

Functional Analysis

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Chapter 1

Metric Spaces

1.0.1 Notes

- "Soft Analysis" instead of "Hard Analysis"
- Harmonic Analysis - Study of Fourier Series
- Hilbert Spaces and Banach Spaces
- Limit Function ? (Major topic for analysis of sequences)
- Spaces of function? It is a linear space equipped with a norm, akin to a vector.
- Complete Space - (Limit actually exists)
- Spaces of functions can be infinite dimensional

1.0.2 Metric Spaces

- In Real Analysis, functions are defined on the Real Line \mathbb{R} .
- Limit and other functions use the distance function

$$d(x, y) = |x - y|$$

- For more general spaces, Real Line \mathbb{R} is replaced with abstract set \mathbb{X} .

Definition 1.0.1 A *Metric Space* is defined as an ordered pair (\mathbb{X}, d) . Where d (distance function) follows following properties:

1. d is a **real-valued, finite** (can also be infinite, if two functions are infinitely placed) and **non-negative** function
2. $d(x, y) = 0$ iff $x = y$
3. $d(x, y) = d(y, x)$

4. *Triangle Inequality:* $d(x, y) \leq d(x, z) + d(z, y)$ or in general,

$$d(x_0, x_n) \leq d(x_0, x_1) + d(x_1, x_2) + \dots + d(x_{n-1}, x_n)$$

Definition 1.0.2 Suppose $Y \subset X$, the **restriction of the Metric in X to Y** is $\tilde{d} = d|_{Y \times Y}$. Where \tilde{d} is the metric induced on Y by d and the new metric space is (Y, \tilde{d})

1.0.3 Examples of Metric Spaces

- Some examples: Real Line \mathbb{R} , Euclidean Plane $\mathbb{R} \times \mathbb{R}$, Sequence Spaces, Unitary Spaces (Complex Numbers)
- Distance function for \mathbb{R}^n , $d = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$
- For \mathbb{C} (complex numbers), it is called n -dimensional unitary Space \mathbb{C}^n consisting of ordered n -tuple complex numbers, having form

$$v = (x_1 + y_1 i, x_2 + y_2 i, \dots, x_n + y_n i) \text{ and } d(\bar{x}, \bar{y}) = |\bar{x} - \bar{y}|$$

Sequences

- Sequence Space l^∞ , consisting of sequences which are bounded.

Definition 1.0.3 Let $\zeta_j \in l^\infty$ which is a **Bounded Sequence** then $\forall j, |\zeta_j| \leq c_x$ where c_x is a real number.

- Example : $\{1, \frac{1}{2}, \frac{1}{2^2}, \frac{1}{2^3}, \dots\}$, here $c_x = 1$ i.e. $|\zeta_j| \leq 1$.

Definition 1.0.4 **Distance Function** for sequences is defined as

$$d(x, y) = \sup_{j \in \mathbb{N}} |\zeta_j - \eta_j|$$

where $x, y \in l^\infty$ with $\zeta_j \in x$ and $\eta_j \in y$. Also, \sup (supremum) is the least upper bound.

Claim. Distance function of a sequence can be defined as \sup (supremum).

Proof 1 Let $x = \{\zeta_j\}$, $y = \{\eta_j\}$ and $z = \{\gamma_j\}$. Then using triangular inequality,

$$|\zeta_j - \eta_j| \leq |\zeta_j - \gamma_j| + |\gamma_j - \eta_j|$$

Taking supremum both sides,

$$\begin{aligned} &= \sup_{j \in \mathbb{N}} |\zeta_j - \eta_j| \leq \sup_{j \in \mathbb{N}} |\zeta_j - \gamma_j| + \sup_{j \in \mathbb{N}} |\gamma_j - \eta_j| \\ &= f(x, y) \leq f(x, z) + f(z, y) \end{aligned}$$

Hence proved.

Chapter 2

Examples of Metric Spaces

2.0.1 Some more Examples of Metric Spaces

Definition 2.0.1 *Function Space* $\mathbb{C}[a, b]$. where \mathbb{C} suggests continuous. It contains the abstract set X of functions which are based on a parameter t i.e. $f(t)$ and distance function (d) is defined as,

$$d(x, y) = \max_{t \in J} |x(t) - y(t)|$$

where J is $[a, b]$.

