5.10 HW6

Question 39

1. Let $u(x,y) = \frac{x^4 + y^4}{x}$, $v(x,y) = \sin x + \cos y$ and f be a function that maps (x,y) to (u,v). Find the point (x,y) where we can solve for x,y in terms of u,v. Also, find $\frac{\partial x}{\partial u}$, $\frac{\partial x}{\partial v}$, $\frac{\partial y}{\partial u}$, $\frac{\partial y}{\partial v}$ at $f\left(\frac{\pi}{2}, \frac{\pi}{2}\right)$.

Proof. Compute

$$[df_{(x,y)}] = \begin{bmatrix} 3x^2 & \frac{4y^3}{x} \\ \cos x & -\sin y \end{bmatrix}$$

By Differentiability Theorem, it is now clear that f is continuously differentiable everywhere on its domain $\{(x,y) \in \mathbb{R}^2 : x \neq 0\}$.

Compute the determinant

$$\det\left(df_{(x,y)}\right) = -3x^2 \sin y - \frac{4y^3 \cos x}{x}$$

Now by Inverse Function Theorem, the set of points (x, y) where we can solve for x, y in terms of u, v is exactly

$$\{(x,y) \in \mathbb{R}^2 : 3x^2 \sin y + \frac{4y^3}{x} \cos x \neq 0 \text{ and } x \neq 0\}$$

Observe

$$[df_{(\frac{\pi}{2},\frac{\pi}{2})}] = \begin{bmatrix} \frac{3\pi^2}{4} & \pi^2 \\ 0 & -1 \end{bmatrix}$$

This implies the local inverse is

$$[d(f^{-1})_{f(\frac{\pi}{2},\frac{\pi}{2})}] = [df_{(\frac{\pi}{2},\frac{\pi}{2})}]^{-1} = \begin{bmatrix} \frac{4}{3\pi^2} & \frac{4}{3} \\ 0 & -1 \end{bmatrix}$$

We now have

$$\frac{\partial u}{\partial x} = \frac{4}{3\pi^2}$$
 and $\frac{\partial u}{\partial y} = \frac{4}{3}$ and $\frac{\partial v}{\partial x} = 0$ and $\frac{\partial v}{\partial y} = -1$

- 2. Let $f: \mathbb{R}^4 \to \mathbb{R}^2$ be given by $f(x, y, u, v) = (xu + yv^2, xv^3 + y^2u^6)$.
 - (a) Show that x, y can be solved in terms of u, v for (u, v) near (1, -1) and (x, y) near (1, -1).
 - (b) From (a), If we write $x = g_1(u, v)$, $y = g_2(u, v)$ for (u, v) near (1, -1) and let $g = (g_1, g_2)$, Find Dg(u, v). (You don't need to calculate explicitly.)

Proof. Compute

$$[df] = \begin{bmatrix} u & v^2 & x & 2yv \\ v^3 & 2yu^6 & 6u^5y^2 & 3xv^2 \end{bmatrix}$$

By Differentiability Theorem, it is now clear that f is continuously differentiable on \mathbb{R}^4 .

Compute

$$[df_{(1,-1,1,-1)}] = \begin{bmatrix} 1 & 1 & 1 & 2 \\ -1 & -2 & 6 & 3 \end{bmatrix}$$

It is clear that the left Jacobian $\begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix}$ is invertible. Now by Implicit Function Theorem, we can conclude (a), and if we write the implicit function by g we have

$$[dg_{(u,v)}] = -\begin{bmatrix} u & v^2 \\ v^3 & 2yu^6 \end{bmatrix}^{-1} \circ \begin{bmatrix} x & 2yv \\ 6u^5y^2 & 3xv^2 \end{bmatrix} = \begin{bmatrix} \frac{6u^5v^2y^2 - 2u^6xy}{v^5 - 2u^7y} & \frac{3v^4x - 4u^6vy^2}{v^5 - 2u^7y} \\ \frac{v^3x - 6u^6y^2}{v^5 - 2u^7y} & \frac{-3uv^2x + 2v^4y}{v^5 - 2u^7y} \end{bmatrix}$$

- 3. Let $f: \mathbb{R}^3 \to \mathbb{R}^2$ be given by $f(x, y, z) = (xe^y + ye^z, xe^z + ze^y)$.
 - (a) Show that y, z can be solved in terms of x for x near -1 and (y, z) near (1, 1).
 - (b) From (a), If we write (y, z) = g(x) for x near -1, Find g'(x). (You don't need to calculate explicitly.)

