

DEFINITION

*metric*

ANALYSIS

DEFINITION

*norm*

ANALYSIS

DEFINITION

*convex*

ANALYSIS

Let  $X$  be a set. A function  $d : X \times X \rightarrow \mathbb{R}$  is a metric on  $X$  if it satisfies the following conditions:

- $d(x, y) \geq 0$  for all  $x, y \in X$  and  $d(x, y) = 0$  if and only if  $x = y$ .
- $d(x, y) = d(y, x)$
- $d(x, z) \leq d(x, y) + d(y, z)$

Let  $X$  be a vector space over  $\mathbb{R}$  or  $\mathbb{C}$ . Then a norm on  $X$ , denoted by  $\|\cdot\|$  if a function  $\|\cdot\| : X \rightarrow [0, \infty)$  with the following properties

- $\|v\| \geq 0$  and  $\|v\| = 0$  if and only if  $v = 0$
- $\|\lambda v\| = |\lambda| \|v\|$  with  $\lambda \in \mathbb{R}$  or  $\mathbb{C}$ .
- $\|v + w\| \leq \|v\| + \|w\|$

A subset  $A$  of a vector space  $X$  is convex if for all  $x, y \in A$

$$tx + (1 - t)y \in A$$

for all  $t \in [0, 1]$

DEFINITION

*sequence*

ANALYSIS

DEFINITION

*converges*

ANALYSIS

DEFINITION

*Cauchy*

ANALYSIS

A sequence in  $X$  is a function  $X : \mathbb{N} \rightarrow X$ , commonly denoted by  $\{x_n\}_{n=1}^\infty$ .

Let  $\{x_n\}_{n=1}^\infty$  be sequence in  $X$  and  $x \in X$ . Then  $\{x_n\}_{n=1}^\infty$  converges to  $x$  provided for all  $\epsilon > 0$ , there exists  $N \in \mathbb{N}$  such that for all  $n \geq N$

$$d(x, x_n) < \epsilon$$

A sequence in a metric space  $X$  is a Cauchy if for every  $\epsilon > 0$ , there exists  $N$  such that

$$d(x_n, x_m) < \epsilon$$

for all  $n, m > N$ .

DEFINITION

*complete*

ANALYSIS

DEFINITION

*open ball*

ANALYSIS

DEFINITION

*closed ball*

ANALYSIS

A metric space  $(X, d)$  is called complete if every Cauchy sequence in  $(X, d)$  converges in  $(X, d)$ .

For all  $x \in X, r > 0$ , we define the open ball

$$B_r(x) := \{y \in X : d(y, x) < r\}$$

For all  $x \in X, r > 0$ , we define the closed ball

$$\bar{B}_r(x) := \{y \in X : d(y, x) \leq r\}$$

DEFINITION

*open*

ANALYSIS

DEFINITION

*closed*

ANALYSIS

DEFINITION

*closure*

ANALYSIS

A set  $U \subset X$  is open provided for every  $x \in U$ , we can identify a  $r > 0$  such that

$$B_r(x) \subset U$$

A set  $F \subset X$  is closed provided  $F^c = X \setminus F$  is open.

For any  $A \subset X$ , the closure of  $A$  is the smallest closed subset of  $X$  that contains  $A$ , denoted by  $\bar{A}$ . Moreover,

$$\bar{A} := \bigcap_{F \text{ closed } \subset X; A \subset F} F$$

Equivalently,

$$\bar{A} := \{x \in X : \text{there exists a sequence in } A \text{ that converges to } x\}$$



DEFINITION

*dense*

ANALYSIS

DEFINITION

*seperable*

ANALYSIS

DEFINITION

*continuous*

ANALYSIS

A set  $A \subset X$  is dense if  $\bar{A} = X$ . In other words, for every point  $x \in X$ , there exists  $\{a_n \in A\}_{n=1}^{\infty}$  such that

$$a_n \rightarrow x$$

A metric space  $(X, d)$  is seperable if it has a countable dense subset.

Let  $x_0 \in X$  and  $f : X \rightarrow Y$  be a function. Then  $f$  is continuous at  $x_0$  if for all  $\epsilon > 0$ ,  $\exists \delta > 0$  such that

$$d_X(x, x_0) < \delta \implies d_Y(f(x), f(x_0)) < \epsilon$$

That is,

$$f(B_\delta(x_0)) \subset B_\epsilon(f(x_0))$$

Further,  $f$  is continuous on a subset  $E \subset X$  provided  $f$  is continuous at every point  $x \in E$ .

DEFINITION

*uniformly continuous*

ANALYSIS

DEFINITION

*upper semi-continuous / lower semi-continuous*

ANALYSIS

DEFINITION

*isometric embedding*

ANALYSIS

A function  $f : X \rightarrow Y$  is uniformly continuous if for all  $\epsilon > 0$ , there exists  $\delta(\epsilon) > 0$  such that

$$d_Y(f(x), f(y)) < \epsilon$$

whenever  $d_X(x, y) < \delta$ .

