

The extended real number system is $\mathbb{R}^1$ with two symbols, ∞ and $-\infty$, adjoined, and with the obvious ordering. If $-\infty \leq a \leq b \leq \infty$, the interval $[a, b]$ and the segment (a, b) are defined to be

$$[a, b] = \{x: a \leq x \leq b\}, \quad (a, b) = \{x: a < x < b\}.$$

We also write

$$[a, b) = \{x: a \leq x < b\}, \quad (a, b] = \{x: a < x \leq b\}.$$

If $E \subset [-\infty, \infty]$ and $E \neq \emptyset$, the least upper bound (supremum) and greatest lower bound (infimum) of E exist in $[-\infty, \infty]$ and are denoted by $\sup E$ and $\inf E$.

Sometimes (but only when $\sup E \in E$) we write $\max E$ for $\sup E$.

The symbol

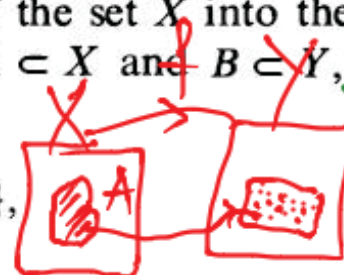
$$f: X \rightarrow Y$$

Always $D_f = X$

means that f is a function (or mapping or transformation) of the set X into the set Y ; i.e., f assigns to each $x \in X$ an element $f(x) \in Y$. If $A \subset X$ and $B \subset Y$, the image of A and the inverse image (or pre-image) of B are

$$f(A) = \{y: y = f(x) \text{ for some } x \in A\},$$

$$f^{-1}(B) = \{x: f(x) \in B\}.$$



Note that $f^{-1}(B)$ may be empty even when $B \neq \emptyset$.

The domain of f is X . The range of f is $f(X)$.

If $f(X) = Y$, f is said to map X onto Y .

We write $f^{-1}(y)$, instead of $f^{-1}(\{y\})$, for every $y \in Y$. If $f^{-1}(y)$ consists of at most one point, for each $y \in Y$, f is said to be one-to-one. If f is one-to-one, then f^{-1} is a function with domain $f(X)$ and range X .

If $f: X \rightarrow [-\infty, \infty]$ and $E \subset X$, it is customary to write $\sup_{x \in E} f(x)$ rather than $\sup f(E)$.

If $f: X \rightarrow Y$ and $g: Y \rightarrow Z$, the composite function $g \circ f: X \rightarrow Z$ is defined by the formula

$$(g \circ f)(x) = g(f(x)) \quad (x \in X).$$

(a) A collection τ of subsets of a set X is said to be a topology in X if τ has the following three properties:

(i) $\emptyset \in \tau$ and $X \in \tau$.

(ii) If $V_i \in \tau$ for $i = 1, \dots, n$, then $V_1 \cap V_2 \cap \dots \cap V_n \in \tau$.

$X = \{a, b, c\}$
 $\tau = \{\emptyset, \{a\}, \{a, b\}, X\}$
 So τ is an upper bound for $\emptyset$

- (iii) If $\{V_\alpha\}$ is an arbitrary collection of members of τ (finite, countable, or uncountable), then $\bigcup_\alpha V_\alpha \in \tau$.
- (b) If τ is a topology in X , then X is called a *topological space*, and the members of τ are called the *open sets* in X .
- (c) If X and Y are topological spaces and if f is a mapping of X into Y , then f is said to be *continuous* provided that $f^{-1}(V)$ is an open set in X for every open set V in Y .

1.3 Definition

- (a) A collection $\mathfrak{M}$ of subsets of a set X is said to be a σ -algebra in X if $\mathfrak{M}$ has the following properties:
- (i) $X \in \mathfrak{M}$. (4) $\Rightarrow \emptyset \in \mathfrak{M}$
 - (ii) If $A \in \mathfrak{M}$, then $A^c \in \mathfrak{M}$, where A^c is the complement of A relative to X .
 - (iii) If $A = \bigcup_{n=1}^{\infty} A_n$ and if $A_n \in \mathfrak{M}$ for $n = 1, 2, 3, \dots$, then $A \in \mathfrak{M}$.
- (b) If $\mathfrak{M}$ is a σ -algebra in X , then X is called a *measurable space*, and the members of $\mathfrak{M}$ are called the *measurable sets* in X . انذار
- (c) If X is a measurable space, Y is a topological space, and f is a mapping of X into Y , then f is said to be *measurable* provided that $f^{-1}(V)$ is a measurable set in X for every open set V in Y .

1.4 Comments on Definition 1.2 The most familiar topological spaces are the *metric spaces*. We shall assume some familiarity with metric spaces but shall give the basic definitions, for the sake of completeness.

A *metric space* is a set X in which a *distance function* (or *metric*) ρ is defined, with the following properties: $\rho: X \times X \rightarrow \mathbb{R}$

- (a) $0 \leq \rho(x, y) < \infty$ for all x and $y \in X$.
- (b) $\rho(x, y) = 0$ if and only if $x = y$.
- (c) $\rho(x, y) = \rho(y, x)$ for all x and $y \in X$.
- (d) $\rho(x, y) \leq \rho(x, z) + \rho(z, y)$ for all x, y , and $z \in X$.

Property (d) is called the *triangle inequality*.

If $x \in X$ and $r \geq 0$, the *open ball* with center at x and radius r is the set $\{y \in X: \rho(x, y) < r\}$.

If X is a metric space and if τ is the collection of all sets $E \subset X$ which are arbitrary unions of open balls, then τ is a topology in X . This is not hard to verify; the intersection property depends on the fact that if $x \in B_1 \cap B_2$, where B_1 and B_2 are open balls, then x is the center of an open ball $B \subset B_1 \cap B_2$. We leave this as an exercise.

For instance, in the real line R^1 a set is open if and only if it is a union of open segments (a, b) . In the plane R^2 , the open sets are those which are unions of open circular discs.

Another topological space, which we shall encounter frequently, is the extended real line $[-\infty, \infty]$; its topology is defined by declaring the following sets to be open: (a, b) , $[-\infty, a)$, $(a, \infty]$, and any union of segments of this type.

1.6 Comments on Definition 1.3 Let $\mathfrak{M}$ be a σ -algebra in a set X . Referring to Properties (i) to (iii) of Definition 1.3(a), we immediately derive the following facts.

- (a) Since $\emptyset = X^c$, (i) and (ii) imply that $\emptyset \in \mathfrak{M}$.
- (b) Taking $A_{n+1} = A_{n+2} = \cdots = \emptyset$ in (iii), we see that $A_1 \cup A_2 \cup \cdots \cup A_n \in \mathfrak{M}$ if $A_i \in \mathfrak{M}$ for $i = 1, \dots, n$.
- (c) Since

$$\bigcap_{n=1}^{\infty} A_n = \left(\bigcup_{n=1}^{\infty} A_n^c \right)^c,$$

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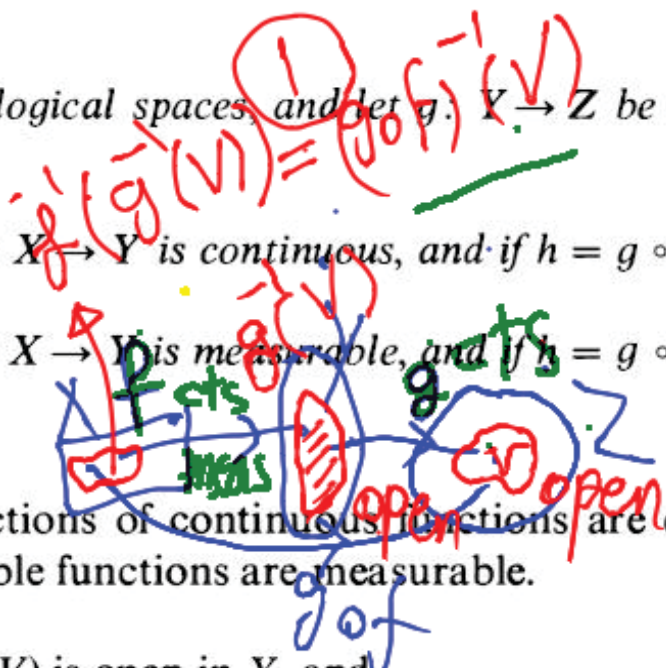
$\mathfrak{M}$ is closed under the formation of countable (and also finite) intersections.

- (d) Since $A - B = B^c \cap A$, we have $A - B \in \mathfrak{M}$ if $A \in \mathfrak{M}$ and $B \in \mathfrak{M}$.

The prefix σ refers to the fact that (iii) is required to hold for all *countable* unions of members of $\mathfrak{M}$. If (iii) is required for finite unions only, then $\mathfrak{M}$ is called an *algebra* of sets.

1.7 Theorem Let Y and Z be topological spaces, and let $g: Y \rightarrow Z$ be continuous.

- (a) If X is a topological space, if $f: X \rightarrow Y$ is continuous, and if $h = g \circ f$, then $h: X \rightarrow Z$ is continuous.
- (b) If X is a measurable space, if $f: X \rightarrow Y$ is measurable, and if $h = g \circ f$, then $h: X \rightarrow Z$ is measurable.



Stated informally, continuous functions of continuous functions are continuous; continuous functions of measurable functions are measurable.

PROOF If V is open in Z , then $g^{-1}(V)$ is open in Y , and

$$h^{-1}(V) = f^{-1}(g^{-1}(V)).$$

If f is continuous, it follows that $h^{-1}(V)$ is open, proving (a).

If f is measurable, it follows that $h^{-1}(V)$ is open, proving (a).
 If f is measurable, it follows that $h^{-1}(V)$ is measurable, proving (b).

////

$E \subseteq \mathbb{R}$, then $+\infty$ is an upper bound
 $-\infty$ " " lower "

E has the sup, since $\begin{cases} E \text{ is bd above, so } \sup E \in \mathbb{R} \\ E \text{ is not bd above, so } \sup E = +\infty \end{cases}$
 Since $+\infty$ is the only upper bound for E

E is called countable if there exists a 1-1 correspondence between E & $\mathbb{N}$.



بگوئید باید دسترسی به کتاب
 نظریه مجموعه‌ها، ب. ب. راجا، آران
 در هند و پاکستان

If $f: X \rightarrow Y$ is a function, then

$$f^{-1}(A \cup B) = f^{-1}(A) \cup f^{-1}(B)$$

$$f^{-1}(C) = \{x \in X : f(x) \in C\}$$

So $x \in f^{-1}(C) \Leftrightarrow f(x) \in C$

$$x \in f^{-1}(A \cup B) \Leftrightarrow f(x) \in A \cup B \Leftrightarrow f(x) \in A \vee f(x) \in B \Leftrightarrow x \in f^{-1}(A) \vee x \in f^{-1}(B) \Leftrightarrow x \in f^{-1}(A) \cup f^{-1}(B)$$

$$f(A \cap B) \subseteq f(A) \cap f(B)$$

$$y \in f(A \cap B) \Leftrightarrow \exists x \in A \cap B; y = f(x) \Rightarrow$$

$$f(D) = \{f(x) : x \in D\}$$

So $x \in D \Rightarrow f(x) \in f(D)$
 So $y \in f(D) \Rightarrow \exists x \in D; y = f(x)$

$x \in A, x \in B, y = f(x) \Rightarrow y \in f(A), y \in f(B)$
 $\Rightarrow y \in f(A) \cap f(B).$



$f(D) = \{\sqrt{2}, \text{اباما}\}$

Recall that a function is ... $f(1) = \text{اباما} \in f(D)$ but $1 \notin D$



نکته

$f \subseteq X \times Y$ و f تابع است اگر و تنها اگر برای هر $x \in X$ دقیقاً یک $y \in Y$ وجود داشته باشد که $(x, y) \in f$

تابع با x و y مشخص می شود: x مجموعه آغاز، y مجموعه مقادیر

تابع با x و y در هر دو طرف معادل مشخص می شود:

ضابطه، D_f ، مجموعه مقادیر، مجموعه آغاز

اما وقتی مثلاً در ریاضی عمومی بخواهیم از $\mathbb{R}$ به $\mathbb{R}$ سر و کار داریم

تابع فقط با ضابطه، اگر مشخص می شود زیرا از فرض ضابطه دامنه به عنوان

بزرگترین زیر مجموعه $\mathbb{R}$ که برابر x در آن $f(x)$ معین است

قابل تعیین است بهمانند

$$f(x) = \frac{1}{x}$$

نکته: دامنه تابع $f(x) = \frac{1}{x}$ تمام اعداد حقیقی غیر از صفر است.

$$\hookrightarrow f(g(V)) = (g \circ f)(V)$$

$$x \in f^{-1}(g^{-1}(V)) \Leftrightarrow f(x) \in g^{-1}(V) \Leftrightarrow \underbrace{g(f(x))}_{(g \circ f)(x)} \in V \Leftrightarrow x \in (g \circ f)^{-1}(V)$$

1.8 Theorem Let u and v be real measurable functions on a measurable space X , let Φ be a continuous mapping of the plane into a topological space Y , and define

$$h(x) = \Phi(u(x), v(x))$$

for $x \in X$. Then $h: X \rightarrow Y$ is measurable.

PROOF Put $f(x) = (u(x), v(x))$. Then f maps X into the plane. Since $h = \Phi \circ f$, Theorem 1.7 shows that it is enough to prove the measurability of f .

If R is any open rectangle in the plane, with sides parallel to the axes, then R is the cartesian product of two segments I_1 and I_2 , and

$$f^{-1}(R) = u^{-1}(I_1) \cap v^{-1}(I_2), \quad 1$$

which is measurable, by our assumption on u and v . Every open set V in the plane is a countable union of such rectangles R_i , and since

$$f^{-1}(V) = f^{-1}\left(\bigcup_{i=1}^{\infty} R_i\right) = \bigcup_{i=1}^{\infty} f^{-1}(R_i),$$

$f^{-1}(V)$ is measurable. ////

1.9 Let X be a measurable space. The following propositions are corollaries of Theorems 1.7 and 1.8:

(a) If $f = u + iv$, where u and v are real measurable functions on X , then f is a complex measurable function on X .

This follows from Theorem 1.8, with $\Phi(z) = z$.

(b) If $f = u + iv$ is a complex measurable function on X , then u , v , and $|f|$ are real measurable functions on X .

This follows from Theorem 1.7, with $g(z) = \operatorname{Re}(z)$, $\operatorname{Im}(z)$, and $|z|$.

(c) If f and g are complex measurable functions on X , then so are $f + g$ and fg .

For real f and g this follows from Theorem 1.8, with

$$\Phi(s, t) = s + t$$

and $\Phi(s, t) = st$. The complex case then follows from (a) and (b).

(d) If E is a measurable set in X and if

$$\chi_E(x) = \begin{cases} 1 & \text{if } x \in E \\ 0 & \text{if } x \notin E \end{cases}$$

then χ_E is a measurable function

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This is obvious. We call χ_E the *characteristic function* of the set E . The letter χ will be reserved for characteristic functions throughout this book.

- (e) If f is a complex measurable function on X , there is a complex measurable function α on X such that $|\alpha| = 1$ and $f = \alpha|f|$.

PROOF Let $E = \{x: f(x) = 0\}$, let Y be the complex plane with the origin removed, define $\varphi(z) = z/|z|$ for $z \in Y$, and put

$$\alpha(x) = \varphi(f(x) + \chi_E(x)) \quad (x \in X).$$

If $x \in E$, $\alpha(x) = 1$; if $x \notin E$, $\alpha(x) = f(x)/|f(x)|$. Since φ is continuous on Y and since E is measurable (why?), the measurability of α follows from (c), (d), and Theorem 1.7. ////

We now show that σ -algebras exist in great profusion.

1.10 Theorem If $\mathcal{F}$ is any collection of subsets of X , there exists a smallest σ -algebra $\mathfrak{M}^*$ in X such that $\mathcal{F} \subset \mathfrak{M}^*$.

This $\mathfrak{M}^*$ is sometimes called the σ -algebra *generated* by $\mathcal{F}$.

PROOF Let Ω be the family of all σ -algebras $\mathfrak{M}$ in X which contain $\mathcal{F}$. Since the collection of all subsets of X is such a σ -algebra, Ω is not empty. Let $\mathfrak{M}^*$ be the intersection of all $\mathfrak{M} \in \Omega$. It is clear that $\mathcal{F} \subset \mathfrak{M}^*$ and that $\mathfrak{M}^*$ lies in every σ -algebra in X which contains $\mathcal{F}$. To complete the proof, we have to show that $\mathfrak{M}^*$ is itself a σ -algebra.

If $A_n \in \mathfrak{M}^*$ for $n = 1, 2, 3, \dots$, and if $\mathfrak{M} \in \Omega$, then $A_n \in \mathfrak{M}$, so $\bigcup A_n \in \mathfrak{M}$, since $\mathfrak{M}$ is a σ -algebra. Since $\bigcup A_n \in \mathfrak{M}$ for every $\mathfrak{M} \in \Omega$, we conclude that $\bigcup A_n \in \mathfrak{M}^*$. The other two defining properties of a σ -algebra are verified in the same manner. ////

1.11 Borel Sets Let X be a topological space. By Theorem 1.10, there exists a smallest σ -algebra $\mathcal{B}$ in X such that every open set in X belongs to $\mathcal{B}$. The members of $\mathcal{B}$ are called the *Borel sets* of X .

In particular, closed sets are Borel sets (being, by definition, the complements of open sets), and so are all countable unions of closed sets and all countable intersections of open sets. These last two are called F_σ 's and G_δ 's, respectively, and play a considerable role. The notation is due to Hausdorff. The letters F and

G were used for closed and open sets, respectively, and σ refers to union (*Summe*), δ to intersection (*Durchschnitt*). For example, every half-open interval $[a, b)$ is a G_δ and an F_σ in R^1 .

Since $\mathcal{B}$ is a σ -algebra, we may now regard X as a measurable space, with the Borel sets playing the role of the measurable sets; more concisely, we consider the measurable space $(X, \mathcal{B})$. If $f: X \rightarrow Y$ is a continuous mapping of X , where Y is any topological space, then it is evident from the definitions that $f^{-1}(V) \in \mathcal{B}$ for every open set V in Y . In other words, *every continuous mapping of X is Borel measurable*.

Borel measurable mappings are often called *Borel mappings*, or *Borel functions*.

1.12 Theorem Suppose $\mathfrak{M}$ is a σ -algebra in X , and Y is a topological space. Let f map X into Y .

- (a) If Ω is the collection of all sets $E \subset Y$ such that $f^{-1}(E) \in \mathfrak{M}$, then Ω is a σ -algebra in Y .
- (b) If f is measurable and E is a Borel set in Y , then $f^{-1}(E) \in \mathfrak{M}$.
- (c) If $Y = [-\infty, \infty]$ and $f^{-1}((\alpha, \infty]) \in \mathfrak{M}$ for every real α , then f is measurable.
- (d) If f is measurable, if Z is a topological space, if $g: Y \rightarrow Z$ is a Borel mapping, and if $h = g \circ f$, then $h: X \rightarrow Z$ is measurable.

Part (c) is a frequently used criterion for the measurability of real-valued functions. (See also Exercise 3.) Note that (d) generalizes Theorem 1.7(b).

PROOF (a) follows from the relations

$$f^{-1}(Y) = X,$$

$$f^{-1}(Y - A) = X - f^{-1}(A),$$

and
$$f^{-1}(A_1 \cup A_2 \cup \cdots) = f^{-1}(A_1) \cup f^{-1}(A_2) \cup \cdots.$$

To prove (b), let Ω be as in (a); the measurability of f implies that Ω contains all open sets in Y , and since Ω is a σ -algebra, Ω contains all Borel sets in Y .

To prove (c), let Ω be the collection of all $E \subset [-\infty, \infty]$ such that $f^{-1}(E) \in \mathfrak{M}$. Choose a real number α , and choose $\alpha_n < \alpha$ so that $\alpha_n \rightarrow \alpha$ as $n \rightarrow \infty$. Since $(\alpha_n, \infty] \in \Omega$ for each n , since

$$[-\infty, \alpha) = \bigcup_{n=1}^{\infty} [-\infty, \alpha_n] = \bigcup_{n=1}^{\infty} (\alpha_n, \infty]^c,$$

and since (a) shows that Ω is a σ -algebra, we see that $[-\infty, \alpha) \in \Omega$. The same

and since (a) shows that $\mathcal{B}$ is a σ -algebra, we see that $[-\infty, \alpha) \in \mathcal{B}$. The same is then true of

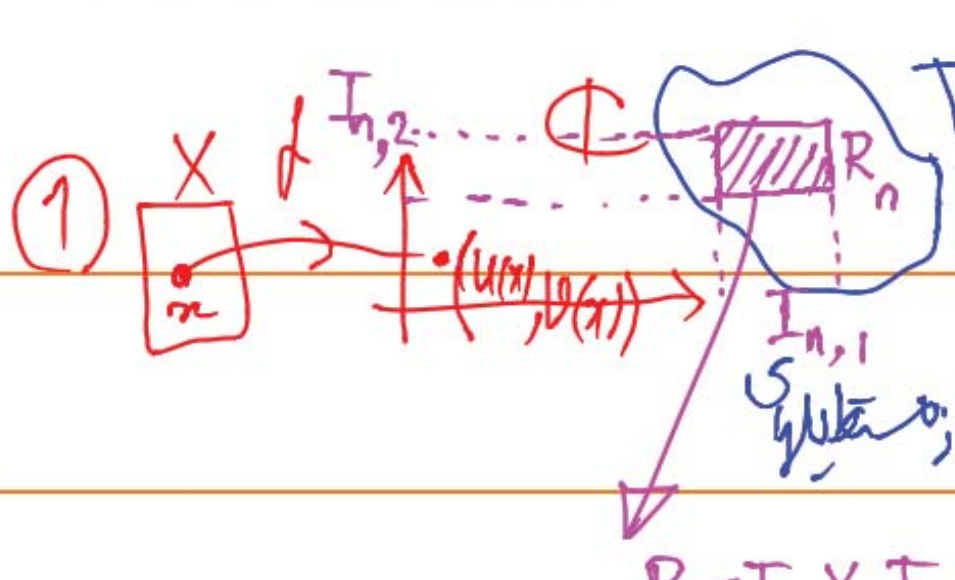
$$(\alpha, \beta) = [-\infty, \beta) \cap (\alpha, \infty].$$

Since every open set in $[-\infty, \infty]$ is a countable union of segments of the above types, Ω contains every open set. Thus f is measurable.

To prove (d), let $V \subset Z$ be open. Then $g^{-1}(V)$ is a Borel set of Y , and since

$$h^{-1}(V) = f^{-1}(g^{-1}(V)),$$

(b) shows that $h^{-1}(V) \in \mathfrak{M}$. ////

① 

$V^{\text{open}} = \bigcup_{n=1}^{\infty} R_n$

$R = I \times I$

$f(V) = f\left(\bigcup_{n=1}^{\infty} R_n\right) = \bigcup_{n=1}^{\infty} f(R_n) = \bigcup_{n=1}^{\infty} (\bar{u}^{-1}(I_{n,1}) \cap \bar{v}^{-1}(I_{n,2}))$ Epp

$\bar{f}(R) = \bar{u}^{-1}(I_1) \cap \bar{v}^{-1}(I_2)$ where $R = I_1 \times I_2$

$u, v: X \rightarrow \mathbb{R}$ are meas. func.

$x \in \bar{f}^{-1}(R) \Rightarrow f(x) \in R = I_1 \times I_2 \Leftrightarrow u(x) \in I_1, v(x) \in I_2 \Leftrightarrow x \in \bar{u}^{-1}(I_1) \cap \bar{v}^{-1}(I_2)$

$x \in \bar{u}^{-1}(I_1) \quad x \in \bar{v}^{-1}(I_2)$

(2) $\Phi(t, s) = t + s$

$\Phi(\text{Re}f, \text{Re}g) = \text{Re}f + \text{Re}g$ is meas, by this $\Rightarrow f+g =$
 $\text{Re}(f+g)$

$\Phi(\text{Im}f, \text{Im}g) = \text{Im}f + \text{Im}g$ is " " " "
 $\text{Im}(f+g)$

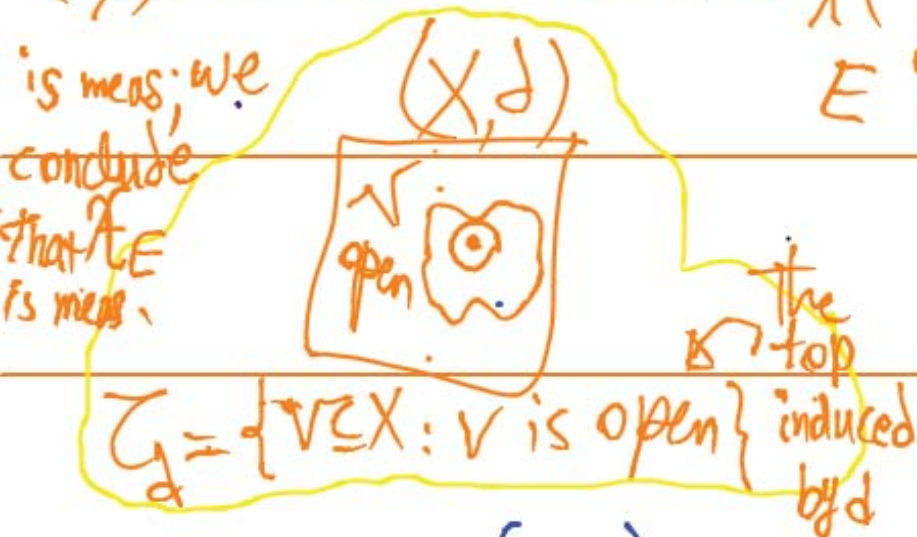
~~$\Phi(f, g) = f + g$~~
 f, g are complex valued.

$\Rightarrow f+g = \text{Re}(f+g) + i \text{Im}(f+g)$
 is meas, by Cor. 1.9(a).

(3) E is a meas set iff $\chi_E(x) = \begin{cases} 1 & x \in E \\ 0 & x \notin E \end{cases}$ is a meas func.

($\Leftarrow$) Let χ_E be meas. Since $E = \chi_E^{-1}(\{1\})$, E is meas.

($\Rightarrow$) Let E be meas. Since $\chi(V) = \begin{cases} E & \text{if } V \subseteq E \\ X \setminus E & \text{if } V \not\subseteq E \end{cases}$ is meas, we conclude that χ_E is meas.



$\tau_Y = \{V \cap Y : V \in \tau\}$ is called the subspace topology.

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- (e) If f is a complex measurable function on X , there is a complex measurable function α on X such that $|\alpha| = 1$ and $f = \alpha |f|$.

$$X - f^{-1}(\{0\}) = f^{-1}(\{0\})^c$$

open

PROOF Let $E = \{x: f(x) \neq 0\}$, let Y be the complex plane with the origin removed, define $\phi(z) = z/|z|$ for $z \in Y$, and put

meas

$$\alpha(x) = \phi(f(x)) = \frac{f(x)}{|f(x)|} \quad (x \in E)$$

If $x \in E$, $\alpha(x) = 1$; if $x \notin E$, $\alpha(x) = 0$. Since ϕ is continuous on Y and since E is measurable (why?), the measurability of α follows from (c), (d), and Theorem 1.7.

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1.11 Borel Sets Let X be a topological space. By Theorem 1.10, there exists a smallest σ -algebra $\mathcal{B}$ in X such that every open set in X belongs to $\mathcal{B}$. The members of $\mathcal{B}$ are called the Borel sets of X .

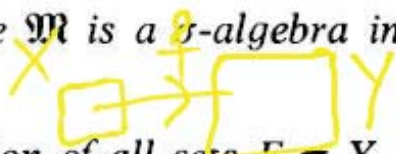
In particular, closed sets are Borel sets (being, by definition, the complements of open sets), and so are all countable unions of closed sets and all countable intersections of open sets. These last two are called F_σ 's and G_δ 's, respectively, and play a considerable role. The notation is due to Hausdorff. The letters F and

G were used for closed and open sets, respectively, and σ refers to union (*Summe*), δ to intersection (*Durchschnitt*). For example, every half-open interval $[a, b)$ is a G_δ and an F_σ in $\mathbb{R}^1$.

Since $\mathcal{B}$ is a σ -algebra, we may now regard X as a measurable space, with the Borel sets playing the role of the measurable sets; more concisely, we consider the measurable space $(X, \mathcal{B})$. If $f: X \rightarrow Y$ is a continuous mapping of X , where Y is any topological space, then it is evident from the definitions that $f^{-1}(V) \in \mathcal{B}$ for every open set V in Y . In other words, every continuous mapping of X is Borel measurable.