Proof. Compute

$$[df] = \begin{bmatrix} e^y & xe^y + e^z & ye^z \\ e^z & ze^y & xe^z + e^y \end{bmatrix}$$

By Differentiability Theorem, it is now clear that f is continuously differentiable on \mathbb{R}^3 . Compute

$$[df_{(-1,1,1)}] = \begin{bmatrix} e & 0 & e \\ e & e & 0 \end{bmatrix}$$

It is clear that the right Jacobian $\begin{bmatrix} 0 & e \\ e & 0 \end{bmatrix}$ is invertible. Now, by Implicit Function Theorem, we can conclude (a), and if we write the implicit function by g, we have

$$[dg_x] = -\begin{bmatrix} xe^y + e^z & ye^z \\ ze^y & xe^z \end{bmatrix}^{-1} \circ \begin{bmatrix} e^y \\ e^z \end{bmatrix}$$

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- 4. Suppose $f: \mathbb{R} \to \mathbb{R}^n$ is differentiable, such that |f(t)| = 1 for every $t \in \mathbb{R}$.
 - (a) Prove that $f'(t) \cdot f(t) = 0$. (Here "\cdot" is the standard inner product in \mathbb{R}^n)
 - (b) In fact, this result has a geometric interpretation. For example, in \mathbb{R}^2 , the function $f(t) = (\cos t, \sin t)$ satisfies |f(t)| = 1. Draw the graph of f(t) and f'(t) on \mathbb{R}^2 , what do you discover?

Proof. If we define $g: \mathbb{R} \to \mathbb{R}$ by

$$g(t) \triangleq f(t) \cdot f(t)$$

We have

g is constant 1 and g' = 0

It remains to prove

$$g'(t) = 2f'(t) \cdot f(t)$$

Observe

$$g(t) = \sum_{k=1}^{n} f_k^2(t)$$

This implies

$$g'(t) = \sum_{k=1}^{n} 2f_k(t)f'_k(t) = 2f'(t) \cdot f(t)$$
 (done)

From the graph below, we see that f(t) and f'(t) orthogonal.

Question 43

- 5. (Second Derivative Test) Let V be an open subset of \mathbb{R}^2 and $(a, b) \in V$, and suppose that $f: V \to \mathbb{R}$ satisfy $\nabla f(a, b) = 0$. Suppose also that $f \in \mathcal{C}^2(V)$, and set $D = f_{xx}(a, b) f_{yy}(a, b) f_{xy}(a, b)^2$. Prove the following statements:
 - (a) If D > 0 and $f_{xx}(a, b) > 0$, then f admits a local minimum at (a, b).
 - (b) If D > 0 and $f_{xx}(a, b) < 0$, then f admits a local maximum at (a, b).
 - (c) If D < 0, then f is a saddle point at (a, b).

Definition 5.10.1. (Definition of Hessian) For all $p \in V$, we define

$$H(p) \triangleq \begin{bmatrix} f_{xx}(p) & f_{xy}(p) \\ f_{yx}(p) & f_{yy}(p) \end{bmatrix}$$
 and $D(p) \triangleq (f_{xx} \cdot f_{yy} - f_{xy}^2)(p)$

Note that $f \in \mathcal{C}^2(V) \implies f_{xy} = f_{yx}$ on $V \implies H$ is symmetric on V.

Then by Spectral Theorem, we know

 ${\cal H}$ has an orthonormal basis consisting of eigenvectors

Definition 5.10.2. (Definition of Saddle) Given real-valued function f exists on an open set $V \subseteq \mathbb{R}^2$ containing p and is differentiable at p, we say f is a saddle point at p if

 $df_p = 0$ and f is neither a local minimum nor a local maximum at p

Lemma 5.10.3. Given a $\alpha(t)$ that lies in V and satisfy $\alpha'' = 0$. We have

$$(f \circ \alpha)'' = \alpha' \cdot H\alpha'$$

Proof. Using Chain Rule and the property of gradient, we can write

$$(f \circ \alpha)' = \nabla f(\alpha) \cdot \alpha'$$

This give us

$$(f \circ \alpha)'' = (\nabla f(\alpha))' \cdot \alpha' + \nabla f(\alpha) \cdot \alpha''$$
$$= (\nabla f(\alpha))' \cdot \alpha'$$