A function  $f : X \rightarrow Y$  is upper semi-continuous if for every sequence  $\{x_n\} \rightarrow x$ , we have that

$$\limsup_{n \rightarrow \infty} f(x_n) \leq f(x)$$

Similarly,  $f : X \rightarrow Y$  is lower semi-continuous if for every sequence  $\{x_n\} \rightarrow x$ , we have that

$$\liminf_{n \rightarrow \infty} f(x_n) \geq f(x)$$

Let  $(X, d)$  and  $(\tilde{X}, \tilde{d})$  be metric spaces. Then an isometric embedding is a one-to-one map  $i : X \rightarrow \tilde{X}$  such that for all  $x, y \in X$

$$\tilde{d}(i(x), i(y)) = d(x, y)$$

DEFINITION

*isometric isomorphism*

ANALYSIS

DEFINITION

*completion of an isometric isomorphism*

ANALYSIS

DEFINITION

*sequentially compact*

ANALYSIS

An isometric isomorphism is a surjective isometric embedding. Moreover, Two Metric Spaces  $(X, d)$  and  $(Y, d')$  are isometric if there exists a bijection  $h : X \rightarrow Y$  that preserves the metric.

$(\tilde{X}, \tilde{d})$  is the completion of  $(X, d)$  isometric isomorphism if

1.  $\exists i : X \rightarrow \tilde{X}$  an isometric embedding.
2.  $i(X)$  is dense in  $\tilde{X}$
3.  $\tilde{X}$  is complete

A subset  $K$  of a metric space is called sequentially compact if every sequence in  $K$  has a convergent subsequence whose limit belongs to  $K$ .

DEFINITION

*cover*

ANALYSIS

DEFINITION

*open cover*

ANALYSIS

DEFINITION

*subcover*

ANALYSIS

Let  $I$  be an index set. A collection  $G = \{G_\alpha : \alpha \in I\}$  of subsets of  $X$  is called a cover of a subset  $A \subset X$  if

$$A \subset \bigcup_{\alpha \in I} G_\alpha$$

If every  $G_\alpha$  in the cover is open, we say that  $\{G_\alpha\}$  is an open cover.

A subcover of  $G = \{G_\alpha : \alpha \in I\}$  is a collection  $\{G_\alpha : \alpha \in I_0\}$  where  $I_0 \subset I$ .



DEFINITION

*finite cover*

ANALYSIS

DEFINITION

*$\epsilon$ -net*

ANALYSIS

DEFINITION

*totally bounded*

ANALYSIS

If  $G = \{G_\alpha : \alpha \in I\}$  where  $I$  is finite, then  $G$  is a finite cover.

For all  $\epsilon > 0, A \subset X$ . An  $\epsilon$ -net for  $A$  is a subset of  $X$  of the form

$$\{x_\alpha : \alpha \in I\}$$

such that

$$\{B_\epsilon(x_\alpha) : \alpha \in I\}$$

$A \subset X$  is totally bounded if it has a finite  $\epsilon$ -net cover for every  $\epsilon > 0$ . That is,  $A$  can be covered by finite many open  $\epsilon$ -balls for any  $\epsilon > 0$ .

DEFINITION

*support of a function*

ANALYSIS

DEFINITION

$C_0(X)$

ANALYSIS

DEFINITION

*Bernstein Polynomials*

ANALYSIS

The support of a function  $f : X \rightarrow \mathbb{R}$  is given by

$$\text{supp}(f) := \overline{\{x \in X : f(x) \neq 0\}}$$

$$C_c(X) := \{f \in C(X) : \text{supp}(f) \text{ is compact}\} \subset C(X)$$

Now define:

$$C_0(X) := \overline{C_c(X)}$$

and equip this set with the  $\|\cdot\|_\infty$ . We see this set fits within the chain:

$$C_c(X) \subset C_0(X) \subset C_b(X) \subset C(X) \cap B(X)$$

The Bernstein Polynomials. We define a basis of unity:

$$\mathcal{X}_{n,k}(x) = \binom{n}{k} x^k (1-x)^{n-k}$$

Then by the binomial theorem:

$$1 = \sum_{k=0}^n \binom{n}{k} x^k (1-x)^{n-k}$$

So for any  $f \in C([0,1])$ , the Bernstein Polynomial of  $f$  is

$$B_n(x; f) := \sum_{k=0}^n f\left(\frac{k}{n}\right) \mathcal{X}_{n,k}(x)$$

DEFINITION

*Hausdorff*

ANALYSIS

DEFINITION

*equibounded*

ANALYSIS

DEFINITION

*equicontinuous*

ANALYSIS

A metric space  $X$  is Hausdorff if for every  $x \neq y$ , then we can identify open subsets  $U_x, V_y \subset X$  such that

$$x \in U_x, y \in V_y \text{ and } U_x \cap V_y = \emptyset$$

We then say  $A$  can separate points if we can identify an  $a \in A$  "between" two points.

