

## 2021 Spring MAS 365: Midterm 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. [35 (5+15+10+5) points] Consider a linear system  $A\mathbf{x} = \mathbf{b}$ , where

$$A = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} \\ \frac{2}{3} & \frac{1}{3} \end{bmatrix} \quad \text{and} \quad \mathbf{b} = \begin{bmatrix} \frac{1}{6} \\ \frac{5}{6} \end{bmatrix}.$$

Suppose that a two-digit rounding was used when storing  $A$  and  $\mathbf{b}$ , and let the corresponding matrices be  $\tilde{A}$  and  $\tilde{\mathbf{b}}$ , respectively. Let  $\mathbf{x} = (x_1, x_2)$  and  $\tilde{\mathbf{x}} = (\tilde{x}_1, \tilde{x}_2)$  be the solutions of  $A\mathbf{x} = \mathbf{b}$  and  $\tilde{A}\tilde{\mathbf{x}} = \tilde{\mathbf{b}}$ , respectively. Note that  $\mathbf{x} = (1.5, -0.5)$ . Use  $l_\infty$  norm through out this problem, and consider the theorem below.

**Theorem 1.** Suppose  $A$  is nonsingular and  $\|\delta A\| < \frac{1}{\|A^{-1}\|}$ . The solution  $\tilde{\mathbf{x}}$  to  $(A + \delta A)\tilde{\mathbf{x}} = \mathbf{b} + \delta \mathbf{b}$  approximates the solution  $\mathbf{x}$  of  $A\mathbf{x} = \mathbf{b}$  with the error estimate

$$\frac{\|\mathbf{x} - \tilde{\mathbf{x}}\|}{\|\mathbf{x}\|} \leq \frac{K(A)\|A\|}{\|A\| - K(A)\|\delta A\|} \left( \frac{\|\delta \mathbf{b}\|}{\|\mathbf{b}\|} + \frac{\|\delta A\|}{\|A\|} \right).$$

- (a) Determine  $\tilde{A}$  and  $\tilde{\mathbf{b}}$ .
- (b) Use Gaussian elimination with a partial pivoting and a two-digit rounding to approximate the solution  $\tilde{\mathbf{x}}$  of  $\tilde{A}\tilde{\mathbf{x}} = \tilde{\mathbf{b}}$ .
- (c) Find an upper bound of  $\frac{\|\mathbf{x} - \tilde{\mathbf{x}}\|}{\|\mathbf{x}\|}$ .
- (d)  $\tilde{x}_1$  approximates  $x_1$  to  $t$  significant digits. Find a tight lower bound of  $t$  using (c).

**Solution:**

- (a) Applying a two-digit rounding to  $A$  (3 points) and  $\mathbf{b}$  (2 points) yields

$$\tilde{A} = \begin{bmatrix} 0.33 & 0.67 \\ 0.67 & 0.33 \end{bmatrix} \quad \text{and} \quad \tilde{\mathbf{b}} = \begin{bmatrix} 0.17 \\ 0.83 \end{bmatrix}.$$

- (b) Since  $|\tilde{a}_{21}| = 0.67$  is greater than  $|\tilde{a}_{11}| = 0.33$ , we interchange rows and the resulting augmented matrix is (5 points)

$$\left[ \begin{array}{cc|c} 0.67 & 0.33 & 0.83 \\ 0.33 & 0.67 & 0.17 \end{array} \right]$$

The multiplier of this system is (2 points)

$$m_{21} = \frac{0.33}{0.67} \approx 0.49$$

Performing the operation  $(E_2 - m_{21}E_1) \rightarrow (E_2)$  gives

$$\left[ \begin{array}{cc|c} 0.67 & 0.33 & 0.83 \\ 0.00 & 0.51 & -0.24 \end{array} \right],$$

where  $0.67 - 0.49 \times 0.33 \approx 0.67 - 0.16 \approx 0.51$  (2 points) and  $0.17 - 0.49 \times 0.83 \approx 0.17 - 0.41 \approx -0.24$  (2 points). Using the backward substitution, we have

$$\begin{aligned} x_2 &\approx \frac{-0.24}{0.51} \approx -0.47, \quad (2 \text{ points}) \\ x_1 &\approx \frac{0.83 - 0.33 \times (-0.47)}{0.67} \approx \frac{0.83 + 0.16}{0.67} \approx 1.5 \quad (2 \text{ points}). \end{aligned}$$

