

### Solutions

1 a) Denote  $r = Ax - b$ ;  $x = \arg \min \|r\|_2 = \arg \min r^T r = \arg \min (x^T A^T Ax - 2(A^T b)^T x + b^T b)$

Differentiating with respect to  $x$  and setting the result to zero, we get  $2A^T Ax - 2A^T b = 0$ , or

$$A^T Ax = A^T b \quad (2)$$

$A$  is of full column rank  $\Rightarrow A^T A$  is nonsingular  $\Rightarrow$  (2) has a unique solution.

$A^T A$  is symmetric and positive definite  $\Rightarrow$  we can compute its Cholesky factorization

$$A^T A = LL^T,$$

where  $L$  - lower triangular. This allows us to reduce (2) to solving the triangular systems

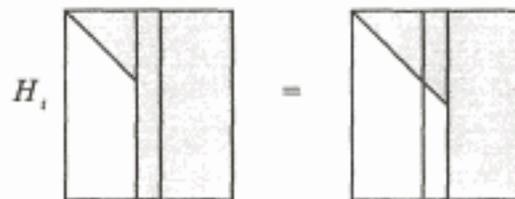
$$Ly = A^T b, \quad L^T x = y.$$

b) First we want to decompose  $A$  into the product  $Q \begin{bmatrix} R \\ 0 \end{bmatrix}$ ,  $Q: m \times m$  orthogonal,  $R: n \times n$  upper-triangular; since  $A$  is of full column rank,  $R$  is nonsingular.

To reduce  $A$  to the upper-triangular form, we successively apply Householder transformations

$$H_n \dots H_1 A = \begin{bmatrix} R \\ 0 \end{bmatrix}, \quad \text{where } H_i = I - 2 \frac{v_i v_i^T}{v_i^T v_i} \Rightarrow \text{orthogonal and symmetric.}$$

$H_i$  reflects a vector against the hyperplane  $v_i^\perp$ . We choose  $v_i$  so that  $H_i$  zeros out the subdiagonal part of the  $i$ th column  $\tilde{a}_i$  of the current state of  $A$  -  $(H_{i-1} \dots H_1 A)$



If  $\tilde{a}'_i = \{\tilde{a}_i \text{ with the upper } i-1 \text{ entries set to } 0\}$ , then  $v_i = \tilde{a}'_i \pm \|\tilde{a}'_i\|_2 e_i$ , where the sign is chosen so as to avoid cancellation.

Now, using this decomposition  $Q^T A = \begin{bmatrix} R \\ 0 \end{bmatrix}$ ,  $Q^T = H_n \dots H_1$ ,

$$x = \arg \min \|Ax - b\|_2 = \arg \min \|Q^T Ax - Q^T b\|_2 = \arg \min \left\| \begin{bmatrix} R \\ 0 \end{bmatrix} x - Q^T b \right\|_2;$$

$$\text{let } Q^T b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}, \quad b_1: n \times 1, \quad b_2: (m-n) \times 1;$$

$$x = \arg \min \left\| \begin{bmatrix} R \\ 0 \end{bmatrix} x - \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \right\|_2^2 = \arg \min (\|Rx - b_1\|_2^2 + \|b_2\|_2^2) = \arg \min \|Rx - b_1\|_2 = R^{-1} b_1$$

c)

	Computations	Accuracy
Normal equations	$A^T A - \approx mn^2$ flops; Cholesky factorization - $\approx \frac{n^3}{3}$ flops Triangular systems - $O(n^2)$ $\approx mn^2 + \frac{n^3}{3}$ flops	relative error in $x$ is proportional to $(\text{cond}(A))^2$
Householder transformations	$\approx 2mn^2 + \frac{2}{3}n^3$ flops	relative error in $x$ is proportional to $\text{cond}(A) + \ r\ _2 (\text{cond}(A))^2$

For nearly square problems,  $m \approx n$ , the two methods require about the same amounts of work, but for  $m \gg n$  normal equations method is about 2 times cheaper than Householder method. On the other hand, the Householder method is more accurate.

2 a) Via elementary symbolic calculations we obtain  $y(t) = e^{\lambda t}$ . As  $t \rightarrow \infty$ ,  $y(t) \rightarrow +0$ .

$$\text{b) } y_{k+1} \left(1 - \frac{\lambda h}{2}\right) = y_k \left(1 + \frac{\lambda h}{2}\right), \quad y_{k+1} = \frac{1 + \lambda h / 2}{1 - \lambda h / 2} y_k, \quad y_k = \left(\frac{1 + \lambda h / 2}{1 - \lambda h / 2}\right)^k y_0$$

$$y_k \rightarrow 0 \Leftrightarrow \left| \frac{1 + \lambda h / 2}{1 - \lambda h / 2} \right| < 1, \quad (2 + \lambda h)^2 < (2 - \lambda h)^2, \quad 4\lambda h < -4\lambda h, \quad \lambda h < 0,$$

and since we assume  $h > 0$ , this is true for all negative  $\lambda$ .

$$\text{c) We have } y_{k+1} = y_k + \frac{f(t_k, y_k) + f(t_{k+1}, y_{k+1})}{2} h \quad (1)$$

For the true solution  $y(t)$  we can write

$$y(t_{k+1}) = y(t_k) + \frac{y'(t_k) + y'(t_{k+1})}{2} h + \xi, \quad (2)$$

where  $\xi$  is some unknown term. From Taylor expansion we have two formulas:

$$\frac{f(x+h) + f(x-h)}{2} = f(x) + \frac{\theta_0 h^2}{2}, \quad |\theta_0| \leq \|f''\|_c;$$

$$\frac{f(x+h) - f(x-h)}{2h} = f'(x) + \frac{\theta_1 h^2}{6}, \quad |\theta_1| \leq \|f^{(3)}\|_c.$$

