

## Computational and Applied Mathematics

*Solve every problem.*

**Problem 1.** Let  $f \in C^{k+1}[-1, 1]$  and  $[-1, 1]$  be partitioned into subintervals  $I_j = [(j-1)h, jh]$  of width  $h$ . Assume  $p$  is a polynomial of degree  $k$  which approximates  $f$  in  $I_j$  with

$$\max_{x \in I_j} |p_j(x) - f(x)| \leq C_0 h^{k+1},$$

where  $C_0$  is a constant independent of  $j$ . Show that there exists another constant  $C$ , independent of  $j$ , such that

$$\max_{x \in I_{j \pm 1}} |p_j(x) - f(x)| \leq C h^{k+1}.$$

(as long as  $I_{j \pm 1} \subset [-1, 1]$ , of course).

**Solution:** Pick points  $0 \leq x_0 < x_1 < \dots < x_k \leq 1$ , and let

$$L_i(x) = \prod_{l \neq i} \frac{x - x_l}{x_i - x_l}$$

be the  $i$ -th Lagrange polynomial. Let

$$\Lambda = \sum_{i=0}^k \max_{x \in [-1, 2]} |L_i(x)|.$$

On  $I_j$  we use rescaled versions with  $x_{ji} = (j-1)h + hx_i$ , and

$$L_{ji}(x) = \prod_{l \neq i} \frac{x - x_{jl}}{x_{ji} - x_{jl}}.$$

Note that  $\Lambda$  is unchanged with

$$\Lambda = \sum_i \max_{x \in I_{j \pm 1}} |L_{ji}(x)|.$$

Let  $f_j$  be the interpolating polynomial on  $I_j$

$$f_j(x) = \sum_{i=0}^k f(x_{ji}) L_{ji}(x),$$

and note that also

$$p_j(x) = \sum_i p(x_{ji}) L_{ji}(x).$$

Then for  $x \in I_{j \pm 1}$ ,

$$\begin{aligned} |p_j(x) - f(x)| &\leq |p_j(x) - f_j(x)| + |f_j(x) - f(x)| \\ &= \left| \sum_{i=0}^k (f(x_{ji}) - p_j(x_{ji})) L_{ji}(x) \right| + |R_k f(x)| \\ &\leq \max_{x \in I_j} |f(x) - p_j(x)| \Lambda + \frac{\|f^{(k+1)}\|}{(k+1)!} \max_{x \in I_{j \pm 1}} \left| \prod_{i=0}^k (x - x_{ji}) \right| \\ &\leq C_0 \Lambda h^{k+1} + \frac{\|f^{(k+1)}\|}{(k+1)!} \max_x |(2h)^{k+1}| \\ &= C h^{k+1}, \end{aligned}$$

$$C = C_0\Lambda + \frac{2^{k+1}\|f^{(k+1)}\|}{(k+1)!}.$$

**Problem 2.** Consider the iteration

$$x_{n+1} = x_n - \left( \frac{x_n - x_0}{f(x_n) - f(x_0)} \right) f(x_n)$$

for finding the roots of a two times continuous differentiable function  $f(x)$ . Assuming the method converges to a simple root  $x^*$ , what is the rate of convergence? Justify your answer.

**Solution:** The iteration may be rewritten as

$$x_{n+1} = \frac{[x_n f(x_n) - x_n f(x_0)] - [x_n f(x_n) - x_0 f(x_n)]}{f(x_n) - f(x_0)} = \frac{x_0 f(x_n) - x_n f(x_0)}{f(x_n) - f(x_0)}.$$

Thus

$$x_{n+1} - x^* = \frac{x_0 f(x_n) - x_n f(x_0)}{f(x_n) - f(x_0)} - x^* = \frac{(x_0 - x^*)f(x_n) - (x_n - x^*)f(x_0)}{f(x_n) - f(x_0)}.$$

