - u^{n}}{\Delta t} = u'(t_{n}) + R^{n},$$

$$R^{n} = \frac{1}{2}u''(t_{n})\Delta t + \mathcal{O}(\Delta t^{2})$$

$$(8)$$

# 1.8 Overview of leading-order error terms in finite difference formulas (2)

$$[\bar{D}_{t}u]^{n+\theta} = \frac{u^{n+1} - u^{n}}{\Delta t} = u'(t_{n+\theta}) + R^{n+\theta},$$

$$R^{n+\theta} = \frac{1}{2}(1 - 2\theta)u''(t_{n+\theta})\Delta t - \frac{1}{6}((1 - \theta)^{3} - \theta^{3})u'''(t_{n+\theta})\Delta t^{2} + \mathcal{O}(\Delta t^{3})$$

$$[D_{t}^{2-}u]^{n} = \frac{3u^{n} - 4u^{n-1} + u^{n-2}}{2\Delta t} = u'(t_{n}) + R^{n},$$

$$R^{n} = -\frac{1}{3}u'''(t_{n})\Delta t^{2} + \mathcal{O}(\Delta t^{3})$$

$$[D_{t}D_{t}u]^{n} = \frac{u^{n+1} - 2u^{n} + u^{n-1}}{\Delta t^{2}} = u''(t_{n}) + R^{n},$$

$$R^{n} = \frac{1}{12}u''''(t_{n})\Delta t^{2} + \mathcal{O}(\Delta t^{4})$$

$$(11)$$

### 1.9 Overview of leading-order error terms in averages

Weighted arithmetic mean:

$$[\overline{u}^{t,\theta}]^{n+\theta} = \theta u^{n+1} + (1-\theta)u^n = u(t_{n+\theta}) + R^{n+\theta},$$

$$R^{n+\theta} = \frac{1}{2}u''(t_{n+\theta})\Delta t^2 \theta (1-\theta) + \mathcal{O}(\Delta t^3).$$
(12)

Standard arithmetic mean:

$$[\overline{u}^t]^n = \frac{1}{2}(u^{n-\frac{1}{2}} + u^{n+\frac{1}{2}}) = u(t_n) + R^n,$$

$$R^n = \frac{1}{8}u''(t_n)\Delta t^2 + \frac{1}{384}u''''(t_n)\Delta t^4 + \mathcal{O}(\Delta t^6).$$
(13)

Geometric mean:

$$u^{n-\frac{1}{2}}u^{n+\frac{1}{2}} = (u^n)^2 + R^n,$$

$$R^n = -\frac{1}{4}u'(t_n)^2 \Delta t^2 + \frac{1}{4}u(t_n)u''(t_n)\Delta t^2 + \mathcal{O}(\Delta t^4).$$
(14)

Harmonic mean:

$$u^{n} = \frac{2}{\frac{1}{u^{n-\frac{1}{2}}} + \frac{1}{u^{n+\frac{1}{2}}}} + R^{n+\frac{1}{2}},$$

$$R^{n} = -\frac{u'(t_{n})^{2}}{4u(t_{n})} \Delta t^{2} + \frac{1}{8}u''(t_{n}) \Delta t^{2}.$$
(15)

#### 1.10 Software for computing truncation errors

Can use sympy to automate calculations with Taylor series.

```
>>> from truncation_errors import TaylorSeries
>>> from sympy import *
>>> u, dt = symbols('u dt')
>>> u_Taylor = TaylorSeries(u, 4)
>>> u_Taylor(dt)
D1u*dt + D2u*dt**2/2 + D3u*dt**3/6 + D4u*dt**4/24 + u
>>> FE = (u_Taylor(dt) - u)/dt
>>> FE
(D1u*dt + D2u*dt**2/2 + D3u*dt**3/6 + D4u*dt**4/24)/dt
>>> simplify(FE)
D1u + D2u*dt/2 + D3u*dt**2/6 + D4u*dt**3/24
```

Notation: D1u for u', D2u for u'', etc. See trunc/truncation\_errors.py<sup>1</sup>.

