

ECE/MATH 520, Spring 2008

Homework Problems 1

Solutions (version: February 5, 2008, 11:7)

- 1.5** Suppose you are shown four cards, laid out in a row. Each card has a letter on one side and a number on the other. On the visible side of the cards are printed the symbols:

$S \quad 8 \quad 3 \quad A$

Determine which cards you should turn over to decide if the following rule is true or false: “If there is a vowel on one side of the card, then there is an even number on the other side.”

Ans.: The cards that you should turn over are 3 and A . The remaining cards are irrelevant to ascertaining the truth or falsity of the rule. The card with S is irrelevant because S is not a vowel. The card with 8 is not relevant because the rule does not say that if a card has an even number on one side, then it has a vowel on the other side.

Turning over the A card directly verifies the rule, while turning over the 3 card verifies the contraposition.

- 2.6** Show that for any two vectors $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, $||\mathbf{x}|| - ||\mathbf{y}|| \leq ||\mathbf{x} - \mathbf{y}||$.

Hint: Write $\mathbf{x} = (\mathbf{x} - \mathbf{y}) + \mathbf{y}$, and use the Triangle inequality. Do the same for \mathbf{y} .

Ans.: We have $||\mathbf{x}|| = ||(\mathbf{x} - \mathbf{y}) + \mathbf{y}|| \leq ||\mathbf{x} - \mathbf{y}|| + ||\mathbf{y}||$ by the Triangle Inequality. Hence, $||\mathbf{x}|| - ||\mathbf{y}|| \leq ||\mathbf{x} - \mathbf{y}||$. On the other hand, from the above we have $||\mathbf{y}|| - ||\mathbf{x}|| \leq ||\mathbf{y} - \mathbf{x}|| = ||\mathbf{x} - \mathbf{y}||$. Combining the two inequalities, we obtain $||\mathbf{x}|| - ||\mathbf{y}|| \leq ||\mathbf{x} - \mathbf{y}||$.

- 3.2** Let $\lambda_1, \dots, \lambda_n$ be the eigenvalues of the matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$. Show that the eigenvalues of the matrix $\mathbf{I}_n - \mathbf{A}$ are $1 - \lambda_1, \dots, 1 - \lambda_n$.

Ans.: Suppose $\mathbf{v}_1, \dots, \mathbf{v}_n$ are eigenvectors of \mathbf{A} corresponding to $\lambda_1, \dots, \lambda_n$, respectively. Then, for each $i = 1, \dots, n$, we have

$$(\mathbf{I}_n - \mathbf{A})\mathbf{v}_i = \mathbf{v}_i - \mathbf{A}\mathbf{v}_i = \mathbf{v}_i - \lambda_i\mathbf{v}_i = (1 - \lambda_i)\mathbf{v}_i$$

which shows that $1 - \lambda_1, \dots, 1 - \lambda_n$ are the eigenvalues of $\mathbf{I}_n - \mathbf{A}$.

Alternatively, we may write the characteristic polynomial of $\mathbf{I}_n - \mathbf{A}$ as

$$\pi_{\mathbf{I}_n - \mathbf{A}}(1 - \lambda) = \det((1 - \lambda)\mathbf{I}_n - (\mathbf{I}_n - \mathbf{A})) = \det(-[\lambda\mathbf{I}_n - \mathbf{A}]) = (-1)^n \pi_{\mathbf{A}}(\lambda),$$

which shows the desired result.

3.12a Consider the matrix

$$\mathbf{Q} = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}.$$

a. Is this matrix positive definite, negative definite, or indefinite?

Ans.: a. The matrix \mathbf{Q} is indefinite, since $\Delta_2 = -1$ and $\Delta_3 = 2$.

5.5 Let $\mathbf{x}(t) = [e^t + t^3, t^2, t+1]^T, t \in \mathbb{R}$, and $f(\mathbf{x}) = x_1^3 x_2 x_3^2 + x_1 x_2 + x_3, \mathbf{x} = [x_1, x_2, x_3]^T \in \mathbb{R}^3$. Find $(d/dt)f(\mathbf{x}(t))$ in terms of t .

