

Math 217 Fall 2013 Homework 7 Solutions

DUE THURSDAY NOV. 7, 2013 5PM

- This homework consists of 6 problems of 5 points each. The total is 30.
- You need to fully justify your answer – prove that your function indeed has the specified property – for each problem.
- Please read this week’s lecture notes before working on the problems.

Question 1. Let $f(x, y) = x^3 + y^3 + x y^2$. Calculate its Taylor expansion to degree 2 with remainder (that is $n = 2$, the remainder involves 3rd order derivatives) at $(1, 0)$.

Solution. We have

$$\frac{\partial f}{\partial x} = 3x^2 + y^2, \quad \frac{\partial f}{\partial y} = 3y^2 + 2xy, \quad (1)$$

$$\frac{\partial^2 f}{\partial x^2} = 6x, \quad \frac{\partial^2 f}{\partial x \partial y} = 2y, \quad \frac{\partial^2 f}{\partial y^2} = 6y + 2x \quad (2)$$

$$\frac{\partial^3 f}{\partial x^3} = 6, \quad \frac{\partial^3 f}{\partial x^2 \partial y} = 0, \quad \frac{\partial^3 f}{\partial x \partial y^2} = 2, \quad \frac{\partial^3 f}{\partial y^3} = 6. \quad (3)$$

Therefore the Taylor expansion with remainder is

$$f(x, y) = 1 + 3(x - 1) + \frac{1}{2}[6(x - 1)^2 + 2y^2] + [(x - 1)^3 + (x - 1)y^2 + y^3]. \quad (4)$$

Question 2. Let $f(x, y) = \frac{x^2}{y}$. Calculate its Taylor polynomial of degree 3 (that is P_3) at $(1, 1)$.

Solution. We have

$$\frac{\partial f}{\partial x} = 2xy^{-1}, \quad \frac{\partial f}{\partial y} = -x^2 y^{-2}; \quad (5)$$

$$\frac{\partial^2 f}{\partial x^2} = 2y^{-1}, \quad \frac{\partial^2 f}{\partial x \partial y} = -2xy^{-2}, \quad \frac{\partial^2 f}{\partial y^2} = 2x^2 y^{-3}; \quad (6)$$

$$\frac{\partial^3 f}{\partial x^3} = 0, \quad \frac{\partial^3 f}{\partial x^2 \partial y} = -2y^{-2}, \quad \frac{\partial^3 f}{\partial x \partial y^2} = 4xy^{-3}, \quad \frac{\partial^3 f}{\partial y^3} = -6x^2 y^{-4}. \quad (7)$$

Thus P_3 at $(1, 1)$ is

$$1 + [2(x - 1) - (y - 1)] + [(x - 1)^2 - 2(x - 1)(y - 1) + (y - 1)^2] + [-(x - 1)^2(y - 1) + 2(x - 1)(y - 1)^2 - (y - 1)^3]. \quad (8)$$

Question 3. Let $f(x, y, z) = \frac{\cos x \cos y}{\cos z}$. Calculate its Hessian matrix at $(0, 0, 0)$.

Solution. We have

$$\frac{\partial f}{\partial x} = -\frac{\sin x \cos y}{\cos z}, \quad \frac{\partial f}{\partial y} = -\frac{\cos x \sin y}{\cos z}, \quad \frac{\partial f}{\partial z} = \frac{\cos x \cos y \sin z}{(\cos z)^2}. \quad (9)$$

Next

$$\frac{\partial^2 f}{\partial x^2} = -\frac{\cos x \cos y}{\cos z}, \quad \frac{\partial^2 f}{\partial y^2} = -\frac{\cos x \cos y}{\cos z}, \quad \frac{\partial^2 f}{\partial z^2} = \frac{\cos x \cos y \cos z}{(\cos z)^2} + 2 \frac{\cos x \cos y (\sin z)^2}{(\cos z)^3}; \quad (10)$$

