



MATH20060

Calculus of Several Variables

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Introduction

These notes are for the course MATH20060, taken at the University College Dublin(UCD), during the months of September to December. The title of the course is Calculus of Several Variables, colloquially known as Calculus 3, or Multi-variable Calculus. These notes have been rewritten in \LaTeX from my own hand-written notes. If you have any questions or feedback, feel free to contact me via my email.

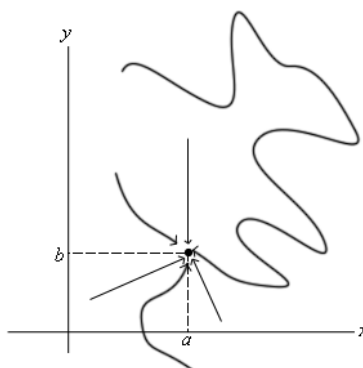
Contents

1	Limits with Several Variables	2
1.1	Two Path Test	3
2	Domains of Multi-Variable Functions	4
3	Level Curves	6
4	Partial Differentiation	6
4.1	Higher Order Partial Derivatives	8
4.2	Tangent Planes	9
4.3	Normal Lines	10
5	Linear Approximation	10
6	Chain Rule (2 variables)	13
6.1	Chain Rule (n variables)	14
7	Directional Derivatives	16
8	The Gradient	18
9	Local Extrema	21
9.1	Second Derivative Test (for functions of 2 variables)	22
9.1.1	Critical Points	22

10 Method of Lagrange Multipliers	24
10.1 Lagrange Multipliers with 2 constraints	26
11 Integration	26
11.1 Line Integrals	26
11.2 Potential Functions	28
11.3 Double Integrals	29
11.3.1 Iterated Integration	29
11.4 X and Y Simple functions	30
11.5 Changing the Order of Integration	31
11.6 Green's Theorem	32

1 Limits with Several Variables

Recall that in 1 variable calculus, we can check either the right or left hand limit. This concept remains the same for limits of several variables in the sense that every path to the point (x, y) , should approach the same value. However with functions of several variables, there are infinite paths that can be taken to approach a point (x, y) .



We can use the same concept from single variable calculus to find the limit. A function is continuous at the point (a, b) if

$$\lim_{(x,y) \rightarrow (a,b)} f(x, y) = f(a, b) \quad (1.0.1)$$

Plugging in the point into the function, $f(x, y)$, will return the value of the limit. You just need to be careful for division by zero, square roots of negative numbers, logarithms of zero or negative numbers, etc. Additionally, if two different paths approach different values for the limit, it can be concluded that the limit does not exist (DNE).

Examples:

1.

$$\begin{aligned} \lim_{(x,y) \rightarrow (3,-4)} (x^2 + y^2) \\ \lim_{x \rightarrow 3} x^2 + \lim_{y \rightarrow -4} y^2 = 25 \end{aligned} \quad (1.0.2)$$

2.

$$\lim_{(x,y) \rightarrow (3,-4)} \sqrt{x^2 + y^2} = \sqrt{25} = 5 \quad (1.0.3)$$

3.

$$\lim_{(x,y) \rightarrow (0,1)} \frac{x - xy + 3}{x^2y + 5xy - y^3} = \frac{0 - (0)(1) + 3}{(0)^2(1) + 5(0)(1) - (1)^3} = -3 \quad (1.0.4)$$

4. Show that the function $f(x, y) = \frac{x^2y}{x^2+y^2}$ has limit 0 at (0,0). $|f(x, y) - \ell|$ becomes arbitrarily small as $(x, y) \rightarrow (0, 0)$.

$$|f(x, y) - \ell| = |f(x, y) - 0| = \left| \frac{x^2y}{x^2+y^2} \right| = \frac{\overbrace{x^2}^{\leq 1}}{x^2+y^2} \cdot |y| \leq |y| \rightarrow 0 \text{ as } (x, y) \rightarrow (0, 0) \quad (1.0.5)$$

So $f(x, y) \rightarrow 0$ as $(x, y) \rightarrow (0, 0)$.

5.

$$f(x, y) = \frac{x^2y - y^3}{x^2 + y^2} \quad (1.0.6)$$

Show that the function $f(x, y)$ has a limit at (0,0). If f has a limit at (0,0) then it must have this limit along any path approaching (0,0).

On the x-axis we have $f(x, 0) = \frac{0}{x^2} = 0$. The limit of f (if it exists) must be zero. The function can be rewritten as

$$\begin{aligned} f(x, y) &= g(x, y) - h(x, y) \\ g(x, y) &= \frac{x^2y}{x^2 + y^2} \\ h(x, y) &= \frac{y^3}{x^2 + y^2} \\ g(x, y) &\rightarrow 0 \text{ as } (x, y) \rightarrow 0 \\ |h(x, y) - 0| &= \left| \frac{y^3}{x^2 + y^2} \right| = \frac{\overbrace{y^2}^{\leq 1}}{x^2 + y^2} \cdot |y| \leq |y| \rightarrow 0 \text{ as } (x, y) \rightarrow (0, 0) \\ f(x, y) &= g(x, y) - h(x, y) \rightarrow 0 - 0 = 0 \text{ as } (x, y) \rightarrow (0, 0) \end{aligned} \quad (1.0.7)$$

1.1 Two Path Test

If $f(x, y)$ has different limits along 2 different paths approaching (x, y) then the limit DNE.

1.

$$f(x, y) = \frac{xy}{x^2 + y^2} \quad (1.1.1)$$

Show that the function's limit does not exist at (0,0). For the limit of f to exist at (0,0), the limit along every curve to (0,0) must exist, and all limits must be equal.

Along the x-axis, $(x, 0)$, we get $f(x, 0) = \frac{x \cdot 0}{x^2 + 0} = 0$.

Along the y-axis, $(0, y)$, we get $f(0, y) = \frac{0 \cdot y}{0 + y^2} = 0$.

Along the line $y = x$, we get $f(x, x) = \frac{x^2}{2x^2} = \frac{1}{2} \neq 0$. \therefore the limit of $f(x, y)$ at (0,0) DNE.

2.

$$f(x, y) = \frac{2x^2y}{x^4 + y^2} \quad (1.1.2)$$

Does the function $f(x, y)$ have a limit at (0,0)?

Along a line, $y = mx$, we get:

$$f(x, mx) = \frac{2mx^3}{x^4 + m^2x^2} = \frac{2mx}{x^2 + m^2} \rightarrow 0 \text{ as } x \rightarrow 0 \quad (1.1.3)$$

Along the curve, $y = kx^2$, we get:

$$f(x, kx^2) = \frac{2kx^4}{x^4 + k^2x^4} = \frac{2k}{1 + k^2} \rightarrow \frac{2k}{1 + k^2} \quad (1.1.4)$$

This limit varies depending on k , \therefore the limit DNE by the two-path test.

2 Domains of Multi-Variable Functions

D is a subset of \mathbb{R}^2 . A function $f(x, y)$ assigns each pair (x, y) in D a value $z = f(x, y) \in \mathbb{R}$. The domain, D , is all pairs (x, y) allowable as input in $f(x, y)$. The set of all outputs, $f(x, y)$, is the range. D is usually represented graphically with the form:

$$D = \{(x, y) : x, y\} \quad (2.0.1)$$

With conditional statements in the space for x and y which keep the function $f(x, y)$ true.

1.

$$\begin{aligned} f(x, y) &= \sqrt{xy} \\ D &= \{(x, y) : xy \geq 0\} \end{aligned} \quad (2.0.2)$$

2.

$$\begin{aligned} f(x, y) &= \frac{x}{x^2 - y^2} \\ D &= \{(x, y) : y > x, y < -x, y \geq 0\} \end{aligned} \quad (2.0.3)$$

3.

$$\begin{aligned} f(x, y, z) &= \frac{1}{x^2 + e^{yz}} \\ D &= \{(x, y, z) : \mathbb{R}^3\} \end{aligned} \quad (2.0.4)$$

4.

$$f(x, y) = \ln\left(1 - \frac{x^2}{y}\right)$$

$$D = \{(x, y) : \frac{x^2}{y} > 1, y \neq 0\}$$
(2.0.5)

5.

$$f(x, y) = \frac{1}{\sqrt{x^2 + y^2 - 9}}$$

$$D = \{(x, y) : \|(x, y)\| > 3\}$$
(2.0.6)

6.

$$f(x, y) = \frac{1}{\sqrt{x^2 + 2y^2 - y}}$$

$$D = \{(x, y) : x > 0 \& y > 0, \sqrt{x^2 + 2y^2} > y\}$$
(2.0.7)

The domain can be split into ∂D , the outer domain, and D° , the inner domain. The two domains create the full domain \overline{D} . Depending on the nature of the domain, the domain can be determined to be open or closed.

7.

$$D = \{(x, y) : x \geq 0, y > 0\}$$

$$\partial D = \{(x, y) : x = 0\}$$

$$D^\circ = \{(x, y) : x > 0, y > 0\}$$

$$\overline{D} = \{(x, y) : x \geq 0, y > 0\}$$

$$D \text{ is open}$$
(2.0.8)

8.

$$D = \{(x, y) : x + y = 1\}$$

$$\partial D = \{(x, y) : x + y = 1\}$$

$$D^\circ = \{(x, y) : x + y = 1\}$$

$$\overline{D} = \{(x, y) : x + y = 1\}$$

$$D \text{ is closed}$$
(2.0.9)

9.

$$D = \{(x, y) : 0 \leq x \leq 1\}$$

$$\partial D = \{(x, y) : x = 0, x = 1\}$$

$$D^\circ = \{(x, y) : 0 < x < 1\}$$

$$\overline{D} = \{(x, y) : 0 \leq x \leq 1\}$$

$$D \text{ is closed}$$
(2.0.10)

10.

$$D = \{(x, y) : x^2 + y^2 < 1, y < 0\}$$

$$\partial D = \{(x, y) : \}$$

$$D^\circ = \{(x, y) : x^2 + y^2 < 1, y < 0\}$$

$$\overline{D} = \{(x, y) : x^2 + y^2 < 1, y < 0\}$$

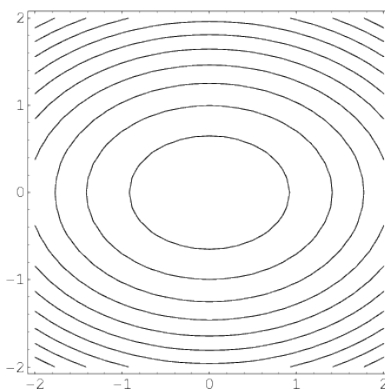
$$D \text{ is open}$$
(2.0.11)

3 Level Curves

Graphs of functions with the form $z = f(x, y)$ are called level curves, and level surfaces for the form $F(x, y, z) = c$. The function $z = f(x, y)$ has exactly one z value for (x, y) in the domain of $f(x, y)$, and has no value for (x, y) outside the domain of $f(x, y)$. The graph of the function $z = f(x, y)$ must pass the vertical line test.

$$\begin{aligned} f(x, y) &= -x^2 - 2y^2 \\ c &= -x^2 - 2y^2 \\ 1 &= \frac{x^2}{-c} + \frac{2y^2}{-c} \end{aligned} \tag{3.0.1}$$

The level curves for the function $f(x, y)$ can be graphed for $c = -1, -2, -3, \dots, -10$.



A level curve $f(x, y) = c$ can be thought of as a horizontal slice of the graph at height $z = c$. This slice is the intersection of the plane $z = c$ and the function $f(x, y)$.

4 Partial Differentiation

Recall one-variable differentiation, $f : \mathbb{R} \rightarrow \mathbb{R}$. Include the graphic from the notes.

Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$. Given (x_0, y_0) , the vertical plane $y = y_0$ cuts the surface $z = f(x, y)$ in the curve $z = f(x, y_0)$ (cross section of the graph). The partial derivative $\frac{\partial f}{\partial x}$ of f with respect to x at (x_0, y_0) is the (ordinary) derivative of the function $x \rightarrow f(x, y_0)$ at $x = x_0$. Geometrically, $\frac{\partial f}{\partial x}$ is the slope of the line tangent to the trace of the graph of f in the plane $y = y_0$. Similarly, $\frac{\partial f}{\partial y}$ is the slope of the line tangent to the trace of the graph of f in the plane $x = x_0$.

The Partial Derivative of $f(x, y)$ with respect to x at (x_0, y_0) is the number

$$\frac{\partial f}{\partial x} = \lim_{h \rightarrow 0} \frac{f(x_0 + h, y_0) - f(x_0, y_0)}{h} \tag{4.0.1}$$

Provided that the limit exists

Similarly, the Partial Derivative of $f(x, y)$ with respect to y at (x_0, y_0) is the number

$$\frac{\partial f}{\partial y} = \lim_{h \rightarrow 0} \frac{f(x_0, y_0 + h) - f(x_0, y_0)}{h} \quad (4.0.2)$$

Provided that the limit exists

The notation f_x and f_y is often used for $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$, respectively.