### 1.11 Symbolic computing with difference operators

A class DiffOp represents many common difference operatorsL

```
>>> from truncation_errors import DiffOp
>>> from sympy import *
>>> u = Symbol('u')
>>> diffop = DiffOp(u, independent_variable='t')
>>> diffop['geometric_mean']
-D1u**2*dt**2/4 - D1u*D3u*dt**4/48 + D2u**2*dt**4/64 + ...
>>> diffop['Dtm']
D1u + D2u*dt/2 + D3u*dt**2/6 + D4u*dt**3/24
>>> diffop.operator_names()
['geometric_mean', 'harmonic_mean', 'Dtm', 'D2t', 'DtDt',
    'weighted_arithmetic_mean', 'Dtp', 'Dt']
```

 $<sup>^{1} \</sup>verb|http://tinyurl.com/jvzzcfn/trunc/truncation_errors.py|$ 

Names in diffop: Dtp for  $D_t^+$ , Dtm for  $D_t^-$ , Dt for  $D_t$ , D2t for  $D_{2t}$ , DtDt for  $D_tD_t$ .

# 2 Truncation errors in an ODE for exponential decay

Model:

$$u'(t) = -au(t), \quad u(0) = I.$$
 (16)

### 2.1 Truncation error of the Forward Euler scheme

The Forward Euler scheme:

$$[D_t^+ u = -au]^n. (17)$$

Definition of the truncation error  $\mathbb{R}^n$ :

$$[D_t^+ u_e + a u_e = R]^n. (18)$$

From (8):

$$[D_t^+ u_e]^n = u'_e(t_n) + \frac{1}{2}u''_e(t_n)\Delta t + \mathcal{O}(\Delta t^2).$$

Inserted in (18):

$$u'_{e}(t_n) + \frac{1}{2}u''_{e}(t_n)\Delta t + \mathcal{O}(\Delta t^2) + au_{e}(t_n) = R^n.$$

Key observation:  $u'_{e}(t_n) + au_{e}^{n} = 0$  since  $u_{e}$  solves the ODE. The remaining terms constitute the residual:

$$R^{n} = \frac{1}{2} u_{\mathrm{e}}^{"}(t_{n}) \Delta t + \mathcal{O}(\Delta t^{2}). \tag{19}$$

Largest term in  $\mathbb{R}^n$  is proportional to  $\Delta t$ , hence a first-order (in  $\Delta t$ ) scheme.

### 2.2 Truncation error of the Crank-Nicolson scheme

Crank-Nicolson:

$$[D_t u = -au]^{n + \frac{1}{2}}, (20)$$

Truncation error:

$$[D_t u_e + a \overline{u_e}^t = R]^{n + \frac{1}{2}}. \tag{21}$$

From (5) and (13):

$$[D_t u_e]^{n+\frac{1}{2}} = u'(t_{n+\frac{1}{2}}) + \frac{1}{24} u_e'''(t_{n+\frac{1}{2}}) \Delta t^2 + \mathcal{O}(\Delta t^4),$$
$$[a\overline{u_e}^t]^{n+\frac{1}{2}} = u(t_{n+\frac{1}{2}}) + \frac{1}{8} u''(t_n) \Delta t^2 + \mathcal{O}(\Delta t^4)$$

Inserted in the scheme we get

$$R^{n+\frac{1}{2}} = \left(\frac{1}{24}u_{\rm e}^{\prime\prime\prime}(t_{n+\frac{1}{2}}) + \frac{1}{8}u^{\prime\prime}(t_n)\right)\Delta t^2 + \mathcal{O}(\Delta t^4) \tag{22}$$

 $R^n = \mathcal{O}(\Delta t^2)$  (second-order scheme)

#### 2.3 Test the understanding!

Analyze the truncation error of the Backward Euler scheme.

#### 2.4 Truncation error of the heta-rule

The  $\theta$ -rule:

$$[\bar{D}_t u = -a\overline{u}^{t,\theta}]^{n+\theta}.$$

Truncation error:

$$[\bar{D}_t u_e + a \overline{u_e}^{t,\theta} = R]^{n+\theta}$$

Use (9) and (12) along with  $u'_{e}(t_{n+\theta}) + au_{e}(t_{n+\theta}) = 0$  to show

$$R^{n+\theta} = (\frac{1}{2} - \theta)u_{\rm e}^{"}(t_{n+\theta})\Delta t + \frac{1}{2}\theta(1 - \theta)u_{\rm e}^{"}(t_{n+\theta})\Delta t^{2} + \frac{1}{2}(\theta^{2} - \theta + 3)u_{\rm e}^{"}(t_{n+\theta})\Delta t^{2} + \mathcal{O}(\Delta t^{3})$$
(23)

Note: 2nd-order scheme if and only if  $\theta = 1/2$ .