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\sin x \sin y}{\cos z}, \quad \frac{\partial^2 f}{\partial y \partial z} = -\frac{\cos x \sin y \sin z}{(\cos z)^2}, \quad \frac{\partial^2 f}{\partial z \partial x} = -\frac{\sin x \cos y \sin z}{(\cos z)^2}. \quad (11)$$

So its Hessian matrix at $(0, 0, 0)$ is

$$H_f(0, 0, 0) = \begin{pmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix}. \quad (12)$$

Question 4. Let $f: \mathbb{R}^N \mapsto \mathbb{R}$ belong to C^2 , that is all of its second order partial derivatives exist and are continuous. Let $\mathbf{x}_0 \in \mathbb{R}^N$. Assume

$$\forall \mathbf{v} \in \mathbb{R}^N, \mathbf{v} \neq \mathbf{0} \quad \mathbf{v}^T H(\mathbf{x}_0) \mathbf{v} > 0 \quad (13)$$

where $H(\mathbf{x}_0)$ is the Hessian matrix of f at \mathbf{x}_0 . Prove that there is $r > 0$ such that for all $\mathbf{x} \in B(\mathbf{x}_0, r)$, there holds

$$\forall \mathbf{v} \in \mathbb{R}^N, \mathbf{v} \neq \mathbf{0} \quad \mathbf{v}^T H(\mathbf{x}) \mathbf{v} > 0. \quad (14)$$

Solution. Since $f \in C^2$, each entry $h_{ij}(\mathbf{x})$ of the Hessian matrix $H(\mathbf{x})$ is continuous. Now define

$$g(\mathbf{x}, \mathbf{v}): \mathbb{R}^{2N} \mapsto \mathbb{R}^{N \times N} \quad (15)$$

as

$$g(\mathbf{x}, \mathbf{v}) = \mathbf{v}^T H(\mathbf{x}) \mathbf{v} = \sum_{i,j=1}^N v_i h_{ij}(\mathbf{x}) v_j. \quad (16)$$

Since g is the sum of products $\sum_{i,j=1}^N v_i h_{ij}(\mathbf{x}) v_j$ and $v_i, v_j, h_{ij}(\mathbf{x})$ are all continuous as functions of (\mathbf{x}, \mathbf{v}) , g is a continuous function of (\mathbf{x}, \mathbf{v}) .

Now consider the bounded closed set $A := \{(\mathbf{x}, \mathbf{v}) \mid \mathbf{x} = \mathbf{x}_0, \|\mathbf{v}\| = 1\}$. By assumption we have

$$g(\mathbf{x}, \mathbf{v}) > 0 \quad (17)$$

for all $(\mathbf{x}, \mathbf{v}) \in A$. By continuity for each point $(\mathbf{x}_0, \mathbf{v}_0)$ in A there is $r_{\mathbf{x}_0, \mathbf{v}_0} > 0$ such that

$$\forall (\mathbf{x}, \mathbf{v}) \in B((\mathbf{x}_0, \mathbf{v}_0), r_{\mathbf{x}_0, \mathbf{v}_0}), \quad g(\mathbf{x}, \mathbf{v}) > 0 \quad (18)$$

By Heine-Borel A is compact, so there are finitely many such balls covering A . Now take r to be the smallest of their radius. We have, in particular,

$$\forall (\mathbf{x}, \mathbf{v}) \text{ with } \mathbf{x} \in B(\mathbf{x}_0, r), \|\mathbf{v}\| = 1, \quad g(\mathbf{x}, \mathbf{v}) > 0. \quad (19)$$

Now for any $\mathbf{u} \in \mathbb{R}^N, \mathbf{u} \neq \mathbf{0}$, we have $\left\| \frac{\mathbf{u}}{\|\mathbf{u}\|} \right\| = 1$ and therefore

$$\mathbf{u}^T H(\mathbf{x}) \mathbf{u} = \|\mathbf{u}\|^2 \left[\left(\frac{\mathbf{u}}{\|\mathbf{u}\|} \right)^T H(\mathbf{x}) \left(\frac{\mathbf{u}}{\|\mathbf{u}\|} \right) \right] > 0. \quad (20)$$