Calculation of f_x involves treating y as a constant and then differentiating with respect to x . This can usually be done with standard rules of differentiating.

$$\begin{aligned} f(x, y) &= x^2 y \\ f_x &= 2xy \\ f_y &= x^2 \end{aligned} \quad (4.0.3)$$

1.

$$\begin{aligned} f(x, y) &= x^2 y^3 + e^x \ln(y) \\ f_x &= 2xy^3 + e^x \ln(y) \\ f_y &= 3x^2 y^2 + \frac{e^x}{y} \\ f_x(4, 1) &= 8 \\ f_y(4, 1) &= 48 + e^4 \end{aligned} \quad (4.0.4)$$

2.

$$\begin{aligned} f(x, y) &= \sqrt{9 - x^2 - y^2} \\ f_x &= \frac{-x}{\sqrt{9 - x^2 - y^2}} \\ f_y &= \frac{-y}{\sqrt{9 - x^2 - y^2}} \\ f_x(1, 2) &= -\frac{1}{2} \\ f_y(1, 2) &= -1 \end{aligned} \quad (4.0.5)$$

3.

$$f(x, y) = \begin{cases} x^2 + y^2 + z^2 = 9 \\ z \geq 0 \end{cases} \quad (4.0.6)$$

If we're on the surface at (1,2,2), then the slope in the x-direction is $-\frac{1}{2}$, and in the y-direction is -1 .

If $f(x, y)$ is a differentiable function of one variable, show that the function $g(x, y) = f\left(\frac{x}{y}\right)$ satisfies the "partial differential equation" $x \frac{\partial g}{\partial x} + y \frac{\partial g}{\partial y} = 0$. By the chain rule:

$$\begin{aligned} g_x &= f' \left(\frac{x}{y} \right) \left(\frac{1}{y} \right) \\ g_y &= f' \left(\frac{x}{y} \right) \left(-\frac{x}{y^2} \right) \\ xg_x + yg_y &= f' \left(\frac{x}{y} \right) \left(\frac{x}{y} \right) - f' \left(\frac{x}{y} \right) \left(\frac{x}{y} \right) \end{aligned} \quad (4.0.7)$$

Partial derivatives are define similarly for functions of more than two variables. In this case, we treat all but one of the variables as constant.

$$\begin{aligned}
 f(x, y, z) &= \sqrt{x} e^{\frac{y}{z}} \\
 f_x &= \frac{e^{\frac{y}{z}}}{2\sqrt{x}} \\
 f_y &= \frac{\sqrt{x}}{z} e^{\frac{y}{z}} \\
 f_z &= \sqrt{x} \cdot e^{\frac{y}{z}} \left(-\frac{y}{z^2} \right)
 \end{aligned} \tag{4.0.8}$$

4.1 Higher Order Partial Derivatives

Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$, then f_x and f_y are also functions from $\mathbb{R}^2 \rightarrow \mathbb{R}$, so we can consider partial derivatives of them:

$$\begin{aligned}
 \frac{\partial^2 f}{\partial x^2} &= \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial x} \right), & f_{xx} &= (f_x)_x \\
 \frac{\partial^2 f}{\partial y^2} &= \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial y} \right), & f_{yy} &= (f_y)_y \\
 \frac{\partial^2 f}{\partial x \partial y} &= \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial y} \right), & f_{yx} &= (f_y)_x \\
 \frac{\partial^2 f}{\partial y \partial x} &= \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial x} \right), & f_{xy} &= (f_x)_y
 \end{aligned} \tag{4.1.1}$$

f_{xy} and f_{yx} are called mixed partial derivatives and f_{xx} and f_{yy} are called pure partial derivatives. We can go further:

$$\frac{\partial^3 f}{\partial x^3} = \frac{\partial}{\partial x} \left(\frac{\partial^2 f}{\partial x^2} \right) \tag{4.1.2}$$

If $f(x, y)$ and all its first and second order partial derivatives are continuous on an open set, then $f_{xy} = f_{yx}$ there.

$$f(x, y) = x^2 y^3 + \cos(x) \sin(y) \tag{4.1.3}$$

$$\begin{aligned}
 f_x &= 2xy^3 - \sin(x) \sin(y) & f_y &= 3x^2 y^2 + \cos(x) \cos(y) \\
 f_{xx} &= 2y^3 - \cos(x) & f_{yy} &= 6x^2 y - \cos(x) \sin(y) \\
 f_{xy} &= 6xy^2 - \sin(x) \cos(y) & f_{yx} &= 6xy^2 - \sin(x) \cos(y)
 \end{aligned}$$

Find f_{yx} of $f(x, y)$

$$\begin{aligned}
 f(x, y) &= x \left(y + \frac{e^y}{y^2 + 1} \right) \\
 f_x &= y + \frac{e^y}{y^2 + 1} \\
 f_{xy} &= 1 + \frac{e^y(y^2 + 1) - e^y(2y)}{(y^2 + 1)^2} \\
 f_{xy} &= f_{yx}
 \end{aligned} \tag{4.1.4}$$

4.2 Tangent Planes

For a function f of one variable, knowledge of $f(x_0)$ and $f'(x_0)$ allows us to write an equation for the line tangent to the graph of f at x_0 :

$$y - f(x_0) = f'(x_0)(x - x_0) \tag{4.2.1}$$

$f_x(x_0, y_0)$ gives you the slope of the line tangent to the trace of the graph in the plane $y = y_0$:

$$U_x = \underline{i} + f_x(x_0, y_0) \cdot \underline{k} \tag{4.2.2}$$

$f_y(x_0, y_0)$ gives the slope of the line tangent to the trace of the graph in the plane $x = x_0$:

$$U_y = \underline{j} + f_y(x_0, y_0) \cdot \underline{k} \tag{4.2.3}$$

The plane with normal vector \underline{n} can be found:

$$\begin{aligned}
 \underline{n} &= U_y \times U_x = \det \begin{pmatrix} \underline{i} & \underline{j} & \underline{k} \\ 0 & 1 & f_y(x_0, y_0) \\ 1 & 0 & f_x(x_0, y_0) \end{pmatrix} \\
 \underline{n} &= (f_x(x_0, y_0) - 0) \cdot \underline{i} + f_y(x_0, y_0) \cdot \underline{j} - 1 \cdot \underline{k}
 \end{aligned} \tag{4.2.4}$$

$z_0 = f(x_0, y_0)$ and $P_0 = (x_0, y_0, z_0)$. The plane with normal vector \underline{n} that contains P_0 consists of all points $P = (x, y, z)$ satisfying $\underline{n} \cdot \overrightarrow{P_0 P} = 0$.

$$\begin{aligned}
 \begin{pmatrix} f_x(x_0, y_0) \\ f_y(x_0, y_0) \\ -1 \end{pmatrix} \cdot \begin{pmatrix} x - x_0 \\ y - y_0 \\ z - z_0 \end{pmatrix} &= 0 \\
 f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) - (z - z_0) &= 0
 \end{aligned} \tag{4.2.5}$$

The equation for a plane Z can be found:

$$Z = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0) \tag{4.2.6}$$

1.

$$\begin{aligned}
 f(x, y) &= x^2 + 4y^2 \\
 f_x &= 2x \quad f_y = 8y
 \end{aligned} \tag{4.2.7}$$

Find the equation for the plane tangent to the graph $f(x, y)$ at the point (2,1):

$$\begin{aligned} f(2, 1) &= 8 \\ f_x(2, 1) &= 4 \\ f_y(2, 1) &= 8 \\ Z &= 4x - 8 + 8y - 8 + 8 \\ Z &= 4x - 8 + 8y \end{aligned} \tag{4.2.8}$$

4.3 Normal Lines

Let $P_0 = (x_0, y_0, z_0)$ be a point on the graph $z = f(x, y)$. We can say that a line through P_0 is normal to the graph of $f(x, y)$ at P_0 if it is normal (ie. orthogonal) to the tangent plane at P_0 . A vector normal to the tangent plane is given by

$$\underline{n} = f_x(x_0, y_0) \cdot \underline{i} + f_y(x_0, y_0) \cdot \underline{j} - \underline{k} \tag{4.3.1}$$

Thus an equation for the normal line is

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x_0 \\ y_0 \\ z_0 \end{pmatrix} + t \cdot \begin{pmatrix} f_x(x_0, y_0) \\ f_y(x_0, y_0) \\ -1 \end{pmatrix} \quad (t \in \mathbb{R}) \tag{4.3.2}$$

Find the normal line to the graph of

$$\begin{aligned} f(x, y) &= e^y - x^2 \text{ at } (1, \ln 2) \\ f_x(1, \ln 2) &= -4e^{-1} \\ f_y(1, \ln 2) &= 2e^{-1} \\ z_0 &= f(1, \ln 2) = 2e^{-1} \end{aligned} \tag{4.3.3}$$

the normal line is given by

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 1 \\ \ln 2 \\ 2e^{-1} \end{pmatrix} + t \cdot \begin{pmatrix} -4e^{-1} \\ 2e^{-1} \\ -1 \end{pmatrix} \quad (t \in \mathbb{R}) \tag{4.3.4}$$

5 Linear Approximation

In one variable calculus, a function is differentiable at a point if and only if its graph has a tangent line there. It is therefore natural to say that $f(x, y)$ is differentiable at (x_0, y_0) if its graph has a tangent plane there. It follows that the trace of this graph in the plane $x = x_0$ has a tangent line at the point $y = y_0$, and so $f_y(x_0, y_0)$ and $f_x(x_0, y_0)$ exists. $z = f(x, y)$ can be approximated well in all directions by the plane containing the two tangent lines.

Equation of the plane:

$$Z = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0) \tag{5.0.1}$$

A function $f(x, y)$ is said to be differentiable at (x_0, y_0) if $f_y(x_0, y_0)$ and $f_x(x_0, y_0)$ exist **AND**

$$\lim_{(x,y) \rightarrow (x_0,y_0)} \frac{f(x,y) - [f_x(x_0,y_0)(x-x_0) + f_y(x_0,y_0)(y-y_0) + f(x_0,y_0)]}{\|(x-x_0, y-y_0)\|} \rightarrow 0 \quad (5.0.2)$$

or equivalently writing $x = x_0 + h$ and $y = y_0 + k$, we get

$$\lim_{(h,k) \rightarrow (0,0)} \frac{f(x_0+h, y_0+k) - [f_x(x_0,y_0)h + f_y(x_0,y_0)k + f(x_0,y_0)]}{\|(h,k)\|} \rightarrow 0 \quad (5.0.3)$$

The equation of a plane is the linear approximation of $f(x, y)$ at (x_0, y_0) as $(h, k) \rightarrow (0, 0)$.

Example:

$$\begin{aligned} f(x, y) &= 2x^2 + 4y^2 \text{ at } (1, 2) \\ f_x(1, 2) &= 4 \\ f_y(1, 2) &= 16 \\ f(1, 2) &= 18 \end{aligned} \quad (5.0.4)$$

then,

$$\begin{aligned} f(1+h, 2+k) &= 2(1+h)^2 + 4(2+k)^2 \\ &= 2(h^2 + 2h + 1) + 4(k^2 + 4k + 4) \\ &= 4k^2 + 2h^2 + 16k + 4h + 18 \\ &= (4k^2 + 2h^2) + f_y(1, 2)k + f_x(1, 2)h + f(1, 2) \end{aligned} \quad (5.0.5)$$

so,

$$\begin{aligned} \frac{f(1+h, 2+k) - [f_x(1, 2)h + f_y(1, 2)k + f(1, 2)]}{\|(h, k)\|} &= \frac{2h^2 + 4k^2}{\sqrt{h^2 + k^2}} \\ 0 \leq \frac{2h^2 + 4k^2}{\sqrt{h^2 + k^2}} &\leq \frac{4(h^2 + k^2)}{\sqrt{h^2 + k^2}} = 4\sqrt{h^2 + k^2} \rightarrow 0 \text{ as } (h, k) \rightarrow (0, 0) \end{aligned} \quad (5.0.6)$$

We can see that $f(x, y)$ is differentiable at $(1, 2)$. Thus it has a tangent plane, such is given by

$$Z = 4(x - 1) + 16(y - 2) + 18 \quad (5.0.7)$$

For a differentiable function $f(x, y)$, the linear approximation to $f(x, y)$ at (x_0, y_0) is given by

$$\begin{aligned} &f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0) \\ &\text{for } f(x_0 + h, y_0 + k) \\ &f_x(x_0, y_0)h + f_y(x_0, y_0)k + f(x_0, y_0) \end{aligned} \quad (5.0.8)$$

Example: Use a linear approximation to estimate

$$\sqrt{(3.04)^2 + (3.95)^2} \quad (5.0.9)$$

$$\begin{aligned}
f(x, y) &= \sqrt{x^2 + y^2} \\
f_x &= \frac{x}{\sqrt{x^2 + y^2}} \\
f_y &= \frac{y}{\sqrt{x^2 + y^2}}
\end{aligned} \tag{5.0.10}$$

$$\begin{aligned}
&f(3, 4) + f_x(3, 4)h + f_y(3, 4)k \\
&h = 0.04 \quad k = -0.05 \\
&5 + \frac{3}{5}(0.004) + \frac{4}{5}(-0.005) = 4.984
\end{aligned} \tag{5.0.11}$$

The change in value of $f(x, y)$ near (x_0, y_0) , namely $f(x_0 + h, y_0 + k) - f(x_0, y_0)$, is approximated by the change in height of the tangent plane, namely $f_x(x_0, y_0)h + f_y(x_0, y_0)k$.

The differential of a function $f(x, y)$ at (x_0, y_0) , we think of it as the “approximate change in $f(x, y)$.” It is denoted by

$$df(h, k) = f_x(x_0, y_0)h + f_y(x_0, y_0)k \quad (h, k \in \mathbb{R}) \tag{5.0.12}$$

It is a “linear function” of (h, k) .