(c) Note that (1 point each)

$$A^{-1} = \begin{bmatrix} -1 & 2 \\ 2 & -1 \end{bmatrix}, \quad \delta A = \tilde{A} - A = \begin{bmatrix} -\frac{1}{300} & \frac{1}{300} \\ \frac{1}{300} & -\frac{1}{300} \end{bmatrix}, \quad \text{and} \quad \delta \mathbf{b} = \tilde{\mathbf{b}} - \mathbf{b} = \begin{bmatrix} \frac{1}{300} \\ -\frac{1}{300} \end{bmatrix}$$

We then have (1 point each)

$$\|A\| = 1, \quad \|A^{-1}\| = 3, \quad \|\delta A\| = \frac{1}{150}, \quad \|\mathbf{b}\| = \frac{5}{6}, \quad \|\delta \mathbf{b}\| = \frac{1}{300}, \quad K(A) = \|A\| \cdot \|A^{-1}\| = 3.$$

Therefore, since  $A$  is nonsingular (*i.e.*,  $A$  has an inverse) and  $\|\delta A\| = \frac{1}{150} \leq \frac{1}{\|A^{-1}\|} = \frac{1}{3}$ , we have the following upper bound (1 point)

$$\frac{\|\mathbf{x} - \tilde{\mathbf{x}}\|}{\|\mathbf{x}\|} \leq \frac{3}{1 - 3 \times \frac{1}{150}} \left( \frac{\frac{1}{300}}{\frac{5}{6}} + \frac{1}{150} \right) = \frac{150}{49} \frac{16}{1500} = \frac{8}{245} \approx 0.0327.$$

(d) Since (3 points)

$$\frac{|x_1 - \tilde{x}_1|}{|x_1|} = \frac{|x_1 - \tilde{x}_1|}{\|\mathbf{x}\|} \leq \frac{\|\mathbf{x} - \tilde{\mathbf{x}}\|}{\|\mathbf{x}\|} \approx 0.0327 \leq 5 \times 10^{-2},$$

$t$  is greater than equal to 2. (2 points)

2. [30 (15+15) points] Let  $D = \{(x_1, x_2) : a \leq x_1 \leq b, c \leq x_2 \leq d\} \subset \mathbb{R}^2$ . Suppose that  $\mathbf{G}(\mathbf{x}) = (g_1(\mathbf{x}), g_2(\mathbf{x}))^t : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  is a continuous function from  $D$  into  $\mathbb{R}^2$  with the property that  $\mathbf{G}(\mathbf{x}) \in D$  whenever  $\mathbf{x} \in D$ . Assume that  $\mathbf{G}(\mathbf{x})$  has a unique fixed point  $\mathbf{p} = (p_1, p_2)^t \in D$ .

- (a) Show that  $\|\mathbf{G}(\mathbf{x}) - \mathbf{G}(\mathbf{y})\|_\infty \leq K\|\mathbf{x} - \mathbf{y}\|_\infty$  for any  $\mathbf{x}, \mathbf{y} \in D$ , under the assumption that all the component functions of  $\mathbf{G}$  have continuous partial derivatives and a constant  $K < 1$  exists with, for  $i = 1, 2$  and  $j = 1, 2$ ,

$$\left| \frac{\partial g_i(\mathbf{x})}{\partial x_j} \right| \leq \frac{K}{2}, \quad \text{whenever } \mathbf{x} \in D.$$

Consider the following Taylor's Theorem in two variables.

