

2020 Spring MAS 365: Final Exam

Write the following Honor Pledge and sign your name under it.

"I have neither given nor received aid on this examination, nor have I concealed a violation of the Honor Code."

1. [40 (10+10+5+5+10) points] Consider the initial-value problem:

$$y' = -f(y), \quad a \leq t \leq b, \quad y(a) = \alpha.$$

- (a) State the condition on f that the initial-value problem become well-posed. Determine whether or not it is well-posed for $f(y) = 2|y| - 3$.
(b) Consider $f(y) = \cos(3y)$. Determine the constants C and γ in

$$|y(t_i) - w_i| \leq Ch[e^{\gamma(t_i-a)} - 1]$$

for $i = 0, 1, 2, \dots, N$, where $y(t)$ denotes the solution to the initial-value problem and w_0, w_1, \dots, w_N denote the approximations generated by Euler's method with step size h for some integer N .

- (c) Determine whether Euler's method is strongly stable, weakly stable, or unstable.
(d) Describe Euler's method with step size h for the m th-order system of the first-order initial-value problem:

$$\begin{aligned} \frac{du_1}{dt} &= -f_1(u_1, u_2, \dots, u_m), \\ \frac{du_2}{dt} &= -f_2(u_1, u_2, \dots, u_m), \\ &\vdots \\ \frac{du_m}{dt} &= -f_m(u_1, u_2, \dots, u_m), \end{aligned}$$

for $a \leq t \leq b$, with the initial conditions

$$u_1(a) = \alpha_1, \quad u_2(a) = \alpha_2, \quad \dots, \quad u_m(a) = \alpha_m.$$

- (e) Consider the following functions for the m th-order system in (d):

$$f_l(u_1, u_2, \dots, u_m) = \sum_{j=1}^m a_{lj}u_j - b_l,$$

for $l = 1, \dots, m$. State the condition of h that guarantees the sequence $\{\mathbf{w}_0, \mathbf{w}_1, \dots\}$ generated by Euler's method with step size h to converge to the solution of $f_l(u_1, u_2, \dots, u_m) = 0$ for $l = 1, \dots, m$, if it exists.

2. [20 (10+10) points] For the initial-value problem

$$y' = f(t, y), \quad a \leq t \leq b, \quad y(a) = \alpha,$$

consider the Taylor method of order n

$$w_0 = \alpha, \\ w_{i+1} = w_i + hT^{(n)}(t_i, w_i, h), \quad \text{for each } i = 0, 1, \dots, N-1,$$

where

$$T^{(n)}(t_i, w_i, h) = f(t_i, w_i) + \frac{h}{2}f'(t_i, w_i) + \dots + \frac{h^{n-1}}{n!}f^{(n-1)}(t_i, w_i).$$

- (a) State a condition when this method becomes consistent.
 (b) State a condition when this method becomes convergent.
3. [35 (10+15+10) points] Consider approximating $f(x) = x^3 - x$ by a polynomial $P_2(x) = a_2x^2 + a_1x + a_0$ that minimizes

$$E_2(a_0, a_1, a_2) = \int_0^1 [f(x) - P_2(x)]^2 dx$$

- (a) Construct the normal equations in a form $H\mathbf{a} = \mathbf{b}$ to find $P_2(x)$, where H is a 3×3 matrix, \mathbf{b} is a 3×1 vector, and $\mathbf{a} = (a_0, a_1, a_2)^t$. [Note: List the normal equations in order, so that the problem (c) becomes invariant for every student.]
 (b) Apply Gaussian elimination with scaled partial pivoting to the system $H\mathbf{a} = \mathbf{b}$ to find the solution \mathbf{a} .
 (c) Find a polynomial P_1 that solves

$$\min_{P_1 \in \Pi_1} \max_{x \in [0,1]} |P_2(x) - P_1(x)|,$$

where Π_1 denotes set of all polynomials degree at most 1.