Proof 2 It satisfies all the first 3 properties of distance functions. So, let's prove **Triangular Inequality**:

$$\begin{aligned} |x(t) - y(t)| &\leq |x(t) - z(t)| + |z(t) - y(t)| \\ &\leq \sup |x(t) - z(t)| + \sup |z(t) - y(t)| \end{aligned}$$

as RHS is independent of t , take sup on both sides

$$d(x, y) = \sup |x(t) - y(t)| \leq d(x, z) + d(z, y)$$

Hence Proved.

- **Note:** Sometimes sup may exist but not max.

Definition 2.0.2 *Discrete Metric Space*. Consider any set X , the discrete metric d ,

$$\begin{aligned} d(x, x) &= 0 \\ d(x, y) &= 1, \text{ for } (x \neq y) \end{aligned}$$

Proof 3 Let a, b, c be three distinct points. It already follows first 3 properties of distance.

$$d(a, c) = 1, d(a, b) = 1 \text{ and } d(b, c) = 1 \quad (2.1)$$

Therefore,

$$d(a, c) = 1 \leq (d(a, b) = 1) + (d(b, c) = 1) = 2 \quad (2.2)$$

Hence Proved.

- If we use the notion of circle i.e. all points which have equal distance from some single point. Then that circle varies according to the Space in consideration.
- **Comparison test** is to bound the series sum, and show that a particular series is convergent.

2.0.2 Some Problems

- **Problem 1:** Consider a distance function between two series $\tilde{x}, \tilde{y} \in l^\infty$:

$$d(\tilde{x}, \tilde{y}) = \sum_{j=1}^{\infty} \frac{1}{2^j} \frac{|x_j - y_j|}{1 + |x_j - y_j|}$$

Prove that it is a valid metric distance.

Proof 4 This function is bounded by $\sum_{j=1}^{\infty} \frac{1}{2^j} = 1$

Let $f(x) = \frac{x}{1+x}$, as $f'(t) = \frac{1}{1+t^2} > 0$ f is monotonically increasing

$$\begin{aligned} t_1 &< t_2 \\ f(t_1) &< f(t_2) \\ \text{as, } |a+b| &\leq |a| + |b| \\ f(|a+b|) &\leq f(|a| + |b|) \\ \frac{|a+b|}{1+|a+b|} &\leq \frac{|a|+|b|}{1+|a|+|b|} \\ &\leq \frac{|a|}{1+|a|+|b|} + \frac{|b|}{1+|a|+|b|} \\ &\leq \frac{|a|}{1+|a|} + \frac{|b|}{1+|b|} \end{aligned}$$

Let $a = x_j - y_j$ and $b = y_j - z_j$, where $z = (z_j)$

$$\begin{aligned} a+b &= x_j - z_j \\ \frac{|x_j - z_j|}{1+|x_j - z_j|} &\leq \frac{|x_j - y_j|}{1+|x_j - y_j|} + \frac{|y_j - z_j|}{1+|y_j - z_j|} \end{aligned}$$

Multiplying by $\frac{1}{2^j}$ on both sides and taking summation,

$$\begin{aligned} \sum_{j=1}^{\infty} \frac{1}{2^j} \frac{|x_j - z_j|}{1+|x_j - z_j|} &\leq \sum_{j=1}^{\infty} \frac{1}{2^j} \frac{|x_j - y_j|}{1+|x_j - y_j|} + \sum_{j=1}^{\infty} \frac{1}{2^j} \frac{|y_j - z_j|}{1+|y_j - z_j|} \\ d(x, z) &\leq d(x, y) + d(y, z) \end{aligned}$$

Hence Proved.

- **(Prob-7 ex-1.1, Kreyzig):** If A is the sub-sequence l^∞ consisting of all sequences of 0's and 1's. What is the induced metric on A ?

Solution: It is the discrete metric distance.

Definition 2.0.3 Hamming Distance: The number of places two ordered tuples (let's of length 3) of 0's and 1's differs.

Proof 5 Hint: We can prove (if this is a valid distance metric) using enumeration for every possible bit possible i.e. 0 or 1.

2.0.3 Space of Bounded Functions

- Space of Bounded Functions is denoted by $B(A)$.
- $d(x, y) = \sup_{t \in A} |x(t) - y(t)|$

Definition 2.0.4 The ℓ^p space where $p \geq 1$ is a fixed real number, is called **Hilbert Sequence Space**. ℓ^p is a sequence space

$$\tilde{x} = (x_1, x_2, \dots, x_n, \dots) \quad (2.3)$$

$|x_1|^p + |x_2|^p + |x_3|^p + \dots$ converges,

$$\sum_{j=1}^{\infty} |x_j|^p < \infty$$

$$d(x, y) = \left(\sum_{j=1}^{\infty} |x_j - y_j|^p \right)^{\frac{1}{p}}$$

Definition 2.0.5 *Hilbert Space, Hölder's Inequality:*

$$\sum_{j=1}^{\infty} |x_j y_j| \leq \left(\sum_{k=1}^{\infty} |x_k|^p \right)^{\frac{1}{p}} \left(\sum_{m=1}^{\infty} |y_m|^q \right)^{\frac{1}{q}} \quad (2.4)$$

where $p > 1$ and $\frac{1}{p} + \frac{1}{q} = 1$.