This reduce the problem into proving

$$(\nabla f(\alpha))' \cdot \alpha' = \alpha' \cdot H\alpha'$$

Express

$$\nabla f(\alpha) = ((\partial_x f)(\alpha), (\partial_y f)(\alpha))$$

Compute

$$(\nabla f(\alpha))' = (\nabla(\partial_x f)(\alpha) \cdot \alpha', \nabla(\partial_y f)(\alpha) \cdot \alpha')$$

Note that

$$\nabla(\partial_x f) = (\partial_{xx} f, \partial_{xy} f)$$
 and $\nabla(\partial_y f) = (\partial_{yx} f, \partial_{yy} f)$

Express $\alpha = (\alpha_1, \alpha_2)$. Compute

$$\nabla(\partial_x f)(\alpha) \cdot \alpha' = \partial_{xx} f(\alpha) \cdot \alpha'_1 + \partial_{xy} f(\alpha) \cdot \alpha'_2$$

$$\nabla(\partial_y f)(\alpha) \cdot \alpha' = \partial_{yx} f(\alpha) \cdot \alpha'_1 + \partial_{yy} f(\alpha) \cdot \alpha'_2$$

We can now compute

$$(\nabla f(\alpha))' \cdot \alpha' = ((\alpha'_1)\partial_{xx}f(\alpha) + (\alpha'_2)\partial_{xy}f(\alpha), (\alpha_1)'\partial_{yx}f(\alpha) + (\alpha'_2)\partial_{yy}f(\alpha)) \cdot \alpha'$$

$$= (\alpha'_1)^2\partial_{xx}f(\alpha) + 2(\alpha'_1\alpha'_2)\partial_{xy}f(\alpha) + (\alpha'_2)^2\partial_{yy}f(\alpha)$$

$$= \alpha' \cdot H\alpha' \text{ (done)}$$

Lemma 5.10.4. (Property of H)

- (a) If D > 0 and $f_{xx} > 0$, then H is positive definite.
- (b) If D > 0 and $f_{xx} < 0$, then H is negative definite.
- (c) If D < 0, then H has a positive and a negative eigenvalue.

Proof. Observe that det(H) = D. This let us write the characteristic polynomial of H by

$$t^2 - (\partial_{xx}f + \partial_{yy}f)t + D$$

Then we know the eigenvalues are

$$t = \frac{(\partial_{xx}f + \partial_{yy}f) \pm \sqrt{(\partial_{xx}f + \partial_{yy}f)^2 - 4D}}{2}$$

Now, if

$$\partial_{xx} f \partial_{yy} f - (\partial_{xy} f)^2 = D > 0$$

Then we have

$$\partial_{xx} f \partial_{yy} f > 0$$

Moreover, if $\partial_{xx}f > 0$, we can deduce

$$\partial_{xx}f + \partial_{yy}f > 0$$

Then from the formula of the eigenvalues t, because D > 0, regardless of the sign of the square-root term, we have

Note that $t \in \mathbb{R}$ by Spectral Theorem.

Similarly, if D > 0 and $\partial_{xx} f < 0$, then t < 0, and if D < 0, then the two eigenvalues are of opposite signs and non-zero.

Now, if D > 0 and $f_{xx} > 0$, the two eigenvalues of H are all positive. Fix the orthonormal basis $\{e_1, e_2\}$ corresponding to eigenvalues $t_1, t_2 > 0$. We see that for all $c_1e_1 + c_2e_2 \neq 0 \in \mathbb{R}^2$

$$(c_1e_1 + c_2e_2) \cdot H(c_1e_2 + c_2e_2) = t_1c_1^2 + t_2c_2^2 > 0$$

Similarly, if D > 0 and $f_{xx} < 0$, the two eigenvalues of H are negative, which leads to

