# Higher Several Variable Calculus Math2111 UNSW

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# 1 Analysis

#### 1.1 Assumed

Assumed Concepts from Real Single-Variable Calculus

- limits
- continuity
- differentiability
- integrability

#### **Assumed Theorems**

- Min/ Max Theorem
- Intermediate Value Theorem
- Mean Value Theorem

#### 1.2 Limits

Recall that  $\lim_{x\to a} f(x) = L$  requires that for all  $\epsilon > 0$ , there exists a  $\delta > 0$  such that if  $|x-a| < \delta$  then

$$|f(x) - L| < \delta.$$

#### 1.3 Metrics

We have metrics (distance functions) as

$$m: \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}$$

satisfying the following 3 axioms.

- Positive Definite such that for all  $x, y \in \mathbb{R}^n$ , m(x, y) > 0 and,  $m(x, y) = 0 \iff x = y$ .
- Symmetric m(x,y) = m(y,x).z
- Triangle Inequality such that for all  $x, y, z \in \mathbb{R}^n$ ,  $m(x, y) + m(y, z) \leq m(x, z)$ .

Euclidian Distance We allow the Euclidian distance to be defined as

$$d_n(x,y) := ||x - y|| = \sqrt{\sum_{i=1}^n (x_i - y_y)^2}$$

We often allow d to be  $d_2$ .

**Norms** Norms will be revisited in the Fourier Series section. They can be thought of as the length of an element in vectors space.

**Equivalent Metrics** Two metrics d and  $\delta$  are considered equal if there exists constants  $0 < c < C < \infty$  such that

$$c\delta(x,y) \le d(x,y) \le C\delta(x,y).$$

#### 1.4 Limits of Sequences

**Balls** A ball around  $\vec{a} \in \mathbb{R}$  is of radius  $\epsilon$  is the set

$$B(\vec{a}, \epsilon) = \{x \in \mathbb{R} : d(\vec{a}, x) < \epsilon\}.$$

**Limit in Sequence** For a sequence  $\{x_i\}$  of points in  $\mathbb{R}^n$ , x is the limit of the sequence if and only if

$$\forall \epsilon > 0 \exists N \text{ such that } n \geq N \implies d(x, x_n) \leq \epsilon.$$

Equivalently,

$$\forall \epsilon > 0 \exists N \text{ such that } n \geq N \implies d(x, x_n) \in B(x, \epsilon).$$

#### Theorems with Limits of Sequences

A sequence  $x_k$  converges to a limit x

 $\Leftrightarrow$  The components of  $x_k$  converge to the components of x  $\Leftrightarrow d(x_k, x) \to 0$ .

**Limits and Equivalent Metrics** Suppose that d and  $\delta$  are two equivalent metrics. That is,  $cd(x,y) \leq \delta(x,y) \leq Cd(x,y)$  for c,C>0.

Considering d as the metric, suppose that

$$x_k \to x$$
 for  $x_k, x \in \mathbb{R}^n$ .

That is,

$$\forall \epsilon > 0, \exists K : k > K \implies d(x_k, x) < \epsilon.$$

Using  $\delta$ , we may make an equivalent statement, choosing  $\epsilon > 0$  such that  $\epsilon' = C\epsilon$ . Considering that  $\epsilon > 0 \implies \exists K : \forall k \geq K \implies d(x_k, x) < \epsilon$  then,

$$\delta(x_k, x) \le Cd(x_k, x) < C\epsilon = \epsilon'.$$

That is,  $\delta(x_k, x) < \epsilon'$ . Hence  $x_k \to x$  using an equivalent metric  $\delta$ .

Cauchy Sequences A sequence  $\{x_K\} \in \mathbb{R}$  is a Cauchy sequence if

$$\exists \epsilon > 0 \text{ such that } k, I > K \implies d(x_k, x_l) < \epsilon.$$

Cauchy Sequences and Convergence The following are equivalent:

A sequence  $\{x_k\}$  converges in  $\mathbb{R}^2 \iff \{x_k\}$  is a Cauchy Sequence.

#### 1.5 Open and Closed Sets

**Definitions** Consider  $x_k$ 

- $x_0 \in \Omega$  is an interior points of  $\Omega$  if there is a ball around x completely contained in  $\Omega$ . That is, there exists a  $\epsilon > 0$  such that  $B(x_0, \epsilon) \subseteq \Omega$ .
- $\Omega$  is open if every point of  $\Omega$  is an interior point.
- $\Omega$  is closed if its complement is open.
- $x_0 \in \Omega$  is a boundary point of  $\Omega$  if every ball around  $x_0$  contains points in  $\Omega$  and points not in  $\Omega$ .

**Closed Sets** A set  $\Omega \subset \mathbb{R}$  is closed iff and only if it contains all of its boundary points.

**Limit Points and Sets**  $x_0$  is a limit point of  $\Omega$  if there is a sequence  $\{x_i\}$  in  $\Omega$  with limit  $x_0$  and  $x_i \neq x$ .

- Every interior points of  $\Omega$  is a limit point of  $\Omega$ .
- $x_0$  is not necessarily in Omega
- A set is closed  $\Leftrightarrow$  it contains all of its limit points.

Variations of a Set Consider the set  $\Omega \in \mathbb{R}^n$ .

- The <u>interior</u> of  $\Omega$  is the set of all its interior points.
- The boundary  $\partial\Omega$ ) of  $\Omega$  is the set of all its boundary points.
- The closure of  $\Omega$ :  $\bar{\Omega} = \Omega \cup \partial \Omega$ .