Chapter 3

Inequalities

3.0.1 Some Inequalities

Definition 3.0.1 *Young's Inequality*

$$\alpha\beta \leq \frac{\alpha^p}{p} + \frac{\beta^q}{q} \quad (3.1)$$

Proof 6 Let $f : [0, \infty) \mapsto \mathbb{R}$

$$\begin{aligned} f(\alpha) &= \frac{\alpha^p}{p} + \frac{\beta^q}{q} - \alpha\beta \text{ for a fixed } \beta \\ f'(\alpha) &= \alpha^{p-1} - \beta = 0 \\ \alpha &= \beta^{\frac{1}{p-1}} \end{aligned}$$

now $\frac{1}{p} + \frac{1}{q} = 1$, therefore $\frac{q}{p} = \frac{1}{p-1}$

$$\begin{aligned} \alpha &= \beta^{\frac{q}{p}} \\ f''(\alpha) &= (p-1)\alpha^{p-2} > 0 \end{aligned}$$

$f(\alpha)$ has a min at $\alpha = \beta^{\frac{q}{p}}$.

$$\begin{aligned} 0 = f(\beta^{\frac{q}{p}}) &\leq f(\alpha) = \frac{\alpha^p}{p} + \frac{\beta^q}{q} - \alpha\beta \\ \alpha\beta &\leq \frac{\alpha^p}{p} + \frac{\beta^q}{q} \end{aligned}$$

Hence Proved.

Definition 3.0.2 *Hölder's inequality*

Proof 7 Let $(x_n) \in \ell^p$ and $(y_n) \in \ell^q$. Then trivial case:

$$\sum_{k=1}^{\infty} |x_k|^p = 0 \text{ or } \sum_{k=1}^{\infty} |y_k|^q = 0 \quad (3.2)$$

Assume both these sums are not equal to 0, and set in **Young's Inequality**

$$\alpha = \frac{|x_k|}{(\sum_{k=1}^{\infty} |x_k|^p)^{\frac{1}{p}}} \text{ and } \beta = \frac{|y_k|}{(\sum_{k=1}^{\infty} |y_k|^q)^{\frac{1}{q}}} \quad (3.3)$$

$$\alpha\beta = \frac{|x_k y_k|}{(\sum_{k=1}^{\infty} |x_k|^p)^{\frac{1}{p}} (\sum_{k=1}^{\infty} |y_k|^q)^{\frac{1}{q}}} \leq \frac{1}{p} \cdot \frac{|x_k|^p}{\sum_{k=1}^{\infty} |x_k|^p} + \frac{1}{q} \cdot \frac{|y_k|^q}{\sum_{k=1}^{\infty} |y_k|^q}$$

Sum both sides from 1 to ∞

$$\begin{aligned} \frac{\sum_{k=1}^{\infty} |x_k y_k|}{(\sum_{k=1}^{\infty} |x_k|^p)^{\frac{1}{p}} (\sum_{k=1}^{\infty} |y_k|^q)^{\frac{1}{q}}} &\leq \frac{1}{p} \cdot \frac{\sum_{k=1}^{\infty} |x_k|^p}{\sum_{k=1}^{\infty} |x_k|^p} + \frac{1}{q} \cdot \frac{\sum_{k=1}^{\infty} |y_k|^q}{\sum_{k=1}^{\infty} |y_k|^q} \\ &= \frac{1}{p} + \frac{1}{q} = 1 \end{aligned}$$

Therefore,

$$\sum_{k=1}^{\infty} |x_k y_k| \leq (\sum_{k=1}^{\infty} |x_k|^p)^{\frac{1}{p}} (\sum_{k=1}^{\infty} |y_k|^q)^{\frac{1}{q}}$$

Hence Proved.