$$(c_1e_1 + c_2e_2) \cdot H(c_1e_1 + c_2e_2) = t_1c_1^2 + t_2c_2^2 < 0$$

Proof. (D > 0 and $\partial_{xx} f(a, b) > 0$ Case) Because $f \in \mathcal{C}^2(V)$, we know there exists $B_{\epsilon}(a, b)$ such that D and $\partial_{xx} f$ are positive on $B_{\epsilon}(a, b)$. We claim

$$f(a,b) = \min_{(x,y)\in B_{\epsilon}(a,b)} f(x,y)$$

Fix $(c,d) \in B_{\epsilon}(a,b)$. We reduce the problem into proving

$$f(a,b) \le f(c,d)$$

Let $\gamma: [-1,1] \to B_{\epsilon}(a,b)$ be the line joining (2a-c,2b-d) and (c,d)

$$\gamma(t) \triangleq \left(a + (c - a)t, b + (d - b)t\right)$$

Note that $\gamma'' = 0$ and $\gamma(0) = (a, b)$. Now by Lemma 5.10.3 and Lemma 5.10.4, we have $(f \circ \gamma)'' > 0$ on [-1, 1]

This implies $f \circ \gamma$ is convex on [-1,1]. Compute using $\nabla f(a,b) = 0$

$$(f \circ \gamma)'(0) = df_{(a,b)}(c - a, d - b) = \nabla f(a,b) \cdot (c - a, d - b) = 0$$

This then implies

$$(f \circ \gamma)(0)$$
 is minimum on $[-1, 1]$ (done)

Proof. (D > 0 and $\partial_{xx} f(a, b) < 0$ Case) Because $f \in \mathcal{C}^2(V)$, we know there exists $B_{\epsilon}(a, b)$ such that D is positive and $\partial_{xx} f$ is negative on $B_{\epsilon}(a, b)$. We claim

$$f(a,b) = \max_{(x,y)\in B_{\epsilon}(a,b)} f(x,y)$$

Fix $(c,d) \in B_{\epsilon}(a,b)$. We reduce the problem into proving

$$f(a,b) \ge f(c,d)$$

Let $\gamma: [-1,1] \to B_{\epsilon}(a,b)$ be the line joining (2a-c,2b-d) and (c,d)

$$\gamma(t) \triangleq \left(a + (c - a)t, b + (d - b)t\right)$$

Note that $\gamma'' = 0$ and $\gamma(0) = (a, b)$. Now by Lemma 5.10.3 and Lemma 5.10.4, we have $(f \circ \gamma)'' < 0$ on [-1, 1]

This implies $f \circ \gamma$ is concave on [-1,1]. Compute using $\nabla f(a,b) = 0$

$$(f \circ \gamma)'(0) = df_{(a,b)}(c - a, d - b) = \nabla f(a,b) \cdot (c - a, d - b) = 0$$

This then implies

$$(f \circ \gamma)(0)$$
 is maximum on $[-1, 1]$ (done)

Proof. (D < 0 Case) From $\nabla f(a,b) = 0$, we know $df_{(a,b)} = 0$. This reduce the problem into proving

$$(a,b)$$
 is not a local extremum

Fix ϵ such that D < 0 on $B_{\epsilon}(a, b)$. We prove

$$\min_{(x,y)\in B_{\epsilon}(a,b)} f(x,y) < f(a,b) < \max_{(x,y)\in B_{\epsilon}(a,b)} f(x,y)$$

Fix two curve $\alpha, \beta: (-\epsilon, \epsilon) \to \mathbb{R}^2$ by

$$\alpha(t) \triangleq (a, b) + te_1 \text{ and } \beta(t) \triangleq (a, b) + te_2$$

where e_1, e_2 are the eigenvectors of H(p), e_1 correspond to the positive eigenvalue and e_2 correspond to the negative eigenvalue.