Let  $\mathcal{F}$  be a family of functions

$$f : (X, d_X) \rightarrow (Y, d_Y)$$

between metric spaces. Then  $\mathcal{F}$  is called equibounded provided for all  $f \in \mathcal{F}$  there exists  $g \in \mathcal{F}$  if

$$\|f\|_\infty \leq M + \|g\|_\infty$$

Let  $\mathcal{F}$  be a family of functions

$$f : (X, d_X) \rightarrow (Y, d_Y)$$

between metric spaces. Then  $\mathcal{F}$  is called equicontinuous at  $x \in X$  if for every  $\epsilon > 0$ , there is a  $\delta = \delta(\epsilon, x) > 0$  such that

$$d_X(x, y) < \delta \implies d_Y(f(x), f(y)) \leq \epsilon$$

for all  $f \in \mathcal{F}$ .

DEFINITION

*dimension of a linear space*

ANALYSIS

DEFINITION

*Schauder Basis*

ANALYSIS

DEFINITION

*linear mapping*

ANALYSIS

The dimension of a linear space is equal to the number of elements that form a (linear ) basis for the space.

Let  $X$  be a separable Banach space. A Schauder Basis is a sequence  $\{x_n\}_{n=1}^{\infty}$  such that for all  $x \in X$ , there exists a unique sequence  $\{c_n\}_{n=1}^{\infty} \subset \mathbb{R}$  such that

$$x = \sum_{i=1}^{\infty} c_n x_n$$

Let  $X, Y$  be linear spaces. Then  $T : X \rightarrow Y$  is linear provided

$$T(ax + by) = aT(x) + bT(y)$$

for any  $x, y \in X, a, b \in \mathbb{F}$ .



DEFINITION

$$\mathcal{L}(X,Y)$$

ANALYSIS

DEFINITION

$$bounded$$

ANALYSIS

DEFINITION

$$\mathcal{B}(X,Y)$$

ANALYSIS

$$\mathcal{L}(X, Y) := \{T : X \rightarrow Y : T \text{ is linear} \}$$

Let  $X, Y$  be normed, linear spaces. A linear map  $T : X \rightarrow Y$  is bounded if there exists  $M \geq 0$  such that

$$\|Tx\|_Y \leq M \|x\|_X$$

$$\mathcal{B}(X, Y) := \{T : X \rightarrow Y : T \text{ is a bounded, linear map from } X \text{ to } Y\}$$

DEFINITION

*operator norm*

ANALYSIS

DEFINITION

*Relation "Stronger" in regards to Norms*

ANALYSIS

DEFINITION

*equivalent norms*

ANALYSIS

Given the space  $B(X, Y)$  we can define the operator norm

$$\|T\| = \inf\{M : \forall x \in X, \|Tx\| \leq M \|x\|\}$$

Let  $\|\cdot\|_1$  and  $\|\cdot\|_2$  be two norms on a linear space  $X$ . We say  $\|\cdot\|_2$  is stronger than  $\|\cdot\|_1$  if

$$\|x_n\|_2 \rightarrow 0 \implies \|x_n\|_1 \rightarrow 0$$

Two norms  $\|\cdot\|_1$  and  $\|\cdot\|_2$  are equivalent if there exists  $c_1 > 0$  and  $c_2 > 0$  such that

$$c_1 \|x\|_1 \leq \|x\|_2 \leq c_2 \|x\|_1$$

for all  $x \in X$ .

DEFINITION

$$Ker(T)$$

ANALYSIS

DEFINITION

$$Range(T)$$

ANALYSIS

DEFINITION

*well-posed problem*

ANALYSIS

$$Ker(T) := \{x \in X : Tx = 0\}$$

It is a subspace of  $X$ , and also called the null space of  $T$ .

$$Range(T) := \{y \in Y : Tx = y \text{ for some } x \in X\}$$

It is a subspace of  $Y$ .

A problem is called well-posed if

- A solution exists
- The solution is unique
- The solution is "stable", i.e. the solution depends continuously on the data.

DEFINITION

*open mapping*

ANALYSIS

DEFINITION

*linear functional*

ANALYSIS

DEFINITION

*dual space*

ANALYSIS

$T : X \rightarrow Y$  is an open mapping if  $TU$  is open in  $Y$  whenever  $U$  is open in  $X$ .

Let  $X$  be a linear space. Then a linear map  $f : X \rightarrow \mathbb{R}$  is called a linear functional on  $X$ .