Borel measurable mappings are often called *Borel mappings*, or *Borel functions*.

1.12 Theorem Suppose $\mathfrak{M}$ is a σ -algebra in X , and Y is a topological space. Let f map X into Y .



- (a) If Ω is the collection of all sets $E \subset Y$ such that $f^{-1}(E) \in \mathfrak{M}$, then Ω is a σ -algebra in Y .
- (b) If f is measurable and E is a Borel set in Y , then $f^{-1}(E) \in \mathfrak{M}$.
- (c) If $Y = [-\infty, \infty]$ and $f^{-1}((\alpha, \infty]) \in \mathfrak{M}$ for every real α , then f is measurable.
- (d) If f is measurable, if Z is a topological space, if $g: Y \rightarrow Z$ is a Borel mapping, and if $h = g \circ f$, then $h: X \rightarrow Z$ is measurable.

Part (c) is a frequently used criterion for the measurability of real-valued functions. (See also Exercise 3.) Note that (d) generalizes Theorem 1.7(b).

PROOF (a) follows from the relations

$$A \in \Omega \iff f^{-1}(A) \in \mathfrak{M} \iff f^{-1}(A) \in \mathfrak{M} \iff f^{-1}(A) \in \mathfrak{M}$$

$$f^{-1}(Y) = X,$$

$$f^{-1}(Y - A) = X - f^{-1}(A),$$

$$\text{and } f^{-1}(A_1 \cup A_2 \cup \cdots) = f^{-1}(A_1) \cup f^{-1}(A_2) \cup \cdots.$$

To prove (b), let Ω be as in (a); the measurability of f implies that Ω contains all open sets in Y , and since Ω is a σ -algebra, Ω contains all Borel sets in Y .

To prove (c), let Ω be the collection of all $E \subset [-\infty, \infty]$ such that $f^{-1}(E) \in \mathfrak{M}$. Choose a real number α , and choose $\alpha_n < \alpha$ so that $\alpha_n \rightarrow \alpha$ as $n \rightarrow \infty$. Since $(\alpha_n, \infty] \in \Omega$ for each n , since

$$[-\infty, \alpha) = \bigcup_{n=1}^{\infty} [-\infty, \alpha_n] = \bigcup_{n=1}^{\infty} (\alpha_n, \infty]^c,$$

and since (a) shows that Ω is a σ -algebra, we see that $[-\infty, \alpha) \in \Omega$. The same

and since (a) shows that $\mathcal{B}$ is a σ -algebra, we see that $[-\infty, \alpha) \in \mathcal{B}$. The same is then true of

$$(\alpha, \beta) = [-\infty, \beta) \cap (\alpha, \infty].$$

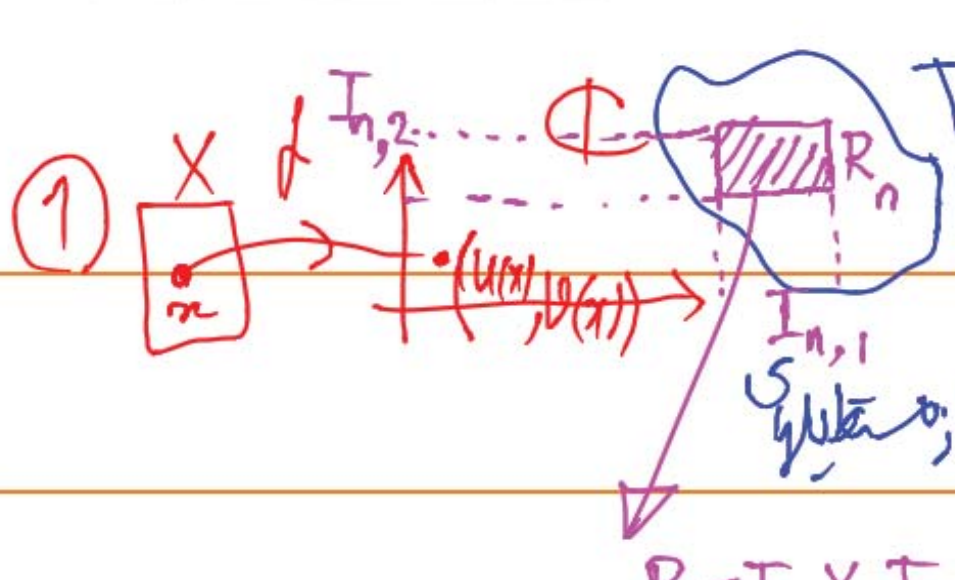
Since every open set in $[-\infty, \infty]$ is a countable union of segments of the above types, Ω contains every open set. Thus f is measurable.

To prove (d), let $V \subset Z$ be open. Then $g^{-1}(V)$ is a Borel set of Y , and since

$$h^{-1}(V) = f^{-1}(g^{-1}(V)),$$

(b) shows that $h^{-1}(V) \in \mathfrak{M}$.

////

① 

$V^{\text{open}} = \bigcup_{n=1}^{\infty} R_n$

$R = I \times I$

$f(V) = f\left(\bigcup_{n=1}^{\infty} R_n\right) = \bigcup_{n=1}^{\infty} f(R_n) = \bigcup_{n=1}^{\infty} (\bar{u}^{-1}(I_{n,1}) \cap \bar{v}^{-1}(I_{n,2}))$ (Epp)

$\bar{f}(R) = \bar{u}^{-1}(I_1) \cap \bar{v}^{-1}(I_2)$ where $R = I_1 \times I_2$

$u, v: X \rightarrow \mathbb{R}$ are meas. func.

$x \in \bar{f}^{-1}(R) \Rightarrow f(x) \in R = I_1 \times I_2 \Leftrightarrow u(x) \in I_1, v(x) \in I_2 \Leftrightarrow x \in \bar{u}^{-1}(I_1) \cap \bar{v}^{-1}(I_2)$

$x \in \bar{u}^{-1}(I_1) \quad x \in \bar{v}^{-1}(I_2)$

(2) $\Phi(t, s) = t + s$

$\Phi(\text{Re}f, \text{Re}g) = \text{Re}f + \text{Re}g$ is meas, by th. 1.8
 $\text{Re}(f+g)$

$\Phi(\text{Im}f, \text{Im}g) = \text{Im}f + \text{Im}g$ is " " " "
 $\text{Im}(f+g)$

~~$\Phi(f, g) = f + g$~~
 f, g are complex valued.

$\Rightarrow f+g = \text{Re}(f+g) + i \text{Im}(f+g)$ is meas, by Cor. 1.9(a).

(3) E is a meas set iff $\chi_E(x) = \begin{cases} 1 & x \in E \\ 0 & x \notin E \end{cases}$ is a meas func.

($\Leftarrow$) Let χ_E be meas. Since $E = \chi_E^{-1}(\{1\})$, E is meas.

($\Rightarrow$) Let E be meas. Since $\chi(V) = \begin{cases} E & \text{if } V \subseteq E \\ X \setminus E & \text{if } V \not\subseteq E \end{cases}$ is meas, we conclude that χ_E is meas.

is meas, we conclude that χ_E is meas.



the top induced by d

$\tau_d = \{V \subseteq X : V \text{ is open}\}$ induced by d



$\tau_Y = \{V \cap Y : V \in \tau\}$ is called the subspace top on Y.

0 since 1/1 well

با فرض اینکه از مراحلی قبلی برهیم

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با فرض اینکه از مراحلی قبلی برهیم

(4) $\mathcal{F} \subseteq 2^X \Rightarrow \bigcap_{\mathcal{F} \in \mathcal{M}} \mathcal{M}$ is a σ -alg
 $\mathcal{F} \in \mathcal{M}$ is σ -alg

and is the smallest σ -alg containing $\mathcal{F}$.

proof: Let $A \in \bigcap \mathcal{M}$. Then $A \in \mathcal{M}$ for all $\mathcal{M}$.

So $A^c \in \mathcal{M}$ for all $\mathcal{M}$. Hence $A^c \in \bigcap \mathcal{M}$.

The rest is clear.

If $\mathcal{N}$ is σ -alg & $\mathcal{F} \subseteq \mathcal{N}$ then

$\mathcal{M} \in \{\mathcal{M} \mid \mathcal{M} \text{ is } \sigma\text{-alg} \& \mathcal{F} \subseteq \mathcal{M}\}$. So $\bigcap \mathcal{M} \subseteq \mathcal{N}$
 $\mathcal{F} \subseteq \mathcal{M}$ σ -alg

If $\mathcal{F} = \mathcal{T}$ (a top on X), then $\langle \mathcal{T} \rangle$ the σ -alg of
 Borel sets. Example: $(a, b), [a, b], (a, b] = \bigcap_{n=1}^{\infty} (a, b + \frac{1}{n}]$ in $\mathbb{R}$

In particular $\bigcap_{n=1}^{\infty} G_n$, $\bigcup_{n=1}^{\infty} F_n \in \langle \tau \rangle$
 $\swarrow$ open $\searrow$ closed
 a G_δ -set an F_σ -set

Recall that a basis for a top is a set $\mathcal{B}$ of subsets of X such that each open set can be written as a union of elements of $\mathcal{B}$.

Example: $\{N(x) : x \in \mathbb{R}, r \in \mathbb{R}^{>0}\}$ is a basis for Euclidean top on $\mathbb{R}$.

Also $\{I_1 \times I_2 : I_1 \text{ \& } I_2 \text{ are open intervals}\}$ is a basis for the Euclidean top on $\mathbb{R}^2$.

In addition, $\{(a, b) : a, b \in \mathbb{R}\} \cup \{(a, \infty) : a \in \mathbb{R}\} \cup \{(-\infty, b) : b \in \mathbb{R}\}$ is a basis for a top on $[-\infty, +\infty]$

1.13 Definition Let $\{a_n\}$ be a sequence in $[-\infty, \infty]$, and put

$$b_k = \sup \{a_k, a_{k+1}, a_{k+2}, \dots\} \quad (k = 1, 2, 3, \dots) \quad (1)$$

and

$$\beta = \inf \{b_1, b_2, b_3, \dots\}. \quad (2)$$

We call β the *upper limit* of $\{a_n\}$, and write

$$\beta^{\wedge} = \limsup_{n \rightarrow \infty} a_n. \quad (3)$$

The following properties are easily verified: First, $b_1 \geq b_2 \geq b_3 \geq \dots$, so that $b_k \rightarrow \beta$ as $k \rightarrow \infty$; secondly, there is a subsequence $\{a_{n_i}\}$ of $\{a_n\}$ such that $a_{n_i} \rightarrow \beta$ as $i \rightarrow \infty$, and β is the largest number with this property.

The *lower limit* is defined analogously: simply interchange sup and inf in (1) and (2). Note that

$$\liminf_{n \rightarrow \infty} a_n = -\limsup_{n \rightarrow \infty} (-a_n). \quad (4)$$

If $\{a_n\}$ converges, then evidently

$$\limsup_{n \rightarrow \infty} a_n = \liminf_{n \rightarrow \infty} a_n = \lim_{n \rightarrow \infty} a_n. \quad (\text{why?}) \quad (5)$$

Suppose $\{f_n\}$ is a sequence of extended-real functions on a set X . Then $\sup_n f_n$ and $\limsup_{n \rightarrow \infty} f_n$ are the functions defined on X by

$$\left(\sup_n f_n\right)(x) = \sup_n (f_n(x)), \quad (6)$$

$$\left(\limsup_{n \rightarrow \infty} f_n\right)(x) = \lim_{n \rightarrow \infty} \sup (f_n(x)). \quad (7)$$

If

$$f(x) = \lim_{n \rightarrow \infty} f_n(x), \quad (8)$$

the limit being assumed to exist at every $x \in X$, then we call f the *pointwise limit* of the sequence $\{f_n\}$.

1.14 Theorem If $f_n: X \rightarrow [-\infty, \infty]$ is measurable, for $n = 1, 2, 3, \dots$, and

$$g = \sup_{n \geq 1} f_n, \quad h = \limsup_{n \rightarrow \infty} f_n,$$

then g and h are measurable.

PROOF $g^{-1}((\alpha, \infty]) \stackrel{(1)}{=} \bigcup_{n=1}^{\infty} f_n^{-1}((\alpha, \infty])$. Hence Theorem 1.12(c) implies that g is measurable. The same result holds of course with $\inf$ in place of $\sup$, and since

$$h = \inf_{k \geq 1} \left\{ \sup_{i \geq k} f_i \right\},$$

it follows that h is measurable. ////

Corollaries

- (a) The limit of every pointwise convergent sequence of complex measurable functions is measurable.
- (b) If f and g are measurable (with range in $[-\infty, \infty]$), then so are $\max\{f, g\}$ and $\min\{f, g\}$. In particular, this is true of the functions

$$f^+ = \max\{f, 0\} \quad \text{and} \quad f^- = -\min\{f, 0\}.$$

1.15 The above functions f^+ and f^- are called the *positive and negative parts* of f . We have $|f| = f^+ + f^-$ and $f = f^+ - f^-$, a standard representation of f as a difference of two nonnegative functions, with a certain minimality property:

Proposition If $f = g - h$, $g \geq 0$, and $h \geq 0$, then $f^+ \leq g$ and $f^- \leq h$.

PROOF $f \leq g$ and $0 \leq g$ clearly implies $\max\{f, 0\} \leq g$. ////

Simple Functions

1.16 Definition A complex function s on a measurable space X whose range consists of only finitely many points will be called a *simple function*. Among these are the nonnegative simple functions, whose range is a finite subset of $[0, \infty)$. Note that we explicitly exclude ∞ from the values of a simple function.

If $\alpha_1, \dots, \alpha_n$ are the distinct values of a simple function s , and if we set $A_i = \{x: s(x) = \alpha_i\}$, then clearly

$$s = \sum_{i=1}^n \alpha_i \chi_{A_i},$$

where χ_{A_i} is the characteristic function of A_i , as defined in Sec. 1.9(d). X

It is also clear that s is measurable if and only if each of the sets A_i is measurable.

1.17 Theorem Let $f: X \rightarrow [0, \infty]$ be measurable. There exist simple measurable functions s_n on X such that

- (a) $0 \leq s_1 \leq s_2 \leq \dots \leq f$.
 (b) $s_n(x) \rightarrow f(x)$ as $n \rightarrow \infty$, for every $x \in X$.

PROOF Put $\delta_n = 2^{-n}$. To each positive integer n and each real number t corresponds a unique integer $k = k_n(t)$ that satisfies $k\delta_n \leq t < (k+1)\delta_n$. Define

$$\varphi_n(t) = \begin{cases} k_n(t)\delta_n & \text{if } 0 \leq t < n \\ n & \text{if } n \leq t \leq \infty. \end{cases} \quad (1)$$

Each φ_n is then a Borel function on $[0, \infty]$,

$$t - \delta_n < \varphi_n(t) \leq t \quad \text{if } 0 \leq t \leq n, \quad (2)$$

$0 \leq \varphi_1 \leq \varphi_2 \leq \dots \leq t$, and $\varphi_n(t) \rightarrow t$ as $n \rightarrow \infty$, for every $t \in [0, \infty]$. It follows that the functions

$$s_n = \varphi_n \circ f \quad (3)$$

satisfy (a) and (b); they are measurable, by Theorem 1.12(d). ///

Elementary Properties of Measures

1.18 Definition

- (a) A positive measure is a function μ , defined on a σ -algebra $\mathfrak{M}$, whose range is in $[0, \infty]$ and which is *countably additive*. This means that if $\{A_i\}$ is a disjoint countable collection of members of $\mathfrak{M}$, then

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

To avoid trivialities, we shall also assume that $\mu(A) < \infty$ for at least one $A \in \mathfrak{M}$.

- (b) A *measure space* is a measurable space which has a positive measure defined on the σ -algebra of its measurable sets.
 (c) A *complex measure* is a complex-valued countably additive function defined on a σ -algebra.

X Note: What we have called a *positive measure* is frequently just called a *measure*; we add the word "positive" for emphasis. If $\mu(E) = 0$ for every $E \in \mathfrak{M}$,

then μ is a positive measure, by our definition. The value ∞ is admissible for a positive measure; but when we talk of a *complex* measure μ , it is understood that $\mu(E)$ is a complex number, for every $E \in \mathfrak{M}$. The *real measures* form a subclass of the complex ones, of course.

1.19 Theorem Let μ be a positive measure on a σ -algebra $\mathfrak{M}$. Then

(a) $\mu(\emptyset) = 0$.

(b) $\mu(A_1 \cup \dots \cup A_n) = \mu(A_1) + \dots + \mu(A_n)$ if $A_1, \dots, A_n$ are pairwise disjoint members of $\mathfrak{M}$.

(c) $A \subset B$ implies $\mu(A) \leq \mu(B)$ if $A \in \mathfrak{M}$, $B \in \mathfrak{M}$.

(d) $\mu(A_n) \rightarrow \mu(A)$ as $n \rightarrow \infty$ if $A = \bigcup_{n=1}^{\infty} A_n$, $A_n \in \mathfrak{M}$, and

$$A_1 \subset A_2 \subset A_3 \subset \dots$$

(e) $\mu(A_n) \rightarrow \mu(A)$ as $n \rightarrow \infty$ if $A = \bigcap_{n=1}^{\infty} A_n$, $A_n \in \mathfrak{M}$,

$$A_1 \supset A_2 \supset A_3 \supset \dots$$

and $\mu(A_1)$ is finite.

As the proof will show, these properties, with the exception of (c), also hold for complex measures; (b) is called *finite additivity*; (c) is called *monotonicity*.

PROOF

(a) Take $A \in \mathfrak{M}$ so that $\mu(A) < \infty$, and take $A_1 = A$ and $A_2 = A_3 = \dots = \emptyset$ in 1.18(1).

(b) Take $A_{n+1} = A_{n+2} = \dots = \emptyset$ in 1.18(1).

(c) Since $B = A \cup (B - A)$ and $A \cap (B - A) = \emptyset$, we see that (b) implies $\mu(B) = \mu(A) + \mu(B - A) \geq \mu(A)$.

(d) Put $B_1 = A_1$, and put $B_n = A_n - A_{n-1}$ for $n = 2, 3, 4, \dots$. Then $B_n \in \mathfrak{M}$, $B_i \cap B_j = \emptyset$ if $i \neq j$, $A_n = B_1 \cup \dots \cup B_n$, and $A = \bigcup_{i=1}^{\infty} B_i$. Hence

$$\mu(A_n) = \sum_{i=1}^n \mu(B_i) \quad \text{and} \quad \mu(A) = \sum_{i=1}^{\infty} \mu(B_i).$$

Now (d) follows, by the definition of the sum of an infinite series.

(e) Put $C_n = A_1 - A_n$. Then $C_1 \subset C_2 \subset C_3 \subset \dots$,

$$\mu(C_n) = \mu(A_1) - \mu(A_n),$$

$A_1 - A = \bigcup C_n$, and so (d) shows that

$$\mu(A_1) - \mu(A) = \mu(A_1 - A) = \lim_{n \rightarrow \infty} \mu(C_n) = \mu(A_1) - \lim_{n \rightarrow \infty} \mu(A_n).$$

1.20 Examples The construction of interesting measure spaces requires some labor, as we shall see. However, a few simple-minded examples can be given immediately:

- (a) For any $E \subset X$, where X is any set, define $\mu(E) = \infty$ if E is an infinite set, and let $\mu(E)$ be the number of points in E if E is finite. This μ is called the *counting measure* on X .
- (b) Fix $x_0 \in X$, define $\mu(E) = 1$ if $x_0 \in E$ and $\mu(E) = 0$ if $x_0 \notin E$, for any $E \subset X$. This μ may be called the *unit mass* concentrated at x_0 .
- (c) Let μ be the counting measure on the set $\{1, 2, 3, \dots\}$, let $A_n = \{n, n+1, n+2, \dots\}$. Then $\bigcap A_n = \emptyset$ but $\mu(A_n) = \infty$ for $n = 1, 2, 3, \dots$. This shows that the hypothesis

$$\mu(A_1) < \infty$$

is not superfluous in Theorem 1.19(e).

$$1) \left(\sup_{n \geq 1} f_n \right)^{-1}((\alpha, \infty]) = \bigcup_{n=1}^{\infty} f_n^{-1}((\alpha, \infty])$$

$$\begin{aligned} x \in \left(\sup_{n \geq 1} f_n \right)^{-1}((\alpha, \infty]) &\Leftrightarrow \sup_{n \geq 1} f_n(x) \in (\alpha, \infty] \Leftrightarrow \alpha < \sup_{n \geq 1} f_n(x) \Leftrightarrow \exists m; \alpha < f_m(x) \\ &\Leftrightarrow \exists m; f_m(x) \in (\alpha, \infty] \Leftrightarrow \exists m; x \in f_m^{-1}((\alpha, \infty]) \Leftrightarrow x \in \bigcup_{n=1}^{\infty} f_n^{-1}((\alpha, \infty]) \end{aligned}$$

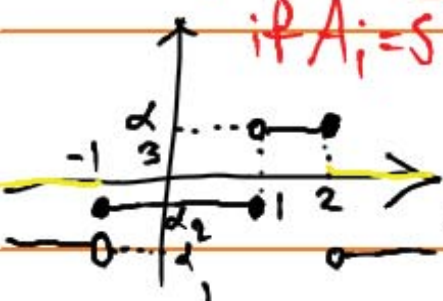
$$2) \forall a \in \mathbb{R} \exists a_+, a_- \in \mathbb{R}^{\geq 0}; \begin{cases} a = a_+ - a_- \\ |a| = a_+ + a_- \end{cases}$$

$\frac{a+|a|}{2} \xrightarrow{|a|=a} \frac{a+a}{2} = a$

$a_+ = \frac{a+|a|}{2}, a_- = \frac{a-|a|}{2}$

$a + |a| = 2a_+$

(3) A function $s: X \rightarrow [0, \infty]$ is called simple if its range, $s(X)$, is finite, say $s(X) = \{\alpha_1, \dots, \alpha_n\}$. Then if $A_i = s^{-1}(\{\alpha_i\})$, then $s = \sum_{i=1}^n \alpha_i \chi_{A_i}$. Note that A_i 's are disjoint.



$$s(x) = \begin{cases} \alpha_1 & x < -1 \text{ \& } x > 2 \\ \alpha_2 & -1 \leq x \leq 1 \\ \alpha_3 & 1 < x \leq 2 \\ 0 & \text{elsewhere} \end{cases}$$

$$x < -1 \text{ \& } x > 2 = s^{-1}(\{\alpha_1\}) = A_1$$

$$-1 \leq x \leq 1 = s^{-1}(\{\alpha_2\}) = A_2$$

$$1 < x \leq 2 = s^{-1}(\{\alpha_3\}) = A_3$$

$$X = \bigcup_{i=1}^3 s^{-1}(\{\alpha_i\}), \quad s(x) = \sum_{i=1}^3 \alpha_i \chi_{A_i}, \text{ where } A_i = s^{-1}(\{\alpha_i\}) \quad (A_i \cap A_j = s^{-1}(\{\alpha_i\}) \cap s^{-1}(\{\alpha_j\}) = s^{-1}(\{\alpha_i\} \cap \{\alpha_j\}) = s^{-1}(\emptyset) = \emptyset)$$

Use induction to prove

(4) $A_n = B_1 \cup \dots \cup B_n$.

Proof: Let $A_{n-1} = B_1 \cup \dots \cup B_{n-1}$.

$$\rightarrow x \in A_n = A_{n-1} \cup B_n \Rightarrow \begin{cases} x \in B_n \subseteq \bigcup_{i=1}^n B_i \\ x \in A_{n-1} = \bigcup_{i=1}^{n-1} B_i \subseteq \bigcup_{i=1}^n B_i \end{cases} \Rightarrow x \in \bigcup_{i=1}^n B_i$$

$$x \in \bigcup_{i=1}^n B_i \Rightarrow \exists j \leq n; x \in B_j = A_j - A_{j-1} \Rightarrow x \in A_j \subseteq A_n \quad \checkmark$$

Integration of Positive Functions

In this section, $\mathfrak{M}$ will be a σ -algebra in a set X and μ will be a positive measure on $\mathfrak{M}$.

1.23 Definition If $s: X \rightarrow [0, \infty)$ is a measurable simple function, of the form

$$s = \sum_{i=1}^n \alpha_i \chi_{A_i} \quad \rightarrow A_i = s^{-1}(\{\alpha_i\}) \quad (1)$$

where $\alpha_1, \dots, \alpha_n$ are the distinct values of s (compare Definition 1.16), and if $E \in \mathfrak{M}$, we define

$$\int_E s \, d\mu = \sum_{i=1}^n \alpha_i \mu(A_i \cap E). \quad (2)$$

The convention $0 \cdot \infty = 0$ is used here; it may happen that $\alpha_i = 0$ for some i and that $\mu(A_i \cap E) = \infty$.

If $f: X \rightarrow [0, \infty]$ is measurable, and $E \in \mathfrak{M}$, we define

$$\int_E f \, d\mu = \sup_{0 \leq s \leq f} \int_E s \, d\mu, \quad (3)$$

the supremum being taken over all simple measurable functions s such that $0 \leq s \leq f$.

The left member of (3) is called the *Lebesgue integral* of f over E , with respect to the measure μ . It is a number in $[0, \infty]$.

Observe that we apparently have two definitions for $\int_E f \, d\mu$ if f is simple, namely, (2) and (3). However, these assign the same value to the integral, since f is, in this case, the largest of the functions s which occur on the right of (3).

1.24 The following propositions are immediate consequences of the definitions. The functions and sets occurring in them are assumed to be measurable:

- (a) If $0 \leq f \leq g$, then $\int_E f \, d\mu \leq \int_E g \, d\mu$.
- (b) If $A \subset B$ and $f \geq 0$, then $\int_A f \, d\mu \leq \int_B f \, d\mu$.
- (c) If $f \geq 0$ and c is a constant, $0 \leq c < \infty$, then

$$\int_E cf \, d\mu = c \int_E f \, d\mu.$$

- (d) If $f(x) = 0$ for all $x \in E$, then $\int_E f d\mu = 0$, even if $\mu(E) = \infty$.
 (e) If $\mu(E) = 0$, then $\int_E f d\mu = 0$, even if $f(x) = \infty$ for every $x \in E$.
 (f) If $f \geq 0$, then $\int_E f d\mu = \int_X \chi_E f d\mu$.

This last result shows that we could have restricted our definition of integration to integrals over all of X , without losing any generality. If we wanted to integrate over subsets, we could then use (f) as the definition. It is purely a matter of taste which definition is preferred.

One may also remark here that every measurable subset E of a measure space X is again a measure space, in a perfectly natural way: The new measurable sets are simply those measurable subsets of X which lie in E , and the measure is unchanged, except that its domain is restricted. This shows again that as soon as we have integration defined over every measure space, we automatically have it defined over every measurable subset of every measure space.

1.25 Proposition *Let s and t be nonnegative measurable simple functions on X . For $E \in \mathfrak{M}$, define*

$$\varphi(E) = \int_E s d\mu. \quad (1)$$

Then φ is a measure on $\mathfrak{M}$. Also

$$\int_X (s + t) d\mu = \int_X s d\mu + \int_X t d\mu. \quad (2)$$

(This proposition contains provisional forms of Theorems 1.27 and 1.29.)

PROOF If s is as in Definition 1.23, and if $E_1, E_2, \dots$ are disjoint members of $\mathfrak{M}$ whose union is E , the countable additivity of μ shows that

$$\begin{aligned} \varphi(E) &= \sum_{i=1}^n \alpha_i \mu(A_i \cap E) = \sum_{i=1}^n \alpha_i \sum_{r=1}^{\infty} \mu(A_i \cap E_r) \\ &= \sum_{r=1}^{\infty} \sum_{i=1}^n \alpha_i \mu(A_i \cap E_r) = \sum_{r=1}^{\infty} \varphi(E_r). \end{aligned}$$

Also, $\varphi(\emptyset) = 0$, so that φ is not identically ∞ .

λ

Next, let s be as before, let $\beta_1, \dots, \beta_m$ be the distinct values of t , and let $B_j = \{x: t(x) = \beta_j\}$. If $E_{ij} = A_i \cap B_j$, then

$$\int_{E_{ij}} (s + t) d\mu = (\alpha_i + \beta_j)\mu(E_{ij})$$

and

$$\int_{E_{ij}} s d\mu + \int_{E_{ij}} t d\mu = \alpha_i \mu(E_{ij}) + \beta_j \mu(E_{ij}).$$

Thus (2) holds with E_{ij} in place of X . Since X is the disjoint union of the sets E_{ij} ($1 \leq i \leq n$, $1 \leq j \leq m$), the first half of our proposition implies that (2) holds. ////

We now come to the interesting part of the theory. One of its most remarkable features is the ease with which it handles limit operations.