The interior is the largest open subset and the closure is the smallest closed set containing  $\Omega$ .

Limit of a Function at a Point For  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$ ,  $\lim_{x\to x_0}$  means that

$$\forall \epsilon \exists \delta > 0 \text{ such that for } x \in \Omega :$$
  
  $0 < d(x, x_0) < \delta \implies d(f(x), b) < \epsilon.$ 

Alternatively,

$$x \in B(x_0, \delta) \setminus \{x_0\} \implies f(x) \in B(b, \epsilon).$$

It is sufficient to consider the limits of the components of a function.

**Limits and sequences** The limit  $\lim_{x\to a} f(x) = b$  exists if and only if,  $\lim_{k\to\infty} f(x_k) = b$  for all sequences  $x_k$  such that  $x_k$  is an element of  $\Omega$  and,  $\lim_{k\to\infty} x_k = a$ .

This is very helpful for showing that a limit does not exists.

#### 1.6 Pinching and IVT Theorem

Pinching Theorem

**IVT** see 1141

# 2 Differentiation

## 2.1 Differentiability, Derivatives and Affine Approximations

**Differentiability in**  $\mathbb{R}$  A function  $f: \mathbb{R} \to \mathbb{R}$  being differentiable at some  $a \in \mathbb{R}$  implies that there exists a *good* straight-line approximation to f at a called a *tangent line*. This function may be found as

$$T(x) = f(a) + f'(a)(x - a) = f(a) - f'(a)a + f'(a)x = y_0 + L(x)$$

where for all a, y = f(a) - f'(a)a and  $L : \mathbb{R} \to \mathbb{R} = f'(a)x$ .

Recall that

$$f'(a) = \lim_{x \to a} \frac{f(x) - f(a)}{x - a}$$

.

**Affine Maps** A function  $T: \mathbb{R}^n \to \mathbb{R}^m$  being affine means that there exists a  $y_0$  such that for all  $x \in \mathbb{R}^n$ 

$$T(x) = y_0 + L(x)$$

.

In  $T: \mathbb{R} \to \mathbb{R}$  this sis of the form y = mx + b.

A function  $f: \mathbb{R} \to \mathbb{R}$  is differentiable if there is a good affine approximation to f of the form

$$T(x) = f(a) - f'(a)a + f'(a)x.$$

In this context good implies that f'(x) is defined in the usual manner and exists.

**Differentiability in**  $\mathbb{R}^n \to \mathbb{R}^n$  A function  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$  is differentiable for some  $a \in \Omega$  if there exists a linear map  $L: \mathbb{R}n \to \mathbb{R}^m$  such that

$$\lim_{x \to a} \frac{||f(x) - f(a) - L(x - a)||}{||L(x - a)||} = 0.$$

Notation: the matrix of the linear map L, the derivative of f at a is denoted by  $D_a f$ .

**Delta Epsilon Definition of Differentiability** A function  $f: \Omega \subset \mathbb{R} \to \mathbb{R}^m$  is differentiable on  $a \in \Omega$  if there is a linear map  $L: \mathbb{R}^n \to \mathbb{R}^m$  such that  $\forall \epsilon > 0 \exists \delta > 0$  such that for all  $x \in \Omega$ 

$$||x-a|| < \delta \Rightarrow ||f(x) - f(a) - L(x-a)|| < \epsilon ||x-a||.$$

Clairaut's Theorem / Mixed Derivative Theorem Suppose  $f, \frac{\partial f}{\partial x_i}, \frac{\partial f}{\partial x_j}, \frac{\partial^2 f}{\partial x_i \partial x_j}, \frac{\partial^2 f}{\partial x_j \partial x_i}$  all exist and are continuous on an open set around a then

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}.$$

That is, the partial derivatives commute.

**Differentiability and Continuity** Differentiability implies continuity. However, continuity does not imply differentiability. The proof of this is contingent on the fact that for  $x \in \mathbb{R}^n$  and a  $m \times n$  matrix L

$$\lim x \to 0 \, ||Lx|| = 0.$$

Partial Derivatives and Differentiability Suppose that  $\Omega \subset \mathbb{R}^n$  is open and  $f:\Omega \to \mathbb{R}^m$ . If all partial derivatives  $\frac{\partial f_j}{\partial x_i}$  exist for integers  $i \in [1,n], j \in [1,m]$  then f is differentiable on  $\Omega$ .

# 2.2 Gradients, Affine Approximations and Matrices

**Jacobian Matrices** Suppose that all partial derivatives of  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$  exist for some  $a \in \Omega$ . Then, the Jacobian matrix of f

$$J_{a}f = \begin{pmatrix} \frac{\partial f_{1}}{\partial x_{1}} & \frac{\partial f_{1}}{\partial x_{2}} & \cdots & \frac{\partial f_{1}}{\partial x_{n}} \\ \frac{\partial f_{2}}{\partial x_{1}} & \frac{\partial f_{2}}{\partial x_{2}} & \cdots & \frac{\partial f_{2}}{\partial x_{n}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_{n}}{\partial x_{1}} & \frac{\partial f_{n}}{\partial x_{2}} & \cdots & \frac{\partial f_{m}}{\partial x_{n}} \end{pmatrix}$$

may be evaluated at a point a. Where f is differentiable, its derivative is given by the Jacobian matrix.

Note however, that the Jacobian Matrix may exist even where f is not differentiable.