When  $X$  is a normed linear space, its (topological) dual space is

$$X^* := B(X, \mathbb{R})$$

the space of all continuous linear functionals on  $X$ .



DEFINITION

*bidual*

ANALYSIS

DEFINITION

*reflexive*

ANALYSIS

DEFINITION

*weak topology*

ANALYSIS

Since  $X^*$  is a Banach space, we can consider its dual space

$$(X^*)^* = X^{**}$$

called the bidual of  $X$ .

If  $X = X^{**}$ , then we say that  $X$  is reflexive

Let  $X$  be a normed, linear space. The weak topology on  $X$  is defined as the weakest topology which makes all of the functionals  $\phi \in X^*$  continuous. Let  $\Phi := \{\phi^{-1}(U) : U \subset \mathbb{R} \text{ open}, \phi \in X^*\}$ . Since

$$\bigcup \phi^{-1}(U) = X$$

there exists a unique topology  $\mathcal{T}_{\text{weak}}$  on  $X$  having  $\Phi$  as its sub-base. Moreover,

$$\mathcal{T}_{\text{weak}} = \bigcup_{\text{arbitrary}} \bigcap_{\text{finite}} \phi^{-1}(U)$$

DEFINITION

*strongly / weakly converging*

ANALYSIS

DEFINITION

*Weak-\* Convergence*

ANALYSIS

DEFINITION

*inner product space*

ANALYSIS

Let  $\{\phi_n\}_{n=1}^\infty$  be a sequence in  $X^*$  and  $\phi \in X^*$ .

1. Strong Convergence:  $\phi_n \rightarrow \phi$  strongly in  $X^*$  provided  $\|\phi_n - \phi\|_{X^*} \rightarrow 0$
2. Weak Convergence:  $\phi_n \rightharpoonup \phi$  weakly in  $X^*$  means for all  $F \in X^{**}$ ,  $F(\phi_n) \rightarrow F(\phi)$ .

Weak-\* Convergence  $\phi_n \xrightarrow{*} \phi$  in  $X$  if  $\phi_n(x) \rightarrow \phi(x)$  for all  $x \in X$ .

An inner product space on a complex linear space  $X$  is a map  $\langle \cdot, \cdot \rangle : X \times X \rightarrow \mathbb{C}$  such that for all  $x, y, z \in X$ , and  $\lambda, \mu \in \mathbb{C}$ :

1.  $\langle x, \lambda y + \mu z \rangle = \lambda \langle x, y \rangle + \mu \langle x, z \rangle$
2.  $\langle x, y \rangle = \overline{\langle y, x \rangle}$  (Hermitian symmetric)
3.  $\langle x, x \rangle \geq 0$  and  $\langle x, x \rangle = 0 \iff x = 0$

A linear space with an inner product is called an inner product space or a pre-Hilbert space.

DEFINITION

*Hilbert Space*

ANALYSIS

DEFINITION

*orthogonal*

ANALYSIS

DEFINITION

*orthogonal complement*

ANALYSIS

A complete inner product space is called a Hilbert Space. So every Hilbert Space is a Banach Space. A common notation for inner products in Hilbert Space  $\langle \cdot, \cdot \rangle$  whereas an inner product space only has  $(\cdot, \cdot)$ .

If  $(x, y) = 0$ , then  $x$  and  $y$  are orthogonal, denoted  $x \perp y$ . Suppose  $M \subset X$ . If  $x \perp y$  for all  $y \in M$ , then we say that  $x$  is orthogonal to  $M$ , and write  $x \perp M$ .

If  $M$  is a subset of an inner product space  $X$ , then we can define the orthogonal complement of  $M$  by

$$M^\perp := \{x \in X : x \perp M\}$$

DEFINITION

*orthogonal projection*

ANALYSIS

DEFINITION

*orthonormal set*

ANALYSIS

DEFINITION

*isomorphic between inner product spaces*

ANALYSIS

A projection  $P : \mathcal{H} \rightarrow \mathcal{H}$  is called orthogonal if for all  $x, y \in \mathcal{H}$ ,

$$\langle x, Py \rangle = \langle Px, y \rangle$$

Let  $X$  be an inner product space. A subset  $S = \{e_\alpha : \alpha \in A\} \subset X$  is called orthonormal if it satisfies

1. (Orthogonality)  $e_\alpha \perp e_\beta$  for all  $\alpha, \beta \in A, \alpha \neq \beta$ .
2. (Normalized)  $\|e_\alpha\| = 1$  for all  $\alpha \in A$

In other words,  $\langle e_\alpha, e_\beta \rangle = \delta_{\alpha, \beta} = \begin{cases} 1 & \alpha = \beta \\ 0 & \alpha \neq \beta \end{cases}$

Let  $(X_1, (\cdot, \cdot)_1)$  and  $(X_2, (\cdot, \cdot)_2)$  be two inner product spaces. If there exists an isomorphism  $T : X_1 \rightarrow X_2$  such that for all  $x, y \in X$  :

$$(Tx, Ty)_2 = (x, y)_1$$

Then we say that the inner product space  $X_1, X_2$  are isomorphic.