1.26 Lebesgue's Monotone Convergence Theorem *Let $\{f_n\}$ be a sequence of measurable functions on X , and suppose that*

- (a) $0 \leq f_1(x) \leq f_2(x) \leq \dots \leq \infty$ for every $x \in X$,
- (b) $f_n(x) \rightarrow f(x)$ as $n \rightarrow \infty$, for every $x \in X$.

Then f is measurable, and

$$\int_X f_n d\mu \rightarrow \int_X f d\mu \quad \text{as } n \rightarrow \infty.$$

PROOF Since $\int f_n \leq \int f_{n+1}$, there exists an $\alpha \in [0, \infty]$ such that

$$\int_X f_n d\mu \rightarrow \alpha \quad \text{as } n \rightarrow \infty. \tag{1}$$

By Theorem 1.14, f is measurable. Since $f_n \leq f$, we have $\int f_n \leq \int f$ for every n , so (1) implies

$$\alpha \leq \int_X f d\mu. \tag{2}$$

Let s be any simple measurable function such that $0 \leq s \leq f$, let c be a constant, $0 < c < 1$, and define

$$E_n = \{x: f_n(x) \geq cs(x)\} \quad (n = 1, 2, 3, \dots). \tag{3}$$

Each E_n is measurable, $E_1 \subset E_2 \subset E_3 \subset \dots$, and $X = \bigcup E_n$. To see this equality, consider some $x \in X$. If $f(x) = 0$, then $x \in E_1$; if $f(x) > 0$, then $cs(x) < f(x)$, since $c < 1$; hence $x \in E_n$ for some n . Also χ

$$\int_X f_n d\mu \geq \int_{E_n} f_n d\mu \geq c \int_{E_n} s d\mu \quad (n = 1, 2, 3, \dots). \quad (4)$$

Let $n \rightarrow \infty$, applying Proposition 1.25 and Theorem 1.19(d) to the last integral in (4). The result is

$$\alpha \geq c \int_X s d\mu. \quad (5)$$

Since (5) holds for every $c < 1$, we have

$$\alpha \geq \int_X s d\mu \quad (6)$$

for every simple measurable s satisfying $0 \leq s \leq f$, so that

$$\alpha \geq \int_X f d\mu. \quad (7)$$

The theorem follows from (1), (2), and (7). ////

1.27 Theorem If $f_n: X \rightarrow [0, \infty]$ is measurable, for $n = 1, 2, 3, \dots$, and

$$f(x) = \sum_{n=1}^{\infty} f_n(x) \quad (x \in X), \quad (1)$$

then

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$$\int_X (f_1 + f_2) d\mu = \int_X f_1 d\mu + \int_X f_2 d\mu. \quad (3)$$

Next, put $g_N = f_1 + \dots + f_N$. The sequence $\{g_N\}$ converges monotonically to f , and if we apply induction to (3) we see that

$$\int g_N d\mu = \sum_{n=1}^N \int f_n d\mu. \quad (4)$$

X

Applying the monotone convergence theorem once more, we obtain (2), and the proof is complete.

If we let μ be the counting measure on a countable set, Theorem 1.27 is a statement about double series of nonnegative real numbers (which can of course be proved by more elementary means):

Corollary If $a_{ij} \geq 0$ for i and $j = 1, 2, 3, \dots$, then

$$\sum_{i=1}^{\infty} \sum_{j=1}^{\infty} a_{ij} = \sum_{j=1}^{\infty} \sum_{i=1}^{\infty} a_{ij}.$$

Let $\beta < \alpha$. So β is not an upper bound for C . So $\exists x \in C; x > \beta$. So $x \in D$ & $x > \beta$. Hence β is not an upper bound for D .

$$C \subseteq D \Rightarrow \sup C \leq \sup D$$

Let $x \in C$. Since $C \subseteq D$, $x \in D$. So $x \leq \beta$. Hence β is an upper bound for C .

(1) $\alpha \leq f \leq g \Rightarrow \int f d\mu \leq \int g d\mu$

$\Rightarrow \sup \int f d\mu \leq \sup \int g d\mu$

$\Rightarrow \int f d\mu \leq \int g d\mu$

Thus α is the least upper bound of C .

$\alpha \geq 0$

(b) $A \subseteq B \Rightarrow \forall 0 \leq f \leq g; \int f d\mu \leq \int g d\mu$

$\sum_{i=1}^n \alpha_i \mu(A_i \cap B) = \int f d\mu \Rightarrow \sup \int f d\mu \leq \sup \int g d\mu$

$\Rightarrow \int f d\mu \leq \int g d\mu$

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$\Rightarrow \int f d\mu \leq \int g d\mu$

an upper bound

$$\begin{aligned}
 (c) \quad \int_E c f d\mu &= c \int_E f d\mu \quad (c > 0) & c t &= \sum_{i=1}^n c \alpha_i \chi_{A_i} \quad t = \sum_{i=1}^n \alpha_i \chi_{A_i} \\
 \int_E c f d\mu &= \sup_{0 \leq s \leq c f} \int_E s d\mu = \sup_{0 \leq \begin{pmatrix} s \\ c \end{pmatrix} \leq f} \int_E s d\mu = \sup_{0 \leq t \leq f} \int_E c t d\mu = \sup_{0 \leq t \leq f} \sum_{i=1}^n c \alpha_i \mu(A_i) \\
 &= \sup_{0 \leq t \leq f} c \left(\sum_{i=1}^n \alpha_i \mu(A_i \cap E) \right) = c \sup_{0 \leq t \leq f} \int_E t d\mu = c \int_E f d\mu
 \end{aligned}$$

$$cA = \{cx : x \in A\} \quad \Rightarrow \quad \underbrace{\sup}_{\alpha} cA = c \underbrace{\sup}_{\beta} A$$

$$\begin{aligned}
 \forall x \in A, x \leq \beta &\Rightarrow \forall x \in A, cx \leq c\beta \Rightarrow \forall x \in cA, \\
 x \leq c\beta &\Rightarrow c\beta \text{ is an upper bound} \Rightarrow \alpha \leq c\beta. \\
 &\text{for } cA
 \end{aligned}$$

$$\forall x \in cA, cx \leq \alpha \Rightarrow \forall x \in A, x \leq \frac{\alpha}{c} \Rightarrow \beta \leq \frac{\alpha}{c} \Rightarrow c\beta \leq \alpha.$$

□

$$(d) f=0 \Rightarrow \int_E f d\mu = \sup \left\{ \int_E s d\mu : 0 \leq s \leq f \right\} = \sup \left\{ \int_E 0 d\mu \right\} = \sup \{0\} = 0$$

$$s=0=0 \cdot \chi$$

$$0 \cdot \chi = 0$$

$$\int_E 0 d\mu = 0 \mu(E \cap X) = 0$$

$$(e) \mu(E)=0 \Rightarrow \int_E f d\mu = \sup \int_E s d\mu = \sup \sum_{i=1}^n \alpha_i \mu(E \cap A_i) = \sup \{0\} = 0$$

$0 \leq s \leq f$ $0 \leq s \leq \sum_{i=1}^n \alpha_i \chi_{A_i}$

$\mu(E \cap A_i) = 0$ since $E \cap A_i \subseteq E$

Integration of Positive Functions

In this section, $\mathfrak{M}$ will be a σ -algebra in a set X and μ will be a positive measure on $\mathfrak{M}$.

1.23 Definition If $s: X \rightarrow [0, \infty)$ is a measurable simple function, of the form

$$s = \sum_{i=1}^n \alpha_i \chi_{A_i}, \quad (1)$$

where $\alpha_1, \dots, \alpha_n$ are the distinct values of s (compare Definition 1.16), and if $E \in \mathfrak{M}$, we define

$$\int_E s \, d\mu = \sum_{i=1}^n \alpha_i \mu(A_i \cap E). \quad (2)$$

The convention $0 \cdot \infty = 0$ is used here; it may happen that $\alpha_i = 0$ for some i and that $\mu(A_i \cap E) = \infty$.

If $f: X \rightarrow [0, \infty]$ is measurable, and $E \in \mathfrak{M}$, we define

$$\int_E f \, d\mu = \sup \int_E s \, d\mu, \quad (3)$$

the supremum being taken over all simple measurable functions s such that $0 \leq s \leq f$.

The left member of (3) is called the *Lebesgue integral* of f over E , with respect to the measure μ . It is a number in $[0, \infty]$.

Observe that we apparently have two definitions for $\int_E f \, d\mu$ if f is simple, namely, (2) and (3). However, these assign the same value to the integral, since f is, in this case, the largest of the functions s which occur on the right of (3).

1.24 The following propositions are immediate consequences of the definitions. The functions and sets occurring in them are assumed to be measurable:

- (a) If $0 \leq f \leq g$, then $\int_E f \, d\mu \leq \int_E g \, d\mu$.
- (b) If $A \subset B$ and $f \geq 0$, then $\int_A f \, d\mu \leq \int_B f \, d\mu$.
- (c) If $f \geq 0$ and c is a constant, $0 \leq c < \infty$, then

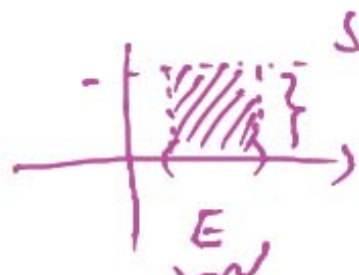
$$\int_E cf \, d\mu = c \int_E f \, d\mu.$$

- (d) If $f(x) = 0$ for all $x \in E$, then $\int_E f d\mu = 0$, even if $\mu(E) = \infty$.
 (e) If $\mu(E) = 0$, then $\int_E f d\mu = 0$, even if $f(x) = \infty$ for every $x \in E$.
 (f) If $f \geq 0$, then $\int_E f d\mu = \int_X \chi_E f d\mu$.

This last result shows that we could have restricted our definition of integration to integrals over all of X , without losing any generality. If we wanted to integrate over subsets, we could then use (f) as the definition. It is purely a matter of taste which definition is preferred.

One may also remark here that every measurable subset E of a measure space X is again a measure space, in a perfectly natural way: The new measurable sets are simply those measurable subsets of X which lie in E , and the measure is unchanged, except that its domain is restricted. This shows again that as soon as we have integration defined over every measure space, we automatically have it defined over every measurable subset of every measure space.

1.25 Proposition Let s and t be nonnegative measurable simple functions on X . For $E \in \mathfrak{M}$, define

$$\varphi(E) = \int_E s d\mu. \quad (1)$$


Then φ is a measure on $\mathfrak{M}$. Also

$$\int_X (s + t) d\mu = \int_X s d\mu + \int_X t d\mu. \quad (2)$$

(This proposition contains provisional forms of Theorems 1.27 and 1.29.)

PROOF If s is as in Definition 1.23, and if $E_1, E_2, \dots$ are disjoint members of $\mathfrak{M}$ whose union is E , the countable additivity of μ shows that

$$\begin{aligned} \varphi(E) &= \int_E s d\mu = \int_E \sum_{i=1}^n \alpha_i \chi_{A_i} d\mu = \sum_{i=1}^n \alpha_i \int_E \chi_{A_i} d\mu \\ &= \sum_{i=1}^n \alpha_i \mu(E \cap A_i) = \sum_{i=1}^n \alpha_i \sum_{r=1}^{\infty} \mu(A_i \cap E_r) = \sum_{r=1}^{\infty} \sum_{i=1}^n \alpha_i \mu(A_i \cap E_r) = \sum_{r=1}^{\infty} \varphi(E_r). \end{aligned}$$

Also, $\varphi(\emptyset) = 0$, so that φ is not identically ∞ .

Next, let t be as before, let $\beta_1, \dots, \beta_m$ be the distinct values of t , and let

$B_j = \{x: t(x) = \beta_j\}$. If $E_{ij} = A_i \cap B_j$, then

$$s+t = \sum_{i=1}^m \sum_{j=1}^n (\alpha_i + \beta_j) \chi_{E_{ij}}$$

$$\int_{E_{ij}} (s+t) d\mu = (\alpha_i + \beta_j) \mu(E_{ij})$$

and

$$\int_{E_{ij}} s d\mu + \int_{E_{ij}} t d\mu = \alpha_i \mu(E_{ij}) + \beta_j \mu(E_{ij}).$$

Thus (2) holds with E_{ij} in place of X . Since X is the disjoint union of the sets E_{ij} ($1 \leq i \leq m, 1 \leq j \leq n$), the first half of our proposition implies that (2) holds.

We now come to the interesting part of the theory. One of its most remarkable features is the ease with which it handles limit operations.

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Then f is measurable, and

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By Theorem 1.14, f is measurable. Since $f_n \leq f$, we have $\int f_n \leq \int f$ for every n , so (1) implies

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$$\alpha \geq c \int_X s d\mu = c \int_{\bigcup E_n} s d\mu = c \int_{E_n} s d\mu \quad (5)$$

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for every simple measurable s satisfying $0 \leq s \leq f$, so that

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The theorem follows from (1), (2), and (7).

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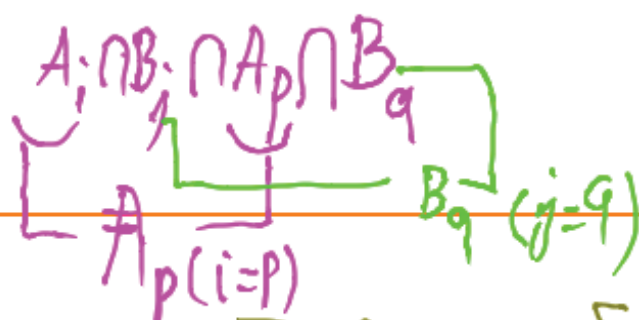
$$\sum_{i=1}^{\infty} \sum_{j=1}^{\infty} a_{ij} = \sum_{j=1}^{\infty} \sum_{i=1}^{\infty} a_{ij}.$$

① $\int f d\mu = \int \sum_X \chi_E d\mu$: It is enough to prove that $\int s d\mu = \int \sum_X \chi_E d\mu$

To this end, let $s = \sum_{i=1}^n \alpha_i \chi_{A_i}$. Then $\int s d\mu = \sum_{i=1}^n \alpha_i \mu(A_i \cap E)$

$$\int \sum_X \chi_E d\mu = \sum_{i=1}^n \alpha_i \int \chi_{A_i} \chi_E d\mu = \sum_{i=1}^n \alpha_i \mu(A_i \cap E). \checkmark$$

$$\textcircled{2} \int_{E_{pq}} s+t d\mu = \sum_{\substack{1 \leq i \leq n \\ 1 \leq j \leq m}} (\alpha_i + \beta_j) \mu(E_{ij} \cap E_{pq}) = (\alpha_p + \beta_q) \mu(E_{pq})$$



$$\textcircled{3} \int (s+t) d\mu = \varphi(X) = \varphi(\bigcup_{i,j} E_{ij}) = \sum_{i,j} \varphi(E_{ij}) = \sum_{i,j} (\varphi_s(E_{ij}) + \varphi_t(E_{ij})) = \varphi_s(X) + \varphi_t(X) = \int s d\mu + \int t d\mu. \square$$

Let $x \in X$.

$$\textcircled{4} \quad 0 \leq s \leq f \Rightarrow c s \leq f(x) \\ 0 \leq c \leq 1 \quad \parallel \quad \sup_n f(x) \\ \Rightarrow x \in E_n \Rightarrow x \in \bigcup_{n=1}^{\infty} E_n \subseteq X.$$

$\Rightarrow CS(x)$ is not an upper bound for $\{f_1(x), f_2(x), \dots\}$
 $\Rightarrow CS(x) \neq f_m(x)$ for some m
 $\therefore \bigcup_{n=1}^{\infty} E_n = X$

$$\alpha = \sup A \quad \left\{ \begin{array}{l} \forall \varepsilon > 0 \exists x \in A; \alpha - \varepsilon < x \\ \nexists u, v \in A \text{ s.t. } u < v \end{array} \right.$$

$\forall \epsilon > 0 \quad \alpha - \epsilon < \alpha$
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an upper
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1.28 Fatou's Lemma If $f_n: X \rightarrow [0, \infty]$ is measurable, for each positive integer n , then

$$\int_X \left(\liminf_{n \rightarrow \infty} f_n \right) d\mu \leq \liminf_{n \rightarrow \infty} \int_X f_n d\mu. \quad (1)$$

Strict inequality can occur in (1); see Exercise 8.

PROOF Put

$$g_k(x) = \inf_{i \geq k} f_i(x) \quad (k = 1, 2, 3, \dots; x \in X).$$

Then $g_k \leq f_k$, so that

$$\int_X g_k d\mu \leq \int_X f_k d\mu \quad (k = 1, 2, 3, \dots).$$

Also, $0 \leq g_1 \leq g_2 \leq \dots$, each g_k is measurable, by Theorem 1.14, and $g_k(x) \rightarrow \liminf_{n \rightarrow \infty} f_n(x)$ as $k \rightarrow \infty$, by Definition 1.13. The monotone convergence theorem shows therefore that the left side of (3) tends to the left side of (1), as $k \rightarrow \infty$. Hence (1) follows from (3). ////

1.29 Theorem Suppose $f: X \rightarrow [0, \infty]$ is measurable, and

$$\left(d\varphi = f d\mu \right) \equiv \varphi(E) = \int_E f d\mu \quad (E \in \mathfrak{M}). \quad (1)$$

Then φ is a measure on $\mathfrak{M}$, and

$$\int_X g d\varphi = \int_X g f d\mu \quad (2)$$

for every measurable g on X with range in $[0, \infty]$.

PROOF Let $E_1, E_2, E_3, \dots$ be disjoint members of $\mathfrak{M}$ whose union is E . Observe that

$$\chi_E f = \sum_{j=1}^{\infty} \chi_{E_j} f \quad (3)$$

and that

$$\varphi(E) = \int_X \chi_E f d\mu, \quad \varphi(E_j) = \int_X \chi_{E_j} f d\mu. \quad (4)$$

It now follows from Theorem 1.27 that

$$\varphi(E) = \sum_{j=1}^{\infty} \varphi(E_j). \quad (5)$$

Since $\varphi(\emptyset) = 0$, (5) proves that φ is a measure.

Next, (1) shows that (2) holds whenever $g = \chi_E$ for some $E \in \mathfrak{M}$. Hence (2) holds for every simple measurable function g , and the general case follows from the monotone convergence theorem. ////

Remark The second assertion of Theorem 1.29 is sometimes written in the form

$$d\varphi = f d\mu. \quad (6)$$

We assign no independent meaning to the symbols $d\varphi$ and $d\mu$; (6) merely means that (2) holds for every measurable $g \geq 0$.

Theorem 1.29 has a very important converse, the Radon-Nikodym theorem, which will be proved in Chap. 6.

Integration of Complex Functions

As before, μ will in this section be a positive measure on an arbitrary measurable space X .

1.30 Definition We define $L^1(\mu)$ to be the collection of all complex measurable functions f on X for which

$$\int_X |f| d\mu < \infty.$$

$: X \rightarrow [0, \infty)$
 $|a+ib| = \sqrt{a^2+b^2}$
 $f = u+iv$

Note that the measurability of f implies that of $|f|$, as we saw in Proposition 1.9(b); hence the above integral is defined.

The members of $L^1(\mu)$ are called *Lebesgue integrable functions* (with respect to μ) or *summable functions*. The significance of the exponent 1 will become clear in Chap. 3.

1.31 Definition If $f = u + iv$, where u and v are real measurable functions on X , and if $f \in L^1(\mu)$, we define

$$\int_E f d\mu = \int_E u^+ d\mu - \int_E u^- d\mu + i \int_E v^+ d\mu - i \int_E v^- d\mu \in \mathbb{C} \quad (1)$$

for every measurable set E .

Here u^+ and u^- are the positive and negative parts of u defined in

Here u^+ and u^- are the positive and negative parts of u , as defined in Sec. 1.15; v^+ and v^- are similarly obtained from v . These four functions are measurable, real, and nonnegative; hence the four integrals on the right of (1) exist, by Definition 1.23. Furthermore, we have $u^+ \leq |u| < |f|$, etc., so that each of these four integrals is finite. Thus (1) defines the integral on the left as a complex number.

Occasionally it is desirable to define the integral of a measurable function f with range in $[-\infty, \infty]$ to be

$$f = f^+ - f^- \quad \int_E f \, d\mu = \int_E f^+ \, d\mu - \int_E f^- \, d\mu, \quad (2)$$

provided that at least one of the integrals on the right of (2) is finite. The left side of (2) is then a number in $[-\infty, \infty]$.

1.32 Theorem Suppose f and $g \in L^1(\mu)$ and α and β are complex numbers. Then $\alpha f + \beta g \in L^1(\mu)$, and

$$\int_X (\alpha f + \beta g) \, d\mu = \alpha \int_X f \, d\mu + \beta \int_X g \, d\mu. \quad (1)$$

PROOF The measurability of $\alpha f + \beta g$ follows from Proposition 1.9(c). By Sec. 1.24 and Theorem 1.27,

$$\begin{aligned} \int_X |\alpha f + \beta g| \, d\mu &\leq \int_X (|\alpha| |f| + |\beta| |g|) \, d\mu \\ &= |\alpha| \int_X |f| \, d\mu + |\beta| \int_X |g| \, d\mu < \infty. \end{aligned}$$

Thus $\alpha f + \beta g \in L^1(\mu)$.

To prove (1), it is clearly sufficient to prove

$$\int_X (f + g) \, d\mu = \int_X f \, d\mu + \int_X g \, d\mu \quad (2)$$

and

$$\int_X (\alpha f) \, d\mu = \alpha \int_X f \, d\mu, \quad (3)$$

and the general case of (2) will follow if we prove (2) for real f and g in $L^1(\mu)$.

Assuming this, and setting $h = f + g$, we have

$$h = h^+ - h^- = f^+ - f^- + g^+ - g^-$$

$$h^+ + f^- + g^- = f^+ + g^+ + h^-. \quad (4)$$

By Theorem 1.27,

$$\int h^+ + \int f^- + \int g^- = \int f^+ + \int g^+ + \int h^-, \quad (5)$$

and since each of these integrals is finite, we may transpose and obtain (2).

That (3) holds if $\alpha \geq 0$ follows from Proposition 1.24(c). It is easy to verify that (3) holds if $\alpha = -1$, using relations like $(-u)^+ = u^-$. The case $\alpha = i$ is also easy: If $f = u + iv$, then

$$\begin{aligned} \int (if) &= \int (iu - v) = \int (-v) + i \int u = - \int v + i \int u = i \left(\int u + i \int v \right) \\ &= i \int f. \end{aligned}$$

Left to the students

Combining these cases with (2), we obtain (3) for any complex α . ////

1.33 Theorem If $f \in L^1(\mu)$, then

$$\left| \int_X f d\mu \right| \leq \int_X |f| d\mu.$$

very interesting

PROOF Put $z = \int_X f d\mu$. Since z is a complex number, there is a complex number α , with $|\alpha| = 1$, such that $\alpha z = |z|$. Let u be the real part of αf . Then $u \leq |\alpha f| = |f|$. Hence

$$\left| \int_X f d\mu \right| = \alpha \int_X f d\mu = \int_X \alpha f d\mu = \int_X u d\mu \leq \int_X |f| d\mu.$$

$\alpha = \frac{|z|}{z}$ $z \neq 0$

The third of the above equalities holds since the preceding ones show that $\int \alpha f d\mu$ is real. ////

1.34 Lebesgue's Dominated Convergence Theorem Suppose $\{f_n\}$ is a sequence of complex measurable functions on X such that

$$f(x) = \lim_{n \rightarrow \infty} f_n(x) \quad (1)$$

exists for every $x \in X$. If there is a function $g \in L^1(\mu)$ such that

$$|f_n(x)| \leq g(x) \quad (n = 1, 2, 3, \dots; x \in X), \quad (2)$$

then $f \in L^1(\mu)$,

$$\lim_{n \rightarrow \infty} \int_X |f_n - f| d\mu = 0, \quad (3)$$

and

$$\lim_{n \rightarrow \infty} \int_X f_n d\mu = \int_X f d\mu.$$

Handwritten notes: $\forall \epsilon \exists N \forall n \geq N; |\int f_n d\mu - \int f d\mu| < \epsilon$ Calculus $\int (f_n - f) d\mu$

$$|f_n| + |f| \leq g + g = 2g \quad (4)$$

PROOF Since $|f| \leq g$ and f is measurable, $f \in L^1(\mu)$. Since $|f_n - f| \leq 2g$, Fatou's lemma applies to the functions $2g - |f_n - f|$ and yields

$$\begin{aligned} \int_X 2g d\mu &\leq \liminf_{n \rightarrow \infty} \int_X (2g - |f_n - f|) d\mu \\ &= \int_X 2g d\mu + \liminf_{n \rightarrow \infty} \left(- \int_X |f_n - f| d\mu \right) \\ &= \int_X 2g d\mu - \limsup_{n \rightarrow \infty} \int_X |f_n - f| d\mu. \end{aligned}$$

Handwritten notes: (4) $|f_n| \leq g(x) \forall n$, $\liminf (2g - |f_n - f|) \geq 0$, F.L., $\liminf_{n \rightarrow \infty} \int (2g - |f_n - f|) d\mu \geq 0$, $|f_n| \leq g(x) \Rightarrow \liminf |f_n| \leq \liminf g(x)$, $|f(x)| \leq g(x)$, (3) $\int |f| \leq \int g < +\infty$

Since $\int 2g d\mu$ is finite, we may subtract it and obtain

$$0 \leq \liminf_{n \rightarrow \infty} \int_X |f_n - f| d\mu \leq \limsup_{n \rightarrow \infty} \int_X |f_n - f| d\mu \leq 0.$$

$$\liminf (x_n + y_n) \leq \liminf x_n + \limsup y_n \quad (5)$$

If a sequence of nonnegative real numbers fails to converge to 0, then its upper limit is positive. Thus (5) implies (3). By Theorem 1.33, applied to $f_n - f$, (3) implies (4). ////

$$g \xrightarrow{\text{increasing}} \liminf_k f = \liminf_{n \rightarrow \infty} f_n \xRightarrow{\text{M.C.T.}} \lim_{k \rightarrow \infty} \int g_k d\mu = \int \liminf_{n \rightarrow \infty} f_n d\mu$$

On the other hand, $\int g_k \leq \int f_n$. So $\liminf_{k \rightarrow \infty} \int g_k d\mu \leq \liminf_{n \rightarrow \infty} \int f_n d\mu$. $\square$

$$1 + 0 = \chi_A(x) + \chi_B(x) \in \text{either } A \text{ or } B \text{ w.l.o.g.}$$

$$\begin{aligned}
 \textcircled{2} \chi_{A \cup B} &= \chi_A + \chi_B : & \chi_{A \cup B} &= \begin{cases} 1 & \text{if } (x \in A \cup B) \text{ } x \in A \text{ or } x \in B \text{ (or } x \in A \text{ \& } x \in B) \\ 0 & \text{if } x \notin A \cup B \text{ neither } x \in A \text{ nor } x \in B \\ \parallel \\ 0 + 0 &= \chi_A(x) + \chi_B(x) \end{cases}
 \end{aligned}$$

1.35 Definition Let P be a property which a point x may or may not have. For instance, P might be the property " $f(x) > 0$ " if f is a given function, or it might be " $\{f_n(x)\}$ converges" if $\{f_n\}$ is a given sequence of functions.

If μ is a measure on a σ -algebra $\mathfrak{M}$ and if $E \in \mathfrak{M}$, the statement " P holds almost everywhere on E " (abbreviated to " P holds a.e. on E ") means that there exists an $N \in \mathfrak{M}$ such that $\mu(N) = 0$, $N \subset E$, and P holds at every point of $E - N$. This concept of a.e. depends of course very strongly on the given measure, and we shall write "a.e. $[\mu]$ " whenever clarity requires that the measure be indicated.

For example, if f and g are measurable functions and if

$$\mu(\{x: f(x) \neq g(x)\}) = 0, \quad (1)$$

we say that $f = g$ a.e. $[\mu]$ on X , and we may write $f \sim g$. This is easily seen to be an equivalence relation. The transitivity ($f \sim g$ and $g \sim h$ implies $f \sim h$) is a consequence of the fact that the union of two sets of measure 0 has measure 0.

Note that if $f \sim g$, then, for every $E \in \mathfrak{M}$,

$$\int_E f d\mu = \int_E g d\mu.$$

To see this, let N be the set which appears in (1); then E is the union of the disjoint sets $E - N$ and $E \cap N$; on $E - N$, $f = g$, and $\mu(E \cap N) = 0$.