Regrouping (2) and applying these formulas, we get:

$$\frac{\xi}{h} = \frac{y(t_{k+1}) - y(t_k)}{h} - \frac{y'(t_{k+1}) + y'(t_k)}{2} = y'(t_{k+1/2}) + \frac{\eta h^2}{6} - y'(t_{k+1/2}) - \frac{\eta_1 h^2}{2} = \left(\frac{\eta}{6} - \frac{\eta_1}{2}\right) h^2, \quad \text{where}$$

$$|\eta| \leq \|y^{(3)}\|_c, \quad |\eta_1| \leq \|y''\|_c. \quad \text{Hence } |\xi| \leq \frac{1}{2} (\|y''\|_c + \|y^{(3)}\|_c) h^3. \quad (3)$$

Now denoting  $\delta_k = y_k - y(t_k)$  and subtracting (1)-(2), we get

$$\delta_{k+1} = \delta_k + \frac{f(t_k, y_k) - f(t_k, y(t_k))}{2} h + \frac{f(t_{k+1}, y_{k+1}) - f(t_{k+1}, y(t_{k+1}))}{2} h - \xi.$$

Then using Lipschitz-continuity of  $f$ ,

$$|\delta_{k+1}| \leq |\delta_k| + \frac{Lh}{2} |\delta_k| + \frac{Lh}{2} |\delta_{k+1}| + |\xi| \quad \left(1 - \frac{Lh}{2}\right) |\delta_{k+1}| \leq \left(1 + \frac{Lh}{2}\right) |\delta_k| + |\xi|,$$

$$|\delta_{k+1}| \leq \left| \frac{1+Lh/2}{1-Lh/2} \right| |\delta_k| + \frac{|\xi|}{1-Lh/2} \leq \left| \frac{1+Lh/2}{1-Lh/2} \right| |\delta_k| + 2|\xi| \quad \text{for } h \leq \frac{1}{L}.$$

To simplify this, consider the general situation:  $x_{k+1} \leq ax_k + b$ . Applying this inequality recursively,

we get  $x_k \leq a^k x_0 + b(a^{k-1} + a^{k-2} + \dots + 1) = a^k x_0 + b \frac{a^k - 1}{a - 1}$ . The first term in our case disappears,

because  $y(0) = y_0 = 1$ ; therefore  $|\delta_k| \leq 2|\xi| \frac{\left| \frac{1+Lh/2}{1-Lh/2} \right|^k - 1}{\left| \frac{1+Lh/2}{1-Lh/2} \right| - 1}$ .

Now since  $k = t_k / h$ ,  $\left| \frac{1+Lh/2}{1-Lh/2} \right|^k \xrightarrow{h \rightarrow 0} \frac{e^{\frac{L t_k}{2}}}{e^{-\frac{L t_k}{2}}} = e^{L t_k}$ , and hence

$$|\delta_k| \leq 2|\xi| \frac{e^{L t_k} + \varepsilon(h) - 1}{Lh} (1 - Lh/2) \leq \frac{2|\xi|}{h} \frac{e^{L t_k} + \varepsilon(h) - 1}{L},$$

where  $\frac{2|\xi|}{h} \leq (\|y''\|_c + \|y^{(3)}\|_c) h^2$ , and  $\varepsilon(h) \rightarrow 0$  as  $h \rightarrow 0$ . Thus  $|\delta_k| \leq Ch^2$  for sufficiently small  $h$ .

3 a) Let  $I(x) = f(a) + (x-a) \frac{f(b)-f(a)}{b-a}$ , then

$$I(f) = \int_a^b I(x) dx = f(a)(b-a) + \frac{(x-a)^2}{2} \Big|_a^b \frac{f(b)-f(a)}{b-a} = (b-a)f(a) + \frac{f(b)-f(a)}{2} = (b-a) \frac{f(a)+f(b)}{2}$$

b) Let  $I_0 = \int_a^b f(x) dx$ ,  $\Delta = b-a$ .

$$f(a+t) = f(a) + f'(a)t + f''(\xi) \frac{t^2}{2}, \quad I_0 = f(a)\Delta + f'(a) \frac{\Delta^2}{2} + A \frac{\Delta^3}{6}, \quad |A| \leq \|f''\|_c$$

$$f(b-t) = f(b) - f'(b)t + f''(\xi_1) \frac{t^2}{2}, \quad I_0 = f(b)\Delta - f'(b) \frac{\Delta^2}{2} + B \frac{\Delta^3}{6}, \quad |B| \leq \|f''\|_c.$$

Averaging the two expressions for  $I_0$ , we obtain

$$I_0 = \frac{f(a)+f(b)}{2} \Delta + \frac{A+B}{2} \frac{\Delta^3}{6} = I + C \frac{\Delta^3}{6}, \quad \text{where } C = \frac{A+B}{2}. \quad \text{This implies } |I_0 - I| \leq \|f''\|_c \frac{\Delta^3}{6}.$$

c) Let  $I_0^{(i)}$ ,  $I^{(i)}$  be the true integral and our estimate correspondingly for the  $i$ th subinterval. Then

$$I_0 = \sum_{i=1}^n I_0^{(i)}, \quad I = \sum_{i=1}^n I^{(i)}.$$

$$|I_0 - I| \leq \sum_{i=1}^n |I_0^{(i)} - I^{(i)}| \leq \sum_{i=1}^n |C_i| \frac{\Delta_i^3}{6} \leq \sum_{i=1}^n \|f''\|_c \frac{1}{6} \left( \frac{\Delta}{n} \right)^3 = \|f''\|_c \frac{\Delta^3}{6n^2} = \frac{\|f''\|_c (b-a)^3}{6n^2}.$$