### 2.5 Using symbolic software

Can use sympy and the tools in trunc/truncation\_errors.py<sup>2</sup>:

The returned dictionary becomes

```
decay: {
  'BE': D2u*dt/2 + D3u*dt**2/6,
  'FE': -D2u*dt/2 + D3u*dt**2/6,
  'CN': D3u*dt**2/24,
}
```

 $\theta$ -rule: see truncation\_errors.py (long expression, very advantageous to automate the math)

 $<sup>^2 \</sup>verb|http://tinyurl.com/jvzzcfn/trunc/truncation\_errors.py|$ 

### 2.6 Empirical verification of the truncation error (1)

Ideas:

- Compute  $\mathbb{R}^n$  numerically
- Run a sequence of meshes, see how  $\mathbb{R}^n$  converges

For the Forward Euler scheme:

$$R^{n} = [D_{t}^{+} u_{e} + a u_{e}]^{n}. (24)$$

Insert correct  $u_e(t)$ , or add source term to fit any choice of  $u_e$ .

### 2.7 Empirical verification of the truncation error (2)

- Assume  $R = C\Delta t^r$
- R varies with n; C and r will vary with n must estimate r for each mesh point
- Use a sequence of meshes:  $N_t = 2^{-k}N_0, k = 1, 2, ...$
- ullet Transform R data to the coarsest mesh and estimate r for each coarse mesh point

See the  $\operatorname{text}^3$  for more details and an implementation.

### 2.8 Empirical verification of the truncation error in the Forward Euler scheme

### 2.9 Increasing the accuracy by adding correction terms

#### Question.

Can we add terms in the differential equation that can help increase the order of the truncation error?

To be precise for the Forward Euler scheme, can we find C to make R  $\mathcal{O}(\Delta t^2)$ ?

$$[D_t^+ u_e + au_e = C + R]^n. (25)$$

Taylor expand:

$$\frac{1}{2}u_{\rm e}''(t_n)\Delta t - \frac{1}{6}u_{\rm e}'''(t_n)\Delta t^2 + \mathcal{O}(\Delta t^3) = C^n + R^n.$$

Observe: choosing

$$C^n = \frac{1}{2} u_{\mathbf{e}}''(t_n) \Delta t,$$

makes

$$R^n = \frac{1}{6} u_{\rm e}^{\prime\prime\prime}(t_n) \Delta t^2 + \mathcal{O}(\Delta t^3).$$

 $<sup>^3</sup>$ http://tinyurl.com/k3sdbuv/pub/

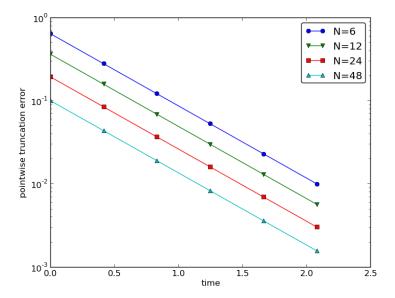


Figure 1: Estimated truncation error at mesh points for different meshes.

# 2.10 Lowering the order of the derivative in the correction term

- $C^n$  contains u''
- Can discretize u'' (one more time level)
- ullet Can also rewrite to u' or u

$$u' = -au, \quad \Rightarrow \quad u'' = -au' = a^u.$$

Result for  $u'' = a^u$ : apply Forward Euler to a perturbed ODE,

$$u' = -\hat{a}u, \quad \hat{a} = a(1 - \frac{1}{2}a\Delta t).$$
 (26)

### 2.11 With a correction term Forward Euler becomes Crank-Nicolson

Use u'' = -au':

$$u' = -au - \frac{1}{2}a\Delta t u' \quad \Rightarrow \quad \left(1 + \frac{1}{2}a\Delta t\right)u' = -au.$$

Apply Forward Euler:

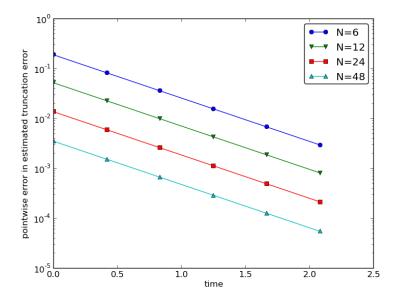


Figure 2: Difference between theoretical and estimated truncation error at mesh points for different meshes.

$$\left(1 + \frac{1}{2}a\Delta t\right)\frac{u^{n+1} - u^n}{\Delta t} = -au^n,$$

which after some algebra can be written as

$$u^{n+1} = \frac{1 - \frac{1}{2}a\Delta t}{1 + \frac{1}{2}a\Delta t}u^n.$$

This is a Crank-Nicolson scheme for u' = -au!

### 2.12 Correction terms in the Crank-Nicolson scheme

Standard Crank-Nicolson scheme:

$$[D_t u = -au]^{n + \frac{1}{2}},$$

Definition of the truncation error R and correction terms C:

$$[D_t u_e + a \overline{u_e}^t = C + R]^{n + \frac{1}{2}}.$$

Must Taylor expand

- the derivative
- the arithmetic mean

$$C^{n+\frac{1}{2}} + R^{n+\frac{1}{2}} = \frac{1}{24} u_{\mathrm{e}}^{\prime\prime\prime}(t_{n+\frac{1}{2}}) \Delta t^2 + \frac{a}{8} u_{\mathrm{e}}^{\prime\prime}(t_{n+\frac{1}{2}}) \Delta t^2 + \mathcal{O}(\Delta t^4) \,.$$

Let  $C^{n+\frac{1}{2}}$  cancel the  $\Delta t^2$  terms:

$$C^{n+\frac{1}{2}} = \frac{1}{24} u_{\rm e}^{\prime\prime\prime}(t_{n+\frac{1}{2}}) \Delta t^2 + \frac{a}{8} u_{\rm e}^{\prime\prime}(t_n) \Delta t^2 \,. \label{eq:cn+2}$$

Using u' = -au, we have  $u'' = a^2u$  and u''' = -a3u.

Result: solve the perturbed ODE by a Crank-Nicolson method,

$$u' = -\hat{a}u, \quad \hat{a} = a(1 - \frac{1}{12}a^2\Delta t^2).$$

Error:  $\mathcal{O}(\Delta t^4)$ .

### 2.13 Extension to variable coefficients

$$u'(t) = -a(t)u(t) + b(t), \quad u(0) = I,$$

Forward Euler:

$$[D_t^+ u = -au + b]^n. (27)$$

The truncation error fulfills

$$[D_t^+ u_e + au_e - b = R]^n. (28)$$

Using (8),

$$u'_{e}(t_n) - \frac{1}{2}u''_{e}(t_n)\Delta t + \mathcal{O}(\Delta t^2) + a(t_n)u_{e}(t_n) - b(t_n) = R^n.$$

Because of the ODE,  $u'_{e}(t_n) + a(t_n)u_{e}(t_n) - b(t_n) = 0$ , and

$$R^{n} = -\frac{1}{2}u_{\mathbf{e}}^{"}(t_{n})\Delta t + \mathcal{O}(\Delta t^{2}). \tag{29}$$

No problems with variable coefficients!

#### 2.14 Exact solutions of the finite difference equations

- One-sided differences:  $u''_{\rm e}\Delta t$  (lowest order)
- Centered differences:  $u_{\rm e}^{\prime\prime\prime}\Delta t^2$  (lowest order)
- $\bullet$  Only harmonic and geometric mean involve  $u_{\rm e}'$  or  $u_{\rm e}$

Consequence:

- $u_e(t) = ct + d$  will very often give exact solution of the discrete equations (R = 0)!
- Ideal for verification
- Centered schemes allow quadratic  $u_e$

# 2.15 Computing truncation errors in nonlinear problems (1)

$$u' = f(u, t), \tag{30}$$

Crank-Nicolson scheme:

$$[D_t u' = \overline{f}^t]^{n + \frac{1}{2}}. \tag{31}$$

Truncation error:

$$[D_t u_e' - \overline{f}^t = R]^{n + \frac{1}{2}}. (32)$$

Using (13) for the arithmetic mean:

$$\begin{split} [\overline{f}^t]^{n+\frac{1}{2}} &= \frac{1}{2} (f(u_{\rm e}^n, t_n) + f(u_{\rm e}^{n+1}, t_{n+1})), \\ & f(u_{\rm e}^{n+\frac{1}{2}}, t_{n+\frac{1}{2}}) + \frac{1}{8} u_{\rm e}''(t_{n+\frac{1}{2}}) \Delta t^2 + \mathcal{O}(\Delta t^4) \,. \end{split}$$