## 2.3 Gradients, Tangent Planes and Affine Approximations

**Gradient** For  $f:\Omega\subset\mathbb{R}^n\to\mathbb{R}$ , if the Jacobian exists, then it is given by the  $1\times n$  matrix

$$Jf = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix}.$$

This is equivalent to the gradient of f. That is,

$$\operatorname{grad}(f) = \mathbf{\nabla} f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix}.$$

**Affine Approximations** Allow  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}$  to be a differentiable function at  $a \in \Omega$ . The best affine approximation to f at a may be written in terms of the gradient vector as

$$T(x) = f(a) + \nabla f(a) \cdot (x - a).$$

**Tangent Planes** The tangent plane to a function z = f(x, y) is given by

$$z = T(x, y)$$
.

# 2.4 Chain Rule, Directional Derivatives and Tangent Planes

**Chain Rule** Suppose that  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$  and  $g: \Omega' \subset \mathbb{R}^m \to \mathbb{R}^P$  where  $f(\Omega) = \Omega'$ . If f and g are both differentiable then, so is  $g \circ f: \Omega \to \mathbb{R}^p$  such that

$$D_a(g \circ f) = D(f(a))gD_af.$$

Equivalently,

$$D(g \circ f)(a) = Dg(f(a))Df(a).$$

**Directional Derivative** The directional derivative of  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$  in the direction of the unit vector u at a point  $a \in \Omega$  is

$$D_u f(a) = f'_u(a) = \lim_{t \to 0} \frac{f(a + tu) - f(a)}{t}.$$

Equivalently, if  $f:\Omega\subset\mathbb{R}^n\to\mathbb{R}$  is differentiable at a then for a unit vector u

$$D_u f(a) = D f(a) \cdot u = \nabla f(a) \cdot u.$$

Alternatively, allowing  $\theta$  to be the angle between  $\nabla f(a)$  and u,

$$D_u f(a) = |\nabla f(a)| \cdot |u| \cdot \cos \theta.$$

**Tangent Planes** Consider the surface in  $\mathbb{R}^3$  defined by  $\phi(x, y, z) = \lambda$ . where  $\lambda$  is constant and  $\phi$  is differentiable.

Let  $c(t) = (c_1(t), c_2(t), c_3(t))$  be a differentiable curve lying on the vector space with a tangent vector given by  $c'(t) = (c'_1(t), c'_2(t), c'_3(t))$ .

Since all points c(t) lie on the surface,  $\phi(c(t)) = \lambda$ . Thus,

$$D(\phi(c(t)))Dc(t) = 0 \Rightarrow \nabla \phi c'(t) = 0.$$

Therefore, all curves passing through a point P on the surface have tangent vector normal to  $\nabla \phi$ . Thus, they all lie in the tangent plane at P.

# 2.5 Taylor Series and Theorem

**Taylor's Theorem** For all continuous and differentiable functions  $f: \mathbb{R} \to \mathbb{R}$ ,

$$f(x) \approx P_k(a) = \sum_{n=0}^k \frac{1}{n!} f^{(n)}(x) (x-a)^n + R$$

where the remainder R is

$$R = \frac{1}{(k+1)!} f^{(k+1)}(z) (x-a)^k.$$

for some z between x and a.

 $P_0, P_1, P_2, P_3$  are the best constant, affine, quadratic, cubic approximations.

**Generalizing Taylor's Theorem** Let  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}$  be  $C^r$  on the open set  $\Omega$ . Let  $a \in \Omega$  be such that the line segment joining a and x lies entirely in  $\Omega$ . Then,

$$f(x) = P_{r,a}(x) + R_{r,a}(a)$$

where for some z on the line segment between a and x

$$P_{r,a}(x) = f(a) + \sum_{k=1}^{r-1} \frac{1}{k!} D^k f(a) \cdot (x-a)^k, \quad , R_{r,a}(a) = \frac{1}{r!} D^r f(z) \cdot (x-a)^r$$

Note that  $\cdot$  is not a dot product.

Second Degree Taylor Series Let  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}$  be  $C^3$  on  $\Omega$ . Then, ignoring the remainder,

$$f(x) = f(a) + \nabla f(a) \cdot (x - a) + \frac{1}{2} ((x - a) \cdot (Hf(a) \cdot (x - a)))$$

where H is the Hessian Matrix.

#### 2.6 Hessian Matrix and Stationary Points

**Hessian Matrix** For  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}$ , the Hessian matrix of f at a point a is the  $n \times n$  matrix

$$H(f,a) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1} & \frac{\partial^2 f}{\partial x_2 \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2 \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_2} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_3} & \frac{\partial^2 f}{\partial x_2 \partial x_3} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1} \end{pmatrix}.$$

**Revision:** Trace and Determinant Recall that the trace is the sum of its diagonal values. The trace of a hessian matrix is also the sum of its eigenvalues. Also, the determinant is the product of the eigenvalues. The eigenvalues of a matrix A can be found by calculating solutions to

$$|A - \lambda I| = 0.$$

**Definite and Semi Definite Matrices** For a  $n \times n$  symmetric matrix H,

- All eigenvalues are  $> 0 \Leftrightarrow$  positive definite
- All eigenvalues are  $< 0 \Leftrightarrow$  negative definite
- All eigenvalues are  $\geq 0 \Leftrightarrow$  positive semidefinite
- All eigenvalues are  $\leq 0 \Leftrightarrow$  negative semidefinite

