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a) pde: $\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2}$ with $0 < x < 1$, $t \geq 0$
 $u(0,t) = u(1,t) = 0$
 and $u(x,0) = \sin(\pi x)$

Taylor:

$$u(x_{i+1}, t) = u(x_i + h, t) = u(x_i, t) + hu_x(x_i, t) + \frac{h^2}{2} u_{xx}(x_i, t) + \frac{h^3}{3!} u_{xxx}(x_i, t) + \frac{h^4}{4!} u_{xxxx}(\xi_i^+, t)$$

$$u(x_{i-1}, t) = u(x_i - h, t) = u(x_i, t) - hu_x(x_i, t) + \frac{h^2}{2} u_{xx}(x_i, t) - \frac{h^3}{3!} u_{xxx}(x_i, t) + \frac{h^4}{4!} u_{xxxx}(\xi_i^-, t)$$

for some $\xi_i^+ \in [x_i, x_{i+1}]$ and $\xi_i^- \in [x_{i-1}, x_i]$

$$u(x_{i+1}, t) - 2u(x_i, t) + u(x_{i-1}, t)$$

$$= h^2 u_{xx}(x_i, t) + \frac{h^4}{12} u_{xxxx}(\eta_i, t) \quad \text{for some } \eta_i \in [x_{i-1}, x_{i+1}]$$

$$u_{xx}(x_i, t) = \frac{u(x_{i+1}, t) - 2u(x_i, t) + u(x_{i-1}, t))}{h^2} - \frac{h^2}{12} u_{xxxx}(\eta_i, t)$$

local truncation error

→ order of accuracy is $O(h^2)$ good for ODE
 non PIV ?

$$b) J = \frac{1}{h^2} \begin{pmatrix} -2 & 1 & 0 & \dots & 0 \\ 1 & -2 & 1 & \dots & 0 \\ 0 & \dots & \dots & \dots & 0 \\ \vdots & \dots & \dots & \dots & \vdots \\ 0 & \dots & 0 & 1 & -2 \end{pmatrix}$$

the eigenvalues are centered around $-\frac{2}{h^2}$
 ~~$\sqrt{2} \pm \sqrt{2} \sqrt{1/h^2}$~~

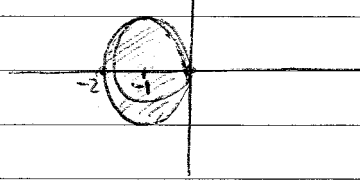
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the eigenvalues are in $R = \{z \in \mathbb{C} \mid |z + \frac{2}{h^2}| \leq \frac{2}{h^2}\}$

and because J is symmetric and real, the eigenvalues are in $[-\frac{4}{h^2}, 0]$

c) Euler: $w_{i+1} = w_i + hf(t_i, w_i) \rightarrow w_{i+1} = w_i + \lambda w_i$
 $= (1 + \lambda h) w_i$
 $R = \{\lambda h \in \mathbb{C} \mid |1 + \lambda h| < 1\}$ (region of abs. stability)

with test equation.



Backward Euler: $w_{i+1} = w_i + hf(t_{i+1}, w_{i+1})$

$$w_{i+1} = w_i + h\lambda w_{i+1}$$

$$w_{i+1}(1 - h\lambda) = w_i$$

$$w_{i+1} = \frac{1}{1 - h\lambda} w_i$$

$$R = \left\{ h\lambda \in \mathbb{C} \mid \underbrace{\left| \frac{1}{1 - h\lambda} \right|}_{< 1} < 1 \right\}$$

$$1 = |1| < |1 - h\lambda|$$

for $h\lambda \in [0, 2]$, we have $|1 - h\lambda| \leq 1$

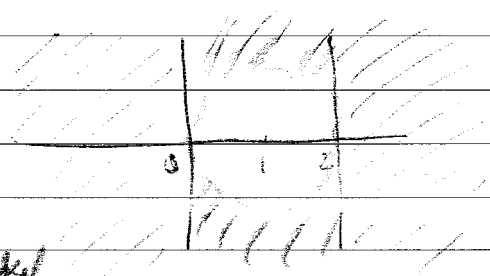
$$|1 - h\lambda| > 1 :$$

$$|1 - 2| \leq 1$$

$$1 - 2 = e^{i\theta}$$

$$z = 1 - e^{i\theta}$$

circle not closed



Trapezoidal method: $w_{i+1} = w_i + \frac{h}{2} (f(t_i, w_i) + f(t_{i+1}, w_{i+1}))$

$$w_{i+1} = w_i + \frac{h}{2} (\lambda w_i + \lambda w_{i+1})$$

$$w_{i+1} \left(1 - \frac{h\lambda}{2}\right) = w_i \left(1 + \frac{h\lambda}{2}\right)$$

$$w_{i+1} = \frac{1 + \frac{h\lambda}{2}}{1 - \frac{h\lambda}{2}} w_i$$

g

$$R = \left\{ h\lambda \in \mathbb{C} \mid \underbrace{\left| \frac{1 + \frac{h\lambda}{2}}{1 - \frac{h\lambda}{2}} \right|}_{< 1} < 1 \right\}$$

$$\left| 1 + \frac{h\lambda}{2} \right| < \left| 1 - \frac{h\lambda}{2} \right|$$