DEFINITION

$\sigma$ -algebra

ANALYSIS

DEFINITION

algebra

ANALYSIS

DEFINITION

measurable space

ANALYSIS

A  $\sigma$ -algebra on a set  $X$  is a collection  $\mathcal{A}$  of subsets of  $X$  such that

1.  $\emptyset \in \Sigma$
2. If  $A \in \Sigma$ , then  $A^c \in \Sigma$
3. If  $A_1, A_2, \dots$ , is a countable family of sets in  $\Sigma$ , then

$$\bigcup_{i=1}^{\infty} A_i \in \Sigma$$

An algebra on a set  $X$  is a collection  $\mathcal{A}$  of subsets of  $X$  such that

1.  $\emptyset \in \Sigma$
2. If  $A \in \Sigma$ , then  $A^c \in \Sigma$
- 3.

$$\bigcup_{i=1}^N A_i \in \Sigma$$

A measurable space  $(X, \mathcal{A})$  is a set  $X$  and a  $\sigma$ -algebra  $\Sigma$  on  $X$ . Elements of  $\Sigma$  are called measurable sets.

DEFINITION

*measure*

ANALYSIS

DEFINITION

*measure space*

ANALYSIS

DEFINITION

*measure*

ANALYSIS

A measure  $\mu$  on  $X$  is a function

$$\mu : \Sigma \rightarrow [0, \infty]$$

such that

1.  $\mu(\emptyset) = 0$
2.  $\mu$  is countably additive:

$$\mu\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mu(A_i)$$

for each countable family of mutually disjoint sets  $\{E_n\}_{n=1}^{\infty} \subset \Sigma$ .

A measure space  $(X, \mathcal{A}, \mu)$  is a set  $X$ , a  $\sigma$ -algebra  $\mathcal{A}$  on  $X$ , and a measure  $\mu : \mathcal{A} \rightarrow [0, \infty]$ .

A measure  $\mu$  is said to be

- finite if  $\mu(X) < \infty$
- $\sigma$ -finite if there exists a sequence  $\{A_i\} \subset \mathcal{A}$  such that  $X = \bigcup_{i=1}^{\infty} A_i$  and  $\mu(A_i) < \infty$  for all  $n$ .
- probability measure if  $\mu(X) = 1$ .

DEFINITION

*Borel measure*

ANALYSIS

DEFINITION

*measure zero*

ANALYSIS

DEFINITION

*complete measure space*

ANALYSIS

Given  $X, \mathcal{A} = \mathcal{B}(X)$  then Borel  $\sigma$ -algebra. Then any measure  $\mu$  on  $(X, \mathcal{B}(X))$  is called a Borel measure.

Let  $(X, \mathcal{A}, \mu)$  be a measure space and  $A \subset X$  is said to have measure zero if it is measurable and  $\mu(A) = 0$ .

A measure space  $(X, \mathcal{A}, \mu)$  is complete if every subset of a measure zero set is measurable.

DEFINITION

$$\overline{\mathcal{A}}$$

ANALYSIS

DEFINITION

*almost everywhere property*

ANALYSIS

DEFINITION

*measurable function*

ANALYSIS

Given a measure space  $(X, \mathcal{A}, \mu)$ , we define  $\overline{\mathcal{A}}$  to be the  $\sigma$ -algebra generated by

$$\mathcal{A} \cup \{\text{subsets of measure zero sets}\}$$

Then

$$\overline{\mathcal{A}} = \{A : \exists E, F \in \mathcal{A} \text{ such that } E \subset A \subset F, \mu(F \setminus E) = 0\}$$

So for all  $A \in \overline{\mathcal{A}}$ , define

$$\overline{\mu}(A) := \mu(E) = \mu(F)$$

Then the complete measure space  $(X, \overline{\mathcal{A}}, \overline{\mu})$  is called the completion of  $(X, \mathcal{A}, \mu)$ .

A property that holds except on a set of measure zero is said to hold almost everywhere or a.e. for short. If the measure is obvious, then explicitly write  $\mu$ -a.e.

Let  $(X, \mathcal{A})$  be a measurable space. A real-valued function  $f : X \rightarrow \mathbb{R}$  is a measurable function (with respect to  $\mathcal{A}$ ) if for all  $t \in \mathbb{R}$ ,

$$f^{-1}((t, \infty)) = \{x \in X : f(x) > t\}$$

is measurable.