# 2.16 Computing truncation errors in nonlinear problems (2)

With (5), (32) leads to

$$u_{\mathrm{e}}'(t_{n+\frac{1}{2}}) + \frac{1}{24} u_{\mathrm{e}}'''(t_{n+\frac{1}{2}}) \Delta t^2 - f(u_{\mathrm{e}}^{n+\frac{1}{2}}, t_{n+\frac{1}{2}}) - \frac{1}{8} u''(t_{n+\frac{1}{2}}) \Delta t^2 + \mathcal{O}(\Delta t^4) = R^{n+\frac{1}{2}} \,.$$

Since  $u_{\rm e}'(t_{n+\frac{1}{2}})-f(u_{\rm e}^{n+\frac{1}{2}},t_{n+\frac{1}{2}})=0$ , the truncation error becomes

$$R^{n+\frac{1}{2}} = (\frac{1}{24} u_{\mathrm{e}}^{\prime\prime\prime}(t_{n+\frac{1}{2}}) - \frac{1}{8} u^{\prime\prime}(t_{n+\frac{1}{2}})) \Delta t^2 \,.$$

The computational techniques worked well even for this nonlinear ODE!

### 3 Truncation errors in vibration ODEs

### 3.1 Linear model without damping

$$u''(t) + \omega^2 u(t) = 0, \quad u(0) = I, \ u'(0) = 0.$$
 (33)

Centered difference approximation:

$$[D_t D_t u + \omega^2 u = 0]^n. (34)$$

Truncation error:

$$[D_t D_t u_e + \omega^2 u_e = R]^n. \tag{35}$$

Use (11) to expand  $[D_t D_t u_e]^n$ :

$$[D_t D_t u_e]^n = u''_e(t_n) + \frac{1}{12} u''''_e(t_n) \Delta t^2,$$

Collect terms:  $u''_{e}(t) + \omega^{2}u_{e}(t) = 0$ . Then,

$$R^{n} = \frac{1}{12} u_{e}^{""}(t_{n}) \Delta t^{2} + \mathcal{O}(\Delta t^{4}).$$
 (36)

### 3.2 Computing correction terms

Can we add terms to the ODE such that the truncation error is improved?

$$[D_t D_t u_e + \omega^2 u_e = C + R]^n,$$

Idea: choose  $C^n$  such that it absorbs the  $\Delta t^2$  term in  $\mathbb{R}^n$ ,

$$C^n = \frac{1}{12} u_e^{\prime\prime\prime\prime}(t_n) \Delta t^2.$$

Downside: u''''

Remedy: use the ODE  $u'' = -\omega^u$  to see that  $u'''' = \omega^4 u$ .

$$u'' + \omega^2 (1 - \frac{1}{12}\omega^2 \Delta t^2) u = 0.$$
 (37)

Just apply the standard scheme to this modified ODE:

$$[D_t D_t u + \omega^2 (1 - \frac{1}{12} \omega^2 \Delta t^2) u = 0]^n,$$

Accuracy is  $\mathcal{O}(\Delta t^4)$ .

### 3.3 Model with damping and nonlinearity

Linear damping  $\beta u'$ , nonlinear spring force s(u), and excitation F:

$$mu'' + \beta u' + s(u) = F(t).$$
 (38)

Central difference discretization:

$$[mD_tD_tu + \beta D_{2t}u + s(u) = F]^n$$
. (39)

Truncation error is defined by

$$[mD_tD_tu_e + \beta D_{2t}u_e + s(u_e) = F + R]^n.$$
(40)

# 3.4 Model with damping and nonlinearity: truncation error analysis

Using (11) and (6) we get

$$[mD_t D_t u_e + \beta D_{2t} u_e]^n = m u_e''(t_n) + \beta u_e'(t_n) + \left(\frac{m}{12} u_e''''(t_n) + \frac{\beta}{6} u_e'''(t_n)\right) \Delta t^2 + \mathcal{O}(\Delta t^4)$$

The terms

$$mu''_{e}(t_n) + \beta u'_{e}(t_n) + \omega^2 u_{e}(t_n) + s(u_{e}(t_n)) - F^n,$$

correspond to the ODE (= zero).