Sylvester's Criterion (for the Definite Property) Allow  $H_k$  to be the upper left  $k \times k$  sub-matrix of h and let  $\Delta_k = \det H_k$ . Then,

- positive definite  $\Leftrightarrow \Delta_k > 0 \forall k$
- positive semidefinite  $\Leftrightarrow \Delta_k \leq 0 \forall k$
- negative definite  $\Leftrightarrow \Delta_k < 0$  for all odd k and  $\Delta > 0$  for all even k
- negative semidefinite  $\Leftrightarrow \Delta_k \leq 0$  for all odd k and  $\Delta \geq 0$  for all even k

The Definite Property and Classification of Stationary Points Suppose that  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}$  is  $C^2$  and  $\nabla f(a) = 0$  at an interior point a of  $\Omega$ . Then

- H(f, a) is a positive definite  $\Rightarrow f$  has a local maximum at a
- H(f, a) is a negative definite  $\Rightarrow f$  has a local minimum at a
- f has a local minimum at  $a \Rightarrow H(f, a)$  is a positive semidefinite
- f has a local maximum at  $a \Rightarrow H(f, a)$  is a negative semidefinite

Observe carefully that the semidefinite cases can also be saddle points.

#### 2.7 Lagrange Multipliers, Implicit and Inverse Function Theorems

**Lagrange Multipliers** Consider two differentiable functions  $f : \mathbb{R}^n \to \mathbb{R}$  and  $f : \mathbb{R}^n \to \mathbb{R}$ . Lagrange multipliers are useful for finding local extrema of f under the constraint  $S = \{x \in \mathbb{R}^n : g(x) = c, c \in \mathbb{R}\}$ .

Then, if a local minimum or maximum of f occurs on  $a \in S$  then,  $\nabla f()$  and  $\nabla g(a)$  are parallel. That is, when  $\nabla g(a) \neq 0$ , there exists  $\lambda \in \mathbb{R}$  such that

$$\nabla f(a) = \lambda \nabla g(a).$$

Note that this theorem will only provide possible candidates for minimum or maximum points. There is no guarantee that there exists minimum or maximums of f on S.

**Inverse Function Theorem in**  $\mathbb{R} \to \mathbb{R}$  Suppose that  $f : \mathbb{R} \to \mathbb{R}$  is differentiable on an interval  $I \subset \mathbb{R}$  and  $f'(x) \neq 0$  for all  $x \in I$ . Then, f is invertible on I and the inverse  $(f^{-1})'(x)$ 

is differentiable such that

$$(f^{-1})'(x) = \frac{1}{f'(f^{-1}(x))}.$$

That is, if y = f(x) and  $f^{-1}$  exist and is differentiable with  $x = f^{-1}(y)$  then

$$\frac{dx}{dy} = \frac{1}{\frac{dy}{dx}}.$$

Generalising the Inverse Function Theorem Let  $\Omega \subset \mathbb{R}^n$  be open and  $f : \Omega \to \mathbb{R}^n$  be  $C^1$ . Suppose that  $a \in \Omega$ .

If the matrix Df(a) is invertible, then f is invertible on an open set U containing a. That is, the inverse exists as

$$f^{-1}: f(U) \to U.$$

Further,  $f^{-1}$  is  $C^1$  and for  $x \in U$ ,

$$D_{f(x)}f^{-1} = (D_x f)^{-1}.$$

Consequently,  $f^{-1}$  has its best affine approximation at f(a) as

$$f^{-1}(x) \approx a + (D_f)^{-1}(x - f(a)).$$

# 3 Integration

#### 3.1 Riemann and Fubini

**Riemann Integral** For a bounded function  $f: R \to \mathbb{R}$ , if there exists a unique number I such that

$$\underline{S}_{\mathcal{P}_1,\mathcal{P}_2}(f) \le I \le \bar{S}_{\mathcal{P}_1,\mathcal{P}_2}(f)$$

for every pair of partitions  $\mathcal{P}_1, \mathcal{P}_2$ . Then, f is Riemann integrable on R and

$$I = \int \int_{R} f = \int \int_{R} f(x, y) dA.$$

I is called the Riemann integral of f over R.

**Properties of the Riemann Integral** The single variable interpretation of a Riemann integral is the (signed) area bound by the graph y = f(x) and the x axis over the interval [a, b]. For two variables,  $\int \int_R f$  is the (signed) volume bounded by the graph z = f(x, y) and the xy-plane over the rectangle R.

 $\overline{\text{If } f, g}$  are integrable on R,

- $\iint_R \alpha f = \beta g = \alpha \iint_r f + \beta \iint_R g$
- If  $f(x) \leq g(x), \forall x \in R$  then  $\iint_R \leq \iint_R g$
- $\left| \int \int_{R} f \right| \le \int \int_{R} |f|$
- If  $R = R_1 \cup R_2$  and interior  $R_1 \cap \text{interior } R_2 = \emptyset$  then  $\iint_R f = \iint_{R_1} f + \iint_{R_2} f$

**Fubini's Theorem** Let  $f: R \to \mathbb{R}$  be continuous on a rectangular domain  $r = [a, b] \times [c, d]$ . Then,  $\int_a^b \int_c^d f(x, y) dy dx = \int_c^d \int_a^b f(x, y) dy, dy$ .

