

Claim: Any convergent sequence in a Metric space must be Cauchy

Claim: Every Cauchy sequence is bounded

Claim: A bounded sequence in  $\mathbb{R}^n$  has at least one limit point: Bolzano Weierstrass Theorem

eg:  $(1, 0, 1, 0, 1, \dots)$

$x \in \mathbb{R}^n$  is said to be a limit point of  $\{x_k\}$  if  $\exists$  a subsequence of  $\{x_k\}$  that converges to  $x$ .

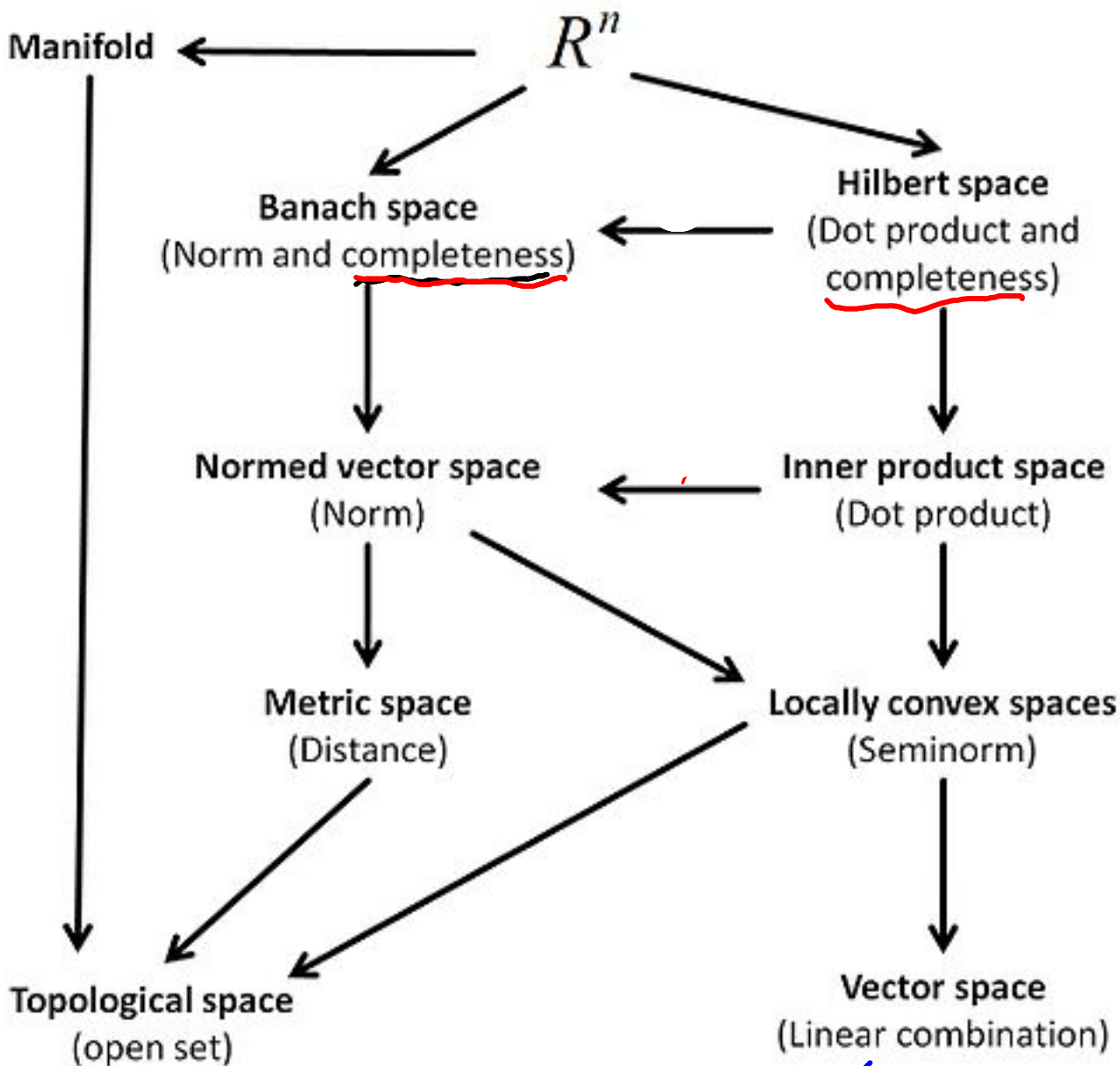
(see Bertsekas  
Convex analysis  
chap 2 & 3)

Claim: GIVEN A METRIC SPACE  $S$ , EVERY CAUCHY SEQUENCE NEED NOT CONVERGE TO A LIMIT POINT IN  $S$ !