Ans.: We have

$$Df(\mathbf{x}) = [3x_1^2 x_2 x_3^2 + x_2, x_1^3 x_3^2 + x_1, 2x_1^3 x_2 x_3 + 1]$$

and

$$\frac{d\mathbf{x}}{dt}(t) = \begin{bmatrix} e^t + 3t^2 \\ 2t \\ 1 \end{bmatrix}.$$

By the chain rule,

$$\begin{aligned} & \frac{d}{dt}f(\mathbf{x}(t)) \\ &= Df(\mathbf{x}(t)) \frac{d\mathbf{x}}{dt}(t) \\ &= [3x_1(t)^2 x_2(t) x_3(t)^2 + x_2(t), x_1(t)^3 x_3(t)^2 + x_1(t), 2x_1(t)^3 x_2(t) x_3(t) + 1] \begin{bmatrix} e^t + 3t^2 \\ 2t \\ 1 \end{bmatrix} \\ &= (3x_1(t)^2 x_2(t) x_3(t)^2 + x_2(t))(e^t + 3t^2) + (x_1(t)^3 x_3(t)^2 + x_1(t))(2t) \\ &\quad + 2x_1(t)^3 x_2(t) x_3(t) + 1 \\ &= 12t(e^t + 3t^2)^3 + 2te^t + 6t^2 + 2t + 1. \end{aligned}$$

(Actually, you're not expected or required to expand out and simplify to the last formula.)

5.8 Let

$$\begin{aligned} f_1(x_1, x_2) &= x_1^2 - x_2^2; \\ f_2(x_1, x_2) &= 2x_1 x_2. \end{aligned}$$

Sketch the level sets associated with $f_1(x_1, x_2) = 12$ and $f_2(x_1, x_2) = 16$ on the same diagram. Indicate on the diagram the values of $\mathbf{x} = [x_1, x_2]^T$ for which $\mathbf{f}(\mathbf{x}) = [f_1(x_1, x_2), f_2(x_1, x_2)]^T = [12, 16]^T$.

Ans.: We have that

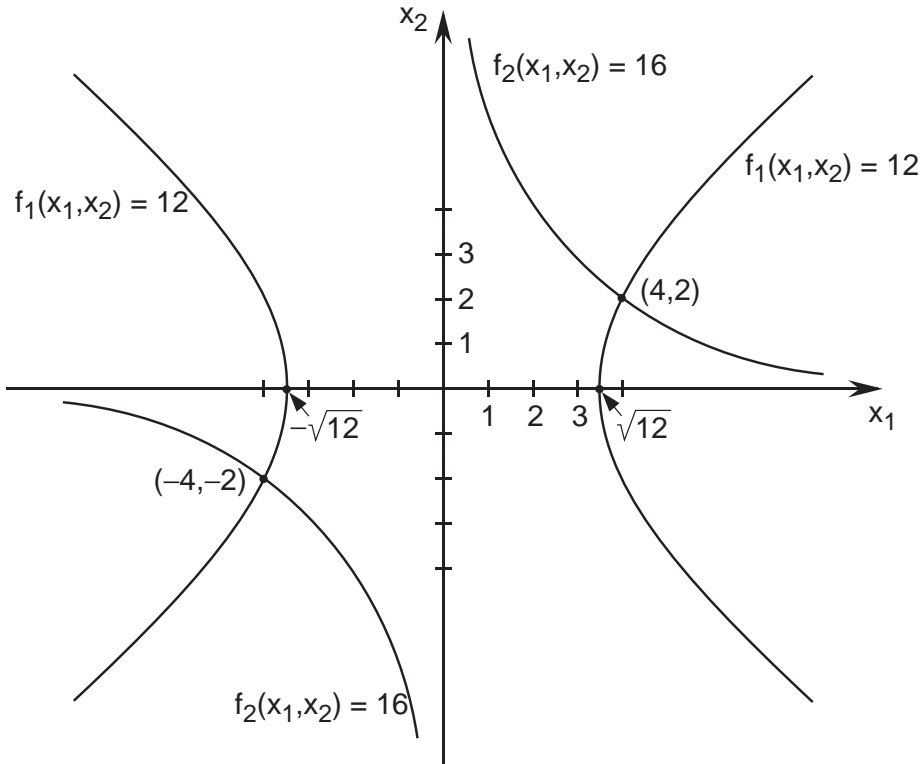
$$\{\mathbf{x} : f_1(\mathbf{x}) = 12\} = \{\mathbf{x} : x_1^2 - x_2^2 = 12\},$$