Example: Ideal gas law

$$PV = nRT \tag{5.0.13}$$

Use the differential to estimate the change in volume of 1000 cm³ of gas at 300 K and pressure of 780 mm mercury, if the gas is heated by 10 K and the pressure is increased by 5 mm.

$$\begin{aligned}
V(T, P) &= \frac{nRT}{P} \\
nR &= \frac{P_0 V_0}{T_0} = \frac{(780)(1000)}{300} = 2600 \\
V(T, P) &= 2600 \frac{T}{P}
\end{aligned} \tag{5.0.14}$$

The change in volume can be estimated by the differential

$$\begin{aligned}
dV(h, k) &= \frac{\partial V}{\partial T}(T_0, P_0)h + \frac{\partial V}{\partial P}(T_0, P_0)k \\
dV(h, k) &= \left(\frac{2600}{P_0}\right)h + \left(-\frac{2600T_0}{P_0^2}\right)k \\
dV(h, k) &= \left(\frac{2600}{780}\right)(10) + \left(-\frac{2600 \cdot 300}{780^2}\right)(5) \\
dV(h, k) &= 26.92 \text{ cm}^3 \\
&\text{Actual change is } 26.75 \text{ cm}^3
\end{aligned} \tag{5.0.15}$$

More generally, for $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and $x_0 \in \mathbb{R}^n$, we define the differential of f at x_0 by

$$df(h_1, h_2, \dots, h_n) = \frac{\partial f}{\partial x_1}(x_0) h_1 + \frac{\partial f}{\partial x_2}(x_0) h_2 + \dots + \frac{\partial f}{\partial x_n}(x_0) h_n \quad (5.0.16)$$

provided that the partial derivatives exist. Our previous definition of differentiability of $f(x, y)$ at (x_0, y_0) can be written as: $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ exist at (x_0, y_0) **AND**

$$\lim_{(h,k) \rightarrow (0,0)} \frac{f(x_0 + h, y_0 + k) - f(x_0, y_0) - df(h, k)}{\|(h, k)\|} \rightarrow 0 \quad (5.0.17)$$

Example: Find the linear approximation to $f(x, y, z) = x^2 - xy + 3 \sin(z)$ at $(2, 1, 0)$.

$$\begin{aligned} f(2, 1, 0) &= 2 \\ f_x(2, 1, 0) &= 3 \\ f_y(2, 1, 0) &= -2 \\ f_z(2, 1, 0) &= 3 \end{aligned} \quad (5.0.18)$$

The linear approximation of f at $(2, 1, 0)$ is:

$$\begin{aligned} &f(2, 1, 0) + df(x - 2, y - 1, z - 0) \\ &f(2, 1, 0) + f_x(2, 1, 0)(x - 2) + f_y(2, 1, 0)(y - 1) + f_z(2, 1, 0)(z) \\ &2 + 3(x - 2) - 2(y - 1) + 3z \\ &3x - 2y + 3z - 2 \end{aligned} \quad (5.0.19)$$

6 Chain Rule (2 variables)

Let $f(x, y, z)$ denote the temperature at point (x, y, z) . Now suppose along the curve with coordinates $(x(t), y(t), z(t))$ where t is time. At time t , temperature $= f(x(t), y(t), z(t))$. To find the rate of change of temperature with time, we need to be able to differentiate with respect to time. Chain rule for one variable calculus:

$$w = f(x(t)) \quad \frac{dw}{dt} = \frac{df}{dx} \cdot \frac{dx}{dt} \quad (6.0.1)$$

If $w = f(x(t), y(t))$ is differentiable and $x(t)$ and $y(t)$ are differentiable functions of t , then w is differentiable of t and

$$\frac{dw}{dt} = \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt} \quad (6.0.2)$$

We know $f(x, y)$ is differentiable. We write the “error term” in the linear approximation to f as

$$\Sigma(h, k) = \underbrace{f(x_0 + h, y_0 + k) - f(x_0, y_0)}_{\text{change in } f} - \underbrace{\left[\frac{\partial f}{\partial x}(x_0, y_0) h + \frac{\partial f}{\partial y}(x_0, y_0) k \right]}_{\text{Approximate change in } f} \quad (6.0.3)$$

Then...

$$\lim_{(h,k) \rightarrow (0,0)} \frac{\Sigma(h, k)}{\sqrt{h^2 + k^2}} \rightarrow 0 \quad \text{to show differentiability} \quad (6.0.4)$$

Now $w(t) = f(x(t), y(t))$. So

$$\begin{aligned}
w(t_0 + \gamma) - w(t_0) &= f[x(t_0 + \gamma), y(t_0 + \gamma)] - f(x(t_0), y(t_0)) \\
&= f(\underbrace{x(t_0)}_{x_0} + \underbrace{[x(t_0 + \gamma) - x(t_0)]}_h, \underbrace{y(t_0)}_{y_0} + \underbrace{[y(t_0 + \gamma) - y(t_0)]}_k) - f(\overbrace{x(t_0)}^{x_0}, \overbrace{y(t_0)}^{y_0}) \\
&= \frac{\partial f}{\partial x} [x(t_0 + \gamma) - x(t_0)] + \frac{\partial f}{\partial y} [y(t_0 + \gamma) - y(t_0)] + \Sigma(h, k) \rightarrow \frac{w(t_0 + \gamma) - w(t_0)}{\gamma} \\
&= \frac{\partial f}{\partial x} \left(\frac{x(t_0 + \gamma) - x(t_0)}{\gamma} \right) + \frac{\partial f}{\partial y} \left(\frac{y(t_0 + \gamma) - y(t_0)}{\gamma} \right) + \frac{\Sigma(h, k)}{\sqrt{h^2 + k^2}} \cdot \sqrt{\frac{h^2}{\gamma^2} + \frac{k^2}{\gamma^2}} \\
&\rightarrow \frac{\partial f}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial f}{\partial y} \cdot \frac{dy}{dt} + 0 \cdot \sqrt{\left(\frac{dx}{dt} \right)^2 + \left(\frac{dy}{dt} \right)^2} \quad \text{as } \gamma \rightarrow 0 \text{ and } (h, k) \rightarrow (0, 0)
\end{aligned} \tag{6.0.5}$$

Example: Find the derivative of $w = xy$ with respect to t along the path $x = \cos(t), y = \sin(t)$. Find the derivative when $t = \frac{\pi}{2}$.

$$\begin{aligned}
f(x, y) &= xy \\
\frac{dw}{dt} &= \frac{\partial w}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial w}{\partial y} \cdot \frac{dy}{dt} \\
&= y(-\sin(t)) + x(\cos(t)) \\
&= -\sin^2(t) + \cos^2 t \\
&= \cos 2t \\
\frac{dw}{dt} \left(\frac{\pi}{2} \right) &= \cos \left(2 \left(\frac{\pi}{2} \right) \right) = \cos(\pi) = 1
\end{aligned} \tag{6.0.6}$$

6.1 Chain Rule (n variables)

If $w = f(\underline{x})$ is differentiable and if x_1, \dots, x_n are all differentiable functions of t then w is a differentiable function of t .

$$\frac{dw}{dt} = \frac{\partial f}{\partial x_1} \cdot \frac{dx_1}{dt} + \frac{\partial f}{\partial x_2} \cdot \frac{dx_2}{dt} + \dots + \frac{\partial f}{\partial x_n} \cdot \frac{dx_n}{dt} \tag{6.1.1}$$

Example: Use chain rule to compute $w'(1)$ where

$$w(t) = f(x(t), y(t), z(t)) \text{ and } f(x, y, z) = \frac{\sqrt{x} \cdot y^2 e^{2z}}{3} \tag{6.1.2}$$

$$\begin{aligned}
x(t) &= 3t^2 + 1 \\
y(t) &= 6t \\
z(t) &= 1 - t^3
\end{aligned} \tag{6.1.3}$$

$$\begin{aligned}
w'(t) &= \frac{y^2 e^{2z}}{6\sqrt{x}} \cdot 6t + \frac{2}{3} \left(y\sqrt{x} \cdot e^{2z} \right) \cdot 6 + \frac{2\sqrt{x} \cdot y^2 e^{2z}}{3} \cdot (-3t^2) \\
&= e^{2z} \left[\frac{y^2 t}{x} + 4y + 2y^2 t^2 \right] \\
&= \sqrt{3t^2 + 1} \cdot e^{2(1-t^3)} \left[\frac{36t^2}{3t^2 + 1} + 24t - 72t^4 \right] \\
&\quad \text{for } t = 1 \\
&= \sqrt{4} \cdot e^0 \left[\frac{36}{4} + 24 - 72 \right] \\
&= -78
\end{aligned} \tag{6.1.4}$$

So far we have considered $w = f(x(t), y(t))$. Next, we'll consider $w(s, t) = f(x(s, t), y(s, t), z(s, t))$. Suppose that $w = g(x, y, z)$ and $x(s, t), y(s, t), z(s, t)$ each depend on two variables (s, t) . Then $w(s, t) = g(x(s, t), y(s, t), z(s, t))$. If we treat s as a constant and then differentiate with respect to t , then the chain rule gives

$$\frac{\partial w}{\partial t} = \frac{\partial g}{\partial x} \cdot \frac{\partial x}{\partial t} + \frac{\partial g}{\partial y} \cdot \frac{\partial y}{\partial t} + \frac{\partial g}{\partial z} \cdot \frac{\partial z}{\partial t} \tag{6.1.5}$$

Similarly, treating t as a constant, we get

$$\frac{\partial w}{\partial s} = \frac{\partial g}{\partial x} \cdot \frac{\partial x}{\partial s} + \frac{\partial g}{\partial y} \cdot \frac{\partial y}{\partial s} + \frac{\partial g}{\partial z} \cdot \frac{\partial z}{\partial s} \tag{6.1.6}$$

Example: Find $\frac{\partial g}{\partial r}$ where $g(x, y) = e^x + y$ and $x = r \cos(\theta), y = r \sin(\theta)$. (Polar coordinates: x and y are functions of (r, θ))

$$\begin{aligned}
\frac{\partial g}{\partial r} &= \frac{\partial g}{\partial x} \cdot \frac{\partial x}{\partial r} + \frac{\partial g}{\partial y} \cdot \frac{\partial y}{\partial r} \\
&= e^x \cos(\theta) + \sin(\theta) \\
&= e^{r \cos(\theta)} \cos(\theta) + \sin(\theta)
\end{aligned} \tag{6.1.7}$$

Example: Suppose $f(x, y)$ has continuous 1st and 2nd order partial derivatives. Express $\frac{\partial^2 f}{\partial r^2}$ in terms of $\frac{\partial^2 f}{\partial x^2}, \frac{\partial^2 f}{\partial x \partial y}, \frac{\partial^2 f}{\partial y^2}$, where $x = r \cos(\theta)$ and $y = r \sin(\theta)$.

$$\begin{aligned}
\frac{\partial f}{\partial r} &= \frac{\partial f}{\partial x} \cdot \frac{\partial x}{\partial r} + \frac{\partial f}{\partial y} \cdot \frac{\partial y}{\partial r} \\
\frac{\partial f}{\partial r} &= \frac{\partial f}{\partial x} \cos(\theta) + \frac{\partial f}{\partial y} \sin(\theta) \\
\frac{\partial^2 f}{\partial r^2} &= \frac{\partial}{\partial r} \left(\frac{\partial f}{\partial r} \right) \\
\frac{\partial^2 f}{\partial r^2} &= \frac{\partial}{\partial r} \left(\frac{\partial f}{\partial x} \right) \cos(\theta) + \frac{\partial}{\partial r} \left(\frac{\partial f}{\partial y} \right) \sin(\theta) \\
\frac{\partial}{\partial r} \left(\frac{\partial f}{\partial x} \right) &= \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial x} \right) \frac{\partial x}{\partial r} + \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial x} \right) \frac{\partial y}{\partial r} \\
\frac{\partial}{\partial r} \left(\frac{\partial f}{\partial x} \right) &= \frac{\partial^2 f}{\partial x^2} \cos(\theta) + \frac{\partial^2 f}{\partial y \partial x} \sin(\theta) \\
\frac{\partial}{\partial r} \left(\frac{\partial f}{\partial y} \right) &= \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial y} \right) \frac{\partial x}{\partial r} + \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial y} \right) \frac{\partial y}{\partial r} \\
\frac{\partial}{\partial r} \left(\frac{\partial f}{\partial y} \right) &= \frac{\partial^2 f}{\partial x \partial y} \cos(\theta) + \frac{\partial^2 f}{\partial y^2} \sin(\theta) \\
\frac{\partial^2 f}{\partial r^2} &= \left[\frac{\partial^2 f}{\partial x^2} \cos(\theta) + \frac{\partial^2 f}{\partial y \partial x} \sin(\theta) \right] \cos(\theta) + \left[\frac{\partial^2 f}{\partial x \partial y} \cos(\theta) + \frac{\partial^2 f}{\partial y^2} \sin(\theta) \right] \sin(\theta) \\
\frac{\partial^2 f}{\partial r^2} &= \frac{\partial^2 f}{\partial x^2} \cos^2(\theta) + \frac{\partial^2 f}{\partial y \partial x} \sin(\theta) \cos(\theta) + \frac{\partial^2 f}{\partial y^2} \sin^2(\theta)
\end{aligned} \tag{6.1.8}$$

7 Directional Derivatives

Suppose that $f(x, y)$ is defined on an open set containing (x_0, y_0) . Then $\frac{\partial f}{\partial x}$ measures the rate of change of f in the direction of the x-axis and $\frac{\partial f}{\partial y}$ measures the rate of change of f in the direction of the y-axis. The directional derivative of f at (x_0, y_0) in the direction of unit vector $\underline{u} = (u_1, u_2)$ is given by

$$\begin{aligned}
D_{\underline{u}} f(x_0, y_0) &= \lim_{t \rightarrow 0} \frac{f((x_0, y_0) + t\underline{u}) - f(x_0, y_0)}{t} \\
D_{\underline{u}} f(x_0, y_0) &= \lim_{t \rightarrow 0} \frac{f(x_0 + tu_1, y_0 + tu_2) - f(x_0, y_0)}{t} t
\end{aligned} \tag{7.0.1}$$

Consider the trace of the graph f in the plane determined by \underline{u} and \underline{k} and the point $(x_0, y_0, 0)$. Then $D_{\underline{u}} f(x_0, y_0)$ is the slope of the tangent line to this curve at $(x_0, y_0, f(x_0, y_0))$. Note that $D_{(1,0)} f = \frac{\partial f}{\partial x}$ and $D_{(0,1)} f = \frac{\partial f}{\partial y}$.