1.36 Theorem Let $(X, \mathfrak{M}, \mu)$ be a measure space, let $\mathfrak{M}^*$ be the collection of all $E \subset X$ for which there exist sets A and $B \in \mathfrak{M}$ such that $A \subset E \subset B$ and $\mu(B - A) = 0$, and define $\mu^*(E) = \mu(A)$ in this situation. Then $\mathfrak{M}^*$ is a σ -algebra, and μ^* is a measure on $\mathfrak{M}^*$.

This extended measure μ^* is called *complete*, since all subsets of sets of measure 0 are now measurable; the σ -algebra $\mathfrak{M}^*$ is called the μ -completion of $\mathfrak{M}$.

1.38 Theorem Suppose $\{f_n\}$ is a sequence of complex measurable functions defined a.e. on X such that

$$\sum_{n=1}^{\infty} \int_X |f_n| d\mu < \infty. \quad (1)$$

Then the series

$$f(x) = \sum_{n=1}^{\infty} f_n(x) \quad (2)$$

converges for almost all x , $f \in L^1(\mu)$, and

$$\int_X f d\mu = \sum_{n=1}^{\infty} \int_X f_n d\mu. \quad (3)$$

PROOF Let S_n be the set on which f_n is defined, so that $\mu(S_n^c) = 0$. Put $\varphi(x) = \sum |f_n(x)|$, for $x \in S = \bigcap S_n$. Then $\mu(S^c) = 0$. By (1) and Theorem 1.27,

$f = g$ a.e.

$$\int_E f d\mu = \int_N f d\mu + \int_{E-N} f d\mu$$

$$\begin{aligned} \mu(A) &= \int_A f d\mu \\ E &= N \cup (E-N) \\ \mu(E) &= \mu(N) + \mu(E-N) \end{aligned}$$

$$\begin{aligned} \mu(A) &= 0 \\ B &\subset A \\ B &\in \mathfrak{M} \\ \text{(so } \mu(B) &= 0) \end{aligned}$$

X

$$\int_S \varphi \, d\mu < \infty. \quad (4)$$

If $E = \{x \in S: \varphi(x) < \infty\}$, it follows from (4) that $\mu(E^c) = 0$. The series (2) converges absolutely for every $x \in E$, and if $f(x)$ is defined by (2) for $x \in E$, then $|f(x)| \leq \varphi(x)$ on E , so that $f \in L^1(\mu)$ on E , by (4). If $g_n = f_1 + \cdots + f_n$, then $|g_n| \leq \varphi$, $g_n(x) \rightarrow f(x)$ for all $x \in E$, and Theorem 1.34 gives (3) with E in place of X . This is equivalent to (3), since $\mu(E^c) = 0$. ////

1.39 Theorem

- (a) Suppose $f: X \rightarrow [0, \infty]$ is measurable, $E \in \mathfrak{M}$, and $\int_E f \, d\mu = 0$. Then $f = 0$ a.e. on E .
 (b) Suppose $f \in L^1(\mu)$ and $\int_E f \, d\mu = 0$ for every $E \in \mathfrak{M}$. Then $f = 0$ a.e. on X .
 (c) Suppose $f \in L^1(\mu)$ and

$$\left| \int_X f \, d\mu \right| = \int_X |f| \, d\mu.$$

Then there is a constant α such that $\alpha f = |f|$ a.e. on X .

Note that (c) describes the condition under which equality holds in Theorem 1.33.

PROOF

- (a) If $A_n = \{x \in E: f(x) > 1/n\}$, $n = 1, 2, 3, \dots$, then

$$\frac{1}{n} \mu(A_n) \leq \int_{A_n} f \, d\mu \leq \int_E f \, d\mu = 0,$$

so that $\mu(A_n) = 0$. Since $\{x \in E: f(x) > 0\} = \bigcup A_n$, (a) follows.

- (b) Put $f = u + iv$, let $E = \{x: u(x) \geq 0\}$. The real part of $\int_E f \, d\mu$ is then $\int_E u^+ \, d\mu$. Hence $\int_E u^+ \, d\mu = 0$, and (a) implies that $u^+ = 0$ a.e. We conclude similarly that

$$u^- = v^+ = v^- = 0 \quad \text{a.e.}$$

- (c) Examine the proof of Theorem 1.33. Our present assumption implies that the last inequality in the proof of Theorem 1.33 must actually be an equality. Hence $\int (|f| - u) \, d\mu = 0$. Since $|f| - u \geq 0$, (a) shows that $|f| = u$ a.e. This says that the real part of αf is equal to $|f|$ a.e., hence $\alpha f = |f| = |f|$ a.e., which is the desired conclusion. ////

1.40 Theorem Suppose $\mu(X) < \infty$, $f \in L^1(\mu)$, S is a closed set in the complex plane, and the averages

$$A_E(f) = \frac{1}{\mu(E)} \int_E f \, d\mu$$

lie in S for every $E \in \mathfrak{M}$ with $\mu(E) > 0$. Then $f(x) \in S$ for almost all $x \in X$. X

PROOF Let Δ be a closed circular disc (with center at α and radius $r > 0$, say) in the complement of S . Since S is the union of a countable number of closed discs,

the complement of S . Since S^c is the union of countably many such discs, it is enough to prove that $\mu(E) = 0$, where $E = f^{-1}(\Delta)$.

If we had $\mu(E) > 0$, then

$$|A_E(f) - \alpha| = \frac{1}{\mu(E)} \left| \int_E (f - \alpha) d\mu \right| \leq \frac{1}{\mu(E)} \int_E |f - \alpha| d\mu \leq r,$$

which is impossible, since $A_E(f) \in S$. Hence $\mu(E) = 0$. ////

1.41 Theorem Let $\{E_k\}$ be a sequence of measurable sets in X , such that

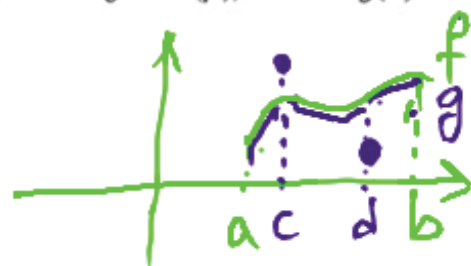
$$\sum_{k=1}^{\infty} \mu(E_k) < \infty. \quad (1)$$

Then almost all $x \in X$ lie in at most finitely many of the sets E_k .

PROOF If A is the set of all x which lie in infinitely many E_k , we have to prove that $\mu(A) = 0$. Put

$$g(x) = \sum_{k=1}^{\infty} \chi_{E_k}(x) \quad (x \in X). \quad (2)$$

For each x , each term in this series is either 0 or 1. Hence $x \in A$ if and only if $g(x) = \infty$. By Theorem 1.27, the integral of g over X is equal to the sum in (1). Thus $g \in L^1(\mu)$, and so $g(x) < \infty$ a.e. ////



$$N = \{x: f(x) \neq g(x)\} = \{c, d\}$$

$$\mu(N) = 0 \quad \therefore \text{So } f = g \text{ a.e.}$$

$$\text{Hence } \int_{[a,b]} f d\mu = \int_{[a,b]} g d\mu.$$

The Lebesgue measure on $\mathbb{R}$ says that $\mu(\{a\}) = 0$.
 So $\mu(\{a_1, a_2, \dots\}) = \mu\left(\bigcup_{i=1}^{\infty} \{a_i\}\right) = \sum_{i=1}^{\infty} \mu(\{a_i\}) = \sum_{i=1}^{\infty} 0 = 0 \checkmark$

1.35 Definition Let P be a property which a point x may or may not have. For instance, P might be the property " $f(x) > 0$ " if f is a given function, or it might be " $\{f_n(x)\}$ converges" if $\{f_n\}$ is a given sequence of functions.

If μ is a measure on a σ -algebra $\mathfrak{M}$ and if $E \in \mathfrak{M}$, the statement " P holds almost everywhere on E " (abbreviated to " P holds a.e. on E ") means that there exists an $N \in \mathfrak{M}$ such that $\mu(N) = 0$, $N \subset E$, and P holds at every point of $E - N$. This concept of a.e. depends of course very strongly on the given measure, and we shall write "a.e. $[\mu]$ " whenever clarity requires that the measure be indicated.

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Note that if $f \sim g$, then, for every $E \in \mathfrak{M}$,

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Then the series

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PROOF Let S_n be the set on which f_n is defined, so that $\mu(S_n^c) = 0$. Put $\varphi(x) = \sum |f_n(x)|$, for $x \in S = \bigcap S_n$. Then $\mu(S^c) = 0$. By (1) and Theorem 1.27,

$$\mu(\cup S_n^c) < \sum \mu(S_n^c) = 0$$

X

$$\int_S \varphi d\mu < \infty. \quad (4)$$

If $E = \{x \in S: \varphi(x) < \infty\}$, it follows from (4) that $\mu(E^c) = 0$. The series (2) converges absolutely for every $x \in E$, and if $f(x)$ is defined by (2) for $x \in E$, then $|f(x)| \leq \varphi(x)$ on E , so that $f \in L^1(\mu)$ on E , by (4). If $g_n = f_1 + \dots + f_n$, then $|g_n| \leq \varphi$, $g_n(x) \rightarrow f(x)$ for all $x \in E$, and Theorem 1.34 gives (3) with E in place of X . This is equivalent to (3), since $\mu(E^c) = 0$.

1.39 Theorem

- (a) Suppose $f: X \rightarrow [0, \infty]$ is measurable, $E \in \mathfrak{M}$, and $\int_E f d\mu = 0$. Then $f = 0$ a.e. on E .
 (b) Suppose $f \in L^1(\mu)$ and $\int_E f d\mu = 0$ for every $E \in \mathfrak{M}$. Then $f = 0$ a.e. on X .
 (c) Suppose $f \in L^1(\mu)$ and

$$\left| \int_X f d\mu \right| = \int_X |f| d\mu.$$

Then there is a constant α such that $\alpha f = |f|$ a.e. on X .

Note that (c) describes the condition under which equality holds in Theorem

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$$\frac{1}{n} \mu(A_n) \leq \int_{A_n} f d\mu \leq \int_E f d\mu = 0,$$

so that $\mu(A_n) = 0$. Since $\{x \in E: f(x) > 0\} = \bigcup_{n=1}^{\infty} A_n$, (a) follows.

- (b) Put $f = u + iv$, let $E = \{x: u(x) \geq 0\}$. The real part of $\int_E f d\mu$ is then $\int_E u^+ d\mu$. Hence $\int_E u^+ d\mu = 0$, and (a) implies that $u^+ = 0$ a.e. We conclude similarly that

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1.40 Theorem Suppose $\mu(X) < \infty$, $f \in L^1(\mu)$, S is a closed set in the complex plane, and the averages

$$A_E(f) = \frac{1}{\mu(E)} \int_E f d\mu$$

lie in S for every $E \in \mathfrak{M}$ with $\mu(E) > 0$. Then $f(x) \in S$ for almost all $x \in X$.

PROOF Let Δ be a closed circular disc (with center at α and radius $r > 0$, say) in the complement of S . Since Δ is the union of an increasing sequence of discs, it

$\sum_{k=1}^n \int f_k \rightarrow \int f$
 $\int g_n \rightarrow \int f$ or
 $\int f = 0$
 But $f \neq 0$

$$\{x: f(x) \neq 0\} = \bigcup_{n=1}^{\infty} \{x: |f(x)| > \frac{1}{n}\} = \bigcup_{n=1}^{\infty} A_n$$

$$0 \leq \mu(\{x: f(x) \neq 0\}) \leq \sum_{n=1}^{\infty} \mu(A_n) = 0$$

$$\int f = \int u + i \int v$$

$$\Re \int f = \int \Re f = \int u = \int u^+ - \int u^-$$

$$= 0 - 0 = 0$$

Since $\bar{u} = 0$ on E

X

the complement of S . Since S^c is the union of countably many such discs, it is enough to prove that $\mu(E) = 0$, where $E = f^{-1}(\Delta)$.

If we had $\mu(E) > 0$, then

$$|A_E(f) - \alpha| = \frac{1}{\mu(E)} \left| \int_E (f - \alpha) d\mu \right| \leq \frac{1}{\mu(E)} \int_E |f - \alpha| d\mu \leq r,$$

which is impossible, since $A_E(f) \in S$. Hence $\mu(E) = 0$.

////

1.41 Theorem Let $\{E_k\}$ be a sequence of measurable sets in X , such that

$$\sum_{k=1}^{\infty} \mu(E_k) < \infty. \quad (1)$$

Then almost all $x \in X$ lie in at most finitely many of the sets E_k .

PROOF If A is the set of all x which lie in infinitely many E_k , we have to prove that $\mu(A) = 0$. Put

$$g(x) = \sum_{k=1}^{\infty} \chi_{E_k}(x) \quad (x \in X). \quad (2)$$

For each x , each term in this series is either 0 or 1. Hence $x \in A$ if and only if $g(x) = \infty$. By Theorem 1.27, the integral of g over X is equal to the sum in (1). Thus $g \in L^1(\mu)$, and so $g(x) < \infty$ a.e.

////

① If $\mu(E^c) \neq 0$, then $\int_X f d\mu \geq \int_{E^c} f d\mu = \int_{E^c} \infty d\mu = \infty \cdot \mu(E^c) = \infty$.
 since $f(x) = \infty$ on E^c .

② $u^+ = 0$ a.e.
 $u^- = 0$ a.e.
 $\vdots$

$$\{x: u(x) \neq 0\} \subseteq \{x: u^+(x) \neq 0\} \cup \{x: u^-(x) \neq 0\}$$

To prove, let $x \notin \{x: u^+(x) \neq 0\} \cup \{x: u^-(x) \neq 0\}$.
 Then $u^+(x) = 0 = u^-(x)$. So $u(x) = 0$

$$\therefore \mu\{x: u(x) \neq 0\} \leq \underbrace{\mu\{x: u^+(x) \neq 0\}}_0 + \underbrace{\mu\{x: u^-(x) \neq 0\}}_0$$

So $\mu\{x: u(x) \neq 0\} = 0$. Hence $u = 0$ a.e.

x

Topological Preliminaries

2.3 Definitions Let X be a topological space, as defined in Sec. 1.2.

- (a) A set $E \subset X$ is *closed* if its complement E^c is open. (Hence $\emptyset$ and X are closed, finite unions of closed sets are closed, and arbitrary intersections of closed sets are closed.)
- (b) The *closure* $\bar{E}$ of a set $E \subset X$ is the smallest closed set in X which contains E . (The following argument proves the existence of $\bar{E}$: The collection Ω of all closed subsets of X which contain E is not empty, since $X \in \Omega$; let $\bar{E}$ be the intersection of all members of Ω .)
- (c) A set $K \subset X$ is *compact* if every open cover of K contains a finite subcover. More explicitly, the requirement is that if $\{V_\alpha\}$ is a collection of open sets whose union contains K , then the union of some finite subcollection of $\{V_\alpha\}$ also contains K .

In particular, if X is itself compact, then X is called a *compact space*.

- (d) A *neighborhood* of a point $p \in X$ is any open subset of X which contains p . (The use of this term is not quite standardized; some use "neighborhood of p " for any set which contains an open set containing p .)
- (e) X is a *Hausdorff space* if the following is true: If $p \in X$, $q \in X$, and $p \neq q$, then p has a neighborhood U and q has a neighborhood V such that $U \cap V = \emptyset$.
- (f) X is *locally compact* if every point of X has a neighborhood whose closure is compact.

$\rightarrow \mathbb{R} \quad a \in \mathbb{R} \Rightarrow (a-1, a+1) = [a-1, a+1]$

2.7 Theorem Suppose U is open in a locally compact Hausdorff space X , $K \subset U$, and K is compact. Then there is an open set V with compact closure such that

$\therefore X \rightarrow \mathbb{R} \quad K \subset V \subset \bar{V} \subset U.$

This condition implies that X is locally compact Hausdorff space.

2.8 Definition Let f be a real (or extended-real) function on a topological space. If

$f^{-1}((\alpha, \infty)) = \{x: f(x) > \alpha\}$ (subinterval of $\mathbb{R}$)

is open for every real α , f is said to be *lower semicontinuous*. If

$\{x: f(x) < \alpha\}$

is open for every real α , f is said to be *upper semicontinuous*.

$\chi^{-1}((\alpha, \infty)) = \begin{cases} \emptyset & \alpha < 0 \\ (0, \infty) & 0 \leq \alpha \end{cases}$

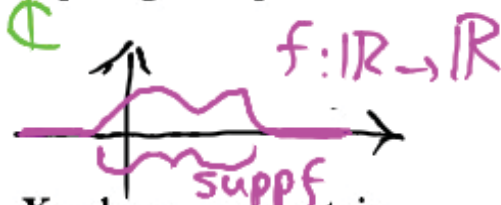
- (a) Characteristic functions of open sets are lower semicontinuous. *what about α ?*
 (b) Characteristic functions of closed sets are upper semicontinuous. *$\alpha=0$ & $\alpha=1$?*

The following property is an almost immediate consequence of the definitions:

- (c) The supremum of any collection of lower semicontinuous functions is lower semicontinuous. The infimum of any collection of upper semicontinuous functions is upper semicontinuous.

2.9 Definition The support of a complex function f on a topological space X is the closure of the set $\{x \in X : f(x) \neq 0\}$. *$: X \rightarrow \mathbb{C}$*

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



The collection of all continuous complex functions on X whose support is compact is denoted by $C_c(X)$.

Observe that $C_c(X)$ is a vector space. This is due to two facts:

- (a) The support of $f + g$ lies in the union of the support of f and the support of g , and any finite union of compact sets is compact.
 (b) The sum of two continuous complex functions is continuous, as are scalar multiples of continuous functions.

$$\text{supp}(f+g) \subseteq \text{supp } f \cup \text{supp } g$$

$x \notin \text{supp } f \cup \text{supp } g \Rightarrow f(x)=0 \text{ \& } g(x)=0 \Rightarrow (f+g)(x)=0 \Rightarrow x \notin \text{supp}(f+g)$

Corollary The range of any $f \in C_c(X)$ is a compact subset of the complex plane.

$$\forall x \in X, f(x) = 0 \Rightarrow f(x) \in \{0\} \\ f(x) \neq 0 \Rightarrow x \in K \Rightarrow f(x) \in f(K) \Rightarrow x \in f(K) \cup \{0\}$$

In fact, if K is the support of $f \in C_c(X)$, then $f(X) \subseteq f(K) \cup \{0\}$. If X is not compact, then $0 \in f(X)$, but 0 need not lie in $f(K)$, as is seen by easy examples.

why $f(X)$ is closed?

2.11 Notation In this chapter the following conventions will be used. The notation

$$K \prec f \tag{1}$$

will mean that K is a compact subset of X , that $f \in C_c(X)$, that $0 \leq f(x) \leq 1$ for all $x \in X$, and that $f(x) = 1$ for all $x \in K$. The notation

$$X \preceq f \tag{2}$$

will mean that V is open, that $f \in C_c(X)$, $0 \leq f \leq 1$, and that the support of f lies in V . The notation

$$f \ll X_V \quad K \prec f \prec V \quad (3)$$

will be used to indicate that both (1) and (2) hold.

2.12 Urysohn's Lemma Suppose X is a locally compact Hausdorff space, V is open in X , $K \subset V$, and K is compact. Then there exists an $f \in C_c(X)$, such that

$$K \prec f \prec V. \quad (1)$$

2.13 Theorem Suppose $V_1, \dots, V_n$ are open subsets of a locally compact Hausdorff space X , K is compact, and

$$K \subset V_1 \cup \dots \cup V_n.$$

Then there exist functions $h_i \prec V_i$ ($i = 1, \dots, n$) such that

$$h_1(x) + \dots + h_n(x) = 1 \quad (x \in K). \quad (1)$$

Because of (1), the collection $\{h_1, \dots, h_n\}$ is called a partition of unity on K , subordinate to the cover $\{V_1, \dots, V_n\}$.

The Riesz Representation Theorem

2.14 Theorem Let X be a locally compact Hausdorff space, and let Λ be a positive linear functional on $C_c(X)$. Then there exists a σ -algebra $\mathfrak{M}$ in X which contains all Borel sets in X , and there exists a unique positive measure μ on $\mathfrak{M}$ which represents Λ in the sense that

(a) $\Lambda f = \int_X f d\mu$ for every $f \in C_c(X)$,

and which has the following additional properties:

(b) $\mu(K) < \infty$ for every compact set $K \subset X$.

(c) For every $E \in \mathfrak{M}$, we have

$$\mu(E) = \inf \{ \mu(V) : E \subset V, V \text{ open} \}.$$

(d) The relation

$$\mu(E) = \sup \{ \mu(K) : K \subset E, K \text{ compact} \}$$

holds for every open set E , and for every $E \in \mathfrak{M}$ with $\mu(E) < \infty$.

$$\Lambda: C_c(X) \rightarrow \mathbb{C}$$

linearity $\Lambda(\alpha f + \beta g) = \alpha \Lambda f + \beta \Lambda g$

Positivity: $f \geq 0 \Rightarrow \Lambda f \geq 0$

$f, g \geq 0 \Rightarrow g - f \geq 0 \Rightarrow \Lambda(g - f) \geq 0 \Rightarrow \Lambda g - \Lambda f \geq 0 \Rightarrow \Lambda f \leq \Lambda g$

outer regularity

inner regularity

✓ (e) If $E \in \mathfrak{M}$, $A \subset E$, and $\mu(E) = 0$, then $A \in \mathfrak{M}$ and so $\mu(A) = 0$.

Let us begin by proving the uniqueness of μ . If μ satisfies (c) and (d), it is clear that μ is determined on $\mathfrak{M}$ by its values on compact sets. Hence it suffices to prove that $\mu_1(K) = \mu_2(K)$ for all K , whenever μ_1 and μ_2 are measures for which the theorem holds. So, fix K and $\epsilon > 0$. By (b) and (c), there exists a $V \supset K$ with $\mu_2(V) < \mu_2(K) + \epsilon$; by Urysohn's lemma, there exists an f so that $K \leq f \leq V$; hence

$$\begin{aligned} \mu_1(K) &= \int_X \chi_K d\mu_1 \leq \int_X f d\mu_1 = \int_X f d\mu_2 \leq \int_X \chi_V d\mu_2 = \mu_2(V) < \mu_2(K) + \epsilon. \end{aligned}$$

$\chi_K \leq f \leq \chi_V$

Thus $\mu_1(K) \leq \mu_2(K)$. If we interchange the roles of μ_1 and μ_2 , the opposite inequality is obtained, and the uniqueness of μ is proved.

Incidentally, the above computation shows that (a) forces (b).

Construction of μ and $\mathfrak{M}$

For every open set V in X , define

$$\mu(V) = \sup \{ \int_X f d\mu : f < V \}.$$

$$\text{Since } \{ \int_X f d\mu : f < V_1 \} \subset \{ \int_X f d\mu : f < V_2 \} \quad (1)$$

If $V_1 \subset V_2$, it is clear that (1) implies $\mu(V_1) \leq \mu(V_2)$. Hence

$$\mu(E) = \inf \{ \mu(V) : E \subset V, V \text{ open} \}, \quad (2)$$

if E is an open set, and it is consistent with (1) to define $\mu(E)$ by (2), for every $E \subset X$.

Note that although we have defined $\mu(E)$ for every $E \subset X$, the countable additivity of μ will be proved only on a certain σ -algebra $\mathfrak{M}$ in X .

Let $\mathfrak{M}_F$ be the class of all $E \subset X$ which satisfy two conditions: $\mu(E) < \infty$, and

$$\mu(E) = \sup \{ \mu(K) : K \subset E, K \text{ compact} \}. \quad (3)$$

Finally, let $\mathfrak{M}$ be the class of all $E \subset X$ such that $E \cap K \in \mathfrak{M}_F$ for every compact K .

Proof that μ and $\mathfrak{M}$ have the required properties

It is evident that μ is monotone, i.e., that $\mu(A) \leq \mu(B)$ if $A \subset B$ and that

$\mu(E) = 0$ implies $E \in \mathfrak{M}_F$ and $E \in \mathfrak{M}$. Thus (e) holds, and so does (c), by definition.

Since the proof of the other assertions is rather long, it will be convenient to divide it into several steps.

Observe that the positivity of Λ implies that Λ is *monotone*: $f \leq g$ implies $\Lambda f \leq \Lambda g$. This is clear, since $\Lambda g = \Lambda f + \Lambda(g - f)$ and $g - f \geq 0$. This monotonicity will be used in Steps II and X.

STEP I If $E_1, E_2, E_3, \dots$ are arbitrary subsets of X , then

$$\mu\left(\bigcup_{i=1}^{\infty} E_i\right) \leq \sum_{i=1}^{\infty} \mu(E_i). \quad (4)$$

PROOF We first show that

$$\mu(V_1 \cup V_2) \leq \mu(V_1) + \mu(V_2), \quad \text{if } V_1, V_2 \text{ are open.} \quad (5)$$

*supp g $\subseteq V_1 \cup V_2$
compact*

if V_1 and V_2 are open. Choose $g \prec V_1 \cup V_2$. By Theorem 2.13 there are functions h_1 and h_2 such that $h_i \prec V_i$ and $h_1(x) + h_2(x) = 1$ for all x in the support of g . Hence $h_i g \prec V_i$, $g = h_1 g + h_2 g$, and so

$$\Lambda g = \Lambda(h_1 g) + \Lambda(h_2 g) \leq \mu(V_1) + \mu(V_2). \quad (6)$$

$\mu(V) = \sup_{k \prec V} \Lambda k$

Since (6) holds for every $g \prec V_1 \cup V_2$, (5) follows.

If $\mu(E_i) = \infty$ for some i , then (4) is trivially true. Suppose therefore that $\mu(E_i) < \infty$ for every i . Choose $\epsilon > 0$. By (2) there are open sets $V_i \supset E_i$ such that

$$\mu(V_i) < \mu(E_i) + 2^{-i} \epsilon \quad (i = 1, 2, 3, \dots).$$

*$\mu(E) = \inf \mu(V)$
 $E \subseteq V$ open*

Put $V = \bigcup_{i=1}^{\infty} V_i$, and choose $f \prec V$. Since f has compact support, we see that $f \prec V_1 \cup \dots \cup V_n$ for some n . Applying induction to (5), we therefore obtain

$$\Lambda f \leq \mu(V_1 \cup \dots \cup V_n) \leq \mu(V_1) + \dots + \mu(V_n) \leq \sum_{i=1}^{\infty} \mu(E_i) + \epsilon.$$

$\sum \mu(E_i) + (\sum 2^{-i}) \epsilon$

Since this holds for every $f \prec V$, and since $\bigcup E_i \subseteq V$, it follows that

$$\mu\left(\bigcup_{i=1}^{\infty} E_i\right) \leq \mu(V) \leq \sum_{i=1}^{\infty} \mu(E_i) + \epsilon,$$

*upper bound
from $\{f \prec V\}$*

which proves (4), since ϵ was arbitrary. $\square$

Note. If $A \subseteq X$ & $B \subseteq X$ are compact, then so is $A \cup B$.

Let $\{G_\alpha\}$ be
an open cover



for $A \cup B$. So $A \subseteq \bigcup_{\alpha} G_\alpha$ & $B \subseteq \bigcup_{\alpha} G_\alpha$. Hence

$$\exists \alpha_1, \dots, \alpha_n; A \subseteq \bigcup_{i=1}^n G_{\alpha_i}$$

$$\exists \alpha_{n+1}, \dots, \alpha_m; B \subseteq \bigcup_{j=n+1}^m G_{\alpha_j}$$

$$\text{Therefore } A \cup B \subseteq \bigcup_{k=1}^m G_{\alpha_k}. \quad \square$$

STEP II *If K is compact, then $K \in \mathfrak{M}_F$ and*

$$\mu(K) = \inf \{ \Lambda f : K \prec f \}. \quad x \in K \Rightarrow f(x) = 1 \Rightarrow x \in V$$

This implies assertion (b) of the theorem.

PROOF If $K < f$ and $0 < \alpha < 1$, let $V_\alpha = \{x: f(x) > \alpha\}$. Then $K \in V_\alpha$, and $\alpha g \leq f$ whenever $g < V_\alpha$. Hence $\forall x: (g)(x) = \alpha g(x) \leq f(x) \Rightarrow x \in V_\alpha$

$$\mu(K) \leq \mu(V_\alpha) = \sup \{ \Lambda g : g < V_\alpha \} \leq \alpha^- \wedge \beta^- = \alpha, \quad \alpha \in V$$

Let $\alpha \rightarrow 1$, to conclude that

$$0 \leq f \leq 1 \Rightarrow \|f\|_p \leq M < \infty, \quad \mu(K) \leq \Lambda f \leq +\infty, \quad \lambda \downarrow 0. \quad (8)$$

Thus $\mu(K) < \infty$. Since K evidently satisfies (3), $K \in \mathfrak{M}_F$.