Taylor's Theorem asserts that there is  $\xi_n$  between  $x_n$  and  $x^*$  such that

$$0 = f(x^*) = f(x_n) + f'(\xi_n)(x^* - x_n) \Rightarrow f(x_n) = f'(\xi_n)(x_n - x^*).$$

This implies

$$x_{n+1} - x^* = \frac{(x_0 - x^*)f'(\xi_n) - f(x_0)}{f(x_n) - f(x_0)}(x_n - x^*).$$

Evaluating the limit as  $n \rightarrow \infty$ ,  $\xi_n \rightarrow x^*$  and

$$\lim_{n \rightarrow \infty} \left| \frac{x_{n+1} - x^*}{x_n - x^*} \right| = \lim_{n \rightarrow \infty} \left| \frac{(x_0 - x^*)f'(\xi_n) - f(x_0)}{f(x_n) - f(x_0)} \right| = \left| \frac{(x_0 - x^*) \lim_{n \rightarrow \infty} f'(\xi_n) - f(x_0)}{0 - f(x_0)} \right|.$$

Applying Taylor's expression one more time, we know there is  $\eta$  between  $x^*$  and  $x_0$  such that

$$f(x_0) = f(x^*) + f'(x^*)(x_0 - x^*) + \frac{f''(\eta)}{2}(x_0 - x^*)^2,$$

So

$$f'(x^*)(x_0 - x^*) - f(x_0) = -\frac{f''(\eta)}{2}(x_0 - x^*)^2.$$

Therefore

$$\lim_{n \rightarrow \infty} \left| \frac{x_{n+1} - x^*}{x_n - x^*} \right| = \left| \frac{f''(\eta)}{2f(x_0)} \right| (x_0 - x^*)^2.$$

Note the right hand side is dependent only upon  $x^*$  and  $x_0$ . Since we know  $x_n \rightarrow x^*$ , this shows the rate of convergence is linear.

**Problem 3.** Suppose  $\mathbf{A}$  is an  $m \times m$  matrix with a complete set of orthonormal eigenvectors  $\mathbf{q}_1, \dots, \mathbf{q}_m$  and corresponding eigenvalues  $\lambda_1, \dots, \lambda_m$ . Assume that  $|\lambda_1| > |\lambda_2| > |\lambda_3|$  and  $\lambda_j \geq \lambda_{j+1}$  for  $j = 3, \dots, m$ . Consider the power method  $\mathbf{v}^{(k)} = \mathbf{A}\mathbf{v}^{(k-1)}/\lambda_1$ , with  $\mathbf{v}^{(0)} = \alpha_1 \mathbf{q}_1 + \dots + \alpha_m \mathbf{q}_m$  where  $\alpha_1$  and  $\alpha_2$  are both nonzero. Show that the sequence  $\{\mathbf{v}^{(k)}\}_{k=0}^\infty$  converges linearly to  $\alpha_1 \mathbf{q}_1$  with asymptotic constant  $C = |\lambda_2/\lambda_1|$ .

**Solution:** Matrix  $\mathbf{A}$  has following eigen-decomposition

$$\mathbf{A} = [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m] \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_m \end{bmatrix} [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m]^{-1},$$

thus

$$\mathbf{A}^k = [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m] \begin{bmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \ddots & \\ & & & \lambda_m^k \end{bmatrix} [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m]^{-1}.$$

The power method reduces to

$$\begin{aligned} \mathbf{v}^{(k)} &= \mathbf{A}^k \frac{\mathbf{v}^{(0)}}{\lambda_1^k} \\ &= [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m] \begin{bmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \ddots & \\ & & & \lambda_m^k \end{bmatrix} [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m]^{-1} [\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_m] \begin{bmatrix} \frac{\alpha_1}{\lambda_1^k} \\ \frac{\alpha_2}{\lambda_1^k} \\ \vdots \\ \frac{\alpha_m}{\lambda_1^k} \end{bmatrix} \\ &= \alpha_1 \mathbf{q}_1 + \sum_{j=2}^m \left( \frac{\lambda_j}{\lambda_1} \right)^k \alpha_j \mathbf{q}_j, \end{aligned}$$

from this we deduce  $\mathbf{v}^{(k)} \rightarrow \alpha_1 \mathbf{q}_1$  as  $k \rightarrow \infty$ , since  $|\lambda_j/\lambda_1| < 1$  for  $j = 2, \dots, m$ .