Definition 3.0.3 Minkowski Inequality for sequences $p > 1$, (x_n) and (y_n) are both sequence in ℓ^p .

$$(\sum_{k=1}^{\infty} |x_k + y_k|^p)^{\frac{1}{p}} \leq (\sum_{k=1}^{\infty} |x_k|^p)^{\frac{1}{p}} + (\sum_{k=1}^{\infty} |y_k|^p)^{\frac{1}{p}} \quad (3.4)$$

Proof 8 Let $q = \frac{p}{p-1}$ and $\sum_{k=1}^{\infty} |x_k + y_k|^p \neq 0$,

$$\begin{aligned} \sum_{k=1}^{\infty} |x_k + y_k|^p &= \sum_{k=1}^{\infty} |x_k + y_k|^{p-1} |x_k + y_k| \\ &\leq \sum_{k=1}^{\infty} |x_k + y_k|^{p-1} |x_k| + \sum_{k=1}^{\infty} |x_k + y_k|^{p-1} |y_k| \text{ using triangle ineq.} \end{aligned}$$

From the **Hölder's inequality** $\sum_{m=1}^{\infty} |x_m y_m| \leq (\sum_{m=1}^{\infty} |x_m|^p)^{\frac{1}{p}} (\sum_{m=1}^{\infty} |y_m|^q)^{\frac{1}{q}}$, taking $|x_k + y_k|^{p-1}$ as y_m and $|x_k|$ as x_m ,

$$\leq (\sum_{k=1}^{\infty} |x_k|^p)^{\frac{1}{p}} (\sum_{k=1}^{\infty} (|x_k + y_k|^{p-1})^q)^{\frac{1}{q}} + (\sum_{k=1}^{\infty} |y_k|^p)^{\frac{1}{p}} (\sum_{k=1}^{\infty} (|x_k + y_k|^{p-1})^q)^{\frac{1}{q}}$$

As $\frac{1}{p} + \frac{1}{q} = 1$, therefore $(p-1)q = p$,

$$\begin{aligned}
 &\leq \left(\sum_{k=1}^{\infty} |x_k|^p \right)^{\frac{1}{p}} \left(\sum_{k=1}^{\infty} |x_k + y_k|^p \right)^{\frac{1}{q}} + \left(\sum_{k=1}^{\infty} |y_k|^p \right)^{\frac{1}{p}} \left(\sum_{k=1}^{\infty} |x_k + y_k|^p \right)^{\frac{1}{q}} \\
 &\leq \left(\left(\sum_{k=1}^{\infty} |x_k|^p \right)^{\frac{1}{p}} + \left(\sum_{k=1}^{\infty} |y_k|^p \right)^{\frac{1}{p}} \right) \left(\sum_{k=1}^{\infty} |x_k + y_k|^p \right)^{\frac{1}{q}} \\
 \left(\sum_{k=1}^{\infty} |x_k + y_k|^p \right)^{1-\frac{1}{q}} &\leq \left(\sum_{k=1}^{\infty} |x_k|^p \right)^{\frac{1}{p}} + \left(\sum_{k=1}^{\infty} |y_k|^p \right)^{\frac{1}{p}}
 \end{aligned}$$

Therefore,

$$\left(\sum_{k=1}^{\infty} |x_k + y_k|^p \right)^{\frac{1}{p}} \leq \left(\sum_{k=1}^{\infty} |x_k|^p \right)^{\frac{1}{p}} + \left(\sum_{k=1}^{\infty} |y_k|^p \right)^{\frac{1}{p}} \quad (3.5)$$

Hence Proved.

3.0.2 Sequence Metric Space

Definition 3.0.4 l^p where $p \geq 1$ is a metric space where,

$$d(x, y) = \left(\sum_{j=1}^{\infty} |x_j - y_j|^p \right)^{\frac{1}{p}} \quad (3.6)$$

provided $\sum_{j=1}^{\infty} |x_j|^p < \infty$.

Proof 9 To prove triangle inequality $d(x, y) \leq d(x, z) + d(z, y)$ for metric distance, i.e. to prove:

$$\left(\sum_{j=1}^{\infty} |x_j - y_j|^p \right)^{\frac{1}{p}} \leq \left(\sum_{j=1}^{\infty} |x_j - z_j|^p \right)^{\frac{1}{p}} + \left(\sum_{j=1}^{\infty} |z_j - y_j|^p \right)^{\frac{1}{p}} \quad (3.7)$$

Hint: Substitute $x_k = x_j - z_j$ and $y_k = z_j - y_j$ in **Minkowski Inequality**.