DEFINITION

$(\mathcal{A}_X, \mathcal{A}_Y)$ -measurable

ANALYSIS

DEFINITION

*product measure*

ANALYSIS

DEFINITION

$L^p$  space

ANALYSIS

A function  $f : (X, \mathcal{A}_X) \rightarrow (Y, \mathcal{A}_Y)$  between two measurable spaces is called  $(\mathcal{A}_X, \mathcal{A}_Y)$ -measurable if  $f^{-1}(\mathcal{A}_Y) \subset \mathcal{A}_X$ .

Let  $(X, \mathcal{A}, \mu)$  and  $(Y, \mathcal{B}, \nu)$  be two  $\sigma$ -finite measure spaces. For any  $E \in \mathcal{A} \times \mathcal{B}$ , define

$$(\mu \times \nu)(E) := \int_X \nu(E_x) d\mu(x) = \int_Y \mu(E_y) d\nu(y)$$

$\mu \times \nu$  is a measure on  $\mathcal{A} \times \mathcal{B}$ , called the product measure.

Let  $1 \leq p < \infty$  and let  $(X, \mathcal{A}, \mu)$  be a measure space. Then

$$\|f\|_{L^p(X)} := \left[ \int_X |f|^p d\mu \right]^{\frac{1}{p}}$$

Then the  $L^p$  space is defined by:

$$L^p(X) := \{f : X \rightarrow [-\infty, \infty] : f \text{ } \mathcal{A} \text{ - measurable with } \|f\|_{L^p(X)} < \infty\}$$

Further,  $L^p(X)$  consists of equivalence classes. That is, given  $f, g \in L^p(X)$ , then

$$f \equiv g \iff f = g \text{ a.e.}$$

DEFINITION

$$L^{\infty}(X)$$

ANALYSIS

DEFINITION

$$L^{p,weak}(X)$$

ANALYSIS

DEFINITION

*distribution function*

ANALYSIS

$$L^\infty(X) := \{f : X \rightarrow [-\infty, \infty] : f \text{ } \mathcal{A} \text{ - measurable with } \|f\|_{L^\infty(X)} < \infty\}$$

where

$$\|f\|_{L^\infty(X)} := \inf\{M : \mu(\{x \in X : |f| > M\}) = 0\}$$

Given  $(X, \mathcal{A}, \mu)$  a measure space with  $1 \leq p < \infty$ . We define the space

$$L^{p, \text{weak}}(X)$$

as the set of functions  $f \in L^{p, \text{weak}}(X)$  provided  $f : X \rightarrow \mathbb{R}$  is a  $\mathcal{A}$ -measurable function and

$$\mu(\{x \in X : |f(x)| > \lambda\}) \leq \frac{C^p}{\lambda^p}$$

for all  $\lambda > 0$  and some choice constant  $C_\lambda$ .

The distribution function  $m_\mu(\lambda) : \mathbb{R}^+ \rightarrow [0, \infty]$  by

$$m_\mu(\lambda) := \mu(\{x \in X : |f(x)| > \lambda\})$$

DEFINITION

$$\|f\|_{L^{p,\text{weak}}(X)}$$

ANALYSIS

DEFINITION

*simple function*

ANALYSIS

DEFINITION

*step function*

ANALYSIS

$$\|f\|_{L^{p,\text{weak}}(X)} = \sup_{\lambda>0} \left\{ \lambda m_{\mu}(\lambda)^{\frac{1}{p}} \right\}$$

A function of the form

$$\phi(x) = \sum_{i=1}^N a_i \chi_{A_i}(x)$$

with  $A_i \subset X$  measurable with  $\mu(A_i) < \infty, |a_i| < \infty$  is a simple function.

On the Lebesgue measure space  $(X, \mathcal{B}(X), \lambda)$ , we define the step function of the form

$$\phi(x) = \sum_{i=1}^N a_i \chi_{Q_i}(x)$$

where  $Q_i$  are pairwise disjoint cubes of the form:

$$Q_i := \prod_{j=1}^n [\alpha_j, \alpha_j + \ell)$$

with  $\alpha_i, \ell \in \mathbb{R}^+$  and  $|a_i| < \infty$ .

DEFINITION

$$C_c(\mathbb{R}^n)$$

ANALYSIS

DEFINITION

*bounded linear functional*

ANALYSIS

DEFINITION

*absolutely continuous*

ANALYSIS

$C_c(\mathbb{R}^n)$  = the space of continuous functions of compact support.