More generally, for a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and unit vector $\underline{u} \in \mathbb{R}^n$, we define $D_{\underline{u}} f(x_0) = \lim_{t \rightarrow 0} \frac{f(x_0 + t\underline{u}) - f(x_0)}{t}$

Example: Compute the directional derivative of $g(x, y)$ at $(0, 0)$ in the direction $(1, 1)$.

$$g(x, y) = \begin{cases} \frac{xy^2}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0) \\ 0 & \text{if } (x, y) = (0, 0) \end{cases} \tag{7.0.2}$$

The unit vector is $\underline{u} = \frac{1}{\sqrt{2}}(1, 1)$ so $u_1 = \frac{1}{\sqrt{2}} = u_2$ and

$$\begin{aligned}
D_{\underline{u}} g(0,0) &= \lim_{t \rightarrow 0} \frac{g(0 + tu_1, 0 + tu_2) - g(0,0)}{t} \\
&= \lim_{t \rightarrow 0} \frac{g\left(\frac{t}{\sqrt{2}}, \frac{t}{\sqrt{2}}\right)}{t} \\
&= \lim_{t \rightarrow 0} \frac{1}{t} \left[\frac{\frac{t^3}{2} \sqrt{2}}{\frac{t^2}{2} + \frac{t^4}{4}} \right] \\
&= \lim_{t \rightarrow 0} \frac{\frac{1}{\sqrt{2}}}{1 + \frac{t^2}{2}} \\
&= \frac{1}{\sqrt{2}}
\end{aligned} \tag{7.0.3}$$

Recall that the directional derivative of a function f at a point \underline{x}_0 in the direction of a unit vector \underline{u} is $D_{\underline{u}} f(\underline{x}_0) = \lim_{t \rightarrow 0} \frac{f(\underline{x}_0 + t\underline{u}) - f(\underline{x}_0)}{t}$. For awkward functions (like a piece-wise) the directional derivative has to be computed from the definition. However for functions that are known to be differentiable, there is a single formula available.

If $f(x, y)$ is differentiable at (x_0, y_0) , and $\underline{u} = (u_1, u_2)$ is a unit vector, then

$$D_{\underline{u}} f(x_0, y_0) = \frac{\partial f}{\partial x}(x_0, y_0) u_1 + \frac{\partial f}{\partial y}(x_0, y_0) u_2 \tag{7.0.4}$$

Proof:

$$\begin{aligned}
D_{\underline{u}} f(x_0, y_0) &= \lim_{t \rightarrow 0} \frac{f(x_0 + tu_1, y_0 + tu_2) - f(x_0, y_0)}{t} \\
x(t) &= x_0 + tu_1 \\
y(t) &= y_0 + tu_2 \\
D_{\underline{u}} f(x_0, y_0) &= \lim_{t \rightarrow 0} \frac{f(x(t), y(t)) - f(x(0), y(0))}{t} \\
w(t) &= f(x(t), y(t)) \\
D_{\underline{u}} f(x_0, y_0) &= \lim_{t \rightarrow 0} \frac{w(t) - w(0)}{t} \\
&= \frac{dw}{dt} \Big|_{t=0} = \overbrace{\left[\frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt} \right]}^{\text{chain rule}} \Big|_{t=0} \\
&= \frac{\partial f}{\partial x}(x_0, y_0) u_1 + \frac{\partial f}{\partial y}(x_0, y_0) u_2
\end{aligned} \tag{7.0.5}$$

For differentiable $f(x, y, z)$ and a unit vector $\underline{u} = (u_1, u_2, u_3)$, we obtain

$$D_{\underline{u}} f(x_0, y_0, z_0) = \frac{\partial f}{\partial x}(x_0, y_0, z_0) u_1 + \frac{\partial f}{\partial y}(x_0, y_0, z_0) u_2 + \frac{\partial f}{\partial z}(x_0, y_0, z_0) u_3 \tag{7.0.6}$$

Example: Find $D_{\underline{u}} f(2, 1)$, where $f(x, y) = x^2 e^{3y}$ and $\underline{u} = \frac{1}{\sqrt{5}} \hat{i} + \frac{2}{\sqrt{5}} \hat{j}$.

$$\begin{aligned}
D_{\underline{u}} f(2, 1) &= \frac{\partial f}{\partial x}(2, 1) \frac{1}{\sqrt{5}} + \frac{\partial f}{\partial y}(2, 1) \frac{2}{\sqrt{5}} \\
D_{\underline{u}} f(2, 1) &= \frac{4e^3}{\sqrt{5}} + \frac{24e^3}{\sqrt{5}} \\
D_{\underline{u}} f(2, 1) &= \frac{28e^3}{\sqrt{5}}
\end{aligned} \tag{7.0.7}$$

Example: Find the directional derivative of $f(x, y, z) = e^x \cos(y) + xz$ at $(-1, \pi, -1)$ in the direction $\underline{w} = \hat{i} - 3\hat{j} + 4\hat{k}$.

The appropriate unit vector is $\underline{u} = \frac{1}{\|\underline{w}\|} \cdot \underline{w} = \frac{1}{\sqrt{1+9+16}} \cdot \underline{w} = \frac{1}{\sqrt{26}}(1, -3, 4)$

$$\begin{aligned}
D_{\underline{u}} f(-1, \pi, -1) &= \left(\frac{\partial f}{\partial x} \frac{1}{\sqrt{26}} + \frac{\partial f}{\partial y} \frac{-3}{\sqrt{26}} + \frac{\partial f}{\partial z} \frac{4}{\sqrt{26}} \right) \Big|_{(-1, \pi, -1)} \\
&= \frac{e^x \cos(y) + z}{\sqrt{26}} - \frac{3e^x \sin(y)}{\sqrt{26}} + \frac{4x}{\sqrt{26}} \\
&= \frac{e(-1) - 1 + 3e(0) + 4}{\sqrt{26}} \\
&= \frac{3 - e}{\sqrt{26}}
\end{aligned} \tag{7.0.8}$$

8 The Gradient

The gradient of a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is defined by

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix} \tag{8.0.1}$$

If $f(x, y) = x^2 \sin(y)$, then $\nabla f = \begin{pmatrix} 2x \sin(y) \\ x^2 \cos(y) \end{pmatrix}$

$$\begin{aligned}
\text{For a differentiable function } f, D_{\underline{u}} f(x_0) &= \frac{\partial f}{\partial x_1}(x_0) u_1 + \cdots + \frac{\partial f}{\partial x_n}(x_0) u_n \\
&= \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix} \cdot \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} = \nabla f(x_0) \cdot \underline{u}
\end{aligned} \tag{8.0.2}$$

Let θ be the angle between $\nabla f(x_0)$ and \underline{u} . Then

$$D_{\underline{u}} f(x_0) = \nabla f(x_0) \cdot \underline{u} = \|\nabla f(x_0)\| \cdot \|\underline{u}\| \cos(\theta) = \|\nabla f(x_0)\| \cos(\theta) \tag{8.0.3}$$

For a fixed x_0 . We have, for any choice of \underline{u} , $- \|\nabla f(x_0)\| \leq D_{\underline{u}} f(x_0) \leq \|\nabla f(x_0)\|$. The max value of $D_{\underline{u}} f(x_0)$ over all possible choices of direction \underline{u} is thus $\|\nabla f(x_0)\|$. It occurs when $\cos(\theta) = 1$, i.e. $\theta = 0$ (that is when \underline{u} has the same direction as $\nabla f(x_0)$)

Hence $\nabla f(x_0)$ points in the direction of the most rapid increase for f at \underline{x}_0 .

Example: Let $f(x, y) = xy - y^3$. Find the unit vector \underline{u} for which $D_{\underline{u}}f(2, 1)$ is a maximum. State this max value.

$$\begin{aligned}\nabla f &= \left(\frac{y}{x - 3y^2} \right)_{\rightarrow(2,1)} = \begin{pmatrix} 1 \\ -1 \end{pmatrix} \\ \underline{u} &= \frac{1}{\sqrt{2}} \cdot \begin{pmatrix} 1 \\ -1 \end{pmatrix}\end{aligned}\tag{8.0.4}$$

The directional derivative in this direction is

$$||\nabla f(2, 1)|| = \sqrt{2}$$

$\nabla f(\underline{x}_0)$ points in the direction of most rapid increase for f at \underline{x}_0 .

Example: Find the direction of steepest ascent at point above (x, y) on the graph of $f(x, y) = 9 - \frac{x^2 + y^2}{4}$.

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{pmatrix} = \begin{pmatrix} -\frac{x}{2} \\ -\frac{y}{2} \end{pmatrix} = -\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}\tag{8.0.5}$$

So ∇f always points towards the origin, since the graph is a paraboloid of revolution.

Theorem: Suppose $f(x, y, z)$ and its 1st order partial derivatives are continuous. If $\nabla f(\underline{x}_0) \neq 0$, then $\nabla f(\underline{x}_0)$ is orthogonal to the level surface of f containing \underline{x}_0 .

Example: $f(x, y, z) = x^2 + y^2 + z^2$ and a given $\underline{x}_0 = (x_0, y_0, z_0)$, the level surface of f containing \underline{x}_0 is a sphere,

$$\nabla f(\underline{x}_0) = \begin{pmatrix} 2x_0 \\ 2y_0 \\ 2z_0 \end{pmatrix} = 2 \begin{pmatrix} x_0 \\ y_0 \\ z_0 \end{pmatrix}\tag{8.0.6}$$

Which is a multiple of the radius to \underline{x}_0 .

Note: There's an obvious analogue for level curves of functions of 2 variables. Let $c = f(\underline{x}_0)$ and let S be the level surface $f(\underline{x}) = c$. Let $\underline{r}(t) = x(t)\underline{i} + y(t)\underline{j} + z(t)\underline{k}$ be a curve in S through \underline{x}_0 . Then:

$$f(x(t), y(t), z(t)) = c \text{ for all } t\tag{8.0.7}$$

We can use the chain rule to differentiate this equation with respect to t :

$$0 = \frac{d}{dt}f(x(t), y(t), z(t)) = \frac{\partial f}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial f}{\partial y} \cdot \frac{dy}{dt} + \frac{\partial f}{\partial z} \cdot \frac{dz}{dt} = \nabla f \cdot \underline{r}'(t)\tag{8.0.8}$$

Now, $\underline{r}'(t)$, the velocity vector, gives a vector tangent to the curve. Thus ∇f is orthogonal to any curve in S through \underline{x}_0 . I.e. ∇f is orthogonal to S (to the tangent plane)

This gives us another way of computing normal lines (and thus tangent planes) to surfaces.

Example: Find equations to the line normal to the ellipsoid: $2x^2 + 4y^2 + z^2 = 21$ at the point $(2, 1, 3)$ and for the tangent plane there.

Let $f(x, y, z) = 2x^2 + 4y^2 + z^2$. Then the ellipsoid is a level surface of f . A vector normal to this surface at $(2, 1, 3)$ is given by

$$\nabla f(2,1,3) = \left. \begin{pmatrix} 4x \\ 8y \\ 2z \end{pmatrix} \right|_{(2,1,3)} = \begin{pmatrix} 8 \\ 8 \\ 6 \end{pmatrix} \quad (8.0.9)$$

and the normal line through $(2,1,3)$ is given by

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2 \\ 1 \\ 3 \end{pmatrix} + t \begin{pmatrix} 8 \\ 8 \\ 6 \end{pmatrix} \quad (t \in \mathbb{R}) \quad (8.0.10)$$

The tangent plane is then given by:

$$\begin{aligned} \begin{pmatrix} 8 \\ 8 \\ 6 \end{pmatrix} \cdot \left[\begin{pmatrix} x \\ y \\ z \end{pmatrix} - \begin{pmatrix} 2 \\ 1 \\ 3 \end{pmatrix} \right] &= 0 \\ 8(x-2) + 8(y-1) + 6(z-3) &= 0 \\ 8x + 8y + 6z - 42 &= 0 \\ 4x + 4y + 3z &= 21 \end{aligned} \quad (8.0.11)$$

Note: If we had used our previous method to answer the above example, we would have to rearrange the equation of the ellipsoid as $z = \sqrt{21 - 2x^2 - 4y^2} = \pm g(x, y)$, say, and then we would have chosen the positive square root, because the point of interest is $(2, 1, +3)$. The normal vector would then be given by:

$$\underline{n} = \left. \begin{pmatrix} \frac{\partial g}{\partial x} \\ \frac{\partial g}{\partial y} \\ -1 \end{pmatrix} \right|_{(2,1)} = \left. \begin{pmatrix} \frac{1}{2} \cdot \frac{-4x}{\sqrt{21-2x^2-4y^2}} \\ \frac{1}{2} \cdot \frac{-8y}{\sqrt{21-2x^2-4y^2}} \\ -1 \end{pmatrix} \right|_{(2,1)} = \begin{pmatrix} -\frac{4}{3} \\ -\frac{4}{3} \\ -1 \end{pmatrix} = \frac{1}{6} \cdot \begin{pmatrix} 8 \\ 8 \\ 6 \end{pmatrix} \quad (8.0.12)$$

However, the newer method works more generally, since we may not always be able to write the surface conveniently in the form $z = g(x, y)$.