Compute using Lemma 5.10.4

$$(f \circ \alpha)''(0) > 0$$
 and $(f \circ \beta)''(0) < 0$

Because $(f \circ \alpha)''$ and $(f \circ \beta)''$ is continuous, we know $(f \circ \alpha)'' > 0$ and $(f \circ \beta)'' < 0$ on a small enough interval I containing 0. In other words

 $(f\circ\alpha)$ is strictly convex and $(f\circ\beta)$ is strictly concave on I

Again, we know $(f \circ \alpha)'(0) = \nabla f(a, b) \cdot e_1 = 0$ and, similarly, $(f \circ \beta)'(0) = 0$. We can now deduce

$$\begin{cases} f \circ \alpha \\ f \circ \beta \end{cases} \quad \text{reach} \quad \begin{cases} \text{strict minimum} \\ \text{strict maximum} \end{cases} \quad \text{at } p \text{ on } I$$

The former implies $f(a,b) < \max_{(x,y) \in B_{\epsilon}(a,b)} f(x,y)$, and the latter implies $f(a,b) > \min_{(x,y) \in B_{\epsilon}(a,b)} f(x,y)$. (done)

6. Let f be a nonnegative and measurable function on E. Prove that

$$\int_{E} f(x) dx = \sup \left[\sum_{j} \left(\inf_{x \in E_{j}} f(x) \right) |E_{j}| \right]$$

where the supremum is taken over all decompositions $E = \bigcup_{j} E_{j}$ of E into the union of a finite number of disjoint measurable sets E_{j} .

Proof. We first show

$$\int_{E} f(x)dx \ge \sup \left[\sum_{j} \left(\inf_{x \in E_{j}} f(x) \right) |E_{j}| \right]$$

Fix a finite disjoint measurable decomposition E_i of E. We reduce the problem into proving

$$\int_{E} f(x)dx \ge \sum_{j} \left(\inf_{x \in E_{j}} f(x) \right) |E_{j}|$$

Define

$$s_0 \triangleq \sum_{i} \left(\inf_{x \in E_j} f(x) \right) \mathbf{1}_{E_j}$$

Because E_j are disjoint, we know

$$s_0(x) = \inf_{t \in E_j} f(t) \text{ if } x \in E_j$$

This implies s_0 is simple. Because E_j , are all measurable, we can deduce s_0 is measurable. In conclusion, $s_0 : E \to \mathbb{R}$ is a simple measurable function.

Given arbitrary $x \in E$, we know x must belong to one and only one of the decomposition. Express $x \in E_i$. This give us

$$s_0(x) = \inf_{t \in E_i} f(t) \le f(x)$$

Because x is arbitrary, we can conclude $s \leq f$ on E.

7. If $\{f_k\}_{k\in\mathbb{N}}$ is a sequence of nonnegative and measurable functions on E, prove that

$$\int_{E} \left(\sum_{k=1}^{\infty} f_k(x) \right) dx = \sum_{k=1}^{\infty} \left(\int_{E} f_k(x) dx \right)$$

Proof. Fix

$$f_n \triangleq \sum_{k=1}^n f_k$$
 and $f \triangleq \sum_{k=1}^\infty f_k$

Because f_k are non-negative, we know f_n are pointwise increasing. Clearly $f_n \to f$ pointwise. Note that f_n are all measurable, since measurable functions are closed under addition.

Now by Lebesgue monotone convergence Theorem, we have

$$\int_{E} \left(\sum_{k=1}^{\infty} f_{k}\right) dx = \int_{E} f dx = \lim_{n \to \infty} \int_{E} f_{n} dx = \lim_{n \to \infty} \int_{E} \sum_{k=1}^{n} f_{k} dx = \lim_{n \to \infty} \sum_{k=1}^{n} \int_{E} f_{k} dx$$
$$= \sum_{k=1}^{\infty} \int_{E} f_{k}(x) dx$$

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8. Let f be nonnegative and measurable on E, prove that

$$\int_{E} f(x) dx = 0 \iff \exists Z \subset E \text{ such that } |Z| = 0 \text{ and } f(x) = 0 \text{ on } E \backslash Z$$

Proof. Note that if |E| = 0, then $\int_E f dx = 0$, and we can just let Z = E. We suppose |E| > 0 from now.