Remark. Alternatively we can prove by contradiction. Assume that for every $r > 0$, there is $\mathbf{x}_r \in B(\mathbf{x}_0, r)$ and nonzero $\mathbf{v}_r \in \mathbb{R}^N$ such that $\mathbf{v}_r^T H(\mathbf{x}_r) \mathbf{v}_r \leq 0$. Then setting $\mathbf{u}_r := \frac{\mathbf{v}_r}{\|\mathbf{v}_r\|}$ we have

$$\mathbf{u}_r^T H(\mathbf{x}_r) \mathbf{u}_r \leq 0. \quad (21)$$

But $\mathbf{u}_r \in S := \{\|\mathbf{x}\| = 1\}$ which is bounded and closed and is therefore compact. Thus there is a cluster point \mathbf{u} such that for some $r_n \rightarrow 0$, $\mathbf{u}_{r_n} \rightarrow \mathbf{u}$. Clearly $\mathbf{x}_r \rightarrow \mathbf{x}_0$. Thus

$$\mathbf{u}_{r_n}^T H(\mathbf{x}_{r_n}) \mathbf{u}_{r_n} \rightarrow \mathbf{u}^T H(\mathbf{x}_0) \mathbf{u} \implies \mathbf{u}^T H(\mathbf{x}_0) \mathbf{u} \leq 0. \quad (22)$$

Finally as $\|\mathbf{u}_{r_n}\| = 1$ for all n , $\mathbf{u} \neq 0$. Contradiction.

Question 5. *Prove*

$$a, b \geq 0, n \geq 1 \implies \left(\frac{a+b}{2}\right)^n \leq \frac{a^n + b^n}{2} \quad (23)$$

through solving $\min f(x, y) = x^n + y^n$ subject to the constraint $x + y = l > 0$.

Solution. Form the Lagrange function

$$L(x, y, \lambda) = (x^n + y^n) - \lambda(x + y - l). \quad (24)$$

Then the necessary conditions are

$$n x^{n-1} - \lambda = \frac{\partial L}{\partial x} = 0 \quad (25)$$

$$n y^{n-1} - \lambda = \frac{\partial L}{\partial y} = 0 \quad (26)$$

$$x + y - l = \frac{\partial L}{\partial \lambda} = 0. \quad (27)$$

Solving this we have $x^{n-1} = y^{n-1}$, $x + y = l > 0$. The only solution is $x = y = l/2$. Now the Hessian matrix at $\left(\frac{l}{2}, \frac{l}{2}\right)$ is $n(n-1)(l/2)^{n-2}I$ where I is the identity matrix. It is easy to check that this matrix is positive definite. Therefore $\left(\frac{l}{2}, \frac{l}{2}\right)$ is the only stationary point and a strict local minimizer.

Now we show that it is the global minimizer. Assume otherwise, that is there is $x_1 + y_1 = l$ such that $f(x_1, y_1) < f\left(\frac{l}{2}, \frac{l}{2}\right)$. Since $\left(\frac{l}{2}, \frac{l}{2}\right)$ is a strict local minimizer, the supreme between $\left(\frac{l}{2}, \frac{l}{2}\right)$ and (x_1, y_1) is reached and has to be different from both (x_1, y_1) and $\left(\frac{l}{2}, \frac{l}{2}\right)$. This point must be a local maximum and is then a stationary point, contradiction.

So we have proved

$$f(x, y) \geq f\left(\frac{l}{2}, \frac{l}{2}\right). \quad (28)$$

This means

$$x^n + y^n \geq \left(\frac{x+y}{2}\right)^n + \left(\frac{x+y}{2}\right)^n \quad (29)$$

and the conclusion follows.

Question 6. Let $f: \mathbb{R}^N \mapsto \mathbb{R}$ belong to C^2 . Let $\mathbf{x}_0 \in \mathbb{R}^N$ be a local *maximizer* for f . Prove

a) $(\text{grad } f)(\mathbf{x}_0) = \mathbf{0}$;

b) $\forall \mathbf{v} \in \mathbb{R}^N, \mathbf{v}^T H(\mathbf{x}_0) \mathbf{v} \leq 0$ where $H(\mathbf{x}_0)$ is the Hessian matrix of f at \mathbf{x}_0 .