If $\epsilon > 0$, there exists $V \supset K$ with $\mu(V) < \mu(K) + \epsilon$. By Urysohn's lemma, $K \prec f \prec V$ for some f . Thus

$$\Lambda f \leq \mu(V) < \mu(K) + \epsilon,$$

which, combined with (8), gives (7).

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STEP III Every open set satisfies (3). Hence $\mathfrak{M}_F$ contains every open set V with $\mu(V) < \infty$. $N(V) = \sup \{ \mu(V) : V \in \mathfrak{M}_F, \mu(V) < \infty \}$

PROOF Let α be a real number such that $\alpha < \mu(V)$. There exists an $f < V$ with $\alpha < \Lambda f$. If W is any open set which contains the support K of f , then $f < W$, hence $\Lambda f \leq \mu(W)$. Thus $\alpha < \Lambda f \leq \mu(K)$. This exhibits a compact $K \subset V$ with $\alpha < \mu(K)$, so that (3) holds for V . ////

STEP IV Suppose $E = \bigcup_{i=1}^{\infty} E_i$, where $E_1, E_2, E_3, \dots$ are pairwise disjoint members of $\mathfrak{M}_F$. Then

$$\mu(E) = \sum_{i=1}^{\infty} \mu(E_i). \quad (9)$$

If, in addition, $\mu(E) < \infty$, then also $E \in \mathfrak{M}_F$.

PROOF We first show that

$$\mu(K_1 \cup K_2) = \mu(K_1) + \mu(K_2) \quad (10)$$

if K_1 and K_2 are disjoint compact sets. Choose $\epsilon > 0$. By Urysohn's lemma, there exists $f \in C(X)$ such that $f(x) = 1$ on K_1 , $f(x) = 0$ on K_2 , and

$0 \leq f \leq 1$. By Step II there exists g such that

$$K_1 \cup K_2 \prec g \quad \text{and} \quad \Lambda g < \mu(K_1 \cup K_2) + \epsilon.$$

Note that $K_1 \prec fg$ and $K_2 \prec (1-f)g$. Since Λ is linear, it follows from (8) that

$$\mu(K_1) + \mu(K_2) \leq \Lambda(fg) + \Lambda(g - fg) = \Lambda g < \mu(K_1 \cup K_2) + \epsilon.$$

Since ϵ was arbitrary, (10) follows now from Step I.

If $\mu(E) = \infty$, (9) follows from Step I. Assume therefore that $\mu(E) < \infty$, and choose $\epsilon > 0$. Since $E_i \in \mathfrak{M}_F$, there are compact sets $H_i \subset E_i$ with

$$\mu(H_i) > \mu(E_i) - 2^{-i}\epsilon \quad (i = 1, 2, 3, \dots). \quad (11)$$

Putting $K_n = H_1 \cup \dots \cup H_n$ and using induction on (10), we obtain

$$\mu(E) \geq \mu(K_n) = \sum_{i=1}^n \mu(H_i) > \sum_{i=1}^n \mu(E_i) - \sum_{i=1}^n 2^{-i}\epsilon > \sum_{i=1}^n \mu(E_i) - \epsilon. \quad (12)$$

Since (12) holds for every n and every $\epsilon > 0$, the left side of (9) is not smaller than the right side, and so (9) follows from Step I.

But if $\mu(E) < \infty$ and $\epsilon > 0$, (9) shows that

$$\mu(E) - \sum_{i=1}^N \mu(E_i) < \epsilon \quad \exists N, \forall n \geq N, \quad \mu(E) \leq \sum_{i=1}^n \mu(E_i) + \epsilon \quad (13)$$

for some N . By (12), it follows that $\mu(E) \leq \mu(K_N) + 2\epsilon$, and this shows that E satisfies (3); hence $E \in \mathfrak{M}_F$.

STEP V If $E \in \mathfrak{M}_F$ and $\epsilon > 0$, there is a compact K and an open V such that $K \subset E \subset V$ and $\mu(V - K) < \epsilon$.

PROOF Our definitions show that there exist $K \subset E$ and $V \supset E$ so that

$$\mu(V) - \frac{\epsilon}{2} < \mu(E) < \mu(K) + \frac{\epsilon}{2}.$$

Since $V - K$ is open, $V - K \in \mathfrak{M}_F$, by Step III. Hence Step IV implies that

$$\mu(K) + \mu(V - K) = \mu(V) < \mu(K) + \epsilon.$$

STEP VI If $A \in \mathfrak{M}_F$ and $B \in \mathfrak{M}_F$, then $A - B$, $A \cup B$, and $A \cap B$ belong to $\mathfrak{M}_F$.

PROOF If $\epsilon > 0$, Step V shows that there are sets K_i and V_i such that $K_1 \subset A \subset V_1$, $K_2 \subset B \subset V_2$, and $\mu(V_i - K_i) < \frac{\epsilon}{2}$, for $i = 1, 2$. Since

$$A - B \subset V_1 - K_2 \subset (V_1 - K_1) \cup (K_1 - V_2) \cup (V_2 - K_2), \quad (2)$$

Step I shows that

$$A - B \subset A \\ \mu(A - B) \leq \mu(A) < \infty$$

$$\begin{aligned} A \subset V_1, B^c \subset K_2^c &\Rightarrow A \cap B^c \subset V_1 \cap K_2^c \\ \mu(A - B) &\leq \frac{\epsilon}{2} + \mu(K_1 - V_2) + \frac{\epsilon}{2} \end{aligned}$$

(14)

Since $K_1 - V_2$ is a compact subset of $A - B$, (14) shows that $A - B$ satisfies (3), so that $A - B \in \mathfrak{M}_F$.

Since $A \cup B = (A - B) \cup B$, an application of Step IV shows that $A \cup B \in \mathfrak{M}_F$. Since $A \cap B = A - (A - B)$, we also have $A \cap B \in \mathfrak{M}_F$. ///

$$x \in A \cap K \Rightarrow x \in A; \& x \in K$$

STEP VII $\mathfrak{M}$ is a σ -algebra in X which contains all Borel sets.

PROOF Let K be an arbitrary compact set in X .

If $A \in \mathfrak{M}$, then $A^c \cap K = K - (A \cap K)$, so that $A^c \cap K$ is a difference of two members of $\mathfrak{M}_F$. Hence $A^c \cap K \in \mathfrak{M}_F$, and we conclude: $A \in \mathfrak{M}$ implies $A^c \in \mathfrak{M}$.

$$\begin{aligned} &\Rightarrow x \in A, \& x \in K \Rightarrow x \in A \cap K \\ &\Rightarrow x \in B \cup \dots \cup B_j \Rightarrow x \in B_n \\ &\Rightarrow x \in B \Rightarrow x \in B_n \end{aligned}$$

Next, suppose $A = \bigcup_{i=1}^{\infty} A_i$, where each $A_i \in \mathfrak{M}$. Put $B_1 = A_1 \cap K$, and

$$B_n = (A_n \cap K) - (B_1 \cup \dots \cup B_{n-1}) \quad (n = 2, 3, 4, \dots). \quad (15)$$

Then $\{B_n\}$ is a disjoint sequence of members of $\mathfrak{M}_F$, by Step VI, and $A \cap K = \bigcup_{i=1}^{\infty} B_n$. It follows from Step IV that $A \cap K \in \mathfrak{M}_F$. Hence $A \in \mathfrak{M}$.

Finally, if C is closed, then $C \cap K$ is compact, hence $C \cap K \in \mathfrak{M}_F$, so $C \in \mathfrak{M}$. In particular, $X \in \mathfrak{M}$.

We have thus proved that $\mathfrak{M}$ is a σ -algebra in X which contains all closed subsets of X . Hence $\mathfrak{M}$ contains all Borel sets in X . ///

STEP VIII $\mathfrak{M}_F$ consists of precisely those sets $E \in \mathfrak{M}$ for which $\mu(E) < \infty$.

$$\mathfrak{M}_F = \{E \subset X \mid \mu(E) < \infty \& \mu(E) = \sup \mu(K)\} = \{E \subset X \mid E \in \mathfrak{M}, \mu(E) < \infty\}$$

This implies assertion (d) of the theorem.

PROOF If $E \in \mathfrak{M}_F$, Steps II and VI imply that $E \cap K \in \mathfrak{M}_F$ for every compact K , hence $E \in \mathfrak{M}$. So $\mathfrak{M}_F \subset \{E \in \mathfrak{M} \mid \mu(E) < \infty\}$

Conversely, suppose $E \in \mathfrak{M}$ and $\mu(E) < \infty$, and choose $\epsilon > 0$. There is an open set $V \supset E$ with $\mu(V) < \infty$; by III and V, there is a compact $K \subset V$ with $\mu(V - K) < \epsilon$. Since $E \cap K \in \mathfrak{M}_F$, there is a compact set $H \subset E \cap K$ with

$$\begin{aligned} \mu(E \cap K) - \epsilon &< \mu(H) \\ \mu(E \cap K) &< \mu(H) + \epsilon \\ \mu(E) &= \inf \mu(V) \end{aligned}$$

Since $E \subset (E \cap K) \cup (V - K)$, it follows that

(3)

$$\mu(E) \leq \mu(E \cap K) + \mu(V - K) < \mu(H) + \epsilon,$$

which implies that $E \in \mathfrak{M}_F$.

$\mu(E) = \sup_{H \subseteq E} \mu(H)$ and so

$\mu(E) = \sum_{i=1}^{\infty} \mu(E_i)$

STEP IX μ is a measure on $\mathfrak{M}$.

PROOF The countable additivity of μ on $\mathfrak{M}$ follows immediately from Steps IV and VIII.

$E_1, E_2, \dots$ are disjoint in $\mathfrak{M} \Rightarrow \{E_i \in \mathfrak{M}_F \mid \mu(E_i) < \infty\}$
 $\mu(\bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^{\infty} \mu(E_i)$

STEP X For every $f \in C_c(X)$, $\Lambda f = \int_X f d\mu$.

This proves (a), and completes the theorem

$\int f d\mu = \int (u+iv) d\mu = \int u d\mu + i \int v d\mu = \Lambda u + i \Lambda v$
 linearity of $\Lambda(u+iv) = \Lambda(f)$

PROOF Clearly, it is enough to prove this for real f . Also, it is enough to prove the inequality

(4)

$$\Lambda f \leq \int_X f d\mu \quad (16)$$

for every real $f \in C_c(X)$. For once (16) is established, the linearity of Λ shows that

$$-\Lambda f = \Lambda(-f) \leq \int_X (-f) d\mu = - \int_X f d\mu$$

which, together with (16), shows that equality holds in (16).

Let K be the support of a real $f \in C_c(X)$, let $[a, b]$ be an interval which contains the range of f (note the Corollary to Theorem 2.10), choose $\epsilon > 0$, and choose y_i , for $i = 0, 1, \dots, n$, so that $y_i - y_{i-1} < \epsilon$ and

$\exists n; \frac{b-a}{n+1} < \epsilon$

$y_0 < a < y_1 < \dots < y_n = b$

(17)

Put

$E_i = \{x: y_{i-1} < f(x) \leq y_i\} \cap K \quad (i = 1, \dots, n)$

Since f is continuous, f is Borel measurable, and the sets E_i are therefore disjoint Borel sets whose union is K . There are open sets $V_i \supset E_i$ such that

$$\mu(V_i) < \mu(E_i) + \frac{\epsilon}{n} \quad (i = 1, \dots, n) \quad (19)$$

and such that $f(x) < y_i + \epsilon$ for all $x \in V_i$. By Theorem 2.13, there are functions $h_i < V_i$ such that $\sum h_i = 1$ on K . Hence $f = \sum h_i f$, and Step II shows that

$$\mu(K) \leq \Lambda(\sum h_i) = \sum \Lambda h_i.$$

Since $h_i f \leq (y_i + \epsilon)h_i$, and since $y_i - \epsilon < f(x)$ on E_i , we have

$$\begin{aligned} \Lambda f &= \sum_{i=1}^n \Lambda(h_i f) \leq \sum_{i=1}^n (y_i + \epsilon) \Lambda h_i \\ &= \sum_{i=1}^n (|a| + y_i + \epsilon) \Lambda h_i - |a| \sum_{i=1}^n \Lambda h_i \\ &\leq \sum_{i=1}^n (|a| + y_i + \epsilon) [\mu(E_i) + \epsilon/n] - |a| \mu(K) \\ &= \sum_{i=1}^n (y_i - \epsilon) \mu(E_i) + 2\epsilon \mu(K) + \frac{\epsilon}{n} \sum_{i=1}^n (|a| + y_i + \epsilon) \\ &\leq \int_X f d\mu + \epsilon [2\mu(K) + |a| + b + \epsilon]. \end{aligned}$$

Since ϵ was arbitrary, (16) is established, and the proof of the theorem is complete. ////

Regularity Properties of Borel Measures

2.15 Definition A measure μ defined on the σ -algebra of all Borel sets in a locally compact Hausdorff space X is called a *Borel measure on X* . If μ is positive, a Borel set $E \subset X$ is *outer regular* or *inner regular*, respectively, if E has property (c) or (d) of Theorem 2.14. If every Borel set in X is both outer and inner regular, μ is called *regular*.

$\Lambda: C_c(X) \rightarrow \mathbb{C}$ is a positive linear functional.

Let f be a ^{real} function in $C_c(X)$. Then $f = \frac{|f|+f}{2} - \frac{|f|-f}{2}$

So $\Lambda f = \underbrace{\Lambda\left(\frac{|f|+f}{2}\right)}_{\geq 0} - \underbrace{\Lambda\left(\frac{|f|-f}{2}\right)}_{\geq 0} \in \mathbb{R}$.

① $\sum_{i=1}^{\infty} a_i$ is $\left\{ \sum_{i=1}^n a_i \right\}_{n=1}^{\infty}$ & its value is $\lim_{n \rightarrow \infty} \sum_{i=1}^n a_i$.

A sequence is a function on $\mathbb{N}$.

② $V_1 - K_2 \subseteq (V_1 - K_1) \cup (K_1 - V_2) \cup (V_2 - K_2)$

$x \in V_1 - K_2 \Rightarrow \begin{cases} x \in V_1 \\ \& \\ x \notin K_2 \end{cases} \begin{cases} \text{or} \\ \text{or} \end{cases} \begin{cases} x \notin K_1 \Rightarrow x \in V_1 - K_1 \\ x \in K_1 \& x \notin V_2 \Rightarrow x \in K_1 - V_2 \\ x \in V_2 \end{cases} \Rightarrow x \in V_2 - K_2 \checkmark$

Describe the Lebesgue measure : ^{پیمائش}
on $\mathbb{R}^n$. ($n=1, 2, \dots$)

1392, 1, 10 in 10: 105

$$(3) \quad x \in E \Rightarrow \begin{cases} x \in E \cap K \Rightarrow x \in (E \cap K) \cup (V - K) & x \in K \\ x \in V - K \Rightarrow x \in (E \cap K) \cup (V - K) & x \notin K \end{cases}$$

$$(4) \quad \text{Let } \Lambda f \leq \int_X f d\mu \text{ for } f \geq 0$$

Let f be a real function. It follows from (1) that

$$\Lambda(-f) \leq \int_X -f d\mu. \text{ Hence } \Lambda f \geq \int_X f d\mu \quad (\dagger).$$

$\parallel$ $\parallel$
 $-\Lambda(f)$ $-\int_X f d\mu$

By (1), (†), we have $\Lambda f = \int_X f d\mu. \quad \square$

Banach Spaces

Def. A function $\|\cdot\|: X \rightarrow \mathbb{R}$ is called a semi-norm

if (i) $x=0 \Rightarrow \|x\|=0$ (ii) $\|\lambda x\|=|\lambda|\|x\|$ (iii) $\|x+y\| \leq \|x\| + \|y\|$.

If $\|x\|=0 \Rightarrow x=0$, then $\|\cdot\|$ is called a norm.

Exercise. $\forall x; \|x\| \geq 0$.

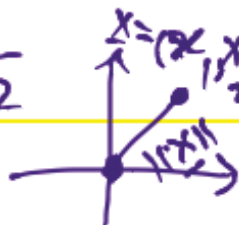
Solution. $\|0\| = \|x-x\| \leq \|x\| + \overbrace{\| -x \|}^{\|x\|} = 2\|x\|$

Then $(X, \|\cdot\|)$ is called a normed space. If $d(x,y) = \|x-y\|$, then d is a metric. If d is complete, then $(X, \|\cdot\|)$ is said to be a Banach space.

Examples. 1) $M_n(\mathbb{C})$, $\| [a_{ij}] \| = \max_m \max_{1 \leq i,j \leq n} |a_{ij}|$, $\| [a_{ij}] \| = \left(\sum_{i,j} |a_{ij}|^p \right)^{1/p}$

$$\| [a_{ij}] \| = \max_{1 \leq i,j \leq n} \left(\sum_{i,j} |a_{ij}|^2 \right)^{1/2}, \quad \| [a_{ij}] \| = \max_{1 \leq i \leq n} \left(\sum_{j=1}^n |a_{ij}|^2 \right)^{1/2}$$

$$2) \mathbb{C}^n, \|(z_1, \dots, z_n)\| = \left(\sum_{i=1}^n |z_i|^2 \right)^{1/2}$$



$$\begin{bmatrix} a_{11} & a_{1j} & a_{1n} \\ \vdots & \vdots & \vdots \\ a_{n1} & a_{nj} & a_{nn} \end{bmatrix}$$

the j-th column

$$3) \ell_0 \subset \ell_1 \subset \ell_2 \subset \ell_\infty = \{ (x_n) \mid \sup_n |x_n| < \infty \}$$

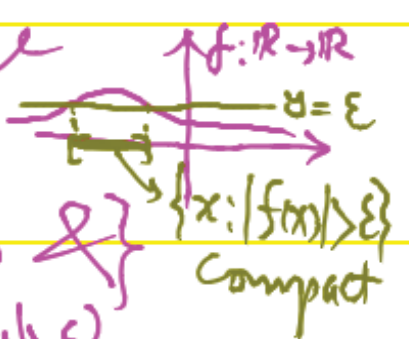
$$\{ (x_n) \mid \sum x_n = 0 \} \quad \{ (x_n) \mid \{x_n\} \text{ convergent} \}$$

$$\lambda \{x_n\} + \{y_n\} = \{\lambda x_n + y_n\}$$

$$4) C_c(X) \subseteq C_0(X) \subseteq l^\infty(X)$$

Let $f \in C_c(X)$. Then

locally compact Hausdorff space $\Rightarrow$ $C_c(X) = \{f: X \rightarrow \mathbb{C} \mid f \text{ is cts } \& \forall \epsilon \{x: |f(x)| \geq \epsilon\} \text{ is compact}\}$



$$\overline{f^{-1}([\epsilon, \infty))} = \{x: |f(x)| \geq \epsilon\} \subseteq \text{supp}(f)$$

closed $\leftarrow$ $f^{-1}([\epsilon, \infty))$ is closed $\rightarrow$ Compact

$\{x: |f(x)| \geq \epsilon\}$ is compact for all $\epsilon > 0$.

$$(\lambda f + g)(x) = \lambda f(x) + g(x)$$

$l^\infty(X)$ = the space of all bd functions
 $\|f\| = \sup_{x \in X} |f(x)|$

Note. Example 3 is a special case of Example 4.

$\mathbb{N}$ is a locally compact Hausdorff space when we consider the relative Euclidean top from $\mathbb{R}$.

$$C_c(\mathbb{N}) = \{f: \mathbb{N} \rightarrow \mathbb{C} \mid \text{supp } f \text{ is compact}\}$$



$$= C_{00}$$

$\mathbb{N}$
 supp f is finite

$$\tau_{\mathbb{N}} = \{ \mathbb{N} \cap G \mid G \text{ is open in } \mathbb{R} \}$$

$$C_0(\mathbb{N}) = C_0$$

$$l^\infty(\mathbb{N}) = l^\infty$$

say $\{n_1, \dots, n_k\}$
 Hence $\forall n \geq n_k: f(n) = 0$
 $\{n\} = \mathbb{N} \cap (n - \frac{1}{2}, n + \frac{1}{2})$ is open, so the top on $\mathbb{N}$ is the discrete top.

Theorem. Let X & Y be normed spaces. Then $(X \times Y, \|(x, y)\| = \max\{\|x\|, \|y\|\})$ is Ban iff so are X, Y .

Proof ($\Leftarrow$) Let X, Y be Ban. Let $\{(x_n, y_n)\}$ be a Cauchy sequence in $X \times Y$. So $\forall \epsilon \exists N \forall m, n \geq N: \|(x_n, y_n) - (x_m, y_m)\| < \epsilon$

$$\frac{\|x_n - x_m\|}{\|y_n - y_m\|} \leq \max\{\|x_n - x_m\|, \|y_n - y_m\|\}$$

Hence both $\{x_n\}$ & $\{y_n\}$ are Cauchy sequences. Since X, Y are Ban, $\exists x \in X \exists y \in Y; x_n \rightarrow x$ & $y_n \rightarrow y$.

Let $\epsilon > 0$ be given. $\exists N_1 \forall n \geq N_1: \|x_n - x\| < \frac{\epsilon}{2}$

$$\exists N_2 \forall n \geq N_2: \|y_n - y\| < \frac{\epsilon}{2} \quad \leftarrow \max\left\{\frac{\epsilon}{2}, \frac{\epsilon}{2}\right\}$$

$$\forall n \geq 2N_1 + 7N_2; \|x_n - x\| < \frac{\epsilon}{2} \text{ \& \& } \|y_n - y\| < \frac{\epsilon}{2}$$

Hence

$$\|(x_n, y_n) - (x, y)\| = \max\{\|x_n - x\|, \|y_n - y\|\} < \epsilon$$

$(x_n, y_n) \rightarrow (x, y)$

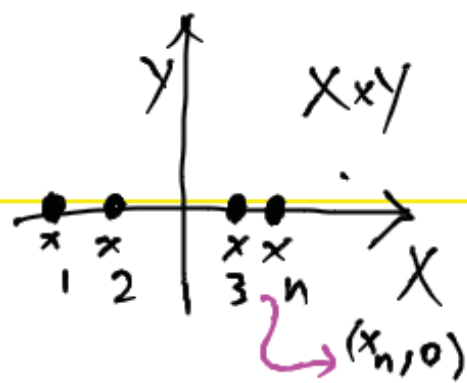
So $X \times Y$ is Ban. //

Conversely, let $X \times Y$ be a Banach space. We shall show that X is Banach.

$$\begin{matrix} a \leq b \\ c \leq d \end{matrix} \Rightarrow \max\{a, c\} \leq \max\{b, d\}$$

Let $\{x_n\}$ be a Cauchy seq. So

$$\forall \varepsilon \exists N \forall m, n \geq N; \underbrace{\|x_n - x_m\|}_{\|(x_n, 0) - (x_m, 0)\|} < \varepsilon$$



Hence $\{(x_n, 0)\}$ is Cauchy in $X \times Y$. Since $X \times Y$ is complete,

$\exists (x_0, y_0) \in X \times Y; (x_n, 0) \rightarrow (x_0, y_0)$. So

$$\forall \varepsilon \exists N \forall n \geq N; \underbrace{\|(x_n, 0) - (x_0, y_0)\|}_{\|y_0\|, \|x_n - x_0\|} < \varepsilon$$

Therefore, $x_n \rightarrow x_0$.

Note: We indeed showed that $\forall \varepsilon > 0; \|y_0\| < \varepsilon$
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If $\|y_0\| \neq 0$, then choose $\varepsilon = \frac{\|y_0\|}{2} > 0$ to get

$$\cancel{\|y_0\|} < \frac{\cancel{\|y_0\|}}{2}$$

$$1 < \frac{1}{2} \times \cdot$$

Thus $\|y_0\| = 0$, so $y_0 = 0$.

Def. A mapping $T: X \rightarrow Y$ is called linear if

$$T(\lambda x + y) = \lambda Tx + Ty.$$

It is called bounded if $\|T\| = \sup_{\|x\| \leq 1} \|Tx\| < \infty$

Exercise. → left to the students

$$1) \|T\| = \sup_{x \neq 0} \frac{\|Tx\|}{\|x\|} = \sup_{\|x\|=1} \|Tx\|$$

$$\alpha \left\{ \frac{\|Tx\|}{\|x\|} = \left\| \frac{1}{\|x\|} Tx \right\| = \left\| T\left(\frac{x}{\|x\|}\right) \right\| \leq \|T\| \right.$$

$\nearrow$ scalar $\nearrow$ T is linear

So $\|T\|$ is an upper bound for $\left\{ \frac{\|Tx\|}{\|x\|} : x \neq 0 \right\}$. Hence $\alpha \leq \|T\|$.

$\left\| \frac{x}{\|x\|} \right\| = \frac{\|x\|}{\|x\|} = 1 \leq 1$
 $\|T\| = \sup_{\|y\| \leq 1} \|Ty\|$

Let $\|x\| \leq 1$. Then $\|Tx\| = \frac{\|Tx\|}{1} \leq \frac{\|Tx\|}{\|x\|} \leq \alpha$

Hence $\sup_{\|x\| \leq 1} \|Tx\| \leq \alpha$, so $\|T\| \leq \alpha$. $\square$

$$2) \sup_{\|x\| < 1} \|Tx\| = \|T\| = \sup_{\|x\| \leq 1} \|Tx\|$$

$A \subseteq B \Rightarrow \sup A \leq \sup B$

Let $x \in A$. Then $x \in B$. So

$\alpha \leq \sup B$
 $\therefore \sup A \leq \sup B$

$$\sup_{\|x\| < 1} \{ \|Tx\| : \|x\| < 1 \} \subseteq \{ \|Tx\| : \|x\| \leq 1 \} = \|T\|$$

$\nearrow$ sup $\nearrow$

Suppose $\|x\| \leq 1$. Let $0 < t < 1$. Then $\|tx\| = t\|x\| < 1$.

So $\|T(tx)\| \leq \beta$. Hence $t\|Tx\| \leq \beta$. Tending t to 1 we get $\|Tx\| \leq \beta$. Therefore $\sup_{\|x\| \leq 1} \|Tx\| \leq \beta$.

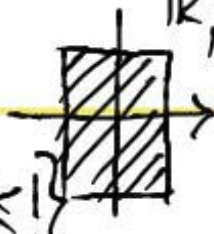
3) Always $\|Tx\| \leq \|T\| \|x\|$ since $\frac{\|Tx\|}{\|x\|} \leq \|T\|$ $\square$

Note. Boundedness in $\left\{ \begin{array}{l} \text{Calculus} \\ \text{operator Theory} \end{array} \right.$

are different. I mean that if we have a linear map $T: X \rightarrow Y$ & $\sup_{x \in X} \|Tx\| < \infty$, then

$\exists x_0; Tx_0 \neq 0$. Then $\forall n \in \mathbb{N}; \|T(nx_0)\| \leq M$. So.

$$\forall n \in \mathbb{N}; n \leq \frac{M}{\|Tx_0\|}.$$

Example. $\mathbb{R}^2, \|(x,y)\| = \sqrt{x^2+y^2}$  $B_1(0) = \{(x,y) | \sqrt{x^2+y^2} \leq 1\}$ $\mathbb{R}^2, \|(x,y)\| = \max\{|x|, |y|\}$  $B_1(0) = \{(x,y) | \max\{|x|, |y|\} \leq 1\}$

Th. T_x is bd iff T is cts at each point of X .
~~linear~~

Proof. ($\Rightarrow$) Let $x_0 \in X$. Let $\varepsilon > 0$. Choose $\delta = \frac{\varepsilon}{\|T\|}$. Then

$$\forall x; \|x - x_0\| < \delta = \frac{\varepsilon}{\|T\|} \Rightarrow \|Tx - Tx_0\| = \|T(x - x_0)\| \leq \|T\| \|x - x_0\| < \|T\| \frac{\varepsilon}{\|T\|} = \varepsilon$$

($\Leftarrow$) T is cts on X . So T is cts at origin 0 .

By the definition of continuity,

$$\exists \delta \forall x; \|x - 0\| < \delta \Rightarrow \|Tx - T0\| < 1$$

$$\begin{aligned} T0 &= T(0+0) \\ &= T0 + T0 \\ 0 &= T0 \end{aligned}$$

Let $\|x\| \leq 1$. Then

$$\frac{\delta}{2} \|x\| \leq \frac{\delta}{2} \Rightarrow \left\| \frac{\delta x}{2} \right\| \leq \frac{\delta}{2} < \delta \Rightarrow \left\| T\left(\frac{\delta x}{2}\right) \right\| < 1 \Rightarrow \|Tx\| < \frac{2}{\delta}$$

$$\therefore \|T\| = \sup_{\|x\| \leq 1} \|Tx\| < \frac{2}{\delta} < \infty. \square$$

Exercise T is cts at x_0 iff T is cts at 0 .