Example: $\nabla f(\underline{x}_0)$ is orthogonal to the level surface of f containing \underline{x}_0 . These surfaces intersect in a curve C . Find an equation for the line tangent to C at the point $(1, 2, 1)$.

$$x^2 + xyz - z^3 = 2 \quad \text{and} \quad x + y^2 + z^3 = 6 \quad (8.0.13)$$

Let $f(x, y, z) = x^2 + xyz - z^3$, $g(x, y, z) = x + y^2 + z^3$, $f(1, 2, 1) = 2$, $g(1, 2, 1) = 6$, so $(1, 2, 1)$ is on C . A vector normal to the level surface $f(x, y, z) = 2$ at $(1, 2, 1)$ is \underline{n}_1 and \underline{n}_2 is a vector normal to the level surface $g(x, y, z) = 6$ at $(1, 2, 1)$.

$$\begin{aligned} \underline{n}_1 &= (\nabla f(1, 2, 1)) = \left. \begin{pmatrix} 2x + yz \\ xz \\ xy - 3z^2 \end{pmatrix} \right|_{(1,2,1)} = \begin{pmatrix} 4 \\ 1 \\ -1 \end{pmatrix} \\ \underline{n}_2 &= (\nabla g(1, 2, 1)) = \left. \begin{pmatrix} 1 \\ 2y \\ 3z^2 \end{pmatrix} \right|_{(1,2,1)} = \begin{pmatrix} 1 \\ 4 \\ 3 \end{pmatrix} \end{aligned} \quad (8.0.14)$$

The tangent line will thus have direction

$$\underline{n}_1 \times \underline{n}_2 = \begin{vmatrix} \underline{i} & \underline{j} & \underline{k} \\ 4 & 1 & -1 \\ 1 & 4 & 3 \end{vmatrix} = \underline{i}(3+4) - \underline{j}(12+1) + \underline{k}(16-1) = \begin{pmatrix} 7 \\ -13 \\ 15 \end{pmatrix} \quad (8.0.15)$$

An equation for this line is therefore:

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} + t \cdot \begin{pmatrix} 7 \\ 13 \\ 5 \end{pmatrix} \quad (t \in \mathbb{R}) \quad (8.0.16)$$

9 Local Extrema

Theorem: Suppose $f(x, y)$ has a local extrema at (a, b) . If $\nabla f(a, b)$ exists, then $\nabla f(a, b) = 0$.

Proof: Suppose that f has a local extrema at (a, b) and that $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$ both exist at (a, b) .

Let $g(x) = f(x, b)$. Then g also has a local extrema at $x = a$, so $g'(a) = 0$, i.e. $\frac{\partial f}{\partial x}(a, b) = 0$. Similarly, the function $h(y) = f(a, y)$ has a local extrema at $y = b$, so $h'(b) = 0$, i.e. $\frac{\partial f}{\partial y}(a, b) = 0$. Hence $\nabla f(a, b) = 0$.

If $\nabla f(x_0) = 0$, then x_0 is called a critical point of f .

Note: We now have a procedure for finding a local extrema of a function f . We need only look for:

- Points where $\nabla f = 0$ (critical points)
- Points where ∇f fails to exist ("singular point")

Note: We use a left and right arrow (\leftrightarrow) to write a logical **IF**. Example: $\nabla f = 0 \leftrightarrow (x, y) = (0, 0)$, translates to: the gradient of f will be 0 **IF** $(x, y) = (0, 0)$.

Example: Find all the local extrema of $f(x, y) = \sqrt{x^2 + y^2}$

$$\nabla f = \begin{pmatrix} \frac{1}{2} \cdot \frac{1}{\sqrt{x^2 + y^2}} \cdot 2x \\ \frac{1}{2} \cdot \frac{1}{\sqrt{x^2 + y^2}} \cdot 2y \end{pmatrix} = \begin{pmatrix} \frac{x}{\sqrt{x^2 + y^2}} \\ \frac{y}{\sqrt{x^2 + y^2}} \end{pmatrix} \quad (9.0.1)$$

Except at $(0, 0)$, where the P.D's fail to exist

$f(x, 0) = \sqrt{x^2} = |x|$, and the function $g(t) = |t|$ is *not* differentiable at 0. Similarly for $f(0, y) = \sqrt{y^2} = |y|$. If $(x, y) \neq (0, 0)$, then at least one of the partial derivatives is non-zero, so $\nabla f(x, y) \neq 0$. Thus, there is only one point left to examine, namely $(0, 0)$. As before $f(0, 0) = 0 \leq f(x, y)$ for all (x, y) , so f has a (local) min at $(0, 0)$ and no other local extrema.

Example: Find **all** the local extrema of $f(x, y) = 2x + 4y - x^2 - y^2 - 1$. $\nabla f = \begin{pmatrix} 2 - 2x \\ 4 - 2y \end{pmatrix}$, so $\nabla f = 0 \leftrightarrow (x, y) = (1, 2)$. So $(1, 2)$ is the only point to examine. $f(x, y) \leq 4 = f(1, 2)$ for all (x, y) . Thus f has a local maximum at $(1, 2)$ and no other local extrema.

Example: Find the local extrema of $f(x, y) = y^2 - x^2$.

$$\begin{aligned} \nabla f &= \begin{pmatrix} -2x \\ 2y \end{pmatrix} \\ \nabla f(x, y) = 0 &\Leftrightarrow x = 0 = y \Leftrightarrow (x, y) = (0, 0) \end{aligned} \quad (9.0.2)$$

So $(0, 0)$ is the only point to examine. We consider $f(x, y) - f(0, 0) = y^2 - x^2 - 0 = y^2 - x^2$, which can be positive or negative arbitrarily close to $(0, 0)$ depending on the relative sizes of $|x|$ and $|y|$. Hence f has **no** local extrema.

Note: In the above example

$$\begin{aligned} f(x, 0) &= -x^2 \\ f(0, y) &= y^2 \end{aligned} \tag{9.0.3}$$

So the function $g(x) = f(x, 0)$ has a local max at 0 and $h(y) = f(0, y)$ has a local min at 0. We call such a point a saddle point. In general, a critical point of f which is neither a local max nor a local min will be called a saddle point of f .

9.1 Second Derivative Test (for functions of 2 variables)

Recall the second derivative test in 1 variable calculus:

Theorem 9.1 (Second Derivative Test for 1 Variable). Suppose that $f : \mathbb{R} \rightarrow \mathbb{R}$ is twice differentiable, and $f'(a) = 0$. If $f''(a) < 0$, then f has a local max at a , if $f''(a) > 0$, then f has a local min at a . If $f''(a) = 0$, then the test is inconclusive.

Suppose that all second order partial derivatives of f are continuous on an open set containing (a, b) , and that $\nabla f(a, b) = 0$. Let $A = \frac{\partial^2 f}{\partial x^2}(a, b)$, $B = \frac{\partial^2 f}{\partial y \partial x}(a, b)$, $C = \frac{\partial^2 f}{\partial y^2}(a, b)$, and $D = AC - B^2$.

1. If $D > 0$ and $A < 0$, then f has a local max at (a, b)
2. If $D > 0$ and $A > 0$, then f has a local min at (a, b)
3. If $D < 0$, then f has a saddle point at (a, b)

Note: If $D = 0$, no conclusion can be drawn from this test.

The matrix of 2nd order partial derivatives is called the Hessian Matrix of f . D is the determinant.

$$\begin{pmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial y \partial x} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} \end{pmatrix} \tag{9.1.1}$$

9.1.1 Critical Points

We'll start by completing the same example using the second derivative test to find the critical points: $f(x, y) = y^2 - x^2$.

As before, the only critical point of f is $(0, 0)$. $f_{xx}|_{(0,0)} = -2$, $f_{xy}|_{(0,0)} = 0$, $f_{yy}|_{(0,0)} = 2$. $D = \begin{vmatrix} -2 & 0 \\ 0 & 2 \end{vmatrix} = -4$.

The second derivative test shows that f has a saddle point at $(0, 0)$.

Example: Find and classify the critical points of the function $f(x, y) = x^4 + y^4 - 4xy$.

$$\begin{aligned} \nabla f &= \begin{pmatrix} 4x^3 - 4y \\ 4y^3 - 4x \end{pmatrix} \\ \nabla f = 0 &\rightarrow \begin{cases} x^3 = y \\ y^3 = x \end{cases} \rightarrow (x^3)^3 = x \rightarrow x^9 = x \rightarrow x^9 - x = 0 \end{aligned} \tag{9.1.2}$$

Hence $x = 0$ or $x = \pm 1$, and (using the equation $y = x^3$) $x^9 - x = 0$. The critical points are $(0, 0), (1, 1), (-1, 1)$.

$$x(x^8 - 1) = 0 \rightarrow \begin{cases} x = 0 \\ x^8 = 1 \end{cases} \tag{9.1.3}$$

$f_{xx} = 12x^2$, $f_{xy} = -4$, $f_{yy} = 12y$. We now tabulate the values A , B , C , and D for each critical point.

Critical Point	$A = 12x^2$	$B = -4$	$C = 12y^2$	$D = AC - B^2$	Classification
$(0, 0)$	0	-4	0	-16	Saddle Point
$(1, 1)$	12	-4	12	128	Local Min
$(-1, -1)$	12	-4	12	128	Local Min

We now look at examples where classifying critical points is less routine. Example: Find and classify the critical points of $f(x, y) = e^{-(x^4+y^4)}$.

$$\begin{aligned}\frac{\partial f}{\partial x} &= -4x^3 e^{-(x^4+y^4)} & \frac{\partial f}{\partial y} &= -4y^3 e^{-(x^4+y^4)} \\ \nabla f = 0 &\Leftrightarrow \begin{cases} x^3 = 0 \\ y^3 = 0 \end{cases} \text{ if } \rightarrow (x, y) = (0, 0)\end{aligned}\tag{9.1.4}$$

Thus the origin is the only critical point. We can now find the value of D for $f(x, y)$.

$$\begin{aligned}A &= \frac{\partial^2 f}{\partial x^2} = (-4x^3)^2 e^{-(x^4+y^4)} + (-12x^2) e^{-(x^4+y^4)} \Big|_{(0,0)} = 0 \\ B &= \frac{\partial^2 f}{\partial x \partial y} = (-4x^3)^2 e^{-(x^4+y^4)} \cdot (-4y^3) \Big|_{(0,0)} = 0 \\ C &= \frac{\partial^2 f}{\partial y^2} = (-4y^3)^2 e^{-(x^4+y^4)} - 12y^2 e^{-(x^4+y^4)} \Big|_{(0,0)} = 0 \\ D &= AC - B^2 = 0, \text{ 2nd derivative test is inconclusive}\end{aligned}\tag{9.1.5}$$

However, $f(x, y) = e^{-(x^4+y^4)} \leq e^0 = f(0, 0)$, so f has a local maximum at $(0, 0)$.

Example: Find and classify critical points of the function $f(x, y) = x^4 - y^4$

$$\begin{aligned}\nabla f &= \begin{pmatrix} 4x^3 \\ -4y^3 \end{pmatrix} \\ \nabla f = 0 &\Leftrightarrow (x, y) = (0, 0)\end{aligned}\tag{9.1.6}$$

Thus, $(0, 0)$ is the only critical point.

$$\begin{aligned}A &= f_{xx} = 12x^2 \Big|_{(0,0)} = 0 \\ B &= f_{xy} = 0 \Big|_{(0,0)} = 0 \\ C &= f_{yy} = -12y^2 \Big|_{(0,0)} = 0 \\ D &= AC - B^2 = 0, \text{ 2nd derivative test is inconclusive}\end{aligned}\tag{9.1.7}$$

However, $f(x, 0) = x^4$, which has a min at $x = 0$, and $f(0, y) = -y^4$, which has a max at $y = 0$. So $(0, 0)$ is a saddle point for f .

Note: For functions of 3 or more variables, it is still the case that local extrema occur among the points where either $\nabla f = 0$ or ∇f fails to exist. However, the second derivative test is more complicated. Suppose f is continuous on a closed bounded set. Then we know that f reaches a max value somewhere on the set, and also f reaches a min value somewhere on the set. If the max(or min) occurs in the interior, then it is a local extrema, and we know how to find it. However we must also look at the boundary points, in case the extrema occurs there.