**Fubini's Theorem - Discontinuous** Let  $f: R \to \mathbb{R}$  be bounded on a rectangular domain  $R = [a, b] \times [c, d]$  with the discontinuities of f confined to a finite union of graphs of continuous functions. If  $\int_c^d$  exists for each  $x \int [a, b]$  then

$$\int \int_{B} = \int_{a}^{b} \left( \int_{x}^{d} f(x, y) dy \right) dx.$$

Since f is not continuous then there is no guarantee that these integrals exist.

# 3.2 Uniform Continuity, Leibniz

**Uniform Continuity** The function  $f: \Omega \subset \mathbb{R}^n \to \mathbb{R}^m$  is uniformly continuous on  $\Omega$  if for all  $x, y \in \Omega$  and for all  $\epsilon > 0$ , there exists  $\delta$  such that

$$d(x,y) < \delta \quad \Rightarrow \quad d(f(x),f(y)) < \epsilon.$$

 $\delta$  may depend on x but, given an  $\epsilon$ , the same  $\delta$  must work for all x.

Continuity and Uniform Continuance A continuous function on a compact  $\Omega$  is uniformly continuous on  $\Omega$ .

**Leibniz' Rule** Consider a function  $f: \mathbb{R}^2 \to \mathbb{R}$  that is continuous on  $\Omega$  and  $\frac{\partial f}{\partial x}$  is uniformly continuous on  $\Omega$ .

If,

$$F(x) = \int_{a}^{b} f(x, y)$$

then

$$F'(x) = \frac{d}{dx} \int_{a}^{b} f(x, y) dy = \int_{a}^{b} \frac{\partial f}{\partial x}(x, y) dy.$$

#### 3.3 Alternate Coordinates

Change of Variable Suppose that  $F: \Omega \in \mathbb{R}^n \to \mathbb{R}^n$  is  $C^1$ ,  $\det(J_x F \neq 0)$  for  $x \in \Omega$  and F is one-to-one. Then, if f is integrable on  $\Omega' = F(\Omega)$ 

$$\int_{\Omega'} f(x,y) = \int_{\Omega} (f \circ F) |\det JF|.$$

As an alternate notation consider,

$$\int_{\Omega'} f(x,y)dxdy = \int_{\Omega} f(x(u,v),y(u,v))|\det JF|dudv.$$

where

$$\det Jf = \det \begin{pmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{pmatrix}.$$

Note that F is the function which maps the change of variable.

# 3.4 Mass, Centre of Mass, Centroid

For the following section, suppose that  $\Omega \subset \mathbb{R}$  with a density function  $p: \Omega to\mathbb{R}$ .

Note that this can be generalised to n dimensions by the n-th integral rather than a double integral

Mass The total mass is

$$M = \int_{\Omega} p(x, y) dx dy.$$

**Center of Mass** The coordinates for centre of mass follow as such:

- $\bar{x} = \int \int_{\Omega} x p(x, y) dx dy$
- $\bar{y} = \iint_{\Omega} y p(x, y) dx dy$

## 4 Fourier Series

**Fourier Series** A Fourier series is the approximation of simple periodic functions by the sum of period functions of the form  $\sin(x)$ ,  $\cos(x)$ . Note that unlike Taylor series, a function f may be discontinuous. However, any lack of continuity leads to an infinite sum in the Fourier series.

#### 4.1 Inner Products

**Inner Products** Let V be a real vector space. An inner product on V is a map that assigns each  $f, g \in V$  a real number  $\langle f, g \rangle$  such that the following properties hold for all  $f, g, h \in V$  and  $\lambda, \mu \in \mathbb{R}$ :

- $\langle f, f \rangle \ge 0$ ,
- $\langle f, f \rangle = 0$  if and only if f is zero,
- $\bullet \langle \lambda f + \mu g, h \rangle, = \lambda \langle f, h \rangle + \mu \langle g, h \rangle,$
- $\langle g, f \rangle = \langle f, g \rangle$ .

#### **Usual Inner Products**

• The vector space  $\mathbb{R}^n$  admits the following inner product

$$\langle u, v \rangle = u \cdot v = \sum_{i=1}^{n} u_i v_i.$$

• The vector space C[a, b] consisting of all continuous function on the interval [a, b] admits the following inner product

$$\langle f, g \rangle = \int_{a}^{b} f(x)g(x)dx.$$

Inner Product and Orthogonality We say functions are orthogonal if  $\langle f, g \rangle = 0$ .

#### 4.2 Norms

A norm on V is a map that assigns each  $f \in V$  a real number ||f|| such that  $\forall f \in V, \lambda \in \mathbb{R}$ 

- ||f|| > 0,
- ||f|| = 0 if and only if f = 0,
- $\bullet \ ||\lambda f|| = \lambda ||f||,$
- $||f + g|| \le ||f|| + ||g||$ ; that is, the triangle inequality holds.

#### **Usual Norms**

• The Euclidian norm  $(L^2$ -norm): is a norm on  $\overline{C[a,b]}$ :

$$||f||_2 = \sqrt{\int_a^b f(x)^2 dx}$$

• The max norm is a norm on C[a, b]:

$$||f||_{\infty} = \max_{a \le x \le b} \{|f(x)|\}$$

# 4.3 Fourier Coefficient and Series

**Fourier Series** Suppose that a function  $f: \mathbb{R} \to \mathbb{R}$  is 2L-periodic, - that is, f(x) = f(x+2L) - and is square integrable - that is,  $\int_{-L}^{L} f(x)^2 dx < \infty$ . Then, f may be represented by a Fourier series of the form

$$f(x) = \frac{a_0}{2} + \sum_{k=1}^{n} \left[ a_k \cos\left(\frac{k\pi}{L}x\right) + b_k \sin\left(\frac{k\pi}{L}x\right) \right] \quad \forall x \in [-\pi, \pi].$$

This series converges to f as  $n \to \infty$ .