Exercise. $\|T\| = \sup_{\|x\| \leq 1} \|Tx\|$ is really a norm.

Hint: $\|x - x_0\| = \|(x - x_0) - 0\|$

Counterexample. There is a metric space whose norm cannot be induced by a norm.

Let S be the set of all complex sequences.

Put $d(\{x_n\}, \{y_n\}) = \sum \frac{|x_n - y_n|}{2^n(1 + |x_n - y_n|)} < \infty$, since

$\frac{|x_n - y_n|}{2^n(1 + |x_n - y_n|)} \leq \left(\frac{1}{2}\right)^n$. There is no norm $\|\cdot\|$ such that

$d(x, y) = \|x - y\|$. Otherwise, $\|\{x_n\}\| = d(\{x_n\}, 0) = \sum \frac{|x_n|}{2^n(1 + |x_n|)}$

Put $x_n = (1, 0, 0, \dots)$. Then $2\|\{x_n\}\| \neq \|\{2x_n\}\|$.

$$\frac{1}{2} = \frac{2}{2(1+1)} \quad \frac{2}{2(1+2)} = \frac{1}{3}$$

Def. A linear mapping $f: X \rightarrow \mathbb{C}$ is called a linear functional

We denote by X^* the space of all bounded linear
functionals on a normed space $(X, \|\cdot\|)$.
↳ the dual of X
(b.d.)

Th. A linear function $f: X \rightarrow \mathbb{C}$ is b.d. iff $\text{Ker } f$ is closed
 $\{x \in X : f(x) = 0\}$

Proof ($\Rightarrow$) If f is cts, then $\text{Ker } f = \overline{f^{-1}(\{0\})}$ is closed.

($\Leftarrow$) We may assume $\text{that } f \neq 0$. Let $\sup_{\|x\|=1} |f(x)| = \infty$. $\forall n \exists x_n$; $\|x_n\|=1$; $|f(x_n)| > n$. So

$\forall n \exists x_n$; $\|x_n\|=1$; $|f(x_n)| > n$. Since $f \neq 0$, there is $e \in X$; $f(e) \neq 0$

If $e_0 = \frac{e}{f(e)}$, then $f(e_0) = \frac{f(e)}{f(e)} = 1$. Put $y_n = e_0 - \frac{x_n}{f(x_n)}$.

Then $f(y_n) = f(e_0) - \frac{f(x_n)}{f(x_n)} = 1 - 1 = 0$. So $y_n \in \text{Ker}(f)$ and

$\|y_n - e_0\| = \left\| \frac{x_n}{f(x_n)} \right\| = \frac{1}{|f(x_n)|} < \frac{1}{n}$. Hence $y_n \rightarrow e_0$. So $e_0 \in \overline{\text{Ker } f}$

$= \text{Ker } f$, since $\text{Ker } f$ is closed. Therefore $f(e_0) = 0 \neq 1$. $\times$

Hence $\sup_{\|x\|=1} |f(x)| < \infty$. Thus f is bd. $\square$ N.P.A.P. (Not Proved)

Exercise 1) $x \in \bar{A} \Leftrightarrow \exists \{x_n\}$ in A such that $x_n \rightarrow x$.

2) If (X, d) is a complete metric space & $Y \subseteq X$, then (X, d) is complete iff Y is closed.

(Not Proved)

Th. $(X, \|\cdot\|)$ is Ban iff $\forall \{x_n\}$ in X ; $\sum_{n=1}^{\infty} \|x_n\| < \infty \Rightarrow \sum_{n=1}^{\infty} x_n$ Converges in X

$$0 \leq \|x_n\| - x_n \leq 2\|x_n\|$$

$$\sum \|x_n\| < \infty \Rightarrow \sum (\|x_n\| - x_n) < \infty$$

$$\Rightarrow \sum (\|x_n\| - (\|x_n\| - x_n)) < \infty$$

$$\sum x_n < \infty$$

Every absolutely convergent series is convergent.

Proof. ($\Rightarrow$) Let X be Ban. Let $\sum_{n=1}^{\infty} \|x_n\| < \infty$. Assume that $s_n = \sum_{k=1}^n \|x_k\|$ & $t_n = \sum_{k=1}^n x_k$. We shall show that $\{t_n\}$ is convergent.

$$\|t_n - t_m\| = \left\| \sum_{k=m+1}^n x_k \right\| \leq \sum_{k=m+1}^n \|x_k\| = |s_n - s_m| \quad n \geq m$$

$\{s_n\}$ converges to ∞
 $\{x_n\}$ is convergent
 Convergence of $\{x_n\}$ is...

Since $\sum_{n=1}^{\infty} \|x_n\|$ is conv., $\{s_n\}$ is conv. So $\{s_n\}$ is Cauchy. Hence

$$\forall \epsilon \exists N \forall n \geq m \geq N, |s_n - s_m| < \epsilon$$

So $\{t_n\}$ is Cauchy. Since $(X, \|\cdot\|)$ is Ban, $\exists x$; $t_n \rightarrow x$.
 $\sum_{n=1}^{\infty} x_n$ is conv.

Conversely, let $\{x_n\}$ be a Cauchy seq. in X .

$$\exists n_1 \forall m, n \geq n_1; \|x_n - x_m\| < \frac{1}{2^1}$$

in Particular; $\|x_{n_1} - x_{n_2}\| < \frac{1}{2^1} \Rightarrow \|x_{n_1} - x_{n_2}\| < \frac{1}{2^1}$

$$\exists \frac{n_1}{2} \forall n \geq \frac{n_1}{2}; \|x_n - x_{\frac{n_1}{2}}\| < \frac{1}{2^2}$$



there is a subseq. $\{x_{n_k}\}$ of $\{x_n\}$ such that

$$\|x_{n_{k+1}} - x_{n_k}\| < \frac{1}{2^k} \quad \left(\sum_{k=1}^{\infty} \left(\frac{1}{2}\right)^k < \infty \right)$$

By the Comparison test, $\sum_{k=1}^{\infty} \|x_{n_{k+1}} - x_{n_k}\| < \infty$. By our assumption, $\sum_{k=1}^{\infty} (x_{n_{k+1}} - x_{n_k})$ converges to, say, x . So

$$\lim_{N \rightarrow \infty} \sum_{k=1}^{N-1} (x_{n_{k+1}} - x_{n_k}) = x. \text{ Hence } \{x_{n_N}\}_{N=1}^{\infty} \text{ is convergent to } \underbrace{x + x_0}_{x_0}$$

Now we shall show that $\{x_n\}$ is convergent to x_0 :

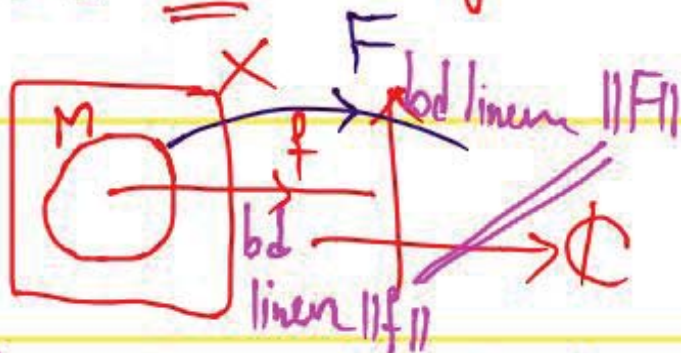
Let $\varepsilon > 0$. $\exists N_0 \forall n \geq N_0; \|x_n - x_0\| < \frac{\varepsilon}{2}$.

$\exists M \geq N_0 \forall m, n \geq M; \|x_n - x_m\| < \frac{\varepsilon}{2}$ (since $\{x_n\}$ is Cauchy)

$$\forall n \geq N; \|x_n - x_0\| \leq \|x_n - x_{N_0}\| + \|x_{N_0} - x_0\| < \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon.$$

So X is Ban. $\square$

Project 2 (Hahn-Banach Theorem). Let M be a subspace of $(X, \|\cdot\|)$ & f be a bd linear functional on M . Then f can be extended to a bd linear functional F to X such that $\|F\| = \|f\|$.

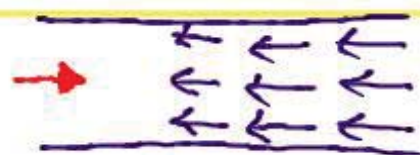


Remark. In linear algebra, $\exists$ a subspace N such that $X = M \oplus N$. Put $F(x) = \begin{cases} f(x) & x \in M \\ 0 & x \in N \end{cases}$. Then F is a linear functional on X , but it may be unbounded.

Note. The space $\langle S \rangle$ generated by $S \subseteq X$ is $\bigcap M$. In particular, $\langle x_0 \rangle = \mathbb{C}x_0 = \{\lambda x_0 \mid \lambda \in \mathbb{C}\}$

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Since $\mathbb{C}x_0 = \bigcap_{x_0 \in M} M$.

$$\{x_1, x_2, \dots, x_n, x_0\}$$



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Corollary If $x_0 \in X$, then $\exists f \in X^*$; $f(x_0) = \|x_0\|$ & $\|f\| = 1$

Proof. $\{g: \langle x_0 \rangle \rightarrow \mathbb{C} \mid g(\lambda x_0) = \lambda \|x_0\| \}$ is a linear functional

$g(\lambda x_0 + \mu x_0) = g((\lambda + \mu)x_0) = (\lambda + \mu)\|x_0\|$
 $= \lambda\|x_0\| + \mu\|x_0\| = g(\lambda x_0) + g(\mu x_0)$

$\|g\| = \sup_{\lambda \neq 0} \frac{|g(\lambda x_0)|}{\|\lambda x_0\|} = \sup_{\lambda \neq 0} \frac{|\lambda| \|x_0\|}{|\lambda| \|x_0\|} = 1$

By the Hahn-Banach theorem, $\exists f \in X^*$; $f|_{\langle x_0 \rangle} = g$ & $\|f\| = \|g\| = 1$

$$f(x_0) = g(x_0) = g(1 \cdot x_0) = 1 \|x_0\| = \|x_0\|. \quad \square$$

$$x_0 \in \langle x_0 \rangle$$

Corollary. Let M be a subspace of $(X, \|\cdot\|)$. Then $x_0 \notin \bar{M}$ if and only if $\exists f \in X^*$; $f(x_0) = 1$ & $f|_M = 0$

Proof ($\Leftarrow$) Let $\exists f \in X^*$; $f(x_0) = 1$ & $f|_M = 0$. We show that $x_0 \notin \bar{M}$. In contrary, assume $x_0 \in \bar{M}$. So $\exists \{x_n\}$ in M such that $x_n \rightarrow x_0$. Since f is continuous, $\underbrace{f(x_n)}_0 \rightarrow \underbrace{f(x_0)}_1$. Hence $x_0 \notin \bar{M}$.

($\Rightarrow$) Let $x_0 \notin \bar{M} \Rightarrow \exists m: N(x) \cap M = \emptyset$



$$\forall x \in M; \|x - x_0\| \geq r \quad (\clubsuit)$$

$$\text{Let } Y = \langle M \cup \{x_0\} \rangle = \{x + \lambda x_0 \mid \lambda \in \mathbb{C}, x \in M\}.$$

$$\alpha(x + \lambda x_0) + (\beta + \mu)x_0 = \alpha x + \beta x_0 + \alpha \lambda x_0 + \mu x_0 = \alpha x + (\beta + \alpha \lambda + \mu)x_0$$

Define $g: Y \rightarrow \mathbb{C}$. g is linear. g is bounded on Y :

$$g(x + \lambda x_0) = \lambda$$

$$r |g(x + \lambda x_0)| = |\lambda| r \leq \| \underbrace{\frac{x}{\lambda} - x_0}_{\in M} \| \leq \|x + \lambda x_0\|$$

$$\|g\| = \sup \frac{|g(x + \lambda x_0)|}{\|x + \lambda x_0\|} \leq \frac{1}{r}$$

By the Hahn-Banach theorem, $\exists f: X \xrightarrow{\text{linear}} \mathbb{C}$; $f|_Y = g$

$\|f\| = \|g\|$. Hence $f(x_0) = g(x_0) = g(0 + 1x_0) = 1$, $f|_M = g|_M = 0$. $\square$

Project. (Banach Open Mapping Theorem)

$$g(x) = g(x + 0x_0)$$

If $T: X \rightarrow Y$ is a b.l. linear mapping from a Ban space X **Onto** a **Banach space** Y , then T is open, i.e., if $U \subseteq X$ is open, then TU is also open in Y . $[9.1, 9.11 \& 9.12]$

Corollary. If $T: X \rightarrow Y$ is bd, 1-1 & onto linear mapping between Banach spaces, then T^{-1} is also a bd linear mapping.

Proof. Since T is 1-1 & surjective, so T^{-1} exists as a mapping. In addition, T^{-1} is linear:

$$T^{-1}(\alpha y_1 + y_2) = \alpha T^{-1}(y_1) + T^{-1}(y_2)$$

$$T(T^{-1}(\alpha y_1 + y_2)) = T(\alpha T^{-1}(y_1) + T^{-1}(y_2)) = \alpha T(T^{-1}(y_1)) + T(T^{-1}(y_2))$$

$$\alpha y_1 + y_2 = \alpha y_1 + y_2$$

(?) $\begin{cases} y = y \\ y = y \end{cases} \Rightarrow y + y = y + y$ ✓
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T maps $\{x: \|x\| < 1\} \stackrel{=}{=} A$ onto an open set in Y , by the open mapping theorem.
 Since TA is open & $0 = T0 \in TA$,

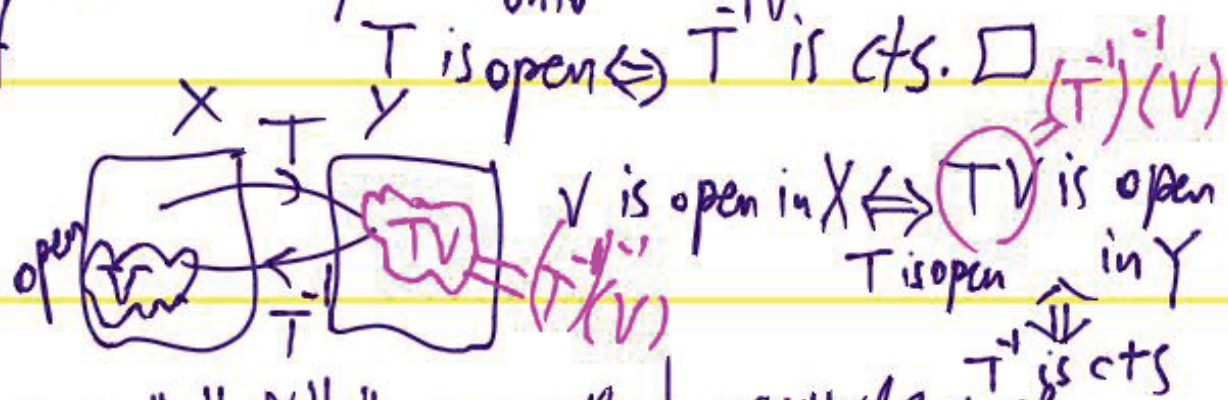
$$\exists r, N(0) \subseteq TA \text{ for } \{Tx: x \in A\}$$



$$\forall y; (\|y\| < r \Rightarrow y \in TA \Rightarrow \exists x \in A; y = Tx) \quad \begin{matrix} \|Tx\| \leq ? \|y\| \\ \|Tx\| \leq \|y\| \end{matrix}$$

Let $y \in Y$. Then $\|\frac{ry}{2\|y\|}\| = \frac{r}{2} < r$. So $\exists x \in X; \|x\| < 1$ & $Tx = \frac{ry}{2\|y\|}$.
 Therefore $\|x\| = \|\frac{rT^{-1}y}{2\|y\|}\| < 1 \Rightarrow \|T^{-1}y\| \leq \frac{2}{r}\|y\|$. So T^{-1} is bd. $\square$

(Second Proof not for exam) T is 1-1 & cts. By a theorem in top, Topological proof



Def. Two norms $\|\cdot\|_1$ & $\|\cdot\|_2$ are called equivalent if

$$\exists \alpha, \beta > 0; \begin{cases} \|x\|_1 \leq \alpha \|x\|_2 \\ \|x\|_2 \leq \beta \|x\|_1 \end{cases} \quad \text{or} \quad \alpha \|x\|_1 \leq \|x\|_2 \leq \beta \|x\|_1$$

Note. 1) $I: (X, \|\cdot\|_1) \rightarrow (X, \|\cdot\|_2)$ is bd iff $\exists \beta; \|I(x)\|_2 \leq \beta \|x\|_1$

2) $I: (X, \|\cdot\|_1) \rightarrow (X, \|\cdot\|_2)$ is bi-continuous iff $\begin{cases} \|x\|_2 \leq \beta \|x\|_1 \\ \|x\|_1 \leq \alpha \|x\|_2 \end{cases}$

$I^{-1} = I$ both I & I^{-1} are cts

for some α, β and all $x \in X$.

Theorem. If $\|\cdot\|_1, \|\cdot\|_2$ are equivalent norms & $(X, \|\cdot\|_1)$ is Ban, then so is $(X, \|\cdot\|_2)$.

Proof. Let $\exists \alpha, \beta; \alpha\|x\|_1 \leq \|x\|_2 \leq \beta\|x\|_1, \forall x \in X$.

Let $\{x_n\}$ be a Cauchy seq in $(X, \|\cdot\|_2)$.

$$\left. \begin{array}{l} \|x_n - x_m\|_1 \leq \frac{1}{\alpha} \|x_n - x_m\|_2 \\ \forall \varepsilon \exists N \forall n, m \geq N; \end{array} \right\} \Rightarrow \|x_n - x_m\|_1 < \varepsilon$$

Hence $\{x_n\}$ is Cauchy in $(X, \|\cdot\|_1)$. So that $\exists x \in X, x_n \xrightarrow{\|\cdot\|_1} x$
 complete norm

Therefore

$$\left. \begin{array}{l} \forall \varepsilon \exists N \forall n \geq N; \|x_n - x\|_1 < \frac{\varepsilon}{\beta} \\ \|x_n - x\|_2 \leq \beta \|x_n - x\|_1 \end{array} \right\} \Rightarrow \|x_n - x\|_2 < \varepsilon$$

Thus $x_n \xrightarrow{\|\cdot\|_2} x. \square$

Note. On $M_n(\mathbb{C})$, all norm are equivalent:

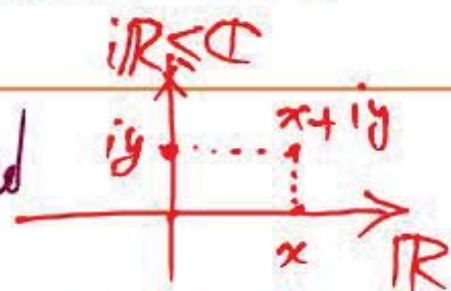
$$\|[a_{ij}]\|_m = \max_{ij} |a_{ij}|, \quad \|[a_{ij}]\|_1 = \sum_{j=1}^n |a_{ij}|$$

$$\|[a_{ij}]\|_m = \max_{ij} |a_{ij}| \leq \sum_{j=1}^n |a_{ij}| = \|[a_{ij}]\|_1, \quad \|[a_{ij}]\|_1 = \sum_{i=1}^n |a_{ij}| \leq n^2 \max_{ij} |a_{ij}| = n^2 \|[a_{ij}]\|_m$$

Let X, Y be linear spaces. Then $X \times Y$ together with $\alpha(x, y) + (x', y') = (\alpha x + x', \alpha y + y')$ in a vector spaces. Then $X \times Y$ is called the external direct sum of X & Y , denoted by $X \oplus Y$.

If X & Y are subspaces of a linear space Z such that $X \cap Y = \{0\}$, then $X + Y = \{x + y \mid x \in X, y \in Y\}$ is a subspace of Z , denoted by $X \oplus Y$, which is called the internal direct sum of X & Y .

Each vector^z of $X + Y$ is uniquely represented as $z = x + y$, where $x \in X, y \in Y$, since



if $z = x + y = x' + y'$, then $\underbrace{x - x'}_{\in X} = \underbrace{y' - y}_{\in Y}$. So $x - x' = y' - y = 0$. Hence $x = x'$ & $y = y'$. $\mathbb{C} = \underbrace{\mathbb{R}}_X \oplus \underbrace{i\mathbb{R}}_Y$

1) If $X \oplus Y$ is internal, then $\begin{cases} \phi: X \oplus Y \rightarrow X \times Y \\ x + y \mapsto (x, y) \end{cases}$ is a vector space isomorphism

2) If X, Y are linear spaces, put $X_1 = \{(x, 0) \mid x \in X\}$, $Y_1 = \{(0, y) \mid y \in Y\}$, then the external direct sum $X \times Y$ is indeed $X_1 \oplus Y_1$ since

$$(x, y) = \underbrace{(x, 0)}_{\in X_1} + \underbrace{(0, y)}_{\in Y_1}, \quad X_1 \cap Y_1 = \{0\}$$

Def If X, Y are normed spaces then $X \oplus Y$ is a

def. If X, Y are normed spaces, then $X \oplus Y$ is a normed space under $\|(x, y)\|_p = (\|x\|^p + \|y\|^p)^{1/p}$

$$\|(x, y)\|_\infty = \max\{\|x\|, \|y\|\}$$

Th. If X, Y are Banach, then so is $(X \oplus Y, \|\cdot\|_\infty)$.

Proof. It is already proved.

If X is a normed space, $X^{**} = (X^*)^*$ is called the second dual of X .

$\hookrightarrow X^*$ is a normed space

$$\begin{cases} (f+g)(x) = f(x) + g(x) \\ (\alpha f)(x) = \alpha f(x) \\ \|f\| = \sup_{x \neq 0} \frac{|f(x)|}{\|x\|} \end{cases}$$

Define $\hat{\cdot} : X \rightarrow X^{**}$
 $x \mapsto \hat{x}, \begin{cases} \hat{x} : X^* \rightarrow \mathbb{C} \\ \hat{x}(f) = f(x) \end{cases}$

Then $\hat{\cdot}$ is a linear map:

$$\widehat{x+y} = \hat{x} + \hat{y} \text{ since } \widehat{x+y}(f) = f(x+y) = f(x) + f(y) = \hat{x}(f) + \hat{y}(f) = (\hat{x} + \hat{y})(f)$$

Similarly, $\widehat{\alpha x} = \alpha \hat{x}$.

We show that $\|\hat{x}\| = \|x\|$:

$$|\hat{x}(f)| = |f(x)| \leq \|f\| \|x\| \Rightarrow \|\hat{x}\| = \sup_{f \neq 0} \frac{|\hat{x}(f)|}{\|f\|} \leq \|x\|.$$

$\hookrightarrow$ $\forall x \in X$. By the Hahn-Banach theorem $\exists f \in X^*$, $\|f\|=1, f(x)=\|x\|$.
 $\|x\| = f(x) = |f(x)| = |\hat{x}(f)| \leq \|\hat{x}\| \|f\| = \|\hat{x}\|$
 $\forall x \geq 0$ $\square$

Def. If $f^{\wedge}: X \hookrightarrow X^{**}$ is surjective (onto), then X is called reflexive.

Exercise 1. What can we say about X^* when X is reflexive and conversely? (Kreyszig's book)

Exercise 2 $\mathbb{C}^n$ is reflexive. (Hint: If $\{e_1, \dots, e_n\}$ is a ^{Standard} basis for $\mathbb{C}^n$, then $f_i: \mathbb{C}^n \rightarrow \mathbb{C}$

$f_j(e_i) = \delta_{ij} = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$ is in $(\mathbb{C}^n)^*$ and $\{f_1, \dots, f_n\}$ is a basis for $(\mathbb{C}^n)^*$.

Hilbert space $\equiv$

$$\mathbb{R}^n \begin{matrix} \nearrow x = (x_1, \dots, x_n) \\ \searrow y = (y_1, \dots, y_n) \\ \text{dot product} \\ x \cdot y = \sum_{i=1}^n x_i y_i \end{matrix}$$

Def. Let H be a vector space. Then $\langle \cdot, \cdot \rangle: H \times H \rightarrow \mathbb{C}$ is called an inner product space if

1) $\langle x, x \rangle \geq 0$ & $\langle x, x \rangle = 0 \Leftrightarrow x = 0$

Without this condition, $\langle \cdot, \cdot \rangle$ is called semi-inner product.

2) $\langle y, x \rangle = \overline{\langle x, y \rangle}$ conjugate

3) $\langle \lambda x + y, z \rangle = \lambda \langle x, z \rangle + \langle y, z \rangle$

Then $(H, \langle \cdot, \cdot \rangle)$ is called an inner product space.

Example. On $([0, 1])$ define $\langle f, g \rangle = \int_0^1 f(x) \overline{g(x)} dx$

$$\langle x, y + \lambda z \rangle = \langle y + \lambda z, x \rangle = \langle y, x \rangle + \lambda \langle z, x \rangle = \overline{\langle x, y \rangle} + \lambda \overline{\langle z, x \rangle} = \overline{\langle x, y \rangle} + \overline{\lambda} \overline{\langle z, x \rangle} = \langle x, y \rangle + \overline{\lambda} \langle x, z \rangle$$

The Cauchy-Schwarz inequality. If $(H, \langle \cdot, \cdot \rangle)$ is an semi-inner product space, then $|\langle x, y \rangle|^2 \leq \langle x, x \rangle \langle y, y \rangle$

$$\text{or } |\langle x, y \rangle| \leq \|x\| \|y\|$$

Proof $\forall \lambda \in \mathbb{C}$
 $0 \leq \langle x - \lambda y, x - \lambda y \rangle = \langle x, x \rangle - \lambda \langle x, y \rangle - \overline{\lambda} \langle y, x \rangle + \lambda \overline{\lambda} \langle y, y \rangle$

If $\lambda = \frac{\langle x, y \rangle}{\langle y, y \rangle}$ then $0 \leq \langle x, x \rangle - \frac{\langle x, y \rangle \langle x, y \rangle}{\langle y, y \rangle} - \frac{\langle x, y \rangle \overline{\langle x, y \rangle}}{\langle y, y \rangle} + \frac{\langle x, y \rangle \overline{\langle x, y \rangle}}{\langle y, y \rangle^2} \langle y, y \rangle$

Hence $\frac{|\langle x, y \rangle|^2}{\langle y, y \rangle} \leq \langle x, x \rangle$, so $|\langle x, y \rangle|^2 \leq \langle x, x \rangle \langle y, y \rangle$.

If $(H, \langle \cdot, \cdot \rangle)$ is an inner product space, then $\|x\| := \langle x, x \rangle^{\frac{1}{2}}$, then $\|\cdot\|$ is really a norm:

$$\|x\| = \langle x, x \rangle^{\frac{1}{2}} \geq 0$$

$$\|x\| = 0 \Leftrightarrow \langle x, x \rangle = 0 \Leftrightarrow x = 0$$

$$\|\lambda x\| = \langle \lambda x, \lambda x \rangle^{\frac{1}{2}} = (|\lambda|^2 \langle x, x \rangle)^{\frac{1}{2}} = |\lambda| \|x\|$$

$$\begin{aligned} \|x + y\|^2 &= \langle x + y, x + y \rangle = \langle x, x \rangle + \langle x, y \rangle + \langle y, x \rangle + \langle y, y \rangle \\ &= \|x\|^2 + \underbrace{\langle x, y \rangle + \overline{\langle x, y \rangle}}_{= 2 \operatorname{Re} \langle x, y \rangle} + \|y\|^2 \end{aligned}$$

$$\begin{aligned} z + \bar{z} &= (a+bi) + (a-bi) \\ &= 2a = 2 \operatorname{Re} z \end{aligned}$$

The Cauchy-Schwarz inequality

$$\therefore \|x + y\|^2 \leq (\|x\| + \|y\|)^2 \quad \checkmark$$

$$\begin{aligned} &\leq 2 |\operatorname{Re} \langle x, y \rangle| \\ &\leq 2 |\langle x, y \rangle| \leq 2 \|x\| \|y\| \end{aligned}$$

$$\|x\|, \|y\| \leq \|x\| \|y\|$$

$$|\operatorname{Re} z| = |a| \leq \sqrt{a^2 + b^2} = |z|$$

If $\|\cdot\|$ on H is complete, then $(H, \langle \cdot, \cdot \rangle)$ is called a Hilbert space.

Counterexamples.

1) $(C[0,1], \|\cdot\|_\infty)$ is a Banach space (why?), but
 $\mathcal{L}^\infty \rightarrow \|f\| = \sup |f(x)|$

its norm can be induced by an inner product.