A bounded linear functional on  $L^p(X)$  is a mapping  $\ell : L^p(X) \rightarrow \mathbb{R}$  with

$$\ell(af + bg) = a\ell(f) + b\ell(g)$$

and

$$|\ell(f)| \leq C \|f\|_{L^p}$$

for some constant  $C$ . Moreover, we define

$$\|\ell\| := \inf\{C : |\ell(f)| \leq C \|f\|_{L^p} \forall f \in L^p\}$$

Suppose we have a measurable space  $(X, \mathcal{A})$  equipped with two finite measures  $\mu, \nu$ .  $\nu$  is absolutely continuous with respect to  $\mu$ , denoted  $\nu \ll \mu$ , provided

$$\mu(B) = 0 \implies \nu(B) = 0$$



DEFINITION

*signed measure*

ANALYSIS

DEFINITION

*Hardy-Littlewood Maximal Function*

ANALYSIS

DEFINITION

$f * g$

ANALYSIS

A function  $\nu$  on a measurable space  $(X, \mathcal{A})$  is a signed measure provided:

1. For any  $E \in \mathcal{A}$ ,  $\nu(E) \in [-\infty, \infty]$
2.  $\nu(\emptyset) = 0$
3.  $\nu$  is countably additive.

Let  $f \in L^1_{loc}(\mathbb{R}^n)$  (i.e.

$$\int_K |f| d\mu < \infty$$

for any compact set  $K \subset \mathbb{R}^n$ ), then we can define the Hardy-Littlewood Maximal Function by

$$\mu f(x) := \sup_{r>0} \frac{1}{\mu(B_r(x))} \int_{B_r(x)} |f(y)| dy$$

If  $f, g : \mathbb{R}^n \rightarrow \mathbb{R}$ , then

$$(f * g)(x) := \int_{\mathbb{R}^n} f(x-y)g(y) dy$$

DEFINITION

*approximation identity*

ANALYSIS

DEFINITION

$W^{k,p}(\Omega)$

ANALYSIS

DEFINITION

$C^1$ -boundary

ANALYSIS

Suppose  $K \in L^1$  satisfies the following properties:

1.  $\lim_{r \rightarrow 0} \int_{|x| > \delta} |K_r(x)| \, dx = 0$
2.  $\int K(x) \, dx = 1$

Then  $K$  is called an approximation identity.

Given  $\Omega \subset \mathbb{R}^n$  open, with  $1 \leq p \leq \infty$  and  $K \in \mathbb{Z}_{\geq 0}$ , then

$$W^{k,p}(\Omega) := \{u : \Omega \rightarrow \mathbb{R}^n : D^\alpha u \text{ exists } \forall |\alpha| \leq k \text{ and } D^\alpha u \in L^p(\mathbb{R}^n) \forall |\alpha| \leq k\}$$

Further, we equip this space with the norm:

$$\|f\|_{W^{k,p}(\Omega)} = \sum_{|\alpha| \leq k} \|D^\alpha f\|_{L^p(\Omega)}$$

Given a region  $\Omega$ , a  $C^1$ -boundary  $\partial\Omega$  is one such that for every  $x_0 \in \partial\Omega$ , there exists  $r > 0$  and an  $C^1$ -embedding-function

$$\gamma : \mathbb{R}^{n-1} \rightarrow \mathbb{R}^n$$

such that

$$B_r(x_0) \cap \Omega = \{(x_1, \dots, x_n) \in B_r(x_0) : x_n > \gamma(x_1, \dots, x_{n-1})\}$$

DEFINITION

*boundary value of  $u \in W^{1,p}(\Omega)$*

ANALYSIS

DEFINITION

*Hölder  $\alpha$ -continuous*

ANALYSIS

DEFINITION

$C^{0,\alpha}(\Omega)$

ANALYSIS

The boundary value of  $u \in W^{1,p}(\Omega)$  is defined as

$$u|_{\partial\Omega} := \lim_{\epsilon \downarrow 0} Eu * \phi_\epsilon|_{\partial\Omega}$$

where  $\phi_\epsilon$  is a standard mollifier.

A function  $f : \Omega \rightarrow \mathbb{R}$  is Hölder  $\alpha$ -continuous provided there exists a constant  $C$  such that

$$|f(x) - f(y)| \leq C|x - y|^\alpha$$

for all  $x, y \in \Omega$ .

The space  $C^{0,\alpha}(\Omega) = C^\alpha(\Omega)$  is the space of Hölder  $\alpha$ -continuous functions equipped with the finite norm

$$\|f\|_{C^\alpha(\Omega)} = \sup_{x \in \Omega} |f(x)| + \sup_{x, y \in \Omega, x \neq y} \frac{|f(x) - f(y)|}{|x - y|^\alpha}$$

DEFINITION

*Fourier Transform*

ANALYSIS

DEFINITION

$H^s(\mathbb{R}^n)$

ANALYSIS

DEFINITION

*function*

ANALYSIS

Suppose  $f \in L^1(\mathbb{R}^n)$ . we define the Fourier Transform by

$$\hat{f}(\xi) = \int_{\mathbb{R}^n} f(x) e^{-2\pi i \xi x} dx$$

Given  $s \in \mathbb{R}$ , define the space

$$H^s(\mathbb{R}^n) := \{f \in \mathcal{S}(\mathbb{R}^n) : (\sqrt{1 + |\xi|^2})^s \hat{f}(\xi) \in L^2(\mathbb{R})\}$$

- $s > 0 \implies$  fractional differentiation.
- $s < 0 \implies$  fractional integration.