To show the convergence is linear with asymptotic constant  $C = |\lambda_2/\lambda_1|$  we need to verify the limit

$$\lim_{k \rightarrow \infty} \frac{\|\mathbf{e}^{(k+1)}\|}{\|\mathbf{e}^{(k)}\|} = \lim_{k \rightarrow \infty} \frac{\|\mathbf{v}^{(k+1)} - \alpha_1 \mathbf{q}_1\|}{\|\mathbf{v}^{(k)} - \alpha_1 \mathbf{q}_1\|} = \left| \frac{\lambda_2}{\lambda_1} \right| \quad (\text{here } \|\cdot\| \text{ denotes the } L_2\text{-norm}).$$

Note that  $\mathbf{e}^{(k)} = \sum_{j=2}^m \left( \frac{\lambda_j}{\lambda_1} \right)^k \alpha_j \mathbf{q}_j$ , using the orthonormality of the eigenvectors we have

$$\|\mathbf{e}^{(k)}\|^2 = \sum_{j=2}^m \left| \frac{\lambda_j}{\lambda_1} \right|^{2k} |\alpha_j|^2 = \left| \frac{\lambda_2}{\lambda_1} \right|^{2k} \left( |\alpha_2|^2 + \sum_{j=3}^m \left| \frac{\lambda_j}{\lambda_2} \right|^{2k} |\alpha_j|^2 \right),$$

similarly

$$\|\mathbf{e}^{(k+1)}\|^2 = \left| \frac{\lambda_2}{\lambda_1} \right|^{2(k+1)} \left( |\alpha_2|^2 + \sum_{j=3}^m \left| \frac{\lambda_j}{\lambda_2} \right|^{2(k+1)} |\alpha_j|^2 \right).$$

Thus

$$\begin{aligned} \lim_{k \rightarrow \infty} \frac{\|\mathbf{e}^{(k+1)}\|}{\|\mathbf{e}^{(k)}\|} &= \lim_{k \rightarrow \infty} \left( \frac{\left| \frac{\lambda_2}{\lambda_1} \right|^{2(k+1)} \left( |\alpha_2|^2 + \sum_{j=3}^m \left| \frac{\lambda_j}{\lambda_2} \right|^{2(k+1)} |\alpha_j|^2 \right)}{\left| \frac{\lambda_2}{\lambda_1} \right|^{2k} \left( |\alpha_2|^2 + \sum_{j=3}^m \left| \frac{\lambda_j}{\lambda_2} \right|^{2k} |\alpha_j|^2 \right)} \right)^{\frac{1}{2}} \\ &= \left| \frac{\lambda_2}{\lambda_1} \right| \frac{|\alpha_2|}{|\alpha_2|} \quad (\text{we have used } |\lambda_2| > |\lambda_3| \geq |\lambda_j| \text{ for } j > 3) \\ &= \left| \frac{\lambda_2}{\lambda_1} \right| \quad (\text{since } \alpha_2 \neq 0). \end{aligned}$$

**Problem 4.** For the initial value problem  $y' = f(t, y)$ ,  $y(0) = y_0$  on the interval  $[0, T]$ , consider the implicit two-step method

$$\begin{aligned} y_{n+1} &= \frac{4}{3}y_n - \frac{1}{3}y_{n-1} + \frac{2h}{3}f(t_{n+1}, y_{n+1}), \\ y_1 &= y_0 + hf(t_1, y_0), \end{aligned}$$

where  $h$  is the step size and  $t_n = nh$ .

- (a) What is the order of the accuracy of the scheme?
- (b) Check the stability of the scheme by analyzing the stability polynomial?
- (c) Find the stability region of the scheme.