Otherwise, let $\|f\|_\infty = \langle f, f \rangle^{\frac{1}{2}}$ for some $\langle \cdot, \cdot \rangle$ on $C([0,1])$. We know that in an inner product space

$$\|x+y\|^2 + \|x-y\|^2 = \langle x+y, x+y \rangle + \langle x-y, x-y \rangle = \dots = 2\|x\|^2 + 2\|y\|^2$$

in particular, for $f(t)=1$, $g(t)=t$, we have



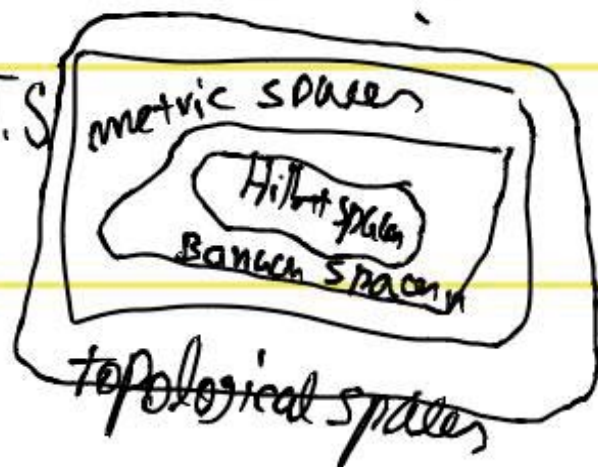
$$\|f+g\|_\infty = \sup_{t \in [0,1]} |1+t| = 2, \quad \|f-g\|_\infty = \sup_{t \in [0,1]} |1-t| = 1, \quad \|f\|_\infty = 1, \quad \|g\|_\infty = 1$$

$$\therefore \|f+g\|_\infty^2 + \|f-g\|_\infty^2 = 5 \neq 4 = 2\|f\|_\infty^2 + 2\|g\|_\infty^2 \quad \times$$

$$H.S. \subsetneq B.S. \subsetneq M.S. \subsetneq T.S.$$

$$2) \mathcal{L}^p = \left\{ \{x_n\} \mid x_n \in \mathbb{C}, \left(\sum_{n=1}^{\infty} |x_n|^p \right)^{\frac{1}{p}} < \infty \right\}$$

is an inner product space iff $p=2$



If ℓ^p is an inner product space i.e. $\|\{x_n\}\|_p = \langle \{x_n\}, \{x_n\} \rangle$ for some $\langle \cdot, \cdot \rangle$ on ℓ^p , then choose $x = (1, 0, 0, \dots)$, $y = (0, 1, 0, \dots)$

Then $\|x+y\|_p^2 + \|x-y\|_p^2 = 2^{\frac{2}{p}} + 2^{\frac{2}{p}} = 2^{1+\frac{2}{p}} = 2^2 = 2\|x\|_p^2 + 2\|y\|_p^2$

$\underbrace{(1, 1, 0, \dots)}_{(1, 1, 0, \dots)} \quad \underbrace{(1, -1, 0, \dots)}_{(1, -1, 0, \dots)}$

So $1 + \frac{2}{p} = 2$, hence $\boxed{p=2}$.

$$\frac{1^p + 1^p}{1^p + 1^p} = \frac{2}{2} = 1$$

Conversely, ℓ^2 is an inner product space under

$$\langle \{x_n\}, \{y_n\} \rangle = \sum_{n=1}^{\infty} x_n \bar{y}_n$$

Then $\|\{x_n\}\|_{\langle \cdot, \cdot \rangle} = \langle \{x_n\}, \{x_n\} \rangle^{\frac{1}{2}} = \left(\sum_{n=1}^{\infty} |x_n|^2 \right)^{\frac{1}{2}} = \|\{x_n\}\|$ embedding

Note that $\mathbb{C}^n = \{(x_1, x_2, \dots, x_n) \mid x_i \in \mathbb{C}, 1 \leq i \leq n\} \xrightarrow[\text{linear}]{\text{isom}} \ell^2$

$(x_1, \dots, x_n) \mapsto (x_1, x_2, \dots, x_n, 0, 0, \dots)$

$\text{Euclidean norm } \|(x_1, \dots, x_n)\| = \left(\sum |x_i|^2 \right)^{\frac{1}{2}} = \|(x_1, x_2, \dots, x_n, 0, 0, \dots)\|_2$

The above inner product on ℓ^2 is well-defined:

$$\forall k; \sum_{n=1}^k |x_n \bar{y}_n| \leq \left(\sum_{n=1}^k |x_n|^2 \right)^{\frac{1}{2}} \left(\sum_{n=1}^k |y_n|^2 \right)^{\frac{1}{2}}$$

Cauchy-Schwarz in $\mathbb{C}^k$

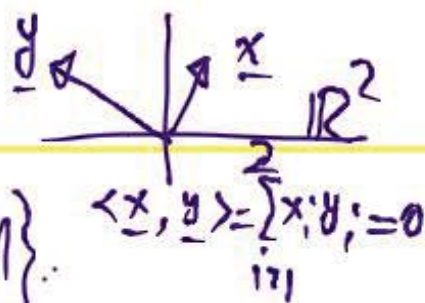
$$\boxed{2+2=3.99999 \dots}$$

$$(\diamond) \leq \left(\sum_{n=1}^{\infty} |x_n|^2 \right)^{\frac{1}{2}} \left(\sum_{n=1}^{\infty} |y_n|^2 \right)^{\frac{1}{2}} < \infty \text{ since } \{x_n\}, \{y_n\} \in \ell^2$$

Since $\left\{ \sum_{n=1}^k |x_n \bar{y}_n| \right\}_{k=1}^{\infty}$ is an increasing seq. and by (♦) it is bounded above. Hence it is convergent. Hence $\sum_{n=1}^{\infty} |x_n \bar{y}_n|$ converges. We know every absolutely convergent seq. is convergent, so $\sum_{n=1}^{\infty} x_n \bar{y}_n$ is convergent.

$$\begin{matrix} x_n \geq 0 \\ \sum_{n=1}^k x_n \leq \sum_{n=1}^{k+1} x_n \end{matrix}$$

Def. We say x is orthogonal to y if $\langle x, y \rangle = 0$ and we write $x \perp y$. For a subset $M \subseteq H$, $M^{\perp} := \{x \in H \mid \langle x, z \rangle = 0 \forall z \in M\}$.



Proposition. 1) $M^{\perp}$ is a closed subspace of H .
2) $M \subseteq M^{\perp\perp}$ 3) $M = M^{\perp\perp}$ if and only if M is closed.
4) $\overline{M^{\perp}} = M^{\perp}$

Proof. 1) Let $x, y \in M^{\perp}$, $\lambda \in \mathbb{C}$.

$$\left\{ \begin{matrix} \langle \lambda x + y, z \rangle = \lambda \langle x, z \rangle + \langle y, z \rangle = 0 \\ \forall z \in M \end{matrix} \right. \Rightarrow \lambda x + y \in M^{\perp}$$

since $x \in M^{\perp}$ and $y \in M^{\perp}$

$\overline{M^{\perp}} = M^{\perp}$. Clearly $M^{\perp} \subseteq \overline{M^{\perp}}$. Let $x \in \overline{M^{\perp}}$. $\exists \{x_n\}$ in $M^{\perp}$ such that $x_n \rightarrow x$. We have

$$\forall z \in M; (\langle x_n, z \rangle = 0 \forall n \Rightarrow \lim_{n \rightarrow \infty} \langle x_n, z \rangle = 0 \Rightarrow \langle \lim_{n \rightarrow \infty} x_n, z \rangle = 0).$$

So $x \in M^\perp$. Thus $M^\perp \subseteq M^\perp$.////

Lemma. $\langle \cdot, z \rangle: H \rightarrow \mathbb{C}$ is a bounded linear function
 $x \mapsto \langle x, z \rangle = f(x)$

Proof. $|\langle x, z \rangle| \leq \|z\| \|x\|$. So f is bdd. $\|f\| = \sup_{x \neq 0} \frac{|f(x)|}{\|x\|} \leq \|z\|$.
by the definition of the inner product
C-S inequality

Moreover, $\frac{|f(z)|}{\|z\|} = \frac{\langle z, z \rangle}{\|z\|} = \frac{\|z\|^2}{\|z\|} = \|z\| \leq \sup_{x \neq 0} \frac{|f(x)|}{\|x\|} \leq \|z\|$.
So $\|f\| = \|z\|$.

Proposition 1) $M \subseteq N \Rightarrow N^\perp \subseteq M^\perp$

$$x \in N^\perp \Rightarrow \forall z \in N; \langle x, z \rangle = 0 \Rightarrow \forall z \in M; \langle x, z \rangle = 0 \Rightarrow x \in M^\perp$$

2) $M \subseteq M^{\perp\perp}$

$$x \in M \Rightarrow \forall z \in M^\perp; \underbrace{z \perp M}_{\text{So } \langle x, z \rangle = 0} \Rightarrow \forall z \in M^\perp; \langle x, z \rangle = 0 \Rightarrow x \in (M^\perp)^\perp$$

3) $\overline{M}^\perp = M^\perp$

First) It follows from (1) & $M \subseteq \overline{M}$ we have $\overline{M}^\perp \subseteq M^\perp$

Second) $M^\perp \subseteq \overline{M}^\perp$: Let $x \in M^\perp$. Let $z \in \overline{M}$. So $\exists \{z_n\}$ in M such

$$\text{that } z_n \rightarrow z. \text{ So } \langle x, z_n \rangle \xrightarrow{0} \langle x, z \rangle \implies \langle x, z \rangle = 0$$

$$\therefore x \in \overline{M}^\perp$$

$\bullet \quad |\langle x, z_n \rangle - \langle x, z \rangle| = |\langle x, z_n - z \rangle| \leq \|x\| \|z_n - z\|$

$\bullet \quad \text{Since } z_n \rightarrow z$

4) M is closed $\Leftrightarrow M = M^{\perp\perp}$

($\Leftarrow$) $M = (M^\perp)^\perp$ is closed since $N^\perp$ is always closed.

($\Rightarrow$) Exercise

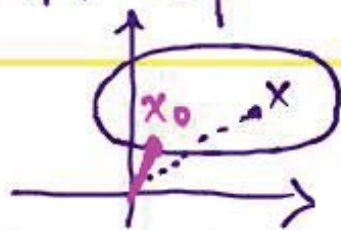
5) $M^\perp = M^{\perp\perp\perp}$ since $M^\perp$ is closed & for all closed set N ; $N^{\perp\perp} = N$.

Def. Let V be a vector space. $E \subseteq V$ is convex when
 $\forall x, y \in E, \forall t \in [0, 1]; tx + (1-t)y \in E$



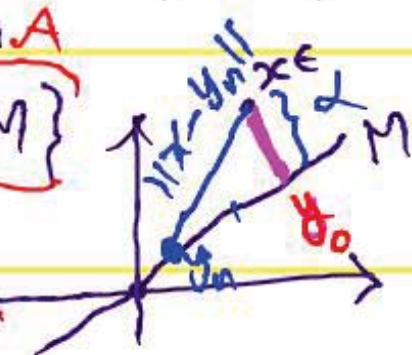
Exercise Every nonempty, closed convex set E in a Hilbert space contains a unique element of smallest norm, i.e. $\exists x_0 \in E; \|x_0\| = \inf\{\|x\| : x \in E\}$

Th. Let M be a closed subspace of a Hilbert space H & $x \in H$. Then



$$\exists y_0 \in M; \|x - y_0\| := \inf\{\|x - y\| : y \in M\}$$

Proof. Let $d = \inf_{y \in M} \|x - y\|$. So $\exists \{y_n\}$ in M ; $\|x - y_n\| \rightarrow d$.



$$\|y_n - y_m\|^2 = \|(y_n - x) + (x - y_m)\|^2$$

$$= 2\|u\|^2 + 2\|v\|^2 - \|u - v\|^2 \quad (\text{law})$$

$$= 2\|y_n - x\|^2 + 2\|x - y_m\|^2 - 4\|x - \frac{y_n + y_m}{2}\|^2$$

$$\leq 2\|y_n - x\|^2 + 2\|x - y_m\|^2 - 4d^2 \rightarrow 0$$

In $\mathbb{R}^2$, subspaces are

Parallel lines through 0
 $\{0\}$
 $\mathbb{R}^2$

Exercise. $\sup A \in \bar{A}$

$\alpha - \epsilon$ $\alpha + \epsilon$



So $\{y_n\}$ is Cauchy in H . Since H is Hilbert, $\exists y_0 \in H$; $y_n \rightarrow y_0$. So $x - y_n \rightarrow x - y_0$. Hence $\|x - y_n\| \rightarrow \|x - y_0\|$

Already $\|x - y_n\| \rightarrow \alpha$. Hence $\|x - y_0\| = \alpha$. Since $y_n \in M$ & $y_n \rightarrow y_0$,
 we have $y_0 \in \overline{M} = M$.
 closed

Next, we show that y_0 is unique. Let y'_0 be another point of M such that $\|x - y'_0\| = \alpha$. Similar to above

$$0 \leq \|y'_0 - y_n\|^2 \leq 2\|y'_0 - x\|^2 + 2\|y_n - x\|^2 - 4\left\|\frac{y'_0 + y_n}{2} - x\right\|^2$$

$$\leq 2\alpha^2 + 2\|y_n - x\|^2 - 4\alpha^2$$

Taking limits, we get $0 \leq \|y'_0 - y_0\|^2 \leq 2\alpha^2 + 2\alpha^2 - 4\alpha^2 = 0$
 $\therefore y'_0 = y_0$. $\square$

Exercise. Norm $\|\cdot\|: X \rightarrow \mathbb{R}$ is cts on X .

Solution. $|\|x\| - \|x_0\|| \leq \|x - x_0\| \iff |\|a\| - \|b\|| \leq \|a - b\| \iff |a - b| \leq \|a\| + \|b\|$

Given $\varepsilon > 0$. Choose $\delta = \varepsilon$. Then $\forall x; \|x - x_0\| < \delta \Rightarrow |\|x\| - \|x_0\|| \leq \|x - x_0\| < \delta = \varepsilon$

Exercise If M is a subspace, then so is $\overline{M}$.

Soln. $x, y \in \overline{M} \Rightarrow \exists \{x_n\}; x_n \xrightarrow{\|\cdot\|} x$
 $\lambda \in \mathbb{C} \quad \exists \{y_n\} y_n \xrightarrow{\|\cdot\|} y \quad \left. \vphantom{\begin{matrix} x_n \xrightarrow{\|\cdot\|} x \\ y_n \xrightarrow{\|\cdot\|} y \end{matrix}} \right\} \Rightarrow \lambda x + y \xrightarrow{\|\cdot\|} \lambda x + y \Rightarrow \lambda x + y \in \overline{M}$

Let $\varepsilon > 0$ be given. Since $x_n \rightarrow x$ & $y_n \rightarrow y$, we have

$$\exists N_1 \forall n \geq N_1; \quad \cancel{\|x_n - x\| < \frac{\varepsilon}{2|\lambda|}} \quad \|\lambda\| \|x_n - x\| < \frac{\varepsilon}{2}$$

$$\exists N_2 \forall n \geq N_2; \quad \|y_n - y\| < \frac{\varepsilon}{2}$$

Let $N = 2N_1 + 700N_2$. Then

$$\forall n \geq N; \quad \|(\lambda x_n + y_n) - (\lambda x + y)\| = \|\lambda(x_n - x) + (y_n - y)\| \leq \underbrace{|\lambda| \|x_n - x\|}_{< \frac{\varepsilon}{2}} + \underbrace{\|y_n - y\|}_{< \frac{\varepsilon}{2}} = \varepsilon$$

Exercise Let $x_n \rightarrow x$ & $y_n \rightarrow y$. Prove that

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

Theorem If M is a closed subspace of a Hilbert space, then $H = M \oplus M^\perp$

$$\begin{aligned} & \{x \in M \cap M^\perp\} = \{0\} \quad \& \\ & \{ \langle x, x \rangle = 0 \} \Rightarrow x = 0 \end{aligned}$$

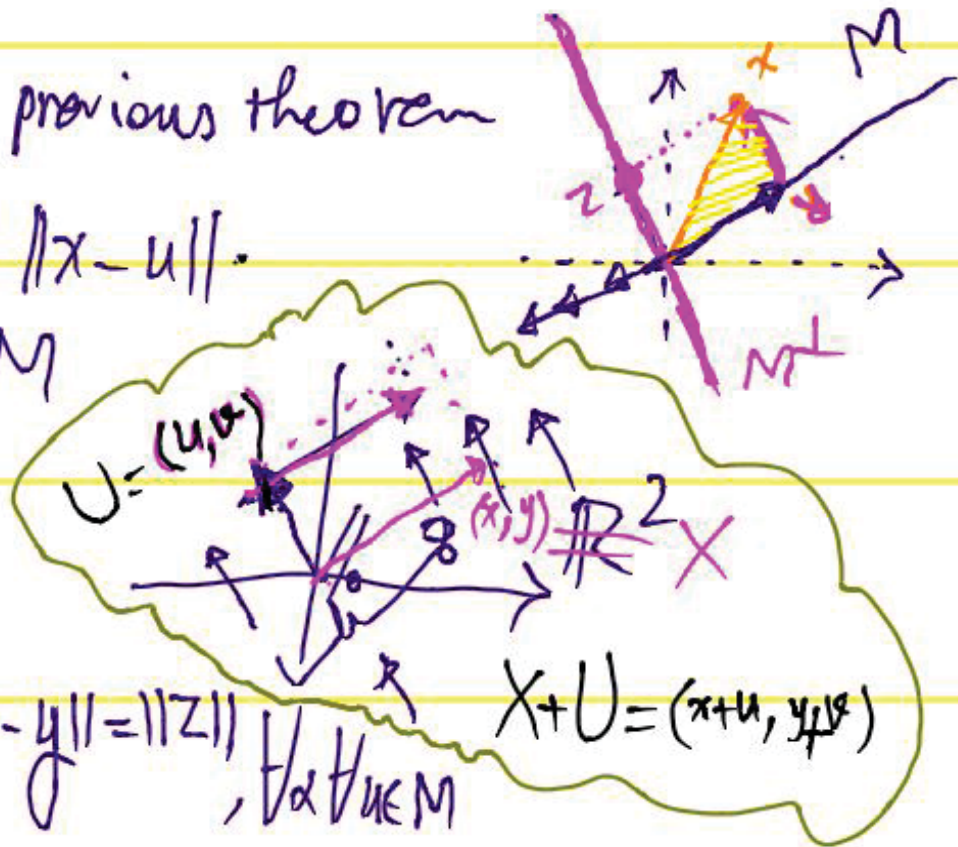
Proof Let $x \in H$. By the previous theorem

$$\exists! y \in M; \|x - y\| = \inf_{u \in M} \|x - u\|.$$

Put $z = x - y$. We show that $z \in M^\perp$:

$$\|z - \alpha u\| = \|x - y - \alpha u\| \geq \|x - y\| = \|z\|, \quad \forall \alpha \forall u \in M$$

$\in M$
subspace



$$\|z - \alpha u\|^2 = \langle z - \alpha u, z - \alpha u \rangle = \|z\|^2 - \alpha \langle z, u \rangle - \alpha \langle u, z \rangle + |\alpha|^2 \|u\|^2 \geq \|z\|^2$$

$2 \operatorname{Re}(\alpha \langle u, z \rangle)$

$$2 \operatorname{Re}(\alpha \langle u, z \rangle) \leq |\alpha|^2 \|u\|^2 \quad \forall \alpha \forall u.$$

E.g. Assume that $\langle u, z \rangle \neq 0$

Put $\alpha = \frac{t \langle u, z \rangle}{\|u\|^2}$ where $t > 0$. So $2 \operatorname{Re} \left(\frac{t \langle u, z \rangle \langle u, z \rangle}{\|u\|^2} \right) \leq t^2 \|u\|^2$

for all $t > 0$. Hence $2 \frac{|\langle u, z \rangle|^2}{\|u\|^2} \leq t^2 \|u\|^2$.

Hence $|\langle u, z \rangle| \leq t \frac{\|u\|^2}{2}$. Let $t \rightarrow 0$. Then $\langle u, z \rangle = 0$.

Thus $\langle u, z \rangle = 0$. ~~*~~ Thus we must have $\langle u, z \rangle = 0 \forall u$

Therefore $z \in M^\perp$. Hence $H = M + M^\perp$. Since $M \cap M^\perp = \{0\}$, we have $H = M \oplus M^\perp$.

Exercise. Let $V = M + N = \{y+z \mid y \in M, z \in N\}$. Then $V = M \oplus N$ if and only if $M \cap N = \{0\}$.

if and only if $\forall x \in V \exists! y \in M \exists! z \in N; x = y + z$

Solution. ($\Rightarrow$) Let $x = y + z = y' + z'$, where $y, y' \in M, z, z' \in N$.

$$\underbrace{y}_{M} - \underbrace{y'}_{M} = \underbrace{z'}_{N} - \underbrace{z}_{N} \implies y - y' = z' - z = 0 \implies y = y' \text{ \& } z = z'.$$

$M \cap N = \{0\}$

$$(\Leftarrow) x \in M \cap N \implies x \in M \text{ \& } x \in N \implies \left. \begin{array}{l} x = \underbrace{x}_{M} + \underbrace{0}_{N} \\ x = \underbrace{0}_{M} + \underbrace{x}_{N} \end{array} \right\} \begin{array}{l} \text{by uniqueness} \\ \implies x = 0. \end{array}$$

Th. If M is closed ^{subspace}, then $M^{\perp\perp} = M$

Proof. Since M is a closed subspace, we have

$$H = M \oplus M^{\perp}$$

We know $M \subseteq M^{\perp\perp}$. Let $x \in M^{\perp\perp} \subseteq H$. So

$\exists y \in M \exists z \in M^{\perp}$ such that $x = y + z$. Then $x - y = z$

So $x - y = z = 0$. Hence $x = y \in M$. $\square$

$$\{0\} = M^{\perp\perp} \cap M^{\perp}$$

Th. Let M be a subspace of H . Then $\overline{M} = H$ iff $M^\perp = \{0\}$.
 $\overline{M}$ is dense in H

Proof. $\overline{M}$ is a closed subspace. So $H = \overline{M} \oplus \overline{M}^\perp$.

($\Leftarrow$) Let $M^\perp = \{0\}$. So $\overline{M}^\perp = M^\perp = \{0\}$. Hence $H = \overline{M} \oplus \{0\} = \overline{M}$.

($\Rightarrow$) Let $\overline{M} = H$. So $\overline{M}^\perp = \overline{M}^\perp \cap \overline{M} = \{0\}$. $\square$

Th. If M is a closed subspace of a Hilbert space, then $H = M \oplus M^\perp$. Put $\begin{cases} P: H \rightarrow H \\ Q: H \rightarrow H \end{cases}$ such that $\begin{cases} Px = y \in M \\ Qx = z \in M^\perp \end{cases}$ where $x = y + z$.

Then

P & Q are linear: $P(x+x') = P(y+z+y'+z') = y+y' = Px + Px'$

$P^2 = P$: $Px = P(Px) = P(y) = P(y+0) = y = Px$

$Q^2 = Q$

$P = I - Q$: $Px = y = x - z = x - Qx = (I - Q)x$

$PQ = QP = 0$: $(PQ)x = P(Qx) = Pz = P(0+z) = 0 = 0(x)$

$\text{ran } P = \{P\tilde{x} : x \in H\} \subseteq M$. Since $y \in M \Rightarrow y = Py \in \text{ran } P$, $M \subseteq \text{ran } P$.

Thus $\text{ran } P = M$.

$$\text{Ker } P = \{x \in H \mid P\tilde{x} = 0\} = M^\perp$$

$$\text{ran } Q = M^\perp, \text{Ker } Q = M$$

$$\begin{aligned} x &= \tilde{y} + \tilde{z} \Rightarrow \|x\|^2 = \langle y+z, y+z \rangle \\ &= \|y\|^2 + \underbrace{\langle y, z \rangle}_0 + \underbrace{\langle z, y \rangle}_0 + \|z\|^2 \end{aligned}$$


P is bd since, $\|P\tilde{x}\| = \|y\| \leq \|x\| \Rightarrow \sup_{x \neq 0} \frac{\|P\tilde{x}\|}{\|x\|} \leq 1$.
 If $x = y \in M$, then $\frac{\|P\tilde{x}\|}{\|x\|} = 1$.

Similarly, $\|Q\| = 1$.

$$\{x \in A; x \leq \alpha\} \Rightarrow \sup A = \alpha$$

We already observed that $\{\langle \cdot, z \rangle : H \rightarrow \mathbb{C} \mid x \mapsto \langle x, z \rangle\}$ is a bd linear functional of norm $\|z\|$.

We want to prove the converse.

Theorem (Riesz Theorem) If $f \in H^*$, then $\exists! z \in H$ s.t. $f(x) = \langle x, z \rangle$ & $\|z\| = \|f\|$.
the dual of H

Proof. If $f \equiv 0$, it is enough to choose $z = 0$.

$$\begin{aligned} |\langle x, 0 \rangle| &\leq \|x\| \|0\| \\ \Downarrow \\ \langle x, 0 \rangle &= 0 \\ \langle x, 0 \rangle &= \langle x, 0+0 \rangle = \langle x, 0 \rangle + \langle x, 0 \rangle \end{aligned}$$

Hence we may assume $\text{Ker} f \neq H$. So $\overline{\text{Ker} f} \neq H$. Then $\text{Ker} f \neq 0$.
 a closed subspace

Let $0 \neq z_0 \in \text{Ker} f$. Let $x \in H$. Put $v = f(x)z_0 - f(z_0)x$. Then

$f(v) = f(x)f(z_0) - f(z_0)f(x) = 0$. So $v \in \text{Ker} f$. Thus

$$\langle v, z_0 \rangle = 0 \Rightarrow f(x)\langle z_0, z_0 \rangle - f(z_0)\langle x, z_0 \rangle = 0 \Rightarrow$$

$$f(x) = \frac{f(z_0)}{\|z_0\|^2} \langle x, z_0 \rangle = \langle x, \underbrace{\frac{f(z_0)}{\|z_0\|^2} z_0}_z \rangle$$

In addition, $\|f\| = \|z\|$.

Let $f(x) = \langle x, z \rangle = \langle x, z' \rangle$. Then $\langle x, z - z' \rangle = 0 \forall x$

So $z - z' = 0$. Hence $z = z'$. $\square$

$$\langle x, w \rangle = 0 \forall x \in H \xrightarrow{x=w} \langle w, w \rangle = 0 \Rightarrow w = 0$$

Def. We say $\{x_\alpha\}$ in H is an orthogonal family if $\langle x_\alpha, x_\beta \rangle = 0 \forall \alpha \neq \beta$. In addition, if $\|x_\alpha\| = 1$, we say $\{x_\alpha\}$ is an orthonormal family.

Proposition. If $\{x_\alpha\}$ is an orthonormal family, then $\{x_\alpha\}$ is linearly independent.

Proof. Let $\{x_1, \dots, x_n\}$ be a finite subset of $\{x_\alpha\}$.

$c_1 x_1 + \dots + c_n x_n = 0$. Then $\forall j$;

$$\begin{aligned} 0 = \langle 0, x_j \rangle &= \langle c_1 x_1 + \dots + c_n x_n, x_j \rangle = c_1 \langle x_1, x_j \rangle + \dots + c_n \langle x_n, x_j \rangle \\ &= 0 + \dots + c_j \underbrace{\langle x_j, x_j \rangle}_{\|x_j\|^2 = 1} + 0 + \dots + 0 = c_j \cdot 1 = c_j. \quad \square \end{aligned}$$

Theorem. If $\{e_n\}$ is an orthonormal seq. in H , then $\sum_{n=1}^{\infty} |\langle x, e_n \rangle|^2 \leq \|x\|^2$ (Bessel ineq.)

Proof. Let $x \in H$ & $z = \sum_{k=1}^n \langle x, e_k \rangle e_k$. We show that $x - z \perp z$:

$$\begin{aligned} \langle x - z, z \rangle &= \langle x, z \rangle - \langle z, z \rangle = \langle x, \sum_{k=1}^n \langle x, e_k \rangle e_k \rangle - \langle \sum_{k=1}^n \langle x, e_k \rangle e_k, \sum_{j=1}^n \langle x, e_j \rangle e_j \rangle \\ &= \sum_{k=1}^n \overline{\langle x, e_k \rangle} \langle x, e_k \rangle - \sum_{k=1}^n \left(\sum_{j=1}^n \langle x, e_k \rangle \overline{\langle x, e_j \rangle} \langle e_k, e_j \rangle \right) = 0 \end{aligned}$$

Handwritten notes:
 - A red circle highlights the term $\sum_{k=1}^n \overline{\langle x, e_k \rangle} \langle x, e_k \rangle$ with the text "مجموع مربعات" (sum of squares) and "مقدار حقيقي" (real quantity).
 - A red bracket under the term $\sum_{k=1}^n \langle x, e_k \rangle \overline{\langle x, e_j \rangle} \langle e_k, e_j \rangle$ is labeled "مجموع = 0" (sum = 0).