We equip this space with the norm:

$$\|f\|_{H^s(\mathbb{R}^n)} := \left\| (\sqrt{1 + |\xi|^2})^s \hat{f}(\xi) \right\|_{L^2(\mathbb{R}^n)}$$

For  $0 < \alpha < n$ , we define the generalized Riesz Potential

$$(I_\alpha f)(x) := \int_{\mathbb{R}^n} \frac{f(y)}{|x - y|^{n-\alpha}} dy$$



DEFINITION

*Space of Schwartz Functions*

ANALYSIS

DEFINITION

*Weak Convergence / Strong Convergence in  $(X, \|\cdot\|)$*

ANALYSIS

DEFINITION

*spectral radius*

ANALYSIS

The Space of Schwartz Functions, denoted  $S(\mathbb{R}^n)$ , is defined as:

$$\mathcal{S}(\mathbb{R}^n) := \{u \in C^\infty(\mathbb{R}^n) : \|u\|_{\alpha,\beta} = \sup_{x \in \mathbb{R}^n} |x^\alpha D^\beta u| < \infty \forall \alpha, \beta \in \mathbb{Z}_{\geq 0}^n\}$$

where we equip this space with a semi-norm:

$$[f]_{\alpha,\beta} := \sup_{x \in \mathbb{R}^n} |x^\alpha D^\beta u|$$

Let  $(X, \|\cdot\|)$  be a normed linear space. Let  $\{x_n\}_{n=1}^\infty$  be a sequence in  $X$ . We say

1.  $x_n$  converges strongly to  $x$  if

$$\|x_n - x\| \rightarrow 0$$

denoted  $x_n \rightarrow x$

2.  $x_n$  converges weakly to  $x$  if

$$\phi(x_n) \rightarrow \phi(x)$$

for all  $\phi \in X^*$ . Denoted  $x_n \rightharpoonup x$ .

For any  $A \in B(X)$ , the number  $r(A) = \sup_{\lambda \in \sigma(A)} |\lambda|$  is called the spectral radius of  $A$ .

DEFINITION

*resolvent*

ANALYSIS

DEFINITION

*spectrum*

ANALYSIS

DEFINITION

*convex combination*

ANALYSIS

Let  $X$  be a Banach space and  $A \in B(X)$ . Then

$$\rho(A) = \{\lambda \in \mathbb{C} : \lambda I - A \text{ is invertible}\}$$

For all  $\lambda \in \rho(A)$ , let

$$R_\lambda = (\lambda I - A)^{-1}$$

be the resolvent of  $A$  at  $\lambda$ .

The spectrum of  $A$  is the set

$$\sigma(A) := \mathbb{C} \setminus \rho(A) = \{\lambda \in \mathbb{C} : \lambda I - A \text{ is not invertible}\}$$

Given  $\{x_1, x_2, \dots, x_n\}$  of a linear space  $X$ , we define a the set

$$C = \left\{ y = \sum_{k=1}^n \lambda_k x_k : \lambda_k \geq 0, \sum_{k=1}^n \lambda_k = 1 \right\}$$

each  $y$  is called a convex combination of  $\{x_1, x_2, \dots, x_n\}$ .

DEFINITION

*weak convergence in a Hilbert Space*

ANALYSIS

DEFINITION

*Isomorphic Hilbert Spaces*

ANALYSIS

DEFINITION

*self-adjoint*

ANALYSIS

A sequence  $\{x_n\}_{n=1}^\infty$  in a Hilbert space  $\mathcal{H}$  converges weakly to  $x \in \mathcal{H}$  if

$$\langle x_n, y \rangle \rightarrow \langle x, y \rangle$$

for all  $y \in \mathcal{H}$ .

Two Hilbert spaces are isomorphic if there is a unitary operator between them.

$A \in B(\mathcal{H}, \mathcal{H})$  is self-adjoint if  $A = A^*$ . That is,

$$\langle x, Ay \rangle = \langle Ax, y \rangle$$

for all  $x, y \in \mathcal{H}$

DEFINITION

*adjoint operator*

ANALYSIS

Let  $A : \mathcal{H} \rightarrow \mathcal{H}$  be a bounded linear operator on a Hilbert space  $\mathcal{H}$ . There exists a unique bounded linear operator  $A^* : \mathcal{H} \rightarrow \mathcal{H}$  such that

$$\langle x, Ay \rangle = \langle A^*x, y \rangle$$

for all  $x, y \in \mathcal{H}$ .  $A^*$  is called the adjoint of the operator.