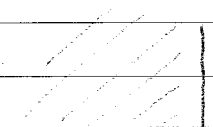
~~$$2 + h\lambda < 2 - h\lambda$$~~

$$|2 + h\lambda| < |2 - h\lambda|$$

$$h\lambda = 0 \rightarrow |2 + h\lambda| = |2 - h\lambda|$$

$$h\lambda > 0 \rightarrow |2 + h\lambda| > |2 - h\lambda|$$

$$h\lambda < 0 \rightarrow |2 + h\lambda| < |2 - h\lambda|$$



1 d) the eigenvalues must be in $(-2, 0)$

$$\lambda \in \left[-\frac{8k}{h^2}, 0 \right] \quad h = \frac{1}{10} = 0,1$$

$$-\frac{8k}{h^2} > -2 \quad \rightarrow \quad \frac{8k}{h} < 2$$

$$8k < 20$$

10 e) no, backward Euler and Trapezoidal method are A-stable (see c)

2 a) A square strictly diagonally dominant matrix \rightarrow Gaussian elimination possible (without row interchanges)

g) for all multiplier m_{ij} ~~absolute~~ : $|m_{ij}| < 1$

\rightarrow unique solution $x \rightarrow A$ non-singular $\} \text{max. unklarheit die hoch}$

b) first A is split up in $A = D - L - U$, where D is ~~str~~ a diagonal matrix, $-L$ is a strictly lower triangular matrix and $-U$ is a strictly upper triangular matrix

$$Ax = b \rightarrow (D - L - U)x = b$$

$$Dx = (L + U)x + b$$

$$x = D^{-1}(L + U)x + D^{-1}b$$

we start with an initial approximation $x^{(0)}$ to x and iterate : $x^{(k)} = T_j x^{(k-1)} + c_j$, where $T_j = D^{-1}(L + U)$ and $c_j = D^{-1}b$

2 (b) the iteration is continued until convergence or until the maximum number of iterations is exceeded.

5 c $\rho(A) \leq \|A\|$ for any natural norm $\|\cdot\|$ analogous!
 $\Rightarrow \rho(A) \leq \|A\|_\infty$

$$\|A\|_\infty = \max_{\|x\|_\infty=1} \|Ax\|_\infty \geq \max_i \max_{\|x\|_\infty=1} \|\lambda_i x\|_\infty = \max_i |\lambda_i| \|x\|_\infty = \max_i |\lambda_i| = \rho(A)$$

0 d Jacobi method converges if $\rho(T_j) < 1$

$\rho(T_j) \leq \|A\|_\infty < 1 \rightarrow$ convergence
reference

2
$$\left. \begin{aligned} x^{(k)} &= T_j x^{(k-1)} + c_j \\ x &= T_j x + c_j \end{aligned} \right\} x - x^{(k)} = T_j (x - x^{(k-1)})$$

and $x^{(k)} - x^{(k-1)} = T_j (x^{(k-1)} - x^{(k-2)})$

$\|T_j\| < 1$

$$\|x - x^{(k)}\| \leq \|T_j\| \|x - x^{(k-1)}\|$$

$$\|x^{(k)} - x^{(k-1)}\| \leq \|T_j\| \|x^{(k-1)} - x^{(k-2)}\|$$

$$\frac{\|x^{(k)} - x^{(k-1)}\|}{\|x^{(k-1)} - x^{(k-2)}\|} \leq \|T_j\|$$

3

a power method: start with initial vector x with

$$\|x\|_\infty = 1$$

iterate: $y = Ax$

$$\mu = y_p \quad \text{where } |y_p| = \|y\|_\infty$$

$$x = y / y_p$$

g

continue until $\|x - y/y_p\|$ is small enough
or until maximum number of iterations is exceeded.

μ gives approximation to λ , x gives approximation to $v^{(1)}$ (eigenvector associated with λ)

b

4

$$\frac{\lambda_2}{\lambda_1} < 1 \rightarrow \left(\frac{\lambda_2}{\lambda_1}\right)^2 < \frac{\lambda_2}{\lambda_1} \rightarrow \text{faster convergence}$$

2x 20 and

$$c \quad w = \frac{1}{\lambda_1 v_i^{(1)}} (a_{i1}, a_{i2}, \dots)$$

2

Householder transformations can be used to bring symmetric matrices to tridiagonal form, such that it can be used to compute the eigenvalues with the QR-algorithm.

a 1. $A - sI = QR$, where Q is orthogonal and R is an upper triangular matrix

$$2. A = RQ + sI$$

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this is iterated until convergence

s is usually chosen close to an eigenvalue of A in the z -method with shift $h = \lambda_{\text{oe}}$

3 (a) to converge when λ_2 is close to λ_1 ($\frac{\lambda_2}{\lambda_1} \approx 1$)

in the standard QR method $s=0$

4 a newton: $x^{(i+1)} = x^{(i)} - \frac{f(x^{(i)})}{f'(x^{(i)})}$

just right of the maximum near 2.5 $f'(x)$ is negative and very close to 0

W $\rightarrow -\frac{f(x)}{f'(x)}$ is very large

\rightarrow the next x lies far from zero

b $f(x) = \frac{\sin(\pi x)}{x}$ $f'(x) = \frac{\pi \cos(\pi x)}{x^2} - \frac{\sin(\pi x)}{x^2}$ 9

$$f'(x) = \frac{\pi x \cos(\pi x) - \sin(\pi x)}{x^2}$$

$$g(\lambda, x) = p(x) + (1-\lambda)f(x_0)$$

W $Dg(\lambda, x) = g_\lambda(\lambda, x) + g_x(\lambda, x) \cdot x'(\lambda) = 0$

$$g_\lambda = -f(x_0)$$

$$g_x = f'(x)$$

$$\rightarrow x'(\lambda) = -\frac{g_\lambda(\lambda, x)}{g_x(\lambda, x)} = \frac{f(x_0) x^2}{\pi x \cos(\pi x) - \sin(\pi x)}$$

$$\text{for } \lambda \in [0, 1], x(0) = x_0$$

c we get an approximation for x

0