Result: accuracy of  $\mathcal{O}(\Delta t^2)$  since

$$R^{n} = \left(\frac{m}{12}u_{e}^{""}(t_{n}) + \frac{\beta}{6}u_{e}^{"'}(t_{n})\right)\Delta t^{2} + \mathcal{O}(\Delta t^{4}),\tag{41}$$

Correction terms: complicated when the ODE has many terms...

#### 3.5 Extension to quadratic damping

$$mu'' + \beta |u'|u' + s(u) = F(t). \tag{42}$$

Centered scheme: |u'|u' gives rise to a nonlinearity.

Linearization trick: use a geometric mean,

$$[|u'|u']^n \approx |[u']^{n-\frac{1}{2}}|[u']^{n+\frac{1}{2}}$$
.

Scheme:

$$[mD_tD_tu]^n + \beta |[D_tu]^{n-\frac{1}{2}}|[D_tu]^{n+\frac{1}{2}} + s(u^n) = F^n.$$
(43)

### 3.6 The truncation error for quadratic damping (1)

Definition of  $\mathbb{R}^n$ :

$$[mD_tD_tu_e]^n + \beta |[D_tu_e]^{n-\frac{1}{2}}|[D_tu_e]^{n+\frac{1}{2}} + s(u_e^n) - F^n = R^n.$$
 (44)

Truncation error of the geometric mean, see (14),

$$|[D_t u_{\mathbf{e}}]^{n-\frac{1}{2}}|[D_t u_{\mathbf{e}}]^{n+\frac{1}{2}} = [|D_t u_{\mathbf{e}}|D_t u_{\mathbf{e}}]^n - \frac{1}{4}u'(t_n)^2 \Delta t^2 + \frac{1}{4}u(t_n)u''(t_n)\Delta t^2 + \mathcal{O}(\Delta t^4).$$

Using (5) for the  $D_t u_e$  factors results in

$$[|D_t u_e|D_t u_e]^n = |u'_e + \frac{1}{24} u'''_e(t_n) \Delta t^2 + \mathcal{O}(\Delta t^4) |(u'_e + \frac{1}{24} u'''_e(t_n) \Delta t^2 + \mathcal{O}(\Delta t^4))|$$

### 3.7 The truncation error for quadratic damping (2)

For simplicity, remove the absolute value. Computing the product leads to

$$[D_t u_e D_t u_e]^n = (u'_e(t_n))^2 + \frac{1}{12} u_e(t_n) u'''_e(t_n) \Delta t^2 + \mathcal{O}(\Delta t^4).$$

With

$$m[D_t D_t u_e]^n = m u_e''(t_n) + \frac{m}{12} u_e''''(t_n) \Delta t^2 + \mathcal{O}(\Delta t^4),$$

and using  $mu'' + \beta(u')^2 + s(u) = F$ , we end up with

$$R^{n} = (\frac{m}{12}u_{\rm e}^{\prime\prime\prime\prime}(t_{n}) + \frac{\beta}{12}u_{\rm e}(t_{n})u_{\rm e}^{\prime\prime\prime}(t_{n}))\Delta t^{2} + \mathcal{O}(\Delta t^{4}).$$

Second-order accuracy! But why?

- Difference approximation with truncation error  $\mathcal{O}(\Delta t^2)$
- Geometric mean approximation with truncation error  $\mathcal{O}(\Delta t^2)$

#### 3.8 The general model formulated as first-order ODEs

$$mu'' + \beta |u'|u' + s(u) = F(t). \tag{45}$$

Rewritten as first-order system:

$$u' = v, (46)$$

$$v' = \frac{1}{m} (F(t) - \beta |v|v - s(u)) . \tag{47}$$

To solution methods:

- Forward-backward scheme
- Centered scheme on a staggered mesh

#### 3.9 The forward-backward scheme

Forward step for u, backward step for v:

$$[D_t^+ u = v]^n, (48)$$

$$[D_t^- v = \frac{1}{m} (F(t) - \beta |v|v - s(u))]^{n+1}.$$
(49)