Example: Find the max and min values of the function $f(x, y) = 2 + 2x + 2y - x^2 - y^2$ on the closed triangular region bounded by the lines $x = 0$, $y = 0$, and $y = 9 - x$. First we look at the boundary

points. On the line segment from $(0,0)$ to $(9,0)$, we have $y = 0$, so $f(x,y) = f(x,0) = 2 + 2x - x^2$ where $(0 \leq x \leq 9)$. Its extreme values may occur at endpoints $x = 0, 9$ or at points where $\frac{d}{dx}(2 + 2x - x^2) = 0$, i.e. $2 - 2x = 0$, or $x = 1$. So the points to examine are: $(0,0), (1,0), (9,0)$. On the line segment from $(0,0)$ to $(0,9)$, $x = 0$, so $f(x,y) = f(0,y) = 2 + 2y - y^2$ where $(0 \leq y \leq 9)$. We need to look at $y = 0, 1, 9$, i.e. the points to examine are $(0,0), (0,1), (0,9)$. On the line from $(0,9)$ to $(9,0)$, we have $y = 9 - x$, so $f(x,y) = f(x,9-x) = 2 + 2x + 2(9-x) - x^2 - (9-x)^2 = -61 + 18x - 2x^2$ where $(0 \leq x \leq 9)$. We need to look at the endpoints $x = 0, 9$ and points where $\frac{d}{dx}(-61 + 18x - 2x^2) = 0$, or $18 - 4x = 0$, where $x = \frac{9}{2}$ and $y = 9 - x = \frac{9}{2}$. So the points to examine are $(0,9), (\frac{9}{2}, \frac{9}{2}), (9,0)$. So now we look at the critical points in the interior:

$$\nabla f = 0 \leftrightarrow \begin{pmatrix} 2-2x \\ 2-2y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \leftrightarrow \begin{cases} x = 1 \\ y = 1 \end{cases} \quad (9.1.8)$$

So, the point to examine is $(1,1)$. In summary,

Point	$(0,0)$	$(1,0)$	$(9,0)$	$(0,1)$	$(0,9)$	$(\frac{9}{2}, \frac{9}{2})$	$(1,1)$
Value of f	2	3	-61	3	-61	$-\frac{41}{2}$	4

Max value of f is 4 at $(1,1)$. Min value of f is -61 at $(9,0)$ and $(0,9)$.

10 Method of Lagrange Multipliers

How do we find the max and min values of a function f over a closed bounded region? Among interior points, we need only examine points where $\nabla f = 0$. To find max and min values on the boundary, in general we need ...

Theorem 10.1 (Method of Lagrange Multipliers). *If x_0 maximizes or minimizes, f subject to the constraint $g(x) = 0$ (which defines a curve in \mathbb{R}^2 , or a surface in \mathbb{R}^3), and if $\nabla g(x_0) \neq 0$ then $\nabla f(x_0) = \lambda \nabla g(x_0)$ for some $\lambda \in \mathbb{R}$. This says that $\nabla f(x_0)$ and $\nabla g(x_0)$ are parallel. The number λ is called a **Lagrange Multiplier**. We're assuming that f and g have continuous 1st order partial derivatives on an open set containing x_0 .*

Sketch Proof of \mathbb{R}^2 : The equation $g(\underline{x}) = 0$ defines a curve which we can parametrize as $r(t) = x(t)\hat{i} + y(t)\hat{j}$, where $r(0) = \underline{x}_0$ and $r'(0) \neq 0$. Now the chain rule tells us that,

$$\frac{d}{dt}g(r(t)) = \frac{\partial g}{\partial x} \frac{dx}{dt} + \frac{\partial g}{\partial y} \frac{dy}{dt} = \nabla g(r(t)) \cdot r'(t) \quad (10.0.1)$$

Since $g(r(t)) = 0$ for all t , we get $\frac{d}{dt}g(r(t)) = 0 \rightarrow \boxed{\nabla g(r(t)) \cdot r'(t) = 0}$. Also, $f(r(t))$ has a max or min at $t = 0$, so $\frac{d}{dt}f(r(t))|_{t=0} = 0$, and we can again use chain rule to see that $\boxed{\nabla f(r(0)) \cdot r'(0) = 0}$. Since $r'(0) \neq 0$, $\boxed{\nabla g(r(t)) \cdot r'(t) = 0}$ and $\boxed{\nabla f(r(0)) \cdot r'(0) = 0}$ show that $\nabla f(x_0)$ and $\nabla g(x_0)$ are parallel. Finally, since $\nabla g(x_0) \neq 0$, we can write $\nabla f(x_0) = \lambda \nabla g(x_0)$ for $\lambda \in \mathbb{R}$.

Example: Find the max and min values of $f(x,y) = 2x^3 + 4y^3$ on the circle $x^2 + y^2 = 1$.

The constraint equation is $g(x,y) = 0$, where $g(x,y) = x^2 + y^2 - 1$. By the method of Lagrange multipliers, we need only look at points on the curve $x^2 + y^2 = 1$ where $\nabla f = \lambda \nabla g$.

$$\begin{pmatrix} 6x^2 \\ 12y^2 \end{pmatrix} = \lambda \begin{pmatrix} 2x \\ 2y \end{pmatrix} \rightarrow \begin{cases} 6x^2 = \lambda \cdot 2x \\ 12y^2 = \lambda \cdot 2y \end{cases} \rightarrow \begin{cases} 2x(3x - \lambda) = 0 \\ 2y(6y - \lambda) = 0 \end{cases} \rightarrow \begin{cases} x = 0 \text{ or } x = \frac{\lambda}{3} \\ y = 0 \text{ or } y = \frac{\lambda}{6} \end{cases} \quad (10.0.2)$$

Case 1 $x = 0 \rightarrow y^2 = 1 \rightarrow y = \pm 1$. Points to consider are: $(0, \pm 1)$

Case 2 $y = 0 \rightarrow x^2 = 1 \rightarrow x = \pm 1$. Points to consider are: $(\pm 1, 0)$

Case 3 Otherwise $x = \frac{\lambda}{3}$ **AND** $y = \frac{\lambda}{6}$, So $x = 2y$ $\xrightarrow{\text{constraint}}$ $4y^2 + y^2 = 1 \rightarrow y = \pm \frac{1}{\sqrt{5}}$. Points to consider are $\pm(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}})$

Now we can tabulate:

Point	$(0, 1)$	$(0, -1)$	$(1, 0)$	$(-1, 0)$	$(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}})$	$(-\frac{2}{\sqrt{5}}, -\frac{1}{\sqrt{5}})$
Value of f	4	-4	2	-2	$\frac{4}{\sqrt{5}}$	$-\frac{4}{\sqrt{5}}$

Max value of 4 at $(0, 1)$. Min value of -4 at $(0, -1)$.

Example: Find the shortest distance from the origin to the curve $x^2y = 16$. Which point(s) on the curve are closest to the origin. The constraint equation is $g(x, y) = 0$, where $g(x, y) = x^2y - 16$. The function we want to minimize is $f(x, y) = \|(x, y)\| = \sqrt{x^2 + y^2}$. However we can find where this function attains its min by instead considering $f_1(x, y) = x^2 + y^2$. So we want to consider: $\nabla f_1 = \lambda \nabla g$.

$$\begin{aligned} \nabla f_1 &= \lambda \nabla g \\ \begin{pmatrix} 2x \\ 2y \end{pmatrix} &= \lambda \begin{pmatrix} 2xy \\ x^2 \end{pmatrix} \rightarrow \begin{cases} x = \lambda xy \\ 2y = \lambda x^2 \end{cases} \end{aligned} \quad (10.0.3)$$

We obtain $x(1 - \lambda y) = 0 \rightarrow x = 0$ or $y = \frac{1}{\lambda}$.

Case 1 $x = 0$. This is impossible because of the constraint $x^2y = 16$

Case 2 $y = \frac{1}{\lambda}$ and $\frac{2}{\lambda^2} = x^2$. Our constraint is $x^2y = 16$, so we can simplify: $\frac{2}{\lambda^2} \cdot \frac{1}{\lambda} \rightarrow \frac{2}{\lambda^3} = 16$. So, $\lambda = \frac{1}{2}$
from $2y = \lambda x^2$

We're thus led to the points $(2\sqrt{2}, 2)$ and $(-2\sqrt{2}, 2)$. Distance from the origin $\sqrt{x^2 + y^2} \rightarrow \sqrt{8 + 4} = 2\sqrt{3}$. Since there must be a shortest distance from the origin to the curve; $2\sqrt{3}$, the closest points are $(\pm 2\sqrt{2}, 2)$.

Example: A rectangular box is to be placed in a cone. Find the dimensions of the box which maximizes its volume. Let the vertex of the box in the first octant be (x, y, z) . Then the volume of the box is $V(x, y, z) = 2x2yz = 4xyz$. The constraint is $g(x, y, z) = x^2 + y^2 - (9 - z)^2$. The method of Lagrange multipliers gives:

$$\begin{aligned} \nabla V &= \lambda \nabla g \\ \begin{pmatrix} 4yz \\ 4xz \\ 4xy \end{pmatrix} &= \lambda \begin{pmatrix} 2x \\ 2y \\ 2(9 - z) \end{pmatrix} \rightarrow \begin{cases} 2yz = \lambda x \\ 2xz = \lambda y \\ 2xy = \lambda(9 - z) \end{cases} \\ \frac{2xyz}{\lambda} &= x^2 = y^2 = z(9 - z) \\ \text{Constraint} \rightarrow 0 &= x^2 = y^2 = z(9 - z) = 2z(9 - z) - (9 - z)^2 \\ &= (9 - z)(2z - 9 + z) = (9 - z)(3z - 9) \rightarrow z = 9 \text{ or } z = 3 \end{aligned} \quad (10.0.4)$$

If $z = 9$, $V = 0$, Hence $z = 3$, and so $\frac{2xyz}{\lambda} = x^2 = y^2 = z(9 - z)$ gives $x^2 = y^2 = (9 - 3)(3)$, so $x = y = 3\sqrt{2}$ since $x > 0$ and $y > 0$. Thus the max volume is attained when its dimensions are: $6\sqrt{2} \times 6\sqrt{2} \times 3$ ($x \times y \times z$).

10.1 Lagrange Multipliers with 2 constraints

It's also possible to use the method of Lagrange multipliers to maximize/minimize a function $f(x, y, z)$ subject to two constraints $g_1(x, y, z) = 0$ and $g_2(x, y, z) = 0$. In this case the formula to use is $\nabla f = \lambda_1 \nabla g_1 + \lambda_2 \nabla g_2$ where $\lambda_1, \lambda_2 \in \mathbb{R}$ (i.e. ∇f is a linear combination of ∇g_1 and ∇g_2 at the relevant points). Note that each constraint equation forces us to be in a surface. We're maximizing f on the curve of intersection of these 2 surfaces. [Technical requirement: $\nabla g_1 \times \nabla g_2 \neq 0$ at the points of interest].

Example: The plane $x + y + z = 1$ cuts the infinite cylinder $x^2 + y^2 = 1$ in an ellipse. Find the points on this ellipse that are closest to and farthest from the origin. The function to maximize/minimize is $\sqrt{x^2 + y^2 + z^2}$, but it's easier to work with $f(x, y, z) = x^2 + y^2 + z^2$. The constrained functions are $g_1(x, y, z) = x + y + z - 1$ and $g_2(x, y, z) = x^2 + y^2 - 1$. The method of Lagrange multipliers gives

$$\begin{aligned} \nabla f &= \lambda_1 \nabla g_1 + \lambda_2 \nabla g_2 \rightarrow \begin{pmatrix} 2x \\ 2y \\ 2z \end{pmatrix} = \lambda_1 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} + \lambda_2 \begin{pmatrix} 2x \\ 2y \\ 0 \end{pmatrix} \\ &\rightarrow \begin{cases} 2x = \lambda_1 + 2\lambda_2 x \\ 2y = \lambda_1 + 2\lambda_2 y \\ 2z = \lambda_1 \end{cases} \rightarrow \begin{cases} 2x = 2z + 2\lambda_2 x \\ 2y = 2z + 2\lambda_2 y \end{cases} \rightarrow \begin{cases} x(1 - \lambda_2) = z \\ y(1 - \lambda_2) = z \end{cases} \end{aligned} \quad (10.1.1)$$

There are two cases for λ_2 :

Case 1 $\lambda_2 = 1$

Case 2 $\lambda_2 \neq 1$

For **Case 1**, $z = 0$,

$$\begin{aligned} \underbrace{\begin{cases} x + y = 1 \\ x^2 + y^2 = 1 \end{cases}}_{\text{2 constraints}} &\rightarrow x^2 + (1 - x)^2 = 1 \\ &2x^2 - 2x + 1 = 0 \\ &x(x - 1) = 0 \\ &x = 0, 1 \text{ and } y = 1, 0 \end{aligned} \quad (10.1.2)$$

So the points to consider are $(0, 1, 0)$ and $(1, 0, 0)$. For **Case 2**

$$\begin{aligned} x &= y = \frac{z}{1 - \lambda_2} \\ \underbrace{\longrightarrow}_{\text{1st constraint}} z &= 1 - 2x = 1 \pm \frac{2}{\sqrt{2}} = 1 \pm \sqrt{2} \\ \underbrace{\longrightarrow}_{\text{2nd constraint}} x^2 &= \frac{1}{2} \rightarrow x = y = \pm \frac{1}{\sqrt{2}} \end{aligned} \quad (10.1.3)$$

So the points to consider are $(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 1 - \sqrt{2})$ and $(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 1 + \sqrt{2})$. Here $f = \frac{1}{2} + \frac{1}{2} + (1 \pm \sqrt{2})^2 = 4 \pm 2\sqrt{2} \approx 6.8$ or 1.2 . $(0, 1, 0)$ and $(1, 0, 0)$ are closest to the origin. $(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 1 + \sqrt{2})$ is furthest from the origin.

11 Integration

11.1 Line Integrals

A curve C in \mathbb{R}^n is the image of a continuous function $\underline{r} : [a, b] \rightarrow \mathbb{R}^n$ i.e. $C = \{\underline{r}(t) : a \leq t \leq b\}$. We call \underline{r} a parametrization of C , and write (when $n = 3$, say) $\underline{r}(t) = (x(t), y(t), z(t))$. We say that C is smooth if we

can choose \underline{r} so that the velocity vector $\underline{r}'(t) = (x'(t), y'(t), z'(t))$ is continuous and never 0.