**Solution:** (a) Let  $y(t)$  be the exact solution, then the truncation error of form

$$h\tau_{n+1} := y(t_{n+1}) - \left( \frac{4}{3}y(t_n) - \frac{1}{3}y(t_{n-1}) + \frac{2h}{3}f(t_{n+1}, y(t_{n+1})) \right)$$

can be estimated by using Taylor expansion to every term involved:

$$\begin{aligned} y(t_{n+1}) &= y_n + hy'_n + \frac{1}{2}h^2y''_n + \frac{1}{6}h^3y'''_n + O(h^4), \\ -\frac{1}{3}y_{n-1} &= -\frac{1}{3}y_n + \frac{1}{3}hy'_n - \frac{1}{6}h^2y''_n + \frac{1}{18}h^3y'''_n + O(h^4), \\ \frac{2h}{3}f(t_{n+1}, y_{n+1}) &= \frac{2h}{3}y'_{n+1} = \frac{2}{3}hy'_n + \frac{2}{3}h^2y''_n + \frac{1}{3}h^3y'''_n + O(h^4). \end{aligned}$$

Hence

$$\begin{aligned} h\tau_{n+1} &= \left[ y_n + hy'_n + \frac{1}{2}h^2y''_n + \frac{1}{6}h^3y'''_n + O(h^4) \right] - \left[ y_n + hy'_n + \frac{1}{2}h^2y''_n + \frac{7}{18}h^3y'''_n + O(h^4) \right] \\ &= -\frac{2}{9}h^3y'''_n + O(h^4). \end{aligned}$$

The method has order of accuracy 2.

(b) Apply the method to the case  $f = 0$ , then

$$y_{n+1} - \frac{4}{3}y_n + \frac{1}{3}y_{n-1} = 0,$$

when for ansatz of form  $y_n = \gamma^n$  gives the stability polynomial

$$\gamma^2 - \frac{4}{3}\gamma + \frac{1}{3} = 0,$$

which has nonzero roots  $\gamma = 1, \frac{1}{3}$ . Since  $|\gamma| \leq 1$  and  $\gamma = 1$  as a single root, the method is stable.

(c) Consider the stiff problem  $y' = \lambda y$ . The method becomes

$$y_{n+1} = \frac{4}{3}y_n - \frac{1}{3}y_{n-1} + \frac{2}{3}h\lambda y_{n+1},$$

which has stability polynomial

$$(3 - 2h\lambda)\gamma^2 - 4\gamma + 1 = 0.$$

So the stability region is given by

$$\left| \frac{4 \pm \sqrt{16 - 4(3 - 2h\lambda)}}{2(3 - 2h\lambda)} \right| < 1,$$

i.e.,

$$R = \left\{ h\lambda \in \mathbb{C} : \left| \frac{2 \pm \sqrt{1 - 2h\lambda}}{3 - 2h\lambda} \right| < 1 \right\}.$$

**Problem 5.** Suppose the difference scheme  $u^{n+1} = Bu^n$  is stable, and  $C(\Delta t)$  is a bounded family of operators. Show that the scheme

$$u^{n+1} = (B + \Delta t C(\Delta t))u^n$$

is stable.

**Solution:** Suppose  $\|B^k\| \leq K_1$  for  $0 \leq k \leq n$  and  $\|C(\Delta t)\| \leq K_2$ . It suffices to show  $(B + \Delta t C(\Delta t))^n$  is bounded for  $n\Delta t \leq T$ . This will consist of  $2^n$  terms, of which  $\binom{n}{j}$  terms involve  $j$  factors  $\Delta t C$  interspersed in  $n - j$  factors  $B$ ; the latter can occur in at most  $j + 1$  sequences of consecutive factors, the norm of each sequence being bounded by  $K_1$ , and hence the norm of each such term by  $K_2^j K_1^{j+1}$ . Thus overall we obtain the bound

$$\begin{aligned} \|(B + \Delta t C(\Delta t))^n\| &\leq \sum_{j=0}^n \binom{n}{j} K_1^{j+1} (\Delta t K_2)^j \\ &= K_1 (1 + \Delta t K_1 K_2)^n \\ &\leq K_1 e^{n\Delta t K_1 K_2} \end{aligned}$$

which is bounded for  $n\Delta t \leq T$ .