#### Fourier Coefficients

- $a_k = \frac{1}{L} \int_{-L}^{L} f(x) \cos\left(\frac{k\pi x}{L}\right)$
- $b_k = \frac{1}{L} \int_{-L}^{L} f(x) \sin\left(\frac{k\pi x}{L}\right)$

# 4.4 Convergence of Fourier Series

**Continuity** Consider a function  $f : \mathbb{R} \to \mathbb{R}$  and a point  $c \in \mathbb{R}$ . Suppose that the one-sided limits  $f(c^+)$  and  $f(c^-)$  exist.

- If  $f^{c^+} = f^{c^-} = f(c)$  then f is continuous at c,
- If  $f^{c^+} = f^{c^-} \neq f(c)$  then f has a removable discontinuity at c,
- If  $f(c^+) \neq f(c^-)$  then, f has a jump discontinuity at at c.

**Piecewise Continuity** A function is piecewise continuous on [a, b] if and only if

- $f(x^+)$  exists  $\forall x \in [a, b]$ ,
- $f(x^-)$  exists  $\forall x \in [a, b]$ ,
- f is continuous on (a, b) except at most a finite number of points.

Note that if f is only piecewise continuous then the partial sum of the Fourier series does not necessarily converge to f for all x.

**Piecewise differentiability** A function f is differentiable on c if and only if  $f(c^+) = f(c^-) = f(c)$  and  $D^+f(c) = D^-f(c)$ 

Note:  $D^+f(c)$  is not necessarily the same as  $\lim_{x\to c^+} f'(x)$ .

A function is piecewise differentiable on [a, b] if and only if

- $D^+f(x)$  exists  $\forall x \in [a,b)$ ,
- $D^-f(x)$  exists  $\forall x \in (a, b],$
- f is differentiable on (a, b) except at most a finite number of points.

**Pointwise convergence** Let  $c \in \mathbb{R}$ . Suppose that a function has the following properties

- f is 2L periodic,
- f is piecewise continuous on [-L, L],
- $D^+f(c), D^-f(c)$  exist.

Then,

$$S_f(c) = \frac{1}{2} [f(c^+) + f(c^-)].$$

Observe that if f is continuous at c then  $S_f(c) = f(c)$ .

**Odd and Evenness** Recall that odd and even functions are defined by the conditions f(-x) = -f(x) and f(x) = f(-x) respectively.

The following elementary properties hold:

- Odd  $\times$  Even = Even,
- Odd  $\times$  Odd = Even,
- Even  $\times$  Even = Even,
- $\int_{-L}^{L} Odd = 0$ .

## 4.5 Convergence of Sequences

**Pointwise convergence** Let  $f_k : \mathbb{R} \to \mathbb{R}$ .  $f_k$  converges to f on [a, b] pointwisely iff and only if for all  $x \in [a, b]$ ,  $f_k(x) \to f(x)$  as  $k \to infty$ .

Epsilon Delta Definition Pointwise Convergence For all  $x \in [a, b]$ ,  $\epsilon > 0, \exists K$  (which will depend on  $\epsilon, x$  such that

$$|f_k(x) - f(x)| \le \epsilon \quad \forall k \ge K.$$

**Uniform Convergence** Let  $f_k : \mathbb{R} \to \mathbb{R}$ .  $f_k$  converges to f on [a, b] uniformly if and only if for all  $\epsilon > 0$ ,  $\exists K$  (depending on  $\epsilon$  only) such that

$$\sup_{x \in [a,b]} |f_k(x) - f(x)| \le \epsilon \quad \forall k \ge K.$$

Weierstrass test Let  $f_k : \mathbb{R} \to \mathbb{R}$  be a sequence of a function f defined on [a, b]. Suppose that there exists a sequence of numbers  $c_k$  such that

$$|f_k(x)| \le c_k \quad \forall x \in [a, b]$$

where  $\sum_{k=1}^{\infty} c_k$  converges to a real number. Then  $\sum_{k=1}^{\infty} f_k$  converges uniformly to a function f on [a, b].

Note that this test also holds for function  $f: \mathbb{R}^n \to \mathbb{R}$  for  $x \in \Omega$  where  $\Omega$  is a closed bounded set in  $\mathbb{R}^n$ .

**Norm Convergence** Using the supremum norm, the definition of uniform convergence can be equivalently written as: for all  $\epsilon > 0, \exists K$  such that

$$||f_k - f|| \le \epsilon \quad \forall k \ge K.$$

Equivalently,

$$\lim_{k \to \infty} ||f_k - f|| = 0.$$

We may extend this to define norm-convergence for any norm.

Extending Norm Convergence to L-2 Recall from the previous paragraph that norm-convergence is defined as follows:

$$\lim_{k \to \infty} ||f_k - f|| = 0.$$

As such,  $L^2$  norm convergence, also knows as mean square convergence is equivalent to the following

$$\lim_{k \to \infty} \int_a^b [f_k(x) - f(x)]^2 dx = 0.$$

**Parseval Theorem** Let f be a  $2\pi$  periodic and bounded function where  $\int_{-pi}^{\pi} f(x)^2 dx < +\infty$ . Then, the Fourier series of f converges to f in the mean square sense. Moreover, the Parseval's identity holds

$$\int_{-pi}^{\pi} f(x)^2 = ||f||_2^2 = \frac{\pi}{2}a_0 + \pi \sum_{k=1}^{\infty} (a_k^2 + b_k^2).$$

This identity continues to hold for 2L periodic functions integrated over [-L, L].