Examples: $\underline{r}(t) = e^t \cos t \hat{i} + e^t \sin t \hat{j} + t \hat{k}$ gives a spiral in \mathbb{R}^3 . $\underline{r}(t) = \cos t \hat{i} + \sin t \hat{j} + t \hat{k}$ gives a helix in \mathbb{R}^3 .

By a vector field we mean a continuous mapping $\underline{F} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ or $\underline{F} : \mathbb{R}^3 \rightarrow \mathbb{R}^3$. The line integral of \underline{F} over a smooth curve C is defined by:

$$\int_C \underline{F} d\underline{r} = \int_a^b \underbrace{\underline{F}(\underline{r}(t)) \cdot \underline{r}'(t)}_{\text{scalar product}} dt \quad (11.1.1)$$

When a constant force moves an object along a straight line, work done = force \times distance. The generalization of this formula for a force field $\underline{F} : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ moving a particle along a curve C (e.g a charged particle moving in an electric field) is the equation, (11.1.1). The notation $d\underline{r} = \underline{r}'(t)dt$ explains the use of $\int_C \underline{F} d\underline{r}$ as a shorthand.

Example: Find $\int_C \underline{F} d\underline{r}$, where $\underline{F}(x, y, z) = 2x\hat{i} + 3y\hat{j} + z\hat{k}$ and C is $\underline{r}(t) = \cos t \hat{i} + \sin t \hat{j} + t \hat{k}$ where $(0 \leq t \leq \pi)$.

$$\begin{aligned} \underline{F}(\underline{r}(t)) &= \underline{F}(\cos t, \sin t, t) = 2 \cos t \hat{i} + 3 \sin t \hat{j} + t \hat{k} \\ \underline{r}'(t) &= -\sin t \hat{i} + \cos t \hat{j} + \hat{k} \\ \rightarrow \int_C \underline{F} d\underline{r} &= \int_0^\pi \begin{pmatrix} 2 \cos t \\ 3 \sin t \\ t \end{pmatrix} \cdot \begin{pmatrix} -\sin t \\ \cos t \\ 1 \end{pmatrix} dt = \int_0^\pi (\sin t \cos t + t) dt \\ &= \left[\frac{\sin^2 t}{2} + \frac{t^2}{2} \right]_0^\pi = \frac{\pi^2}{2} \end{aligned} \quad (11.1.2)$$

Example: Evaluate $\int_C \underline{F} d\underline{r}$, where $\underline{F}(x, y) = 2xy\hat{i} + x^2\hat{j}$ and C is the curve $y = \sqrt{x}$ from $(1, 1)$ to $(4, 2)$.

$$\begin{aligned} y(t) &= \sqrt{t} \quad x(t) = t \\ \underline{r}(t) &= t\hat{i} + \sqrt{t}\hat{j} \quad (1 \leq t \leq 4) \\ \underline{F}(\underline{r}(t)) &= \underline{F}(t, \sqrt{t}) = 2t^{\frac{3}{2}}\hat{i} + t^2\hat{j} \\ \underline{r}'(t) &= \hat{i} + \frac{1}{2\sqrt{t}}\hat{j} \\ \rightarrow \int_C \underline{F} d\underline{r} &= \int_1^4 \begin{pmatrix} 2t^{\frac{3}{2}} \\ t^2 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ \frac{1}{2\sqrt{t}} \end{pmatrix} dt = \int_1^4 \frac{5}{2} t^{\frac{3}{2}} dt = \left[t^{\frac{5}{2}} \right]_1^4 = 31 \end{aligned} \quad (11.1.3)$$

Note: We could have parametrized the above curve differently as $y(t) = t, x(t) = t^2$ where $(1 \leq t \leq 2)$, so $\underline{r}(t) = t^2\hat{i} + t\hat{j}$ where $(1 \leq t \leq 2)$.

$$\begin{aligned} \underline{F}(\underline{r}(t)) &= \underline{F}(t^2, t) = 2t^3\hat{i} + t^4\hat{j} \\ \underline{r}'(t) &= 2t\hat{i} + \hat{j} \\ \rightarrow \int_C \underline{F} d\underline{r} &= \int_1^2 \begin{pmatrix} 2t^3 \\ t^4 \end{pmatrix} \cdot \begin{pmatrix} 2t \\ 1 \end{pmatrix} dt = \int_1^2 5t^4 dt = \left[t^5 \right]_1^2 = 31 \end{aligned} \quad (11.1.4)$$

We get the same answer as before. This illustrates a general fact that the value of $\int_C \underline{F} d\underline{r}$ is independent of the way we parametrize C . We can switch here from one parametrization to the other using integration by

substitution.

Alternative Notation: In \mathbb{R}^2 , we can write $F(x, y) = M(x, y)\hat{i} + N(x, y)\hat{j}$ and $dr = dx\hat{i} + dy\hat{j}$ and then $\int_C Fdr = \int_C M(x, y)dx\hat{i} + N(x, y)dy\hat{j}$. Similarly in \mathbb{R}^3 , we can write $F(x, y, z) = M(x, y, z)\hat{i} + N(x, y, z)\hat{j} + P(x, y, z)\hat{k}$ and $dr = dx\hat{i} + dy\hat{j} + dz\hat{k}$ and then $\int_C Fdr = \int_C M(x, y, z)dx\hat{i} + N(x, y, z)dy\hat{j} + P(x, y, z)dz\hat{k}$.

Example: Evaluate $\int_C xydx - 2y^2dy$, where C is the arc of the unit circle from $(1, 0)$ to $(0, 1)$, traversed anti-clockwise. A parametrization for C is: $x(t) = \cos t, y(t) = \sin t$ ($0 \leq t \leq \frac{\pi}{2}$).

$$\begin{aligned} dx &= -\sin t dt \\ dy &= \cos t dt \\ \rightarrow \int_C xydx - 2y^2dy &= \int_0^{\frac{\pi}{2}} -\cos t \sin^2 t - 2\sin^2 t \cos t dt \\ &= \int_0^{\frac{\pi}{2}} -3\sin^2 t \cos t dt \xrightarrow{\substack{u = \sin t \\ du = \cos t dt}} -3 \int_0^1 u^2 du = \left[-u^3\right]_0^1 = -1 \end{aligned} \quad (11.1.5)$$

Note: If C is a curve from P_0 to P_1 , then $-C$ denotes the reverse path, starting at P_1 and ending at P_0 . In above example, $-C$ goes clockwise instead of ACW. $-C$ is parametrized by $x(t) = \cos(\frac{\pi}{2} - t), y(t) = (\frac{\pi}{2} - t)$ where ($0 \leq t \leq \frac{\pi}{2}$). In this case, a direct calculation(check) shows that $\int_{-C} xydx - 2y^2dy = 1 = -\int_C xydx - 2y^2dy$. This illustrates the general fact that $\int_{-C} Fdr = -\int_C Fdr$. This can be proved in general by using integration by substitution.

A curve C , parametrized by $r(t)$ where ($a \leq t \leq b$) is called piece-wise smooth if there's a partition $a = t_0 < t_1 < t_2 < \dots < t_n = b$ of $[a, b]$ such that r is smooth on each sub-interval $[t_{j-1}, t_j]$. A piece-wise smooth curve is also called a path. We evaluate a line integral by evaluating it along each of the smooth curves corresponding to the sub-intervals, and then summing the answers.

Example: Evaluate the line integral $\int_C (x+y)dx + xydy$ over the path consisting of line segments $C_1, (0, 0) \rightarrow (1, 0), C_2, (1, 0) \rightarrow (1, 2), C_3, (1, 2) \rightarrow (0, 0)$.

$$\begin{aligned} C_1 : x(t) &= t, y(t) = 0 \quad (0 \leq t \leq 1) \\ \int_{C_1} (x+y)dx + xydy &= \int_0^1 t dt = \frac{1}{2} \\ C_2 : x(t) &= 1, y(t) = t \quad (0 \leq t \leq 2) \\ \int_{C_2} (x+y)dx + xydy &= \int_0^2 t dt = 2 \\ -C_3 : x(t) &= t, y(t) = 2t \quad (0 \leq t \leq 1) \\ \int_{-C_3} (x+y)dx + xydy &= -\int_0^1 (3t + 2t^2 \cdot 2)dt = -\frac{17}{6} \\ \int_C &= \int_{C_1} + \int_{C_2} + \int_{C_3} = \frac{1}{2} + 2 - \frac{17}{6} = -\frac{1}{3} \end{aligned} \quad (11.1.6)$$

11.2 Potential Functions

Recall in 1 variable calculus, that a primitive(or antiderivative) of a function f is another function g such that $g' = f$. Also $\int_a^b f(t)dt = g(b) - g(a)$. Here is an analogous result for line integrals that we can *sometimes* use.

Theorem 11.1. Let $C = \{r(t) : a \leq t \leq b\}$ be a path in an open set D , and let \underline{F} be a vector field on D . If $\underline{F} = \nabla\phi$ for some ϕ , then $\int_C \underline{F} dr = \phi(r(b)) - \phi(r(a))$.

Proof: It's enough to consider smooth curves. The chain rule tells us that $\frac{d}{dt}\phi(r(t)) = \nabla\phi(r(t)) \cdot r'(t)$. Hence $\int_C \underline{F} dr = \int_a^b \nabla\phi(r(t)) \cdot r'(t) dt = \int_a^b \underbrace{(\nabla\phi(r(t)) \cdot r'(t))}_{\text{from chain rule}} dt = [\phi(r(t))]_a^b$. If $\nabla\phi = \underline{F}$, then the scalar function ϕ is called a potential for \underline{F} . Vector field \underline{F} which has a potential ϕ is called conservative.

Example: Find a potential for $F(x, y) = 2xye^{x^2}\hat{i} + e^{x^2}\hat{j}$, and use it to evaluate $\int_C F dr$, where C is a path from $(0, 1) \rightarrow (1, 2)$. We want to find a function ϕ such that $\nabla\phi = F$.

$$\begin{aligned} \frac{\partial\phi}{\partial x} &= 2xye^{x^2} \\ \frac{\partial\phi}{\partial y} &= e^{x^2} \rightarrow \phi(x, y) = e^{x^2}y + \underbrace{C(x)}_{\text{constant}} \rightarrow \frac{\partial\phi}{\partial x} = 2xye^{x^2} + \underbrace{C'(x)}_{C'(x)=0} \end{aligned} \quad (11.2.1)$$

Which we can choose to be 0, since we just need a potential. Thus a potential for F is $\phi(x, y) = e^{x^2}y$. Thus, by our theorem, $\int_C F dr = \phi(1, 2) - \phi(0, 1) = 2e - 1$.

Notes: In the example, we didn't need to know the details of the curve C , because there was a potential. We only needed to know the initial and final points. A path C is called closed if $r(b) = r(a)$. For a conservative vector field \underline{F} on an open region D , our theorem says that $\int_C \underline{F} dr = \phi(r(b)) - \phi(r(a)) = 0$ for any closed path C lying in D , since $r(b) = r(a)$. The vector field $F(x, y) = (x + y)\hat{i} + xy\hat{j}$ must not be conservative, since we didn't get zero for the integral around the triangle. In fact, we can also see directly that there is no function ϕ satisfying $\nabla\phi = \underline{F}$, $\left\{ \begin{array}{l} \frac{\partial\phi}{\partial x} = x + y \\ \frac{\partial\phi}{\partial y} = xy \end{array} \right.$ For, if there were, $\frac{\partial^2\phi}{\partial y\partial x} = 1$ and $\frac{\partial^2\phi}{\partial x\partial y} = y$ should be equal, but are not.

11.3 Double Integrals

In 1 variable calculus the (Riemann) integrals $\int_a^b f(x)dx$ of a continuous function f gives the area under the curve $y = f(x)$ between $x = a, b$ if $f \geq 0$. Analogously, if R is the rectangle $R = \{(x, y) : a \leq x \leq b, c \leq y \leq d\}$ and f is continuous on R , then the double integral $\iint_R f(x, y)dA$ represents the volume under the surface $z = f(x, y)$ lying above R if $f \geq 0$.

11.3.1 Iterated Integration

In practice, we evaluate double integrals as follows (assuming f is continuous).

$$\begin{aligned} F(x) &= \int_c^d f(x, y)dy \quad (\text{partial integration}) \\ \iint_R f(x, y)dA &= \int_a^b F(x)dx \\ \iint_R f(x, y)dA &= \int_a^b \left(\int_c^d f(x, y)dy \right) dx \end{aligned} \quad (11.3.1)$$

Alternatively, we could integrate first with respect to x (treating y as a constant) and then with respect to y .