**Theorem 2.** *Supposed that  $g(x_1, x_2) : \mathbb{R}^2 \rightarrow \mathbb{R}$  and all its partial derivatives are continuous on  $D = \{(x_1, x_2) : a \leq x_1 \leq b, c \leq x_2 \leq d\}$  and  $(\bar{x}_1, \bar{x}_2) \in D$ . For every  $(x_1, x_2)$ , there exists  $\xi_1$  between  $x_1$  and  $\bar{x}_1$ , and  $\xi_2$  between  $x_2$  and  $\bar{x}_2$  with*

$$g(x_1, x_2) = g(\bar{x}_1, \bar{x}_2) + (x_1 - \bar{x}_1) \frac{\partial g(\xi_1, \xi_2)}{\partial x_1} + (x_2 - \bar{x}_2) \frac{\partial g(\xi_1, \xi_2)}{\partial x_2}.$$

- (b) Using (a), show that the sequence defined by  $\mathbf{x}^{(k)} = \mathbf{G}(\mathbf{x}^{(k-1)})$  for  $k \geq 1$  satisfies  $\|\mathbf{x}^{(k)} - \mathbf{p}\|_\infty \leq \frac{K^k}{1-K} \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty$ , for any  $\mathbf{x}^{(0)}$  in  $D$ , where  $\{\mathbf{x}^{(k)}\}$  is assumed to converge to  $\mathbf{p}$ .

**Solution:**

- (a) Applying the theorem to  $g_i$  (3 points) for  $i = 1, 2$  and for any  $\mathbf{x}, \mathbf{y} \in D$ , we have

$$\begin{aligned} |g_i(\mathbf{x}) - g_i(\mathbf{y})| &= \left| \sum_{j=1}^2 \frac{\partial g_i(\xi_1, \xi_2)}{\partial x_j} (x_j - y_j) \right| \quad (2 \text{ points}) \\ &\leq \sum_{j=1}^2 \left| \frac{\partial g_i(\xi_1, \xi_2)}{\partial x_j} \right| |x_j - y_j| \leq \|\mathbf{x} - \mathbf{y}\|_\infty \sum_{j=1}^2 \left| \frac{\partial g_i(\xi_1, \xi_2)}{\partial x_j} \right| \leq K\|\mathbf{x} - \mathbf{y}\|_\infty \quad (5 \text{ points}), \end{aligned}$$

where  $\xi_1$  is a number between  $x_1$  and  $y_1$ , and  $\xi_2$  is a number between  $x_2$  and  $y_2$ . Then, (5 points)

$$\|\mathbf{G}(\mathbf{x}) - \mathbf{G}(\mathbf{y})\|_\infty = \max_{i=1,2} |g_i(\mathbf{x}) - g_i(\mathbf{y})| \leq K\|\mathbf{x} - \mathbf{y}\|_\infty$$

for any  $\mathbf{x}, \mathbf{y} \in D$ .

- (b) For  $k \geq 1$ , we have (5 points)

$$\|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\|_\infty = \|\mathbf{G}(\mathbf{x}^{(k)}) - \mathbf{G}(\mathbf{x}^{(k-1)})\|_\infty \leq K\|\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}\|_\infty \leq \dots \leq K^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty$$

Thus for  $m > k \geq 1$ , (5 points)

$$\begin{aligned} \|\mathbf{x}^{(m)} - \mathbf{x}^{(k)}\|_\infty &= \|\mathbf{x}^{(m)} - \mathbf{x}^{(m-1)} + \mathbf{x}^{(m-1)} - \dots + \mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\|_\infty \\ &\leq \|\mathbf{x}^{(m)} - \mathbf{x}^{(m-1)}\|_\infty + \dots + \|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\|_\infty \\ &\leq K^{m-1} \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty + \dots + K^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty \\ &= K^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty (1 + K + K^2 + \dots + K^{m-k-1}). \end{aligned}$$

Then, (5 points)

$$\begin{aligned} \|\mathbf{p} - \mathbf{x}^{(k)}\| &= \lim_{m \rightarrow \infty} \|\mathbf{x}^{(m)} - \mathbf{x}^{(k)}\|_\infty = \lim_{m \rightarrow \infty} K^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty \sum_{i=0}^{m-k-1} K^i \\ &= K^k \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty \sum_{i=0}^{\infty} K^i = \frac{K^k}{1-K} \|\mathbf{x}^{(1)} - \mathbf{x}^{(0)}\|_\infty. \end{aligned}$$