- Note:
  - step u forward with known v in (48)
  - step v forward with known u in (49)

- Problem: |v|v gives nonlinearity  $|v^{n+1}|v^{n+1}$ .
- Remedy: linearized as  $|v^n|v^{n+1}$

$$[D_t^+ u = v]^n, (50)$$

$$[D_t^- v]^{n+1} = \frac{1}{m} (F(t_{n+1}) - \beta |v^n| v^{n+1} - s(u^{n+1})).$$
 (51)

### 3.10 Truncation error analysis

- Aim (as always): turn difference operators into derivatives + truncation error terms
- One-sided forward/backward differences: truncation error  $\mathcal{O}(\Delta t)$
- Linearization of  $|v^{n+1}|v^{n+1}$  to  $|v^n|v^{n+1}$ : truncation error  $\mathcal{O}(\Delta t)$
- All errors are  $\mathcal{O}(\Delta t)$
- First-order scheme? No!
- "Symmetric" use of the  $\mathcal{O}(\Delta t)$  building blocks yields in fact a  $\mathcal{O}(\Delta t^2)$  scheme (!)
- Why? See next slide...

### A centered scheme on a staggered mesh. Staggered mesh:

- u is computed at mesh points  $t_n$
- v is computed at points  $t_{n+\frac{1}{2}}$

Centered differences in (46)-(46):

$$[D_t u = v]^{n - \frac{1}{2}},\tag{52}$$

$$[D_t v = \frac{1}{m} (F(t) - \beta |v|v - s(u))]^n.$$
 (53)

- Problem:  $|v^n|v^n$ , because  $v^n$  is not computed directly
- Remedy: Geometric mean,

$$|v^n|v^n \approx |v^{n-\frac{1}{2}}|v^{n+\frac{1}{2}}$$
.

Resulting scheme:

$$[D_t u]^{n-\frac{1}{2}} = v^{n-\frac{1}{2}}, (54)$$

$$[D_t v]^n = \frac{1}{m} (F(t_n) - \beta | v^{n - \frac{1}{2}} | v^{n + \frac{1}{2}} - s(u^n)).$$
 (55)

A centered scheme on a staggered mesh: truncation error analysis.

The truncation error in each equation fulfills

$$\begin{split} [D_t u_{\mathbf{e}}]^{n-\frac{1}{2}} &= v_{\mathbf{e}}(t_{n-\frac{1}{2}}) + R_u^{n-\frac{1}{2}}, \\ [D_t v_{\mathbf{e}}]^n &= \frac{1}{m} (F(t_n) - \beta |v_{\mathbf{e}}(t_{n-\frac{1}{2}})| v_{\mathbf{e}}(t_{n+\frac{1}{2}}) - s(u^n)) + R_v^n. \end{split}$$

Using (5) for derivatives and (14) for the geometric mean:

$$u_{\mathrm{e}}'(t_{n-\frac{1}{2}}) + \frac{1}{24} u_{\mathrm{e}}'''(t_{n-\frac{1}{2}}) \Delta t^2 + \mathcal{O}(\Delta t^4) = v_{\mathrm{e}}(t_{n-\frac{1}{2}}) + R_u^{n-\frac{1}{2}},$$

and

$$v'_{\rm e}(t_n) = \frac{1}{m} (F(t_n) - \beta |v_{\rm e}(t_n)| v_{\rm e}(t_n) + \mathcal{O}(\Delta t^2) - s(u^n)) + R_v^n.$$

Resulting truncation error is  $\mathcal{O}(\Delta t^2)$ :

$$R_u^{n-\frac{1}{2}} = \mathcal{O}(\Delta t^2), \quad R_v^n = \mathcal{O}(\Delta t^2).$$

#### Observation.

Comparing The schemes (54)-(55) and (50)-(51) are equivalent. Therefore, the forward/backward scheme with ad hoc linearization is also  $\mathcal{O}(\Delta t^2)$ !

### Index

```
correction terms, 9
decay ODE, 7
finite differences
   backward, 2
   centered, 4
   forward, 3

truncation error
   Backward Euler scheme, 2
   correction terms, 9
   Crank-Nicolson scheme, 4
   Forward Euler scheme, 3
   general, 1
   table of formulas, 5

verification, 12
```