Solution.

a) Assume $\text{grad } f \neq 0$ at \mathbf{x}_0 . Denote $\mathbf{v} := (\text{grad } f)(\mathbf{x}_0)$. Since $f \in C^2$ it is in particular differentiable at \mathbf{x}_0 and therefore

$$\frac{\partial f}{\partial \mathbf{v}} = (\text{grad } f)(\mathbf{x}_0) \cdot \mathbf{v} = \|(\text{grad } f)(\mathbf{x}_0)\|^2 > 0. \quad (30)$$

By definition

$$\lim_{h \rightarrow 0} \frac{f(\mathbf{x}_0 - h \mathbf{v}) - f(\mathbf{x}_0)}{h} > 0. \quad (31)$$

and consequently there is $\delta > 0$ such that

$$\forall |h| < \delta, \quad \frac{f(\mathbf{x}_0 - h \mathbf{v}) - f(\mathbf{x}_0)}{h} > 0 \quad (32)$$

which gives

$$\forall h \in (0, \delta), \quad f(\mathbf{x}_0 - h \mathbf{v}) > f(\mathbf{x}_0). \quad (33)$$

Now for any $r > 0$, take $\mathbf{x} = \mathbf{x}_0 - h \mathbf{v}$ with $0 < h < \min\left(\frac{r}{\|\mathbf{v}\|}, \delta\right)$. Then

$$\|h \mathbf{v}\| < r \implies \mathbf{x} \in B(\mathbf{x}_0, r) \quad (34)$$

but we have $f(\mathbf{x}) > f(\mathbf{x}_0)$. Contradiction.

- b) Assume there is $\mathbf{v} \in \mathbb{R}^N$ such that $\mathbf{v}^T H(\mathbf{x}_0) \mathbf{v} > 0$. Then since $f \in C^2$, each $h_{ij}(\mathbf{x})$ of the Hessian matrix is continuous. Consequently the function

$$g(\mathbf{x}) := \mathbf{v}^T H(\mathbf{x}) \mathbf{v} = \sum_{i,j=1}^N v_i h_{ij}(\mathbf{x}) v_j \quad (35)$$

is continuous. Thus there is $\delta > 0$ such that

$$\forall \mathbf{x} \in B(\mathbf{x}_0, \delta), \quad \mathbf{v}^T H(\mathbf{x}) \mathbf{v} > 0. \quad (36)$$

Now for any $r > 0$ consider

$$\mathbf{x} := \mathbf{x}_0 + h \mathbf{v} \quad (37)$$

with

$$h = \frac{\min(r, \delta)}{2 \|\mathbf{v}\|}. \quad (38)$$

Clearly $\mathbf{x} \in B(\mathbf{x}_0, r) \cap B(\mathbf{x}_0, \delta)$.

Taylor expansion gives

$$\begin{aligned} f(\mathbf{x}) &= f(\mathbf{x}_0) + (\text{grad } f)(\mathbf{x}_0) \cdot (\mathbf{x} - \mathbf{x}_0) + \frac{1}{2} (\mathbf{x} - \mathbf{x}_0)^T H(\boldsymbol{\xi}) (\mathbf{x} - \mathbf{x}_0) \\ &= f(\mathbf{x}_0) + \frac{h^2}{2} \mathbf{v}^T H(\boldsymbol{\xi}) \mathbf{v} \\ &> f(\mathbf{x}_0). \end{aligned} \quad (39)$$

Here the last inequality is because $\mathbf{x}, \mathbf{x}_0 \in B(\mathbf{x}_0, \delta) \implies \boldsymbol{\xi} \in B(\mathbf{x}_0, \delta) \implies \mathbf{v}^T H(\boldsymbol{\xi}) \mathbf{v} > 0$. This contradicts \mathbf{x}_0 being a local maximizer.