

THE CHINESE UNIVERSITY OF HONG KONG
Department of Mathematics
MATH3310 2022-2023
Homework Assignment 5
Due Date: December 9 before 11:59 PM

1. Consider the linear system $A\mathbf{x} = \mathbf{k}$, where

$$A = \begin{pmatrix} 1 & 0 & -\frac{1}{4} & -\frac{1}{4} \\ 0 & 1 & -\frac{1}{4} & -\frac{1}{4} \\ -\frac{1}{4} & -\frac{1}{4} & 1 & 0 \\ -\frac{1}{4} & -\frac{1}{4} & 0 & 1 \end{pmatrix} \text{ and } \mathbf{k} = \frac{1}{2} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}.$$

Let $\mathbf{x}^* = (1, 1, 1, 1)^T$ be the solution of the linear system. Suppose $\{\mathbf{x}^{(m)}\}_{m=1}^{\infty}$ and $\{\mathbf{y}^{(m)}\}_{m=1}^{\infty}$ are the sequences of vectors obtained by the Jacobi method and Gauss-Seidel method respectively to solve the linear system with initialization $\mathbf{x}^{(0)} = \mathbf{y}^{(0)} = (0, 0, 0, 0)^T$. Let $\mathbf{e}_J^{(m)} := \mathbf{x}^{(m)} - \mathbf{x}^*$ and $\mathbf{e}_{GS}^{(m)} := \mathbf{y}^{(m)} - \mathbf{x}^*$ be the error vectors at the m -th iteration for the Jacobi and Gauss-Seidel method respectively.

- (a) Show that: $\mathbf{e}_J^{(m)} = -2^{-m}(1, 1, 1, 1)^T$ for $m \geq 1$.
 (b) Show that: $\mathbf{e}_{GS}^{(m)} = -4^{-m}(2, 2, 1, 1)^T$ for $m \geq 1$.
 (c) Show that $\|\mathbf{e}_{GS}^{(m)}\|_2 < \|\mathbf{e}_J^{(m)}\|_2$ for $m \geq 1$. Hence, the Gauss-Seidel method converges faster than the Jacobi method.

2. Consider the linear system $A\mathbf{x} = \mathbf{b}$, where $A \in \mathbb{R}^{n \times n}$ is symmetric positive definite. Consider:

$$f(\boldsymbol{\eta}) = \frac{1}{2} \boldsymbol{\eta}^T A \boldsymbol{\eta} - \mathbf{b}^T \boldsymbol{\eta}$$

- (a) For an iterative scheme $\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \alpha_k \mathbf{p}_k$, where \mathbf{p}_k is a fixed direction, find α_k such that $f(\mathbf{x}^{(k+1)})$ is minimized.
 (b) The conjugate gradient method is given by

$$\begin{aligned} \mathbf{x}^{(k+1)} &= \mathbf{x}^{(k)} + \alpha_k \mathbf{p}_k, \\ \mathbf{p}_{k+1} &= -\mathbf{r}_{k+1} - \beta_k \mathbf{p}_k, \\ \beta_k &= -\frac{\langle \mathbf{r}_{k+1}, \mathbf{p}_k \rangle_A}{\langle \mathbf{p}_k, \mathbf{p}_k \rangle_A} \end{aligned}$$

where α_k is in the form given by (a), $\mathbf{r}_k = A\mathbf{x}^{(k)} - \mathbf{b}$, $\mathbf{p}_0 = -\mathbf{r}_0$, $\langle \mathbf{x}, \mathbf{y} \rangle_A = \mathbf{x} \cdot A\mathbf{y}$, $\mathbf{x}^{(0)} \in \mathbb{R}^n$ is an arbitrary initial guess. Provided that $\mathbf{r}_i \cdot \mathbf{r}_j = 0$ and $\langle \mathbf{p}_i, \mathbf{p}_j \rangle_A = 0$ for all $i \neq j$, show that $\beta_k = -\frac{\mathbf{r}_{k+1} \cdot \mathbf{r}_{k+1}}{\mathbf{r}_k \cdot \mathbf{r}_k}$.

3. Consider the gradient descent method for solving $A\mathbf{x} = \mathbf{b}$ with some $\alpha \in \mathbb{R}$ and A to be a symmetric positive definite matrix:

$$\begin{aligned} \mathbf{x}^{k+1} &= \mathbf{x}^k + \alpha \mathbf{d}^k \\ \mathbf{d}^k &= -(A\mathbf{x}^k - \mathbf{b}) \end{aligned}$$

Prove that the method converges if and only if $\alpha < \frac{2}{\lambda_j}$ for all j where λ_j are the eigenvalues of A .

Hint: suppose $\boldsymbol{\eta}$ is the solution to $A\mathbf{x} = \mathbf{b}$, then $\boldsymbol{\eta}$ satisfies

$$\boldsymbol{\eta} = \boldsymbol{\eta} + \alpha(A\boldsymbol{\eta} - \mathbf{b}).$$

Using this equation, start with the error vector $\mathbf{e}^k = \mathbf{x}^k - \boldsymbol{\eta}$ to make some observation.