3. [25 (5+10+10) points] Consider solving a linear system  $A\mathbf{x} = \mathbf{b}$ , where  $A$  is a  $n \times n$  matrix, by a method  $\mathbf{R}$ :

$$\mathbf{x}^{(k)} = \mathbf{x}^{(k-1)} + w(\mathbf{b} - A\mathbf{x}^{(k-1)})$$

for  $\mathbf{x}^{(0)} \in \mathbb{R}^n$  and for some positive  $w$ .

- State a condition on  $w$  that makes the sequence  $\{\mathbf{x}^{(k)}\}$  of the method  $\mathbf{R}$  converge to the unique solution of  $A\mathbf{x} = \mathbf{b}$ .
- Assume that  $A$  is a diagonal matrix with diagonal elements  $\lambda_1 > \lambda_2 > \dots > \lambda_n > 0$ . Simplify the condition on  $w$  in (a) into a form  $c < w < d$ . (Specify  $c$  and  $d$ .)
- Considering the relationship between the (simultaneous) Jacobi method and the (sequential) Gauss-Seidel Method, derive a (sequential) Gauss-Seidel-like method of the (simultaneous) method  $\mathbf{R}$  in a form  $\mathbf{x}^{(k)} = T\mathbf{x}^{(k-1)} + \mathbf{c}$ .

**Solution:**

- The iteration can be rewritten as

$$\mathbf{x}^{(k)} = (I - wA)\mathbf{x}^{(k-1)} + w\mathbf{b}.$$

By Theorem 7.19, if  $\rho(I - wA) < 1$ , the sequence converges to the unique solution of  $\mathbf{x} = (I - wA)\mathbf{x} + w\mathbf{b}$ , which is the unique solution of  $A\mathbf{x} = \mathbf{b}$ . Therefore, the condition we are looking for is  $\rho(I - wA) < 1$ .

- The condition in (a) reduces to  $\rho(I - wA) = \max_{i=1,\dots,n} |1 - w\lambda_i| < 1$ , (4 points) which is equivalent to  $0 < w < \frac{2}{\lambda_i}$  for all  $i$ . (3 points) This is exactly the condition  $0 < w < \frac{2}{\lambda_1}$ . (3 points)
- The method  $R$  can be rewritten as (2 points)

$$x_i^{(k)} = x_i^{(k-1)} + w \left[ b_i - \sum_{j=1}^n a_{ij}x_j^{(k-1)} \right]$$

for each  $i = 1, \dots, n$ . Its Gauss-Seidel-like modification yields (4 points)

$$x_i^{(k)} = x_i^{(k-1)} + w \left[ b_i - \sum_{j=1}^{i-1} a_{ij}x_j^{(k)} - \sum_{j=i}^n a_{ij}x_j^{(k-1)} \right]$$

for each  $i = 1, \dots, n$ . This is equivalent to

$$x_i^{(k)} + w \sum_{j=1}^{i-1} a_{ij}x_j^{(k)} = x_i^{(k-1)} + w \left[ b_i - \sum_{j=i}^n a_{ij}x_j^{(k-1)} \right]$$

for each  $i = 1, \dots, n$ . Let  $D$  be the diagonal matrix whose diagonal entries are those of  $A$ ,  $-L$  be the strictly lower-triangular part of  $A$ , and  $-U$  be the strictly upper-triangular part of  $A$ . Then, we can rewrite the method into a matrix-vector form as (4 points)

$$\begin{aligned} (I - wL)\mathbf{x}^{(k)} &= (I - w(D - U))\mathbf{x}^{(k-1)} + w\mathbf{b}, \\ \mathbf{x}^{(k)} &= (I - wL)^{-1}(I - w(D - U))\mathbf{x}^{(k-1)} + w(I - wL)^{-1}\mathbf{b}. \end{aligned}$$