and

$$\{\mathbf{x} : f_2(\mathbf{x}) = 16\} = \{\mathbf{x} : x_2 = 8/x_1\}.$$

To find the intersection points, we substitute $x_2 = 8/x_1$ into $x_1^2 - x_2^2 = 12$ to get $x_1^4 - 12x_1^2 - 64 = 0$. Solving gives $x_1^2 = 16, -4$. Clearly, the only two possibilities for x_1 are $x_1 = +4, -4$, from which we obtain $x_2 = +2, -2$. Hence, the intersection points are located at $[4, 2]^T$ and $[-4, -2]^T$.

The level sets associated with $f_1(x_1, x_2) = 12$ and $f_2(x_1, x_2) = 16$ are shown as follows.



5.9 Write down the Taylor series expansion of the following functions about the given points \mathbf{x}_0 . Neglect terms of order three or higher.

- a. $f(\mathbf{x}) = x_1 e^{-x_2} + x_2 + 1, \mathbf{x}_0 = [1, 0]^T$
- b. $f(\mathbf{x}) = x_1^4 + 2x_1^2 x_2^2 + x_2^4, \mathbf{x}_0 = [1, 1]^T$
- c. $f(\mathbf{x}) = e^{x_1 - x_2} + e^{x_1 + x_2} + x_1 + x_2 + 1, \mathbf{x}_0 = [1, 0]^T$.

Ans.: a. We have

$$f(\mathbf{x}) = f(\mathbf{x}_o) + Df(\mathbf{x}_o)(\mathbf{x} - \mathbf{x}_o) + \frac{1}{2}(\mathbf{x} - \mathbf{x}_o)^T D^2 f(\mathbf{x}_o)(\mathbf{x} - \mathbf{x}_o) + \cdots$$

We compute

$$\begin{aligned} Df(\mathbf{x}) &= [e^{-x_2}, -x_1 e^{-x_2} + 1], \\ D^2 f(\mathbf{x}) &= \begin{bmatrix} 0 & -e^{-x_2} \\ -e^{-x_2} & x_1 e^{-x_2} \end{bmatrix}. \end{aligned}$$

Hence,

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

b. We compute

$$\begin{aligned} Df(\mathbf{x}) &= [4x_1^3 + 4x_1 x_2^2, 4x_1^2 x_2 + 4x_2^3], \\ D^2 f(\mathbf{x}) &= \begin{bmatrix} 12x_1^2 + 4x_2^2 & 8x_1 x_2 \\ 8x_1 x_2 & 4x_1^2 + 12x_2^2 \end{bmatrix}. \end{aligned}$$

Expanding f about the point \mathbf{x}_o yields

$$\begin{aligned} f(\mathbf{x}) &= 4 + [8, 8] \begin{bmatrix} x_1 - 1 \\ x_2 - 1 \end{bmatrix} + \frac{1}{2} [x_1 - 1, x_2 - 1] \begin{bmatrix} 16 & 8 \\ 8 & 16 \end{bmatrix} \begin{bmatrix} x_1 - 1 \\ x_2 - 1 \end{bmatrix} + \cdots \\ &= 8x_1^2 + 8x_2^2 - 16x_1 - 16x_2 + 8x_1 x_2 + 12 + \cdots. \end{aligned}$$

c. We compute

$$\begin{aligned} Df(\mathbf{x}) &= [e^{x_1 - x_2} + e^{x_1 + x_2} + 1, -e^{x_1 - x_2} + e^{x_1 + x_2} + 1], \\ D^2 f(\mathbf{x}) &= \begin{bmatrix} e^{x_1 - x_2} + e^{x_1 + x_2} & -e^{x_1 - x_2} + e^{x_1 + x_2} \\ -e^{x_1 - x_2} + e^{x_1 + x_2} & e^{x_1 - x_2} + e^{x_1 + x_2} \end{bmatrix}. \end{aligned}$$