4. [20 (10+10) points] Consider an $n \times n$ matrix

$$A = \begin{bmatrix} 2 & \frac{1}{4} & 0 & 0 & 0 & \cdots & 0 \\ \frac{1}{4} & 4 & \frac{1}{2} & 0 & 0 & \cdots & 0 \\ 0 & \frac{1}{2} & 8 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \ddots & \ddots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & 2^{n-5} & 0 \\ 0 & 0 & \cdots & 0 & 2^{n-5} & 2^{n-1} & 2^{n-4} \\ 0 & 0 & \cdots & 0 & 0 & 2^{n-4} & 2^n \end{bmatrix}.$$

- (a) Describe a strategy that iteratively approximates the i th smallest eigenvalue and its associated eigenvector of the matrix A , without using the deflation technique.
 (b) Compute two iterations of the method in (a) with $\mathbf{x}^{(0)} = (\frac{1}{4}, 1)^t$ to approximate the smallest eigenvalue and its approximated eigenvector of A for $n = 2$.

5. [35 (10+15+10) points] Consider the positive definite linear system $A\mathbf{x} = \mathbf{0}$, where $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n > 0$ are the eigenvalues of $n \times n$ matrix A , and $\mathbf{v}_1, \dots, \mathbf{v}_n$ are associated orthogonal eigenvectors. Conjugate gradient method finds the solution $\mathbf{x}^* = \mathbf{0}$ to the system, which produces the minimal value of

$$g(\mathbf{x}) = \mathbf{x}^t A \mathbf{x}.$$

The conjugate gradient method is equivalent to the followings:

$$\begin{aligned} \mathbf{x}^{(k)} &= \arg \min_{\mathbf{x}} \left\{ g(\mathbf{x}) : \mathbf{x} \in \mathbf{x}^{(0)} + \text{span}\{\nabla g(\mathbf{x}^{(0)}), \dots, \nabla g(\mathbf{x}^{(k-1)})\} \right\} \\ &= \arg \min_{\mathbf{x}} \left\{ g(\mathbf{x}) : \mathbf{x} \in \mathbf{x}^{(0)} + \text{span}\{A\mathbf{x}^{(0)}, \dots, A^k \mathbf{x}^{(0)}\} \right\} \\ &= \arg \min_{\mathbf{x}} \left\{ g(\mathbf{x}) : \mathbf{x} = P_k(A)\mathbf{x}^{(0)}, P_k \in \hat{\Pi}_k \right\}, \end{aligned}$$

where $\hat{\Pi}_k$ is defined as

$$\hat{\Pi}_k := \{P_k \in \Pi_k : P_k(0) = 1\},$$

and Π_k denotes the set of all polynomials degree at most k .

- (a) By expressing $\mathbf{x}^{(0)}$ as a linear combination of the orthogonal eigenvectors, *i.e.*, $\mathbf{x}^{(0)} = \sum_{j=1}^n \beta_j \mathbf{v}_j$, show that

$$g(\mathbf{x}^{(k)}) \leq \min_{P_k \in \hat{\Pi}_k} \max_{\lambda \in [\lambda_n, \lambda_1]} [P_k(\lambda)]^2 g(\mathbf{x}^{(0)}).$$

- (b) Show that the following polynomial

$$\hat{T}_k(\lambda) = \frac{T_k\left(\frac{\lambda_1 + \lambda_n - 2\lambda}{\lambda_1 - \lambda_n}\right)}{T_k\left(\frac{\lambda_1 + \lambda_n}{\lambda_1 - \lambda_n}\right)}$$

minimizes the right-hand side of the inequality in (a) among $\hat{\Pi}_k$, where $T_k(x)$ is the Chebyshev polynomial for $x \in [-1, 1]$.

- (c) Using the fact that the Chebyshev polynomial can be written as

$$T_k(x) = \frac{1}{2} \left[(x + \sqrt{x^2 - 1})^k + (x - \sqrt{x^2 - 1})^k \right] \quad |x| \geq 1,$$

show that

$$g(\mathbf{x}^{(k)}) \leq 4 \left(\frac{\sqrt{K} - 1}{\sqrt{K} + 1} \right)^{2k} g(\mathbf{x}^{(0)}), \quad \text{where } K = \frac{\lambda_1}{\lambda_n}.$$