4. [30 (10+10+10) points] Consider a system of nonlinear equations

$$f_1(x_1, x_2) = x_1^2 + x_2^2 - 2 = 0,$$

$$f_2(x_1, x_2) = x_1 - x_2 = 0.$$

- (a) Show that one iteration of Newton's method gives  $\mathbf{x}^{(1)} = (x_1^{(1)}, x_2^{(1)})^t$  with

$$x_1^{(1)} = x_2^{(1)} = \frac{(x_1^{(0)})^2 + (x_2^{(0)})^2 + 2}{2(x_1^{(0)} + x_2^{(0)})},$$

starting from  $\mathbf{x}^{(0)} = (x_1^{(0)}, x_2^{(0)})^t$ .

- (b) Show that the iteration converges to a fixed point  $(1, 1)^t$ , if  $1 \leq x_1^{(0)} + x_2^{(0)} \leq M$ , possibly using the theorem below.

**Theorem 3.** Let  $g \in C[a, b]$  be such that  $g(x) \in [a, b]$ , for all  $x$  in  $[a, b]$ . Suppose in addition, that  $g'$  exists on  $(a, b)$  and that a constant  $0 < k < 1$  exists with

$$|g'(x)| \leq k, \quad \text{for all } x \in (a, b).$$

Then, for any number  $p_0$  in  $[a, b]$ , the sequence defined by

$$p_n = g(p_{n-1}), \quad n \geq 1,$$

converges to the unique fixed point  $p$  in  $[a, b]$ .

- (c) Verify that the convergence of  $\{x_1^{(n)}\}$  is quadratic, if  $1 \leq x_1^{(0)} + x_2^{(0)} \leq M$ . (Do not directly use any theorem in the textbook.)

**Solution:**

- (a) Let  $\mathbf{F}(x_1, x_2) = (f_1(x_1, x_2), f_2(x_1, x_2))^t$ . The Jacobian matrix  $J(\mathbf{x})$  for this system is (3 points)

$$J(x_1, x_2) = \begin{bmatrix} 2x_1 & 2x_2 \\ 1 & -1 \end{bmatrix},$$

and its inverse is (3 points)

$$[J(x_1, x_2)]^{-1} = \frac{1}{-2x_1 - 2x_2} \begin{bmatrix} -1 & -2x_2 \\ -1 & 2x_1 \end{bmatrix} = \frac{1}{2(x_1 + x_2)} \begin{bmatrix} 1 & 2x_2 \\ 1 & -2x_1 \end{bmatrix}.$$

So one iteration of Newton's method is (4 points)

$$\begin{aligned} \mathbf{x}^{(1)} &= \mathbf{x}^{(0)} - [J(\mathbf{x}^{(0)})]^{-1} \mathbf{F}(\mathbf{x}^{(0)}) \\ &= \begin{bmatrix} x_1^{(0)} \\ x_2^{(0)} \end{bmatrix} - \frac{1}{2(x_1^{(0)} + x_2^{(0)})} \begin{bmatrix} (x_1^{(0)})^2 + (x_2^{(0)})^2 - 2 + 2x_1^{(0)}x_2^{(0)} - 2(x_2^{(0)})^2 \\ (x_1^{(0)})^2 + (x_2^{(0)})^2 - 2 - 2(x_1^{(0)})^2 + 2x_1^{(0)}x_2^{(0)} \end{bmatrix} \\ &= \frac{1}{2(x_1^{(0)} + x_2^{(0)})} \begin{bmatrix} (x_1^{(0)})^2 + (x_2^{(0)})^2 + 2 \\ (x_1^{(0)})^2 + (x_2^{(0)})^2 + 2 \end{bmatrix} \end{aligned}$$

- (b) Let  $p_n = x_1^{(n+1)} = x_2^{(n+1)}$  for  $n \geq 0$ . Then, for  $n \geq 1$ , (2 points)

$$p_n = x_1^{(n+1)} = x_2^{(n+1)} = g(p_{n-1}) = g(x_1^{(n)}) = g(x_2^{(n)}), \quad \text{where } g(x) = \frac{x^2 + x^2 + 2}{4x} = \frac{1}{2} \left( x + \frac{1}{x} \right).$$