- **(Ex - 1.2, Kreyzig):** Consider a distance function between two series $\tilde{x}, \tilde{y} \in l^\infty$:

$$d(\tilde{x}, \tilde{y}) = \sum_{j=1}^{\infty} \frac{1}{2^j} \frac{|x_j - y_j|}{1 + |x_j - y_j|}$$

Prove that by replacing $\frac{1}{2^j}$ by $\mu_j > 0$, such that $\sum \mu_j$ converges, it is a metric space.

Proof 10 We need to show that these infinite terms converge. As $\sum \mu_j$ converges and the term multiplied is < 1 therefore, the series converge.

Chapter 4

Convergence, Cauchy Sequence and Completeness

4.0.1 Notes

- As n increases the distance between the terms (at very large n) comes closer and closer.

4.0.2 Some Definitions

Definition 4.0.1 Limit of a sequence. A sequence in (X, d) is said to converge or a convergent sequence if \exists an $x \in X$ such that $\lim_{n \rightarrow \infty} d(x_n, x) = 0$. x is called the limit of the sequence. If x doesn't exist in X (the metric space) then limit does not exist.

Definition 4.0.2 Bounded Set. A non-empty set $M \subset X$ is said to be bounded if its diameter is finite. Where diameter (δ) is -

$$\delta(M) = \sup_{x, y \in M} d(x, y) \quad (4.1)$$

Definition 4.0.3 Bounded Sequence. A convergent sequence in X is bounded if:

1. Its limit is unique.
2. If $x_n \rightarrow x$ (at ∞) and $y_n \rightarrow y$ then $d(x_n, y_n) = d(x, y)$

Definition 4.0.4 Cauchy Sequence. A sequence is a Cauchy Sequence in a metric space (X, d) if $\forall \epsilon \exists$ an N such that $d(x_m, x_n) < \epsilon \forall m, n > N$. **Note:** ϵ may not belong to the sequence and can be any arbitrarily small number.

Definition 4.0.5 Complete Metric Space. A metric space is complete if every Cauchy sequence converge.

Example 4.1: $X = \mathbb{R} - \{a\}$ is an incomplete metric space because we can find any cauchy sequence which converges at a but $a \notin X$.

Example 4.2: Real Numbers after removal of irrational numbers i.e. a set \mathbb{Q} of rational numbers.

Example 4.3: (a, b) is not complete metric space because there can exist a cauchy sequence whose limit is a or b which are not included.

Note: $d(x, y) = |x - y|$ is said to be **usual metric**.

Example 4.4: $(0, 1]$ with usual metric, and let the cauchy sequence be $x_n = \frac{1}{n}$ which converges to 0 and 0 does not exist in the metric space. So convergence is a property of metric space "more".

Lemma 1 *Every convergent sequence is a cauchy sequence.*

Proof 11 *If $x_n \rightarrow x \forall \epsilon > 0$ and $x_n > N$ the sequence converges.*

$$\begin{aligned} d(x_n, x) &< \frac{\epsilon}{2} \text{ where } n > N \\ d(x_m, x_n) &\leq d(x_m, x) + d(x_n, x) \\ d(x_m, x_n) &\leq \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon \text{ where } n, m > N \end{aligned}$$

Problem 4.1: Prove that \mathbb{R}^n and \mathbb{C}^n are complete metric spaces.

Proof 12 *Hint: Use the fact that \mathbb{R} is complete.*

Problem 4.2: Show that the set \mathbb{X} of all the integers with metric d defined as $d(x, y) = |x - y|$ is a complete metric space.

Proof 13 *To proof completeness,*

1. **Find all cauchy sequence.** Note: $1, 2, 3, \dots$ is not a cauchy sequence. A constant sequence is a cauchy sequence or a sequence which eventually becomes a constant sequence is a cauchy sequence (these are the only cauchy sequence).

Note: Discrete Metric is a complete metric space, as the only cauchy sequence is the sequence of constants which is convergent.

Chapter 5

Normed Linear Spaces

5.1 Vector Space

Definition 5.1.1 (Vector Space). A vector space (or linear space) over a field K is a non-empty set X of elements x, y, \dots (called vectors) together with two algebraic operations. These operations are called vector addition and multiplication of vectors by scalars, that is, by elements of K .