So $\|x\|^2 = \|(x-z)+z\|^2 = \|x-z\|^2 + \|z\|^2$, $\|z\|^2 = \left\langle \sum_{k=1}^n \langle x, e_k \rangle e_k, \sum_{j=1}^n \langle x, e_j \rangle e_j \right\rangle$
 $= \dots = \sum_{k=1}^n |\langle x, e_k \rangle|^2$.

Hence $\left\{ \sum_{k=1}^n |\langle x, e_k \rangle|^2 \right\}$ is an increasing seq, which is
 bd above. Hence it is convergent and

$$\sum_{k=1}^{\infty} |\langle x, e_k \rangle|^2 \leq \|x\|^2. \quad \square$$

تعريف (تعريف) : إذا كان عائلة متعامدة في H و هي ماكسималь فإنها تسمى عائلة متعامدة كاملة.

Def. If $\{x_\alpha\}$ is an orthonormal family for H & it is maximal then it is called a complete orthonormal family.

$$\{x_\alpha\} \subseteq \{y_\beta\}^{\text{orth}} \Rightarrow \{x_\alpha\} = \{y_\beta\}$$

Th. Let $\{e_n\}$ be an orthonormal seq., then the following are equivalent:

(i) $\{e_n\}$ is maximal.

(ii) $x = \sum_{n=1}^{\infty} \langle x, e_n \rangle e_n, \forall x \in H$

(iii) $\langle x, y \rangle = \sum_{n=1}^{\infty} \langle x, e_n \rangle \overline{\langle y, e_n \rangle}, \forall x, y \in H$

(iv) $\|x\|^2 = \sum_{n=1}^{\infty} |\langle x, e_n \rangle|^2, \forall x \in H$

(v) The set of all finite linear combinations of $\{e_n\}$ are dense in H .

Project (مشارعة)

Then $\{\langle x, e_n \rangle\}$ is called the Fourier coefficients of x with respect to $\{e_n\}$.

Project (مشارعة)

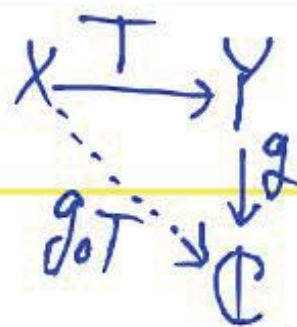
Proof. (1) $\square$

If X, Y are normed spaces, $T \in B(X, Y)$, then

$$T^*: Y^* \rightarrow X^*$$

$$g \mapsto g \circ T$$

is called the dual of T .



فرض
نیز
فرض
که از حالت قبلی
اصل

Th. T^* is well-defined & $\|T^*\| = \|T\|$.

Proof. $g_1 = g_2 \Rightarrow \forall y \in Y; g_1(y) = g_2(y) \Rightarrow g(Tx) = g_2(Tx) \forall x \Rightarrow g \circ T = g_2 \circ T$

$$\begin{aligned} (g \circ T)(\lambda x + x') &= g(T(\lambda x + x')) = g(\lambda Tx + Tx') = \lambda g(Tx) + g(Tx') \\ &= \lambda (g \circ T)(x) + (g \circ T)(x') \end{aligned}$$

$$|(g \circ T)(x)| = |g(Tx)| \leq \|g\| \|Tx\| \leq \|g\| \|T\| \|x\|$$

$$\therefore \|g \circ T\| \leq \|g\| \|T\|$$

$g \in Y^*$

$T \in B(X, Y)$

$$\therefore g \circ T \in X^*$$

$$\left\{ \begin{array}{l} T^*: Y^* \rightarrow X^* \\ g \mapsto g \circ T \end{array} \right. \Rightarrow \|T^*g\| = \|g \circ T\| \leq \|g\| \|T\| \Rightarrow \|T^*\| \leq \|T\|$$

$$= \sup_{g \neq 0} \frac{\|T^*g\|}{\|g\|}$$

Let $\|x\| = 1$. By a consequence of the Hahn-Banach theorem $\Rightarrow g \in Y^*$ such that $\|g\| = 1$ & $g(Tx) = \|Tx\|$. So $\|Tx\| = |g(Tx)| = |(g \circ T)(x)| \leq \|g \circ T\| \|x\| = \|T^*g\| \leq \|T^*\| \|g\|$. Hence $\|Tx\| \leq \|T^*\|$. Taking sup, we get $\|T\| \leq \|T^*\|$.

Closed Graph Theorem. Let $T: \underbrace{D(T) \subseteq X}_{\text{domain of } T} \rightarrow Y$ is a linear operator between Banach spaces & $D(T)$ is closed in X . Then T is bounded if and only if $\text{Gr}(T) = \{(x, Tx) \mid x \in D(T)\}$ is closed in $X \times Y$ equipped with the product top $\|(x, y)\| = \max\{\|x\|, \|y\|\}$

Proof. ($\Rightarrow$) Let T be bounded & $(x_n, Tx_n) \rightarrow (x, y) \in X \times Y$.

So $x_n \rightarrow x$ & $Tx_n \rightarrow y$. Since T is cts, so $Tx_n \rightarrow Tx$. $y = Tx$. Thus $(x, y) = (x, Tx) \in \text{Gr}(T)$.

($\Leftarrow$) Since $\text{Gr}(T)$ is closed in the Banach space $X \times Y$, $\text{Gr}(T)$ is also a Banach space

Define $\varphi: \text{Gr}(T) \rightarrow D(T)$. Clearly φ

$(x, Tx) \mapsto x$
is linear. In addition, $\|\varphi(x, Tx)\| = \|x\| \leq \max\{\|x\|, \|Tx\|\} = \|(x, Tx)\|$
so φ is bd. φ is also onto, since if $x \in D(T)$, the $\varphi(x, Tx) = x$.

φ is 1-1, since $\varphi(x_1, Tx_1) = \varphi(x_2, Tx_2) \Rightarrow x_1 = x_2 \Rightarrow Tx_1 = Tx_2 \Rightarrow (x_1, Tx_1) = (x_2, Tx_2)$

By a consequence of open mapping theorem we φ^{-1} is bd

$(x_n, y_n) \rightarrow (x, y)$
in $X \times Y$
 $\Downarrow$
 $x_n \xrightarrow{\text{in } X} x$
& $y_n \xrightarrow{\text{in } Y} y$

Let (X, d) be a complete metric space. Then $F \subseteq X$ is closed iff (F, d) is complete.

$\exists M; \|\varphi'(x)\| \leq M \|x\|$. Hence T is bd. $\square$

$$\|Tx\| \leq \max\{\|x\|, \|Tx\|\}$$

Proposition. If $T: \mathcal{D}(T) \subseteq X \rightarrow Y$ is a bd linear mapping, & $\text{Gr}(T)$ is closed, then $\mathcal{D}(T)$ is closed in X .

Proof. Let $\{x_n\}$ be a seq. in $\mathcal{D}(T)$ such that $x_n \rightarrow x \in X$. We shall show that $x \in \mathcal{D}(T)$:

Note that $(x_n, Tx_n) \in \text{Gr}(T)$.

$$\begin{aligned} \|(x_n, Tx_n) - (x_m, Tx_m)\| &= \|(x_n - x_m, Tx_n - Tx_m)\| = \max\{\|x_n - x_m\|, \|Tx_n - Tx_m\|\} \\ &= \max\{1, \|T\|\} \|x_n - x_m\| \end{aligned}$$

Since $\{x_n\}$ is Cauchy (since it is convergent), we conclude that $\{(x_n, Tx_n)\}$ is Cauchy. Due to $X \times Y$ is Banach, $\{(x_n, Tx_n)\}$ is convergent in $\text{Gr}(T)$.

Thus $\exists (u, v) \in X \times Y; (x_n, Tx_n) \rightarrow (u, v)$. Since $\text{Gr}(T)$ is closed, $(u, v) \in \text{Gr}(T)$. Hence $u \in \mathcal{D}(T)$. On the other hand, $x_n^{(**)} \rightarrow u$ & $Tx_n \rightarrow v$. By $(**)$, $(**) x = u \in \mathcal{D}(T)$. $\square$

Exercise. If $\forall x \in H, \langle Tx, x \rangle = 0$, then $T = 0$ → Hilbert space over the field $\mathbb{C}$

Th. If $\{e_n\}$ is an orth. seq in H , the T.F. A.E.:

(i) $\{e_n\}$ is maximal (مکمل بودن به این معنیست که هیچ عنصری را از H اضافه نمی توانیم به $\{e_n\}$ اضافه کرد به طوری که تمام عناصر به باقی بماند)

(ii) the set of all finite linear combinations of $\{e_n\}$ is dense in H

(iii) $x = \sum_{n=1}^{\infty} \langle x, e_n \rangle e_n$

$$P = \left\{ \sum_{i=1}^n \lambda_i e_i \mid \lambda_i \in \mathbb{C} \right\}$$

(iv) $\langle x, y \rangle = \sum_{n=1}^{\infty} \langle x, e_n \rangle \langle y, e_n \rangle$

(v) $\|x\|^2 = \sum_{n=1}^{\infty} |\langle x, e_n \rangle|^2$

$$\begin{aligned} e_2 + 30e_4 &= 0e_1 + 1e_2 \\ &\quad + 0e_3 + 30e_4 \end{aligned}$$

Proof: (i $\Rightarrow$ ii). Let $\bar{P} \neq H$. By a theorem, $\bar{P}^\perp \neq \{0\}$, (عبارت دیگر $\bar{P} \neq H$)

$\exists \frac{x_0}{\|x_0\|} \in \bar{P}^\perp$. Then $\frac{x_0}{\|x_0\|}$ is orthogonal to each element of P , in

Particular, $\frac{x_0}{\|x_0\|} \perp e_n$ for all n . Then $\left\{ \frac{x_0}{\|x_0\|} \right\} \cup \{e_1, e_2, \dots\}$ is orthonormal

which is impossible by (i). (چون $\{e_1, e_2, \dots\} \subsetneq$ یک پایه اورتونرمال است)

(ii $\Rightarrow$ iii) Consider the linear map $\theta: P \rightarrow \ell^2 = \left\{ (x_n) \mid \sum_{n=1}^{\infty} |x_n|^2 < \infty \right\}$

$$\sum_{i=1}^n \lambda_i e_i \mapsto (\lambda_i)_{i=1}^n = (\lambda_1, \dots, \lambda_n, 0, 0, \dots)$$

$$\langle \sum_{i=1}^n \lambda_i e_i, \sum_{j=1}^m \mu_j e_j \rangle = \dots = \sum_{i=1}^{\infty} \lambda_i \bar{\mu}_i = \langle (\lambda_i), (\mu_j) \rangle \quad (n \leq m)$$

$$= \langle \theta(v), \theta(v) \rangle. \text{ So } \|\theta(\sum_{i=1}^n \lambda_i e_i)\| = \|(\lambda_i)\| (\sum_{i=1}^n |\lambda_i|^2)^{\frac{1}{2}}$$

لا بد ان يكون θ متساوي القياس
 لان θ يحفظ القياس
 لان $\|\theta(v)\| = \|v\|$

^{Ban}
 T: $D \subset X \rightarrow Y$ is a
 bd linear map, then
 T has an extension
 $\bar{T}: \bar{D} \rightarrow Y$ such that
 $\|\bar{T}\| = \|T\|$. If T
 is isometry, so is
 $\bar{T}$

So $\exists \bar{\theta}: \bar{P} \rightarrow \ell^2$ is an isometry
 $\bar{\theta}(\sum_{i=1}^{\infty} \lambda_i e_i) = (\lambda_i)$
_{(iii) x}

$$\langle x, e_i \rangle = \langle \sum_{j=1}^{\infty} \lambda_j e_j, e_i \rangle = \sum_{j=1}^{\infty} \langle \lambda_j e_j, e_i \rangle = \lambda_i \langle e_i, e_i \rangle = \lambda_i \Rightarrow$$

$$x = \sum_{i=1}^{\infty} \langle x, e_i \rangle e_i.$$

(iii) $\Rightarrow$ (iv) & (iv) $\Rightarrow$ (v) are clear. We prove (vi) $\Rightarrow$ (i).
 Assume $\{e_n\}$ is not maximal. So $\exists \{e_j\} \subset S$ orthonormal.
 Let $x_0 \in S \setminus \{e_n\}$.
 So $\langle x_0, e_n \rangle = 0 \forall n$. Hence $\|x_0\| = (\sum_{n=1}^{\infty} |\langle x_0, e_n \rangle|^2)^{\frac{1}{2}} = 0$.
 So $x_0 = 0$. Here $\|x_0\| = 0$. On the other hand $\|x_0\| = 1$. $\times$

2.8 Definition Let f be a real (or extended-real) function on a topological space. If

$$\bar{f}(\alpha, \infty) = \{x: f(x) > \alpha\} \quad \{x: \sup_i f_i(x) > \alpha\} = \bigcup \{x: f_i(x) > \alpha\}$$

is open for every real α , f is said to be *lower semicontinuous*. If

$$\{x: f(x) < \alpha\}$$

is open for every real α , f is said to be *upper semicontinuous*.

$\downarrow$
 $\sup f_i$ is l.s.c + s
 if so are f_i 's.

A real function is obviously continuous if and only if it is both upper and lower semicontinuous.

The simplest examples of semicontinuity are furnished by characteristic functions:

- (a) Characteristic functions of *open* sets are *lower* semicontinuous.
- (b) Characteristic functions of *closed* sets are *upper* semicontinuous.

The following property is an almost immediate consequence of the definitions:

- (c) The supremum of any collection of lower semicontinuous functions is lower semicontinuous. The infimum of any collection of upper semicontinuous functions is upper semicontinuous.

5.8 The Banach-Steinhaus Theorem Suppose X is a Banach space, Y is a normed linear space, and $\{\Lambda_\alpha\}$ is a collection of bounded linear transformations of X into Y , where α ranges over some index set A . Then either there exists an $M < \infty$ such that

$$\|\Lambda_\alpha\| \leq M$$

$$\sup_\alpha \|\Lambda_\alpha\| < \infty \quad (1)$$

for every $\alpha \in A$ or

$$\sup_{\alpha \in A} \|\Lambda_\alpha x\| = \infty \quad (2)$$

for all x belonging to some dense G_δ in X .

PROOF Put

$$\varphi(x) = \sup_{\alpha \in A} \|\Lambda_\alpha x\| \quad (x \in X) \quad (3)$$

and let

$$V_n = \{x: \varphi(x) > n\} \quad (n = 1, 2, 3, \dots). \quad (4)$$

Since each Λ_α is continuous and since the norm of Y is a continuous function on Y (an immediate consequence of the triangle inequality, as in the proof of Theorem 4.6), each function $x \rightarrow \|\Lambda_\alpha x\|$ is continuous on X . Hence φ is lower

semicontinuous, and each V_n is open.

5.6 Baire's Theorem If X is a complete metric space, the intersection of every countable collection of dense open subsets of X is dense in X .

Two cases happen: (i) If one of the sets V_n is not dense in X , then $\exists x_0 \in X; x_0 \notin \bar{V}_N \Rightarrow \exists r; N(x_0) \cap \bar{V}_N = \emptyset$ or equivalently, if $\|z - x_0\| < r \Rightarrow z \notin V_N$ (or equiv. $\varphi(x+x_0) \leq N$ or equiv. $\|\Lambda_\alpha(x+x_0)\| \leq N \forall \alpha$).

We have $\|\Lambda_\alpha x\| = \|\Lambda_\alpha(x+x_0) - \Lambda_\alpha(x_0)\| \leq \|\Lambda_\alpha(x+x_0)\| + \|\Lambda_\alpha(x_0)\| \leq N + N = 2N$ ($\|x\| < r$)

So $\forall x \in X; \left\| \frac{rx}{2} \right\| = \frac{r}{2} < r \Rightarrow \left\| \Lambda_\alpha \left(\frac{r}{2} x \right) \right\| \leq 2N \Rightarrow \|\Lambda_\alpha(x)\| \leq \frac{4N}{r}$

$\therefore \|\Lambda_\alpha\| = \sup_{\|x\|=1} \|\Lambda_\alpha x\| \leq \frac{4N}{r}$, so $\sup_\alpha \|\Lambda_\alpha\| < \infty$.

(ii) All V_n 's are dense in X . By the Baire theorem $\bigcap_{n=1}^\infty V_n$ is dense in X . $\forall x \in \bigcap_{n=1}^\infty V_n; (\forall n; x \in V_n) \Rightarrow (\forall n; \sup_{\alpha} \|\Lambda_\alpha x\| > n) \Rightarrow \sup_{\alpha} \|\Lambda_\alpha x\| = \infty$. $\square$

Application ①

L^p -SPACES

3.1 Definition A real function φ defined on a segment (a, b) , where $-\infty \leq a < b \leq \infty$, is called *convex* if the inequality

$$\varphi((1-\lambda)x + \lambda y) \leq (1-\lambda)\varphi(x) + \lambda\varphi(y) \quad (1)$$

holds whenever $a < x < b, a < y < b$, and $0 \leq \lambda \leq 1$.

to be open, is important.

3.2 Theorem If φ is convex on (a, b) then φ is continuous on (a, b) .

3.4 Definition If p and q are positive real numbers such that $p + q = pq$, or equivalently

$$\frac{1}{p} + \frac{1}{q} = 1,$$

$$p > 1 \Rightarrow \frac{1}{q} = 1 - \frac{1}{p} < 1 \Rightarrow q > 1 \quad (1)$$

then we call p and q a pair of *conjugate exponents*. It is clear that (1) implies $1 < p < \infty$ and $1 < q < \infty$. An important special case is $p = q = 2$.

As $p \rightarrow 1$, (1) forces $q \rightarrow \infty$. Consequently 1 and ∞ are also regarded as a pair of conjugate exponents. Many analysts denote the exponent conjugate to p by p' , often without saying so explicitly.

3.5 Theorem Let p and q be conjugate exponents, $1 < p < \infty$. Let X be a measure space, with measure μ . Let f and g be measurable functions on X , with range in $[0, \infty]$. Then

Holder inequality

$$\int_X fg \, d\mu \leq \left\{ \int_X f^p \, d\mu \right\}^{1/p} \left\{ \int_X g^q \, d\mu \right\}^{1/q} \quad (1)$$

and

$$\left\{ \int_X (f+g)^p \, d\mu \right\}^{1/p} \leq \left\{ \int_X f^p \, d\mu \right\}^{1/p} + \left\{ \int_X g^p \, d\mu \right\}^{1/p} \quad (2)$$

Minkowski inequality

PROOF Let A and B be the two factors on the right of (1). If $A = 0$, then $f = 0$ a.e. (by Theorem 1.39); hence $fg = 0$ a.e., so (1) holds. If $A > 0$ and $B = \infty$, (1) is again trivial. So we need consider only the case $0 < A < \infty$, $0 < B < \infty$. Put

$$F = \frac{f}{A}, \quad G = \frac{g}{B} \quad (3)$$

This gives

$$F^p = \frac{f^p}{A^p} \Rightarrow \int F^p = \frac{1}{A^p} \int f^p = \frac{A^p}{A^p} = 1$$

$$\int_X F^p \, d\mu = \int_X G^q \, d\mu = 1 \quad (4)$$

$$\begin{aligned} F(x) &= e^{\frac{s}{p}} \\ \ln F(x) &= \frac{s}{p} \\ \uparrow \\ s &= p \ln F(x) \end{aligned}$$

If $x \in X$ is such that $0 < F(x) < \infty$ and $0 < G(x) < \infty$, there are real numbers s and t such that $F(x) = e^{s/p}$, $G(x) = e^{t/q}$. Since $1/p + 1/q = 1$, the convexity of the exponential function implies that

$$e^{s/p + t/q} \leq p^{-1} e^s + q^{-1} e^t \quad (5)$$

It follows that

$$F(x)G(x) \leq p^{-1} F(x)^p + q^{-1} G(x)^q \quad (6)$$

for every $x \in X$. Integration of (6) yields

$$\int_X FG \, d\mu \leq p^{-1} + q^{-1} = 1 \quad (7)$$

by (4); inserting (3) into (7), we obtain (1).

$$\int \frac{fg}{AB} \leq 1 \Rightarrow \int fg \leq AB$$

To prove (2), we write

$$(f+g)^p = f \cdot (f+g)^{p-1} + g \cdot (f+g)^{p-1} \quad (8)$$

Hölder's inequality gives

$$\int f \cdot (f+g)^{p-1} \leq \left\{ \int f^p \right\}^{1/p} \left\{ \int (f+g)^{(p-1)q} \right\}^{1/q} \quad (9)$$

$$\frac{1}{p} + \frac{1}{q} = 1$$

Let (9') be the inequality (9) with f and g interchanged. Since $(p-1)q = p$, addition of (9) and (9') gives

$$\int (f+g)^p \leq \left\{ \int (f+g)^p \right\}^{1/q} \left[\left\{ \int f^p \right\}^{1/p} + \left\{ \int g^p \right\}^{1/p} \right] \quad (10)$$

Clearly, it is enough to prove (2) in the case that the left side is greater than 0 and the right side is less than ∞ . The convexity of the function t^p for $0 < t < \infty$ shows that

$$\left(\frac{f+g}{2} \right)^p \leq \frac{1}{2} (f^p + g^p) \quad \sim \quad \int \left(\frac{f+g}{2} \right)^p \leq \frac{1}{2} \int f^p + \frac{1}{2} \int g^p < \infty$$

for almost all x ; integrating (2), we obtain (1). If $p = 1$, then $q = \infty$, and the same argument applies. ////

3.9 Theorem Suppose $1 \leq p \leq \infty$, and $f \in L^p(\mu)$, $g \in L^p(\mu)$. Then $f + g \in L^p(\mu)$, and

$$\|f + g\|_p \leq \|f\|_p + \|g\|_p. \quad (1)$$

PROOF For $1 < p < \infty$, this follows from Minkowski's inequality, since

$$\int_X |f + g|^p d\mu \leq \int_X (|f| + |g|)^p d\mu.$$

For $p = 1$ or $p = \infty$, (1) is a trivial consequence of the inequality $|f + g| \leq |f| + |g|$.

$$\begin{aligned} \int |f+g| &\leq \int (|f|+|g|) \leq \int |f| + \int |g| \\ \Rightarrow \|f+g\|_1 &\leq \|f\|_1 + \|g\|_1 \end{aligned}$$

$$\begin{aligned} |f+g| &\leq \|f\|_\infty + \|g\|_\infty \text{ a.e.} \Rightarrow \|f+g\|_\infty \leq \|f\|_\infty + \|g\|_\infty. \end{aligned}$$

3.11 Theorem $L^p(\mu)$ is a complete metric space, for $1 \leq p \leq \infty$ and for every positive measure μ .

① Theorem. Let $\{x_n\}$ be a sequence in a Hilbert space H and $x_0 \in H$ such that $\langle x_n, x \rangle \rightarrow \langle x_0, x \rangle$ (we say $\{x_n\}$ weakly tends to x_0)

Then $\{x_n\}$ is bd, i.e., $\exists M$ s.t. $\|x_n\| \leq M$.

Proof. Put $\{\langle x_n, \cdot \rangle\}^\infty$. $\forall x \in H$, $\{\langle x_n, x \rangle\}$ is convergent, so it is bd. By

the Banach-Steinhaus theorem, it is uniformly bounded, so $\exists M$ s.t. $\|x_n\| \leq M$. $\square$

$$\begin{aligned} x_n \xrightarrow{\|\cdot\|} x &\Rightarrow \|x_n - x\| \rightarrow 0 \Rightarrow \langle x_n, y \rangle - \langle x, y \rangle = \langle x_n - x, y \rangle \leq \|x_n - x\| \|y\| \Rightarrow \langle x_n, y \rangle \rightarrow \langle x, y \rangle \\ &\Rightarrow x_n \xrightarrow{\text{weakly}} x \end{aligned}$$

An immediate consequence of the Banach-Steinhaus theorem is that

if $\{\lambda_\alpha\}$ is pointwise bounded, then $\{\lambda_\alpha\}$ is uniformly bd

$$\begin{aligned} \forall x; \{\lambda_\alpha x\} \text{ is bd} &\quad \exists M \forall \alpha; \|\lambda_\alpha\| \leq M \\ \exists M \forall x \forall \alpha; \|\lambda_\alpha x\| \leq M &\quad \exists M \forall \alpha; \|\lambda_\alpha\| \leq M \end{aligned}$$

3.11 Theorem $L^p(\mu)$ is a complete metric space, for $1 \leq p \leq \infty$ and for every positive measure μ .

PROOF Assume first that $1 \leq p < \infty$. Let $\{f_n\}$ be a Cauchy sequence in $L^p(\mu)$. There is a subsequence $\{f_{n_i}\}$, $n_1 < n_2 < \dots$, such that

$$\|f_{n_{i+1}} - f_{n_i}\|_p < 2^{-i} \quad (i = 1, 2, 3, \dots). \quad (1)$$

Put

$$g_k = \sum_{i=1}^k |f_{n_{i+1}} - f_{n_i}|, \quad g = \sum_{i=1}^{\infty} |f_{n_{i+1}} - f_{n_i}|. \quad (2)$$

Since (1) holds, the Minkowski inequality shows that $\|g_k\|_p < 1$ for $k = 1, 2, 3, \dots$. Hence an application of Fatou's lemma to $\{g_k^p\}$ gives $\|g\|_p \leq 1$. In particular, $g(x) < \infty$ a.e., so that the series

$$f_{n_1}(x) + \sum_{i=1}^{\infty} (f_{n_{i+1}}(x) - f_{n_i}(x)) \quad (3)$$

converges absolutely for almost every $x \in X$. Denote the sum of (3) by $f(x)$, for those x at which (3) converges; put $f(x) = 0$ on the remaining set of measure zero. Since

$$f_{n_1} + \sum_{i=1}^{k-1} (f_{n_{i+1}} - f_{n_i}) = f_{n_k}, \quad (4)$$

we see that

$$f(x) = \lim_{i \rightarrow \infty} f_{n_i}(x) \quad \text{a.e.} \quad (5)$$

Having found a function f which is the pointwise limit a.e. of $\{f_{n_i}\}$, we now have to prove that this f is the L^p -limit of $\{f_n\}$. Choose $\epsilon > 0$. There exists an N such that $\|f_n - f_m\|_p < \epsilon$ if $n > N$ and $m > N$. For every $m > N$, Fatou's lemma shows therefore that

$$\int_X |f - f_m|^p d\mu \leq \liminf_{i \rightarrow \infty} \int_X |f_{n_i} - f_m|^p d\mu \leq \epsilon^p. \quad (6)$$

We conclude from (6) that $f - f_m \in L^p(\mu)$, hence that $f \in L^p(\mu)$ [since $f = (f - f_m) + f_m$], and finally that $\|f - f_m\|_p \rightarrow 0$ as $m \rightarrow \infty$. This completes the proof for the case $1 \leq p < \infty$.

In $L^\infty(\mu)$ the proof is much easier. Suppose $\{f_n\}$ is a Cauchy sequence in $L^\infty(\mu)$, let A_k and $B_{m,n}$ be the sets where $|f_k(x)| > \|f_k\|_\infty$ and where $|f_n(x) - f_m(x)| > \|f_n - f_m\|_\infty$, and let E be the union of these sets, for $k, m, n = 1, 2, 3, \dots$. Then $\mu(E) = 0$, and on the complement of E the sequence $\{f_n\}$ converges uniformly to a bounded function f . Define $f(x) = 0$ for $x \in E$. Then $f \in L^\infty(\mu)$, and $\|f_n - f\|_\infty \rightarrow 0$ as $n \rightarrow \infty$.

L^p is the set of all equivalence classes:

$$[f] = \{g \mid g \sim f\}$$

$$g \sim f \Leftrightarrow g = f \text{ a.e.}$$

$$\|f\|_p = 0 \Rightarrow \int |f|^p = 0$$

$$\Rightarrow |f|^p = 0 \text{ a.e.} \Rightarrow$$

$$f = 0 \text{ a.e.} \Rightarrow [f] = [0]$$



کسی f و g برابری
فقط آنی که تقریباً
با هم برابرند

$$L^p = \text{مجموعه کلاس‌های معادله}$$

$$\| [f] \|_p = \left(\int |f|^p \right)^{1/p}$$

$$E = (A_k) \cup (B_{m,n})$$

$$\mu(E) \leq \sum \mu(A_k) + \sum \mu(B_{m,n}) = 0$$

Last statement:
در این بخوانید