(We saw several examples:  $\left(x_{n+1} = \frac{x_n + \frac{2}{x_n}}{2}\right)$ )

Claim: In  $\mathbb{R}^n$ , every Cauchy sequence converges to a limit point in  $\mathbb{R}^n$

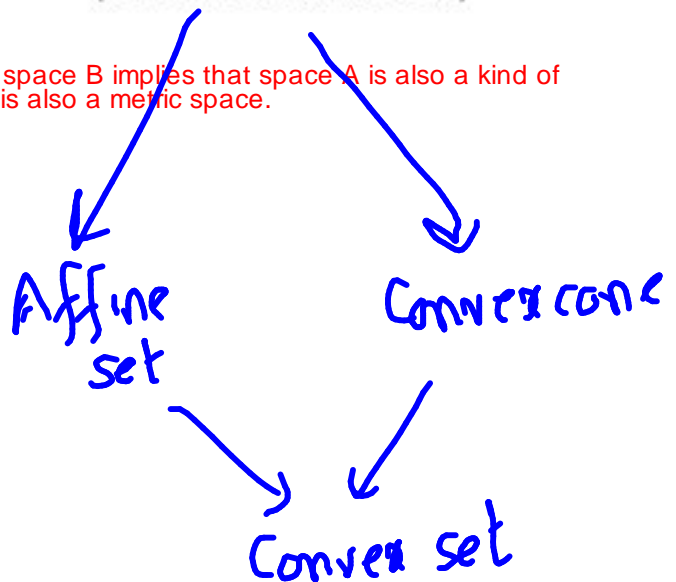
Such spaces are called complete spaces



Overview of types of abstract spaces. An arrow from space A to space B implies that space A is also a kind of space B. That means, for instance, that a normed vector space is also a metric space.

### Complete metric space

A metric space  $S$  in which every Cauchy sequence in  $S$  is convergent in  $S$



Show that the following are vector spaces (assuming scalars come from a set  $S$ ), and then answer questions that follow for each of them: Set of all matrices on  $S$ , set of all polynomials on  $S$ , set of all sequences of elements of  $S$ . (HINT: You can refer to [this book](#) for answers to most questions in this homework.) How would you understand the concepts of independence, span, basis, dimension and null space (chapter 2 of [this book](#)), eigenvalues and eigenvectors (chapter 5), inner product and orthogonality (chapter 6)? EXTRA: Now how about set of all random variables and set of all functions. **Deadline:** January 23 2015.

Examples: Let  $S$  be a field. Good examples can be found at [http://en.wikipedia.org/wiki/Field\\_\(mathematics\)#Examples](http://en.wikipedia.org/wiki/Field_(mathematics)#Examples)

(a)  $S^\infty$ : Space of infinite sequences of elements from  $S$ :  $(x_1, x_2, x_3, \dots)$

↳ Only finitely many non-zero elements

$x \in \text{Span}(V)$   
if  $x$  can be  
obtained by linear  
combination of  
finite  $A$  of elements from  $V$

⇒ Basis =  $\left\{ \underbrace{(1, 0, \dots)}_{e_1}, \underbrace{(0, 1, 0, \dots)}_{e_2}, \dots, \underbrace{(0, \dots, 1, 0, \dots)}_{e_i}, \dots \right\}$

Dimensionality = countably infinite

↳ No restriction on non-zero elements

⇒ Basis exists (enumerating basis is open)

Dimensionality = uncountably infinite

(Banach space)  $l^p$  ↳ With bounded  $p$ -norm:  $\|x\|_p = \left( \sum_{i=1}^{\infty} |x_i|^p \right)^{1/p} < \infty$   
Dimensionality = countably infinite  $p > 0$