Vector addition associates with every ordered pair (x, y) of vectors a vector $x + y$, called the sum of x and y , in such a way that the following properties hold. Vector addition is commutative and associative, that is for all vectors we have

$$\begin{aligned}x + y &= y + x \\x + (y + z) &= (x + y) + z;\end{aligned}$$

furthermore, there exists a vector 0 , called the zero vector, and for every x there exists a vector $-x$, such that for all vectors we have

$$\begin{aligned}x + 0 &= x \\x + (-x) &= 0\end{aligned}$$

Multiplication by scalars associates with every vector x and scalar α a vector αx (also written as $x\alpha$), called the product of α and x , in such a way that for all vectors x, y and scalars α, β we have

$$\begin{aligned}\alpha(\beta x) &= (\alpha\beta)x \\1x &= x\end{aligned}$$

and the distributive laws

$$\begin{aligned}\alpha(x + y) &= \alpha x + \alpha y \\(\alpha + \beta)x &= \alpha x + \beta x\end{aligned}$$

Definition 5.1.2 Field. A field is a set F , containing at least two elements, on which two operations $+$ and \cdot (called addition and multiplication, respectively) are defined so that for each pair of elements x, y in F there are unique elements $x + y$ and $x \cdot y$ (often written as xy) in F for which the following conditions hold for all elements x, y, z in F :

1. $x + y = y + x$ (commutativity of addition)
2. $(x + y) + z = x + (y + z)$ (associativity of addition)
3. There is an element $0 \in F$, called zero, such that $x + 0 = x$. (existence of an additive identity)
4. For each x , there is an element $-x \in F$ such that $x + (-x) = 0$. (existence of additive inverses)
5. $xy = yx$ (commutativity of multiplication)
6. $(x \cdot y) \cdot z = x \cdot (y \cdot z)$ (associativity of multiplication)
7. $(x + y) \cdot z = x \cdot z + y \cdot z$ and $x \cdot (y + z) = x \cdot y + x \cdot z$ (distributivity)
8. There is an element $1 \in F$, such that $1 \neq 0$ and $x1 = x$ (existence of a multiplicative identity)
9. If $x \neq 0$, then there is an element $x^{-1} \in F$ such that $x \cdot x^{-1} = 1$ (existence of multiplicative inverses)

From the definition we see that vector addition is a mapping $X \times X \rightarrow X$, whereas multiplication by scalars is a mapping $K \times X \rightarrow X$.

K is called the **scalar field** (or coefficient field) of the vector space X , and X is called a **real vector space** if $K = \mathbb{R}$ (the field of real numbers), and a **complex vector space** if $K = \mathbb{C}$ (the field of complex numbers).

The use of 0 for the scalar 0 as well as for the zero vector should cause no confusion, in general. If desirable for clarity, we can denote the zero vector by θ .

Definition 5.1.3 Eigen Value of Transformation. Let $T(x) = \alpha x$ where T is transformation and x is vector then α is eigen value of transformation.

5.2 Closed Set and Closure

Definition 5.2.1 Closed Set. A set containing all of its limit points is called closed set.

Definition 5.2.2 Closure of a Set. A set with all of its limit points \bar{M} is called closed set.

If there is a set, then a sequence can always be constructed such that its limit is some arbitrary limit point. For e.g. take a $x_i \in \text{Neighbour}(\text{dis}(\frac{1}{2i}))$ for all $i \in \mathbb{N}$.

Lemma 2 A subspace M of a complete metric space X is itself complete iff M is closed.

5.3 Vector Subspaces

Definition 5.3.1 A *subspace of a vector space* X is a non-empty subset Y i.e. $Y \subset X$ if $\forall y_1, y_2 \in Y$, $\alpha y_1 + \beta y_2 \in Y$.

Definition 5.3.2 A *span of M X* is the set of all linear combination of vectors of M , i.e. $\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_n x_n$.

Definition 5.3.3 A set of vectors is called *linear independent* iff $\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_n x_n = 0 \implies \alpha_1 = \alpha_2 = \dots = \alpha_n = 0$ where α_i s are scalars. So parallel vectors are not independent.

If the number of dimensions (or $\dim(X)$) is n , then $n + 1$ vectors are surely linearly dependent.

Definition 5.3.4 Hamel Basis. If X is any vector space not necessarily finite dimensional and B is a linearly independent subset of X which spans X then B is called Hamel Basis. "Can be infinitely dimensional."

Definition 5.3.5 A *Normed Space* X is a vector space with a norm defined on it.

Note: Functional is different from Function.

Definition 5.3.6 A *norm* is a real valued function X whose value at any $x \in X$ is denoted by $\|\bar{x}\|$ with following properties :-

$$N_1 : \|\bar{x}\| \geq 0$$

$$N_2 : \|\bar{x}\| = 0 \iff \bar{x} = 0$$

$$N_3 : \|\alpha \bar{x}\| = \alpha \|\bar{x}\|$$

$$N_4 : \|\bar{x} + \bar{y}\| \geq \|\bar{x}\| + \|\bar{y}\|$$

where \bar{x}, \bar{y} are arbitrary vectors.