**Problem 6.** Let  $A$  be an  $m \times m$  nonsingular matrix. Suppose  $\inf_{p_n \in P^n} \|p_n(A)\| = \|p^*(A)\| > 0$  where  $P^n$  denotes the set of all degree- $n$  monic polynomials:

$$P^n = \{p : p \text{ is a polynomial of degree } n, p(z) = z^n + \dots\}.$$

Prove that  $p^*$  is unique.

**Solution:** We prove by contradiction. Assuming there are two  $p_i$  for  $i = 1, 2$  as minimizers, then  $p = (p_1 + p_2)/2$  shares the same 2-norm,

$$\|p_1\| = \|p_2\| = \|p\| = \sigma_1,$$

where  $\sigma_1$  is the largest singular value. Let the SVD of  $p$  be

$$p(A) = U \text{diag}(\sigma_1, \dots, \sigma_n) V^*.$$

Suppose  $\sigma_1$  is  $J$ -fold, with left and right singular vectors  $u_1, \dots, u_J$  and  $v_1, \dots, v_J$ , respectively.

By convexity of the 2-norm, we have

$$\sigma_1 = \|p(A)v_j\| \leq \frac{1}{2} (\|p_1(A)v_j\| + \|p_2(A)v_j\|) \leq \sigma_1,$$

which implies that

$$\|p_1(A)v_j\| = \|p_2(A)v_j\| = \sigma_1$$

and

$$(p_1 - p_2)v_j = 0, 1 \leq j \leq J.$$

Similarly we can show that  $(p_1^* - p_2^*)u_j = 0$ .

Construct  $q \in P^n$  using  $p_1 - p_2$  so that  $qv_j = 0$  and  $q^*u_j = 0$ . For a fixed  $\epsilon \in (0, 1)$ , define

$$p_\epsilon = (1 - \epsilon)p + \epsilon q \in P^n.$$

Hence

$$p_\epsilon^* p_\epsilon v_j = (1 - \epsilon)p_\epsilon^* p(A)v_j = (1 - \epsilon)\sigma_1 p_\epsilon^* u_j = (1 - \epsilon)^2 \sigma_1^2 v_j.$$

This says that  $p_\epsilon$  has right singular vector  $v_1, \dots, v_J$  corresponding to the singular value  $(1 - \epsilon)\sigma_1$ .

There are two cases to consider:

- (1)  $\|p_\epsilon\| = (1 - \epsilon)\sigma_1 < \sigma_1$  is not the largest singular value, we see a contradiction.
- (2) None of  $v_1, \dots, v_J$  correspond to the largest singular value of  $p_\epsilon$ . Using this fact and

$$p(A) = U\Sigma V^*,$$

we have

$$\begin{aligned} \|p_\epsilon(A)\| &= \|p_\epsilon(A)V\| = \|p_\epsilon(A)[v_{J+1}, \dots, v_n]\| \\ &= \|(1 - \epsilon)p(A)[v_{J+1}, \dots, v_n] + \epsilon q(A)[v_{J+1}, \dots, v_n]\| \\ &\leq (1 - \epsilon)\|[u_{J+1}, \dots, u_n]\text{diag}(\sigma_{J+1}, \dots, \sigma_n)\| + \epsilon\|q(A)[v_{J+1}, \dots, v_n]\| \\ &\leq (1 - \epsilon)\sigma_{J+1} + \epsilon\|q(A)[v_{J+1}, \dots, v_n]\| \rightarrow \sigma_{J+1} < \sigma_J = \sigma_1 \end{aligned}$$

for  $\epsilon$  small. This again leads to a contradiction. The uniqueness proof is complete.