# 5 Vector Fields

# 5.1 Vector Fields and Flows, Divergence and Curl

**Flow Lines** If F is a vector field, a flow line for F is a path c(t) such that

$$c'(t) = F(c(t)).$$

That is, that F yields the velocity field of the path c(t).

The Del  $\nabla$  operator The vector differential operator  $\nabla$  may be considered a symbolic vector. The differential operator may be written as

$$\nabla = \frac{\partial}{\partial x}i + \frac{\partial}{\partial y}j + \frac{\partial}{\partial z}k.$$

**Divergence** Given a field  $F = (f_1, f_2, \dots, f_n)$ , the divergence of F is

$$\operatorname{div} F = \nabla \cdot F = \sum_{i=1}^{n} \nabla f_i.$$

Divergence may be thought as a type of derivative that describes the measure at which a vector field *spreads away* from a certain point. If the divergence is positive, then there is a net outflow while there is net inflow if the divergence is negative.

Observe that the divergence of a vector field will be real-valued.

Curl If F is a vector field, then the curl may be defined as

$$\operatorname{curl} F = \nabla \times F$$
.

Curl is also analogous to a type of derivative for vector fields. The curl may be though as the measure at which the vector field *swirls* around a point. A positive swirl can be thought of as a counter clockwise rotation.

Observe that the curl of a vector field is also a vector field.

#### 5.2 Vector Identities

**Basic Vector Identities** 

- 1.  $\nabla (f+g) = \nabla f + \nabla g$
- 2.  $\nabla(\lambda f) = \lambda \nabla f$  where  $\lambda \in \mathbb{R}$
- 3.  $\nabla(fg) = g\nabla f + f\nabla g$ . You may draw analogies to the product.
- 4.  $\nabla \frac{f}{g} = \frac{f \nabla g g \nabla f}{g^2}$  where  $g \neq 0$ . This is analogous to the quotient rule.
- 5.  $\nabla \cdot (F+G) = \nabla \cdot F + \nabla \cdot G$
- 6.  $\nabla \times (F+G) = \nabla \times F + \nabla \times G$
- 7.  $\nabla \cdot (fF) = f \nabla \cdot F = F \cdot \nabla f$
- 8.  $\nabla \cdot (F \times G) = G \cdot (\nabla \times F) F \cdot (\nabla \times G)$
- 9.  $\nabla \cdot (\nabla \times F) = 0$
- 10.  $\nabla \times (fF) = f \nabla \times F = \nabla f \times F$
- 11.  $\nabla \times (\nabla f) = 0$
- 12.  $\nabla^2(fg) = f\nabla^2g + 2((\nabla f \cdot \nabla g)) + g\nabla^2f$
- 13.  $\nabla \cdot (\nabla \times \nabla g) = 0$
- 14.  $\nabla \cdot (f\nabla g g\nabla f) = f\nabla^2 g g\nabla f^2$

# 6 Path Integrals

# 6.1 Path Integrals

**Path (Scalar Line) Integrals** Suppose that a vector-valued function c(t) parametrises a curve C for  $t \in [a, b]$ . The scalar line integral may be thought as the integral of along c.

Computing a Scalar Line Integral Let c(t) parametrise a curve C for  $t \in [a, b]$ . Assume that f(x, y, z) and c(t) are continuous. Then,

$$\int_{C} f(x, y, z)ds = \int_{a}^{b} f(c(t)) \cdot ||c(t)||dt.$$

**Elementary Properties of Path Integrals** 

- $\int_C f_1 ds + \int_C f_2 ds = \int_C (f_1 + f_2) ds$ ,
- $\int_C \lambda f ds = \lambda \int_C f ds$ ,  $\lambda \in \mathbb{R}$ .

# 6.2 Applications Of Path Integrals

Suppose that  $\delta = \delta(x, y, z)$  which is a density function.

Mass

$$M = \int_C \delta(x, y, z) ds.$$

First Moments About the Coordinate Plane

- $M_{yz} = \int_C x \delta ds$
- $\bullet \ \overline{M_{xz}} = \int_C y \delta ds$
- $M_{xy} = \int_C z \delta ds$

Coordinates of Center of Mass

- $\bar{x} = \frac{M_{yz}}{M}$
- $\bar{y} = \frac{M_{xz}}{M}$
- $\bar{z} = \frac{M_{xy}}{M}$

Moments of Inertia About Axes

- $I_x = \int_C (y^2 + z^2) \delta ds$
- $I_x = \int_C (x^2 + z^2) \delta ds$
- $I_x = \int_C (x^2 + y^2) \delta ds$

# 7 Vector Line Integrals

#### 7.1 Vector Line Integrals

**Vector Line Integrals** Vector line integrals are different from scalar line integrals in the sense that to define a vector line integral, we must specify a direction along the path or curve C.

Computing a Vector Line Integral Let c(t) parametrise an oriented curve C for  $t \in [a, b]$ . Then,

$$\int_{C} F \cdot ds = \int_{a}^{b} F(c(t)) \cdot c'(t).$$

**Link to Path Integrals** Suppose that C is a smooth curve with a parametrisation c(t) for  $t \in [a, b]$  where c(t) is continuously differentiable and  $c'(t) \neq 0$  for all  $t \in [a, b]$ .