Expanding f about the point \mathbf{x}_o yields

$$\begin{aligned} f(\mathbf{x}) &= 2 + 2e + [2e + 1, 1] \begin{bmatrix} x_1 - 1 \\ x_2 \end{bmatrix} + \frac{1}{2} [x_1 - 1, x_2] \begin{bmatrix} 2e & 0 \\ 0 & 2e \end{bmatrix} \begin{bmatrix} x_1 - 1 \\ x_2 \end{bmatrix} + \cdots \\ &= 1 + x_1 + x_2 + e(1 + x_1^2 + x_2^2) + \cdots. \end{aligned}$$

6.2 Show that, if \mathbf{x}^* is a global minimizer of f over Ω , and $\mathbf{x}^* \in \Omega' \subset \Omega$, then \mathbf{x}^* is a global minimizer of f over Ω' .

Ans.: Suppose \mathbf{x}^* is a global minimizer of f over Ω , and $\mathbf{x}^* \in \Omega' \subset \Omega$. Let $\mathbf{x} \in \Omega'$. Then, $\mathbf{x} \in \Omega$ and therefore $f(\mathbf{x}^*) \leq f(\mathbf{x})$. Hence, \mathbf{x}^* is a global minimizer of f over Ω' .

6.5 Consider the function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ given below:

$$f(\mathbf{x}) = \mathbf{x}^T \begin{bmatrix} 1 & 2 \\ 4 & 7 \end{bmatrix} \mathbf{x} + \mathbf{x}^T \begin{bmatrix} 3 \\ 5 \end{bmatrix} + 6$$

- a. Find the gradient and Hessian of f at the point $[1, 1]^T$.
- b. Find the directional derivative of f at $[1, 1]^T$ in the direction of maximal rate of increase.
- c. Find a point that satisfies the FONC (interior case) for f . Does this point satisfy the SONC (for a minimizer)?

Ans.: a. The gradient and Hessian of f are

$$\begin{aligned}\nabla f(\mathbf{x}) &= 2 \begin{bmatrix} 1 & 3 \\ 3 & 7 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 3 \\ 5 \end{bmatrix} \\ \mathbf{F}(\mathbf{x}) &= 2 \begin{bmatrix} 1 & 3 \\ 3 & 7 \end{bmatrix}.\end{aligned}$$

Hence, $\nabla f([1, 1]^T) = [11, 25]^T$, and $\mathbf{F}([1, 1]^T)$ is as shown above.

b. The direction of maximal rate of increase is the direction of the gradient. Hence, the directional derivative with respect to a unit vector in this direction is

$$\left(\frac{\nabla f(\mathbf{x})}{\|\nabla f(\mathbf{x})\|} \right)^T \nabla f(\mathbf{x}) = \frac{(\nabla f(\mathbf{x}))^T \nabla f(\mathbf{x})}{\|\nabla f(\mathbf{x})\|} = \|\nabla f(\mathbf{x})\|.$$

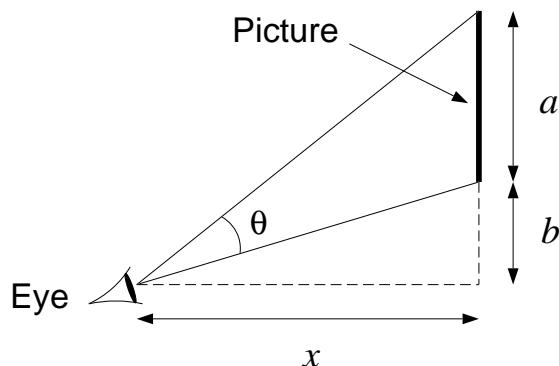
At $\mathbf{x} = [1, 1]^T$, we have $\|\nabla f([1, 1]^T)\| = \sqrt{11^2 + 25^2} = 27.31$.

c. The FONC in this case is $\nabla f(\mathbf{x}) = \mathbf{0}$. Solving, we get

$$\mathbf{x} = \begin{bmatrix} 3/2 \\ -1 \end{bmatrix}.$$

The point above does not satisfy the SONC because the Hessian is not positive semidefinite (its determinant is negative).