$p_0$  is lower bounded by 1, so let  $a = 1$  in the theorem (1 point), since  $p_0 = \frac{t^2+(m-t)^2+2}{2m}$  is minimized when  $t = \frac{m}{2}$ , and  $p_0 = \frac{m^2+4}{4m}$  is minimized with a value 1 when  $m = 2$ . Let  $b = p_0$  in the theorem (1 point).  $g$  is continuous on  $[1, p_0]$  (1 point), since it is differentiable on  $[1, p_0]$  with its derivative  $g'(x) = \frac{1}{2} \left(1 - \frac{1}{x^2}\right)$ . Since  $g'(x) \geq 0$  for  $x \geq 1$ ,  $g$  is a nondecreasing function on an interval  $[1, p_0]$  (1 point), where  $g(1) = 1$  and  $g(p_0) = \frac{1}{2} \left(p_0 + \frac{1}{p_0}\right) \leq p_0$ . Therefore,  $g(x) \in [1, p_0]$ , for all  $x$  in  $[1, p_0]$  (1 point). In addition, we have that  $|g'(x)| \leq \frac{1}{2}$  for all  $x \in (1, p_0)$ , (2 points) so, by the theorem (1 points), the sequence defined by  $p_n = g(p_{n-1})$  for any number  $p_0 \geq 1$  converges to the unique fixed point  $p = 1$  in  $[1, p_0]$ .

(c) Since, for  $n \geq 0$ , (5 points)

$$\frac{|p_{n+1} - p|}{|p_n - p|^2} = \frac{\left|\frac{p_n^2+1}{2p_n} - 1\right|}{|p_n - 1|^2} = \frac{\frac{|p_n-1|^2}{2p_n}}{|p_n - 1|^2} = \frac{1}{2|p_n|},$$

we have that (5 points)

$$\lim_{n \rightarrow \infty} \frac{|p_{n+1} - p|}{|p_n - p|^2} = \lim_{n \rightarrow \infty} \frac{1}{2|p_n|} = \frac{1}{2}.$$

Therefore, the sequence  $\{x_1^{(n)}\}$  converges quadratically to 1.

5. [30 (15+5+10) points] The directional derivative of  $g$  at  $\mathbf{x}$  in the direction of  $\mathbf{v}$  is defined by

$$D_{\mathbf{v}}g(\mathbf{x}) = \lim_{h \rightarrow 0} \frac{1}{h} [g(\mathbf{x} + h\mathbf{v}) - g(\mathbf{x})] = \mathbf{v}^t \nabla g(\mathbf{x}).$$

The steepest direction with respect to  $\ell_2$  norm is found by solving

$$\min_{\mathbf{v} : \|\mathbf{v}\|_2=1} D_{\mathbf{v}}g(\mathbf{x}).$$

- (a) Determine the steepest direction with respect to  $\|D^{1/2} \cdot\|_2$  by solving

$$\min_{\mathbf{v} : \|D^{1/2}\mathbf{v}\|_2=1} D_{\mathbf{v}}g(\mathbf{x}),$$

where  $D$  is a diagonal matrix.

- (b) State a condition for a nonzero vector  $\mathbf{v}$  to be called a descent direction of  $g$  at  $\mathbf{x}$ .  
(c) Show that a direction  $\mathbf{v}^{(k+1)}$  of a conjugate gradient method