Definition 5.3.7 The norm defines a metric d on X .

$$d(x, y) = \|x - y\| \tag{5.1}$$

which is the metric induced by the norm $(X, \|\cdot\|)$.

Note: $\|x\|$ is a generalisation of length of a vector $\|x\| = |x|$

Note: Normed and Banach spaces are metric spaces.

Some examples:

$$1. \mathbb{R}^n \text{ with } \|x\| = (\sum_{i=1}^n |x_i|^2)^{\frac{1}{2}}$$

2. l^p with $\|x\| = (\sum_{j=1}^{\infty} |x_j|^p)^{\frac{1}{p}}$
3. l^{∞} with $\|x\| = \sup_j |x_j|$
4. $C[a, b]$ with $\|x\| = \max_{t \in [a, b]} |x_t|$

Definition 5.3.8 A metric d induced by a norm in a normed space X satisfies

$$\begin{aligned} d(x + a, y + a) &= d(x, y) \\ d(\alpha x, \alpha y) &= |\alpha| d(x, y) \end{aligned}$$

$\forall x, y, a \in X$ and every scalar α , as,

$$\begin{aligned} d(x + a, y + a) &= \|(x + a) - (y + a)\| = \|x - y\| = d(x, y) \\ d(\alpha x, \alpha y) &= \|\alpha x - \alpha y\| \\ &= |\alpha| \|x - y\| = |\alpha| d(x, y) \end{aligned}$$

Definition 5.3.9 A subspace Y of a normed space X is a subspace of X considered as a vector space with the norm obtained by restricting the norm on X to the subset Y . The norm is the **induced norm**.

Problem 5.3.1: Describe the span of $M = \{(1, 1, 1), (0, 0, 2)\}$ in \mathbb{R}^3 .

Solution 5.3.1: Span of $M = \{\alpha \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + \beta \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} : \alpha, \beta \in \mathbb{R}\} = \begin{bmatrix} \alpha \\ \alpha \\ \alpha + 2\beta \end{bmatrix} : \alpha, \beta \in \mathbb{R}$. Or the subspace $x = y$.

Problem 5.3.2: Which of the subsets of \mathbb{R}^3 consists of a subspace of \mathbb{R}^3 . All: x with $x_1 = x_2$ & $x_3 = 0$.

Solution 5.3.2: Let $x = \begin{bmatrix} x_1 \\ x_1 \\ 0 \end{bmatrix}$ and $y = \begin{bmatrix} y_1 \\ y_1 \\ 0 \end{bmatrix}$ where $x, y \in w$. Then $\alpha x + \beta y = \begin{bmatrix} \alpha x_1 + \beta y_1 \\ \alpha x_1 + \beta y_1 \\ 0 \end{bmatrix} \in w$. Therefore, $\alpha x + \beta y \in w$ hence it is a subspace.

Chapter 6

Vector Basis

Note: If a vector under study is ∞ dimensional, then the Hamel Basis is **uncountable**.

Definition 6.0.1 Schauder Basis. e_1, e_2, \dots is an infinite basis.

6.1 Incomplete Normed Space

Definition 6.1.1 An incomplete normed space and its completion $L^2[a, b]$ (square integrable norm). The vector space of all continuous real-valued functions on $[a, b]$ where the norm is defined as

$$\|x\| = \left\{ \int_a^b (x(t))^2 dt \right\}^{\frac{1}{2}} \quad (6.1)$$

Distance induced by this norm

$$\|x_n - x_m\|^2 = \int_0^1 (x_n(t) - x_m(t))^2 dt \quad (6.2)$$

Note: Here basically prove that the given normed space is incomplete by finding a cauchy sequence which doesn't have a limit inside the metric space X . And then complete it using $L^p[a, b]$.

6.2 Finite Dimensional Normed Space

Theorem 1 Let X be a finite dimensional linear space with basis $\{x_1, x_2, \dots, x_n\}$. Then there is a constant $m > 0$ such that for every choice of scalars $\alpha_1, \alpha_2, \dots$

$$m \sum_{j=0}^n |\alpha_j| \leq \left\| \sum_{j=1}^n \alpha_j x_j \right\| \quad (6.3)$$

Proof 14 If $\sum_{j=0}^n |\alpha_j| = 0$ then $|\alpha_j| = 0$ for all $j = 1, 2, \dots, n$. Assume that $\sum_{j=0}^n |\alpha_j| \neq 0$. Let's prove the theorem for scalars $\{\alpha_1, \alpha_2, \dots\}$ with the condition that $\sum_{j=1}^n |\alpha_j| = 1$. Let $A = \{(\alpha_1, \alpha_2, \dots, \alpha_n) \in \mathbb{R}^n | \sum_{j=1}^n |\alpha_j| = 1\}$.