- 6.11** An art collector stands at distance of x feet from the wall where a piece of art (picture) of height a feet is hung, b feet above his eyes, as shown below.



Find the distance from the wall for which the angle θ subtended by the eye to the picture is maximized.

Hint: (1) Maximizing θ is equivalent to maximizing $\tan(\theta)$;

(2) If $\theta = \theta_2 - \theta_1$, then $\tan(\theta) = (\tan(\theta_2) - \tan(\theta_1))/(1 + \tan(\theta_2)\tan(\theta_1))$.

Ans.: Let θ_1 be the angle from the horizontal to the bottom of the picture, and θ_2 the angle from the horizontal to the top of the picture. Then, $\tan(\theta) = (\tan(\theta_2) - \tan(\theta_1))/(1 + \tan(\theta_2)\tan(\theta_1))$. Now, $\tan(\theta_1) = b/x$ and $\tan(\theta_2) = (a + b)/x$. Hence, the objective function that we wish to maximize is

$$f(x) = \frac{(a + b)/x - b/x}{1 + b(a + b)/x^2} = \frac{a}{x + b(a + b)/x}.$$

We have

$$f'(x) = -\frac{a^2}{(x + b(a + b)/x)^2} \left(1 - \frac{b(a + b)}{x^2}\right).$$

Let x^* be the optimal distance. Then, x^* must satisfy the FONC. Now, the point $x^* = 0$ does not satisfy the FONC (why?). Therefore, x^* must be an interior point of the constraint set $\Omega = \{x : x \geq 0\}$. Hence, we have $f'(x^*) = 0$, which gives

$$\begin{aligned} 1 - \frac{b(a + b)}{(x^*)^2} &= 0 \\ \Rightarrow x^* &= \sqrt{b(a + b)}. \end{aligned}$$

6.20 Line Fitting. Let $[x_1, y_1]^T, \dots, [x_n, y_n]^T$, $n \geq 2$, be points on the \mathbb{R}^2 plane (each $x_i, y_i \in \mathbb{R}$).

We wish to find the straight line of “best fit” through these points (“best” in the sense that the average squared error is minimized); that is, we wish to find $a, b \in \mathbb{R}$ to minimize

$$f(a, b) = \frac{1}{n} \sum_{i=1}^n (ax_i + b - y_i)^2$$

a. Let

$$\begin{aligned} \bar{X} &= \frac{1}{n} \sum_{i=1}^n x_i \\ \bar{Y} &= \frac{1}{n} \sum_{i=1}^n y_i \\ \overline{X^2} &= \frac{1}{n} \sum_{i=1}^n x_i^2 \\ \overline{Y^2} &= \frac{1}{n} \sum_{i=1}^n y_i^2 \\ \overline{XY} &= \frac{1}{n} \sum_{i=1}^n x_i y_i. \end{aligned}$$

Show that $f(a, b)$ can be written in the form $\mathbf{z}^T \mathbf{Q} \mathbf{z} - 2\mathbf{c}^T \mathbf{z} + d$, where $\mathbf{z} = [a, b]^T$, $\mathbf{Q} = \mathbf{Q}^T \in \mathbb{R}^{2 \times 2}$, $\mathbf{c} \in \mathbb{R}^2$, and $d \in \mathbb{R}$, and find expressions for \mathbf{Q} , \mathbf{c} , and d in terms of \bar{X} , \bar{Y} , $\overline{X^2}$, $\overline{Y^2}$, and \overline{XY} .

- b. Assume that the x_i , $i = 1, \dots, n$, are not all equal. Find the parameters a^* and b^* for the line of best fit in terms of \bar{X} , \bar{Y} , $\overline{X^2}$, $\overline{Y^2}$, and \overline{XY} . Show that the point $[a^*, b^*]^T$ is the only local minimizer of f .