$$\begin{aligned} \mathbf{r}^{(0)} &= \mathbf{b} - A\mathbf{x}^{(0)}, \quad \mathbf{v}^{(1)} = \mathbf{r}^{(0)}, \\ \text{For } k &= 1, 2, \dots, n \\ t_k &= \frac{\langle \mathbf{v}^{(k)}, \mathbf{r}^{(k-1)} \rangle}{\langle \mathbf{v}^{(k)}, A\mathbf{v}^{(k)} \rangle}, \quad \mathbf{x}^{(k)} = \mathbf{x}^{(k-1)} + t_k \mathbf{v}^{(k)}, \quad \mathbf{r}^{(k)} = \mathbf{r}^{(k-1)} - t_k A\mathbf{v}^{(k)} \\ s_k &= \frac{\langle \mathbf{r}^{(k)}, \mathbf{r}^{(k)} \rangle}{\langle \mathbf{r}^{(k-1)}, \mathbf{r}^{(k-1)} \rangle} \quad \mathbf{v}^{(k+1)} = \mathbf{r}^{(k)} + s_k \mathbf{v}^{(k)}, \end{aligned}$$

for a positive definite matrix  $A$ , is a descent direction of some function at some point under some condition. Specify the corresponding function, point and condition. (Do not directly use any theorem in the textbook.)

### Solution:

- (a) Using the Cauchy-Schwarz inequality (5 points), we have

$$\mathbf{v}^t \nabla g(\mathbf{x}) = (D^{1/2}\mathbf{v})^t (D^{-1/2} \nabla g(\mathbf{x})) \geq -\|D^{1/2}\mathbf{v}\|_2 \|D^{-1/2} \nabla g(\mathbf{x})\|_2 = -\|D^{-1/2} \nabla g(\mathbf{x})\|_2$$

for all  $\mathbf{v}$  such that  $\|D^{1/2}\mathbf{v}\|_2 = 1$ . The lower bound is attained for (10 points)

$$\mathbf{v} = -\frac{D^{-1} \nabla g(\mathbf{x})}{\|D^{-1/2} \nabla g(\mathbf{x})\|_2}.$$

If you discuss about the equality condition of the Cauchy-Schwarz inequality, you will receive 5 points out of 10 points assigned for  $\mathbf{v}$ . The answer  $\mathbf{v} = \frac{D^{-1} \nabla g(\mathbf{x})}{\|D^{-1/2} \nabla g(\mathbf{x})\|_2}$  will receive 7 points.

- (b) A nonzero vector  $\mathbf{v}$  is called a descent direction of  $g$  at  $\mathbf{x}$  if  $D_{\mathbf{v}}g(\mathbf{x}) < 0$ .  
(c) Let  $g(\mathbf{x}) = \langle \mathbf{x}, A\mathbf{x} \rangle - 2\langle \mathbf{x}, \mathbf{b} \rangle$ . Then, (2 points)

$$D_{\mathbf{v}^{(k+1)}}g(\mathbf{x}^{(k)}) = [\mathbf{v}^{(k+1)}]^t \nabla g(\mathbf{x}^{(k)}) = [\mathbf{r}^{(k)} + s_k \mathbf{v}^{(k)}]^t (-2\mathbf{r}^{(k)})$$

Since  $\langle \mathbf{r}^{(k)}, \mathbf{v}^{(k)} \rangle = \langle \mathbf{r}^{(k-1)}, \mathbf{v}^{(k)} \rangle - \frac{\langle \mathbf{v}^{(k)}, \mathbf{r}^{(k-1)} \rangle}{\langle \mathbf{v}^{(k)}, A\mathbf{v}^{(k)} \rangle} \langle A\mathbf{v}^{(k)}, \mathbf{v}^{(k)} \rangle = 0$ , (1 point) we have (2 points)

$$D_{\mathbf{v}^{(k+1)}}g(\mathbf{x}^{(k)}) = -2\|\mathbf{r}^{(k)}\|_2^2 < 0.$$

So,  $\mathbf{v}^{(k+1)}$  is a descent direction of  $g(\mathbf{x}) = \langle \mathbf{x}, A\mathbf{x} \rangle - 2\langle \mathbf{x}, \mathbf{b} \rangle$  (2 points) at  $\mathbf{x}^{(k)}$  (2 points) when  $\mathbf{r}^{(k)} = \mathbf{b} - A\mathbf{x}^{(k)}$  is not a zero vector (1 point) (or equivalently, when  $\mathbf{x}^{(k)}$  is not a solution).