Define $f(\alpha_1, \alpha_2, \dots, \alpha_n) = \left\| \sum_{j=1}^n \alpha_j x_j \right\|$. For any $(\alpha_1, \alpha_2, \dots, \alpha_n)$ and $(\beta_1, \beta_2, \dots, \beta_n)$ consider $|f(\alpha_1, \alpha_2, \dots, \alpha_n) - f(\beta_1, \beta_2, \dots, \beta_n)|$ which is

$$\begin{aligned} \left| \left\| \sum_{j=1}^n \alpha_j x_j \right\| - \left\| \sum_{j=1}^n \beta_j x_j \right\| \right| &\leq \left\| \sum_{j=1}^n \alpha_j x_j - \sum_{j=1}^n \beta_j x_j \right\| = \left\| \sum_{j=1}^n (\alpha_j - \beta_j) x_j \right\| \\ &\leq \sum_{j=1}^n \|(\alpha_j - \beta_j) x_j\| = \sum_{j=1}^n |\alpha_j - \beta_j| \|x_j\| \\ &\leq \max_{1 \leq j \leq n} \|x_j\| \sum_{j=1}^n |\alpha_j - \beta_j| \end{aligned}$$

Let $(\mu_1, \mu_2, \dots, \mu_n) \in A$ be such that $f(\mu_1, \mu_2, \dots, \mu_n) = \inf \{f(\alpha_1, \alpha_2, \dots, \alpha_n) | (\alpha_1, \alpha_2, \dots, \alpha_n) \in A\}$. Let $m = f(\mu_1, \mu_2, \dots, \mu_n)$. In case this minimum is 0.

$$\left\| \sum_{j=1}^n \mu_j x_j \right\| = 0 \implies \sum_{j=1}^n \mu_j x_j = 0$$

but x_1, x_2, \dots are linearly independent therefore $\mu_j = 0$ for all $j = 1, 2, \dots, n$ i.e. it is a zero vector also summation must have been 1, hence a contradiction. Therefore $0 < m \leq f(\alpha_1, \alpha_2, \dots, \alpha_n)$.

$$\left\| \sum_{j=1}^n \mu_j x_j \right\| \leq \left\| \sum_{j=1}^n \alpha_j x_j \right\|$$

Let $\alpha_1, \alpha_2, \dots, \alpha_n$ be any collection of scalars, set $\beta = \sum_{j=1}^n |\alpha_j|$. Let $\beta > 0$, then $(\frac{\alpha_1}{\beta}, \frac{\alpha_2}{\beta}, \dots, \frac{\alpha_n}{\beta}) \in A$ as

$$\frac{\alpha_i}{\beta} = \frac{\alpha_i}{\sum_{j=1}^n |\alpha_j|}$$

taking summation on both sides, gives 1 hence it is in A. Therefore,

$$f(\alpha_1, \alpha_2, \dots, \alpha_n) = \left\| \sum_{j=1}^n \alpha_j x_j \right\| \tag{6.4}$$

$$= \left\| \sum_{j=1}^n \frac{\alpha_j}{\beta} x_j \right\| \beta \tag{6.5}$$

$$= \beta f\left(\frac{\alpha_1}{\beta}, \frac{\alpha_2}{\beta}, \dots, \frac{\alpha_n}{\beta}\right) > m\beta \tag{6.6}$$

Therefore,

$$m\beta = m \sum_{j=1}^n |\alpha_j| \leq \left\| \sum_{j=1}^n \alpha_j x_j \right\|$$

Hence Proved.

Note: Unit sphere is $S(0) = \{x \in X : \|x\| = 1\}$.

Definition 6.2.1 A subset A of a vector space is said to be convex if for any $x, y \in A \implies M = \{z \in X : z = \alpha x + (1 - \alpha)y, 0 \leq \alpha \leq 1\} \subset A$. M is a closed segment with boundary points x, y .

Definition 6.2.2 Equivalence of Norm. A norm $\|\cdot\|$ on a vector space X is said to be equivalent to $\|\cdot\|_0$ on X , if there are positive numbers a, b such that $\forall x \in X$ we have,

$$a\|x\|_0 \leq \|x\| \leq b\|x\|_0$$