Hint: $\overline{X^2} - (\bar{X})^2 = (1/n) \sum_{i=1}^n (x_i - \bar{X})^2$.

- c. Show that if a^* and b^* are the parameters of the line of best fit, then $\bar{Y} = a^* \bar{X} + b^*$ (and hence once we have computed a^* , we can compute b^* using the formula $b^* = \bar{Y} - a^* \bar{X}$).

Ans.: a. We write

$$\begin{aligned} f(a, b) &= \frac{1}{n} \sum_{i=1}^n a^2 x_i^2 + b^2 + y_i^2 + 2x_i a b - 2x_i y_i a - 2y_i b \\ &= a^2 \left(\frac{1}{n} \sum_{i=1}^n x_i^2 \right) + b^2 + 2 \left(\frac{1}{n} \sum_{i=1}^n x_i \right) a b \\ &\quad - 2 \left(\frac{1}{n} \sum_{i=1}^n x_i y_i \right) a - 2 \left(\frac{1}{n} \sum_{i=1}^n y_i \right) b + \left(\frac{1}{n} \sum_{i=1}^n y_i^2 \right) \\ &= [a \ b] \begin{bmatrix} \frac{1}{n} \sum_{i=1}^n x_i^2 & \frac{1}{n} \sum_{i=1}^n x_i \\ \frac{1}{n} \sum_{i=1}^n x_i y_i & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} \\ &\quad - 2 \left[\frac{1}{n} \sum_{i=1}^n x_i y_i, \frac{1}{n} \sum_{i=1}^n y_i \right] \begin{bmatrix} a \\ b \end{bmatrix} + \frac{1}{n} \sum_{i=1}^n y_i^2 \\ &= \mathbf{z}^T \mathbf{Q} \mathbf{z} - 2\mathbf{c}^T \mathbf{z} + d, \end{aligned}$$

where \mathbf{z} , \mathbf{Q} , \mathbf{c} and d are defined in the obvious way.

- b. If the point $\mathbf{z}^* = [a^*, b^*]^T$ is a solution, then by the FONC, we have $\nabla f(\mathbf{z}^*) = 2\mathbf{Q}\mathbf{z}^* - 2\mathbf{c} = \mathbf{0}$, which means $\mathbf{Q}\mathbf{z}^* = \mathbf{c}$. Now, since $\overline{X^2} - (\bar{X})^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{X})^2$, and the x_i are not all equal, then $\det \mathbf{Q} = \overline{X^2} - (\bar{X})^2 \neq 0$. Hence, \mathbf{Q} is nonsingular, and hence

$$\mathbf{z}^* = \mathbf{Q}^{-1} \mathbf{c} = \frac{1}{\overline{X^2} - (\bar{X})^2} \begin{bmatrix} 1 & -\bar{X} \\ -\bar{X} & \overline{X^2} \end{bmatrix} \begin{bmatrix} \overline{XY} \\ \bar{Y} \end{bmatrix} = \begin{bmatrix} \frac{\overline{XY} - (\bar{X})(\bar{Y})}{\overline{X^2} - (\bar{X})^2} \\ \frac{(\overline{X^2})(\bar{Y}) - (\bar{X})(\overline{XY})}{\overline{X^2} - (\bar{X})^2} \end{bmatrix}.$$

Since $\mathbf{Q} > 0$, then by the SOSC, the point \mathbf{z}^* is a strict local minimizer. Since \mathbf{z}^* is the only point satisfying the FONC, then \mathbf{z}^* is the only local minimizer.

- c. We have

$$a^* \bar{X} + b^* = \left(\frac{\overline{XY} - (\bar{X})(\bar{Y})}{\overline{X^2} - (\bar{X})^2} \right) \bar{X} + \frac{(\overline{X^2})(\bar{Y}) - (\bar{X})(\overline{XY})}{\overline{X^2} - (\bar{X})^2} = \bar{Y}.$$