Then, c'(t) is a non-zero tangent vector pointing in the forward direction and the unit tangent vector is

$$T(c(t)) = \frac{c'(t)}{||c'(t)||}.$$

Then,

$$\int_C F \cdot ds = \int_C F \cdot T ds.$$

**Summing Paths** Suppose that C is made of n finitely many paths  $C_i$ . Then,  $C = \sum_{i=1}^{n} C_i$ . Note that all the curves must be joined end to end. Then,

$$\int_{C} F \cdot ds = \sum_{i}^{n} \int_{C_{i}} F \cdot ds.$$

Work and Other Alternative Notations Suppose that c(t) = (x(t), y(t), z(t)) and F = (M, N, P). Then, we denote work as any of the following notations

$$W = \int_{C} F \cdot ds$$

$$= \int_{a}^{b} \left( M \frac{dx}{dt} + N \frac{dy}{dt} + P \frac{dz}{dt} \right) dt$$

$$= \int_{C} M ds + N ds + P ds.$$

#### Properties fo Line Integrals

- Linearity
- Reversing Orient
- Additivity

Flow Integrals and Circulation Suppose that F represents a velocity field of a fluid flowing through a region in space. Then, the flow across a curve may be defined as the following

Flow = 
$$\int_{a}^{b} F \cdot \hat{T} ds$$
.

This integral is called the flow integral. If the curve is a closed loop then this is called the *circulation* around the curve.

Flux in the Plane If C is a smooth closed curve in the domain of a continuous vector field F = M(x, y)i + N(x, y)j + N(x, y)j and, n is the outward pointing unit-normal on C then, the flux of f across C is the following expression

$$\int_C F \cdot \hat{n} ds.$$

Calculating Flux Across a Smooth Closed Plane Curve Suppose that F = Mi = Nj. Let G = -N, M

Flux of 
$$F$$
 across  $C = \oint_C M dy - N dx = \oint_C G ds$ 

# 7.2 Fundamental Theorem of Line Integrals

**Gradient Fields** A vector field F is called a gradient vector field if there exists a real-valued function  $\phi$  such that  $F = \nabla \phi$ . That is,

$$\begin{pmatrix} M \\ N \\ P \end{pmatrix} = \begin{pmatrix} \frac{\partial \phi}{\partial x} \\ \frac{\partial \phi}{\partial y} \\ \frac{\partial \phi}{\partial z} \end{pmatrix}.$$

If such a function  $\phi$  exists then  $\phi$  is called the potential function of F where, F is conservative.

Fundamental Theorem for Gradient Vector Fields If  $F = \nabla \phi$  on a domain  $\mathcal{D}$ . Then, for all oriented curves C in  $\mathcal{D}$  with an initial point P and a terminal point Q,

$$\int_C F \cdot ds = \phi(Q) - \phi(P).$$

The integral is independent of the path.

Cross Partials of Gradient Vector Fields are Equal Let  $F = (F_1, F_2, F_3)$  be a gradient vector field whose components have continuous partial derivatives. Then, the cross partials are equal. That is,

$$\frac{\partial F_1}{\partial y} = \frac{\partial F_2}{\partial x},$$
$$\frac{\partial F_2}{\partial z} = \frac{\partial F_3}{\partial y},$$
$$\frac{\partial F_3}{\partial x} = \frac{\partial F_1}{\partial z}.$$

Equivalently,

$$\nabla \times F = 0.$$

#### 7.3 Green's Theorem

Green's Theorem connects double integrals and line integrals.

Green's Theorem: Flux Divergence or Normal Form Let D be a bounded simple region in  $\mathbb{R}^2$  with a nonempty interior whose boundary consists of a finite number of smooth curves. Let C be the boundary of D with a positive (counter-clockwise) direction. Let F = Mi + Nj be a continuously different boundary vector field on D.

Then, the outward flux of F across the curve C equals the double integral of divergence  $\nabla \cdot F$  over D. That is,

$$\oint_C (F \cdot \hat{n}) ds = \oint_C -N dx + M dy = \int \int_D \left( \frac{\partial M}{\partial x} + \frac{\partial N}{\partial Y} \right).$$

Once again, note the assumptions:

- D is bounded and simple with a non-empty interior,
- $\bullet$  The boundary C is oriented in the positive (counter-clockwise) direction, and is the finite union of smooth curves,
- The vector field F is continuously differentiable on D.

Green's Theorem: Circulation-Curl or Tangential Form Let D be a bounded simple region in  $\mathbb{R}^2$  with a nonempty interior whose boundary consists of a finite number of smooth curves. Let C be the boundary of D with a positive (counter-clockwise) direction. Let F = Mi + Nj be a continuously different boundary vector field on D.

Then, the counter-clockwise circulation of F around C equals the double integral of  $(\nabla \times F) \cdot k$  over D. That is,

$$\oint_C (F \cdot \hat{T}) ds = \oint_C M dx + N dy = \int \int_D \left( \frac{\partial N}{\partial x} - \frac{\partial M}{\partial y} \right) dx dy.$$

**Area of a Region** Let D be a simple and bounded region with a non-empty interior and let C be the boundary of D which is the finite union of smooth curves. Then, the area of D can be calculated as such

$$Area(D) = \frac{1}{2} \oint_C (-ydx + xdy).$$