 (\longleftarrow)

Decompose E into arbitrary disjoint finite measurable union E_j . Because the decomposition is arbitrary, using the result of the sixth question, we only have to prove

$$\sum_{j} (\inf_{E_j} f) |E_j| = 0$$

We reduce the problem into proving

$$\forall j, |E_j| > 0 \implies \inf_{E_j} f = 0$$

Fix $|E_j| > 0$. Notice that f = 0 on Z^c . This allow us to reduce the problem into proving $|E_i \cap Z^c| > 0$

Note that $|Z^c| = |E| - |Z| = |E|$. Observe

$$|E| \ge |E_j \cup Z^c| \ge |Z^c| = |E|$$

This implies $|E_j \cup Z^c| = |E|$. This give us

$$|E_j| + |Z^c| - |E_j \cap Z^c| = |E_j \cup Z^c| = |E|$$

Now because $|Z^c| = |E|$, we can deduce

$$|E_j \cap Z^c| = |E_j| > 0$$
 (done)

 (\longrightarrow)

Fix

$$B_n \triangleq \{x \in E : f(x) \ge \frac{1}{n}\}$$

Because f is measurable, we know B_n are measurable. Note that each B_n must be of zero-measure, otherwise $\int_E f \geq \int_{B_n} f \geq \frac{|B_n|}{n} > 0$.

Now let

$$Z \triangleq \bigcup_{n \in \mathbb{N}} B_n$$

Observe

$$|Z| \le \sum_{n \in \mathbb{N}} |B_n| = 0$$

and observe

$$x \in Z^c \implies \forall n \in \mathbb{N}, f(x) \notin B_n \implies \forall n \in \mathbb{N}, 0 \le f(x) < \frac{1}{n} \implies f(x) = 0$$

9. Let
$$g(x) = \begin{cases} 0, & 0 \le x \le 1/2 \\ 1, & 1/2 < x \le 1 \end{cases}$$
 and $f_{2k}(x) = g(x), f_{2k-1}(x) = g(1-x), x \in [0, 1],$ $k \in \mathbb{N}$. Show that

$$\liminf_{n\to\infty} f_n(x) = 0 \text{ on } [0, 1]$$

but $\int_0^1 f_n(x) dx = \frac{1}{2}$. (This gives an example of strict inequality in **Fatou lemma**)

Proof. Compute

$$f_{2k}(x) = \begin{cases} 0 & \text{if } x \in [0, \frac{1}{2}] \\ 1 & \text{if } x \in (\frac{1}{2}, 1] \end{cases} \text{ and } f_{2k-1}(x) = \begin{cases} 1 & \text{if } x \in [0, \frac{1}{2}) \\ 0 & \text{if } x \in [\frac{1}{2}, 1] \end{cases}$$

Observe

For all
$$x \in [0, \frac{1}{2}), \{f_{2k}(x)\}_{k \in \mathbb{N}} = 0 \implies \liminf_{n \to \infty} f_n = 0 \text{ on } [0, \frac{1}{2})$$

Observe

$$\{f_k(x)\}_{k\in\mathbb{N}} = 0 \implies \liminf_{n\to\infty} f_n = 0 \text{ on } \{\frac{1}{2}\}$$

Observe

For all
$$x \in (\frac{1}{2}, 1], \{f_{2k-1}(x)\}_{k \in \mathbb{N}} = 0 \implies \liminf_{n \to \infty} f_n = 0 \text{ on } (\frac{1}{2}, 1]$$

This now conclude

$$\liminf_{n\to\infty} f_n = 0 \text{ on } [0,1]$$

Compute

$$\int_0^1 f_{2k} dx = \int_{\frac{1}{2}}^1 1 dx = \frac{1}{2}$$

Compute

$$\int_0^1 f_{2k-1} dx = \int_0^{\frac{1}{2}} 1 dx = \frac{1}{2}$$

This conclude $\int_0^1 f_n dx = \frac{1}{2}$ for all $n \in \mathbb{N}$.

10. Let
$$f(x) = \begin{cases} 1/n & , |x| \leq n \\ 0 & , |x| > n \end{cases}$$
. Show that $f_n(x)$ uniformly converges to 0 on \mathbb{R} but
$$\int_{-\infty}^{\infty} f_n(x) \, dx = 2 \text{ for all } n \in \mathbb{N}.$$

Proof. Fix ϵ . Let $N > \frac{1}{\epsilon}$. We now see that for each n > N, we have

$$\sup_{\mathbb{R}} |f| = \frac{1}{n} < \epsilon$$

This conclude that f_n uniformly converge to \mathbb{R} .

Compute

$$\int_{\mathbb{R}} f_n(x)dx = \int_{-n}^n \frac{1}{n}dx = 2$$