Example: Compute the volume of the solid region determined by the graph of the function $f(x, y) = 8 - x^2 - y^2$ over the square $R = \{(x, y) : -1 \leq x \leq 1, -1 \leq y \leq 1\}$.

$$\begin{aligned}
\iint_R f(x, y) dA &= \int_{-1}^1 \left(\int_{-1}^1 (8 - x^2 - y^2) dx \right) dy \\
&= \int_{-1}^1 \left[8x - \frac{x^3}{3} - y^2 x \right]_{-1}^1 dy \\
&= \int_{-1}^1 (8 - \frac{1}{3} - y^2) - (-8 + \frac{1}{3} + y^2) dy \\
&= \int_{-1}^1 \frac{46}{3} - 2y^2 dy \\
&= \left[\frac{46}{3} y - \frac{2}{3} y^3 \right]_{y=-1}^{y=1} \rightarrow \left(\frac{46}{3} - \frac{2}{3} \right) - \left(-\frac{46}{3} + \frac{2}{3} \right) = \frac{88}{3}
\end{aligned} \tag{11.3.2}$$

Example: Compute $\iint_R x \cos(xy) dA$, where $R = \{(x, y) : 0 \leq x \leq \frac{\pi}{4}, 0 \leq y \leq 2\}$.

$$\begin{aligned}
\iint_R f(x, y) dA &= \int_0^{\frac{\pi}{4}} \left(\int_0^2 x \cos(xy) dy \right) dx \\
&= \int_0^{\frac{\pi}{4}} [\sin(xy)]_0^2 dx \rightarrow \int_0^{\frac{\pi}{4}} (\sin(2x) - \sin(0)) dx \\
&= \int_0^{\frac{\pi}{4}} \sin(2x) dx \\
&= \left[-\frac{1}{2} \cos(2x) \right]_0^{\frac{\pi}{4}} \rightarrow -\frac{1}{2} \cos\left(\frac{\pi}{2}\right) + \frac{1}{2} \cos(0) = \frac{1}{2}
\end{aligned} \tag{11.3.3}$$

11.4 X and Y Simple functions

- A region Q in the plane is called **y-simple** if there are continuous functions g_1 and g_2 such that

$$Q = \{(x, y) : a \leq x \leq b, g_1(x) \leq y \leq g_2(x)\} \tag{11.4.1}$$

- A region Q is called **x-simple** if there are continuous functions h_1 and h_2 such that

$$Q = \{(x, y) : c \leq y \leq d, h_1(y) \leq x \leq h_2(y)\} \tag{11.4.2}$$

Theorem 11.2. Let $f(x, y)$ be continuous on region Q .

- If Q is **y-simple** and given by (11.4.1), then

$$\iint_Q f(x, y) dA = \int_a^b \left(\int_{g_1(x)}^{g_2(x)} f(x, y) dy \right) dx \tag{11.4.3}$$

- If Q is **x-simple** and given by (11.4.2), then

$$\iint_Q f(x, y) dA = \int_c^d \left(\int_{h_1(y)}^{h_2(y)} f(x, y) dx \right) dy \tag{11.4.4}$$

A region may be neither x or y simple. However it's wise to check if it is either in order to simplify the integral.

Example: Find $\iint_Q (2xy + y^2) dA$, where Q is the triangle with vertices: $(0,0), (1,0), (1,2)$. Q is y-simple (and x-simple).

$$\begin{aligned}
 Q &= \{(x, y) : 0 \leq x \leq 1, \underset{g_1(x)}{0} \leq y \leq \underset{g_2(x)}{2x}\} \\
 \iint_Q (2xy + y^2) dA &= \int_0^1 \left(\int_0^{2x} (2xy + y^2) dy \right) dx \\
 &= \int_0^1 \left[xy^2 + \frac{y^3}{3} \right]_0^{2x} dx \\
 &= \int_0^1 4x^3 + \frac{8x^3}{3} dx \rightarrow \left[\frac{5}{3} x^4 \right]_0^1 = \frac{5}{3}
 \end{aligned} \tag{11.4.5}$$

Example: Evaluate $\iint_Q 4xy dA$, where Q is the region bounded by the curves: $y = x + 1, x = 1 - y^2$. The curves intersect where $y - 1 = 1 - y^2$.

$$\begin{aligned}
 0 &= y^2 + y - 2 \\
 0 &= (y - 1)(y + 2) \\
 y &= 1 \quad y = -2 \\
 \downarrow \quad \downarrow \\
 x &= 0 \quad x = -3
 \end{aligned} \tag{11.4.6}$$

The region Q is thus x-simple and $Q = \{(x, y) : -2 \leq y \leq 1, y - 1 \leq x \leq 1 - y^2\}$.

$$\begin{aligned}
 \iint_Q 4xy dA &= \int_{-2}^1 \left(\int_{y-1}^{1-y^2} 4xy dx \right) dy \\
 &= \int_{-2}^1 \left[2x^2 y \right]_{y-1}^{1-y^2} dy \\
 &= \int_{-2}^1 (2(1 - y^2)y - 2(y - 1)^2 y) dy \\
 &= \left[\frac{y^3}{3} - \frac{y^4}{4} \right]_{-2}^1 = 27
 \end{aligned} \tag{11.4.7}$$

Note: Observe that $\iint_Q 1 dA = \text{area}(Q)$ since the volume of the generalized cylinder with base Q and height 1 is the same as the area of Q .

11.5 Changing the Order of Integration

Sometimes we need to change the order of integration, because a primitive can't easily be found for the inner integral, eg. $\int_0^1 \int_{y^2}^1 y e^{x^2} dx dy = \int_0^1 y \left(\int_{y^2}^1 e^{x^2} dx \right) dy$. We cannot write a simple primitive for e^{x^2} . This problem can sometimes be overcome as follows, to reverse the order of integration in $\int_c^d \int_{h_1(y)}^{h_2(y)} f(x, y) dx dy$, Identify the region Q such that the iterated integral is $\iint_Q f(x, y) dA$. Express Q as $\{(x, y) : a \leq x \leq b, g_1(x) \leq y \leq g_2(x)\}$ (y-simple). Then evaluate $\int_a^b \int_{g_1(x)}^{g_2(x)} f(x, y) dy dx$. This procedure can be applied when Q is both x-simple and y-simple. It can be used to change from $\iint \dots dx dy \rightarrow \iint \dots dy dx$ and vice-versa.

Example: Evaluate $\int_0^1 \int_{y^2}^1 ye^{x^2} dx dy$.

$$\begin{aligned}
 Q &= \underbrace{\{(x, y) : 0 \leq y \leq 1, y^2 \leq x \leq 1\}}_{\text{(x-simple)}} \rightarrow \underbrace{\{(x, y) : 0 \leq x \leq 1, 0 \leq y \leq \sqrt{x}\}}_{\text{(y-simple)}} \\
 &\rightarrow \int_0^1 \int_{y^2}^1 ye^{x^2} dx dy = \iint_Q ye^{x^2} dA = \int_0^1 \int_0^{\sqrt{x}} ye^{x^2} dy dx \\
 &= \int_0^1 \left[\frac{y^2}{2} e^{x^2} \right]_0^{\sqrt{x}} dx = \int_0^1 \frac{x}{2} e^{x^2} dx = \frac{1}{4} [e^{x^2}]_0^1 = \frac{e-1}{4}
 \end{aligned} \tag{11.5.1}$$

Example: Evaluate $\int_0^1 \int_0^{\sqrt{1-x}} xy^2 dy dx$, by first reversing the order of integration.

$$\begin{aligned}
 Q &= \underbrace{\{(x, y) : 0 \leq x \leq 1, 0 \leq y \leq \sqrt{1-x}\}}_{\text{(y-simple)}} \rightarrow \underbrace{\{(x, y) : 0 \leq y \leq 1, 0 \leq x \leq 1-y^2\}}_{\text{(x-simple)}} \\
 &\rightarrow \int_0^1 \int_0^{\sqrt{1-x}} xy^2 dy dx = \iint_Q xy^2 dA = \int_0^1 \int_0^{1-y^2} xy^2 dx dy \\
 &= \int_0^1 \left[\frac{1}{2} x^2 y^2 \right]_0^{1-y^2} dy = \int_0^1 \frac{1}{2} (1-y^2)^2 y^2 dy \\
 &= \frac{1}{2} \int_0^1 y^2 (1-y^2) dy = \frac{1}{2} \left[\frac{y^3}{3} - \frac{2y^5}{5} + \frac{y^7}{7} \right]_0^1 = \frac{4}{105}
 \end{aligned} \tag{11.5.2}$$

11.6 Green's Theorem

Recall that a closed curve $C = \{r(t) : a \leq t \leq b\}$ is one where $r(a) = r(b)$. We call C simple if $r(t_1) \neq r(t_2)$ whenever $t_1 \neq t_2$, where $t_1, t_2 \in [a, b]$. Thus, a simple closed curve is one which doesn't cross itself.

Note: The fundamental theorem of calculus tells us that $\int_a^b g'(x) dx = g(b) - g(a)$. The integral of g' is determined by the values of g at the end points of the interval $[a, b]$. Green's theorem relates the value of a double integral \iint_Q involving derivatives to an associated line integral around the boundary of Q .

Theorem 11.3 (Green's Theorem). *Let Q be the region enclosed by a piece-wise smooth simple closed curve C , oriented anti-clockwise. If $M(x, y)$ and $N(x, y)$ have continuous 1st order partial derivatives in an open region containing Q and C then*

$$\int_C M(x, y) dx + N(x, y) dy = \iint_Q \left(\frac{\partial N}{\partial x} - \frac{\partial M}{\partial y} \right) dA \tag{11.6.1}$$

Proof(for nice regions): First suppose that Q is both x and y simple.

$$\begin{aligned}
 Q &= \{(x, y) : a \leq x \leq b, g_1(x) \leq y \leq g_2(x)\} \text{ and } Q = \{(x, y) : c \leq y \leq d, h_1(y) \leq x \leq h_2(y)\} \\
 \int_C M(x, y) dx &= \int_{C_1} + \int_{C_2} = \int_a^b M(x, g_1(x)) dx - \int_a^b M(x, g_2(x)) dx \\
 \iint_Q \frac{\partial M}{\partial y} dA &= \int_a^b \int_{g_1(x)}^{g_2(x)} \frac{\partial}{\partial y} M(x, y) dy dx = \int_a^b [M(x, y)]_{y=g_1(x)}^{y=g_2(x)} dx \\
 \int_C M(x, y) dx &= \iint_Q \frac{\partial M}{\partial y} dA
 \end{aligned} \tag{11.6.2}$$

We can argue similarly on the basis of (x-simple) to obtain

$$\int_C N(x, y) dy = \iint_Q \frac{\partial N}{\partial x} dA \quad (11.6.3)$$

We can add (11.6.2) and (11.6.3) to obtain (11.6.1) for such regions Q . More general regions may be dealt with by using “cross-cuts”. If a region is neither x-simple or y-simple, but if you can draw a line from p_1 to p_2 , and obtain 2 regions Q_1 and Q_2 which are both x-simple and y-simple, you obtain:

$$\begin{aligned} \int_{C_1} M(x, y) dx + N(x, y) dy &= \iint_{Q_1} \left(\frac{\partial N}{\partial x} - \frac{\partial M}{\partial y} \right) dA \\ \int_{C_2} M(x, y) dx + N(x, y) dy &= \iint_{Q_2} \left(\frac{\partial N}{\partial x} - \frac{\partial M}{\partial y} \right) dA \\ \underbrace{\int_C M(x, y) dx + N(x, y) dy}_{\text{Adding } C_1 \text{ and } C_2} &= \iint_Q \left(\frac{\partial N}{\partial x} - \frac{\partial M}{\partial y} \right) dA \end{aligned} \quad (11.6.4)$$

Adding returns the same value, since the two integrals along p_1, p_2 are in opposite directions, and so cancel.

Example: Use Green’s Thm. to evaluate $\int_C xy^2 dx + 2x^3 y dy$, where CC is the closed path around the boundary of the half-disc $\{(x, y) : x^2 + y^2 \leq 1, y \geq 0\}$ oriented anti-clockwise.

$$\begin{aligned} M(x, y) &= xy^2 & N(x, y) &= 2x^3 y \\ Q &= \{(x, y) : -1 \leq x \leq 1, 0 \leq y \leq \sqrt{1 - x^2}\} \end{aligned} \quad (11.6.5)$$

Using Green’s Thm., we obtain

$$\begin{aligned} \int_C \underbrace{xy^2}_{M(x, y)} dx + \underbrace{2x^3 y}_{N(x, y)} dy &= \iint_Q \left(\frac{\partial N}{\partial x} - \frac{\partial M}{\partial y} \right) dA \\ &= \int_{-1}^1 \int_0^{\sqrt{1-x^2}} (6x^2 y - 2xy) dy dx \\ &= \int_{-1}^1 \left[3x^2 y^2 - xy^2 \right]_0^{\sqrt{1-x^2}} dx \\ &= \int_{-1}^1 3x^2(1-x^2) - x(1-x^2) dx \\ &= \int_{-1}^1 3x^2 - 3x^4 - x + x^3 dx \\ &= \left[x^3 - \frac{3x^5}{5} - \frac{x^2}{2} + \frac{x^4}{4} \right]_{-1}^1 = \frac{4}{5} \end{aligned} \quad (11.6.6)$$

Example: Use Green’s Thm. to evaluate $\int_C xy dx + e^y dy$, where C is the path from $(0, 0) \rightarrow (1, 1)$ along the curve $y = x^2$, and then from $(1, 1) \rightarrow (0, 0)$ along the curve $y = \sqrt{x}$. Here, $Q = \{(x, y) : 0 \leq x \leq 1, x^2 \leq y \leq \sqrt{x}\}$

$$\begin{aligned}
\int_C \underbrace{xy}_{M(x,y)} dx + \underbrace{e^y}_{N(x,y)} dy &= \iint_Q (0 - x) dA \\
&= \int_0^1 \int_{x^2}^{\sqrt{x}} -x dy dx \\
&= \int_0^1 [-xy]_{x^2}^{\sqrt{x}} dx \\
&= \int_0^1 -x^{\frac{3}{2}} + x^3 dx = \left[\frac{x^4}{4} - \frac{2x^{\frac{5}{2}}}{5} \right]_0^1 = -\frac{3}{20}
\end{aligned}
\tag{11.6.7}$$