Q: For  $\mathbb{R}^n$

$$\left\{ \|x - x_0\|_p \leq r \right\} \stackrel{?}{\subseteq} \left\{ \|x - x_0\|_{p'} \leq r \right\}$$

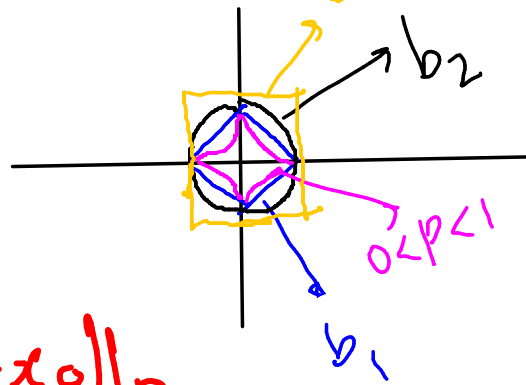
$$p' \geq p$$

Ans: Yes

Because

$$\|x - x_0\|_{p'} \leq \|x - x_0\|_p$$

Same relation holds with  $l_p$  &  $l_{p'}$   
for infinite sequences



Eg: For  $p=2$  you have  $l^2$ : Square summable sequences

$$l^p = \left\{ (x_1, x_2, \dots) \mid \left( \sum_{i=1}^{\infty} |x_i|^p \right)^{1/p} < \infty \right\}$$

Note:  $l^p \subseteq l^{p'}$  for  $p' > p$

Does this hold for  $\mathbb{R}^n$ : Yes with =

$$\sum_{i=1}^{\infty} \left| \frac{1}{i} \right|^p$$

Eg:  $(1, \frac{1}{2}, \dots, \frac{1}{n}, \frac{1}{n+1}, \dots) \notin l^1$

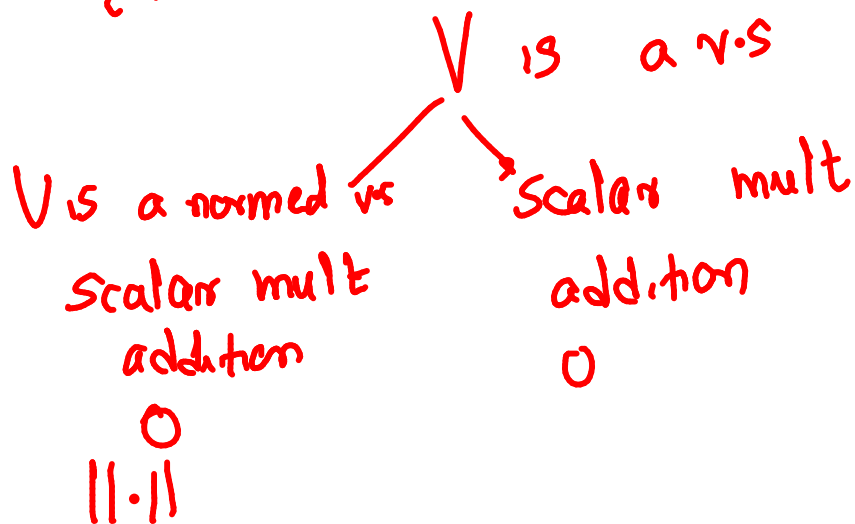
But:  $(1, \frac{1}{2}, \dots, \frac{1}{n}, \frac{1}{n+1}, \dots) \in l^p$  for  $p > 1$   $\left( \sum_{i=1}^{\infty} \left| \frac{1}{i} \right|^p \right)^{1/p}$

Q: Is  $l^p$  Hilbert space?

Ans: Only when  $p=2$

Show that there does not exist  $(x, y \in \mathbb{R}^n)$   
 $\langle x, y \rangle$  inner product s.t

$$\langle x, x \rangle = \left( \sum_{i=1}^n |x_i|^p \right)^{2/p}$$



Solution:

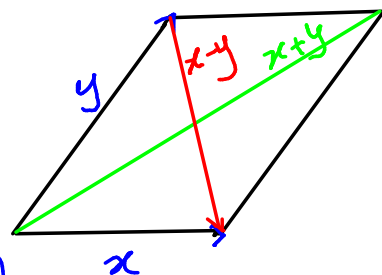
If  $\|x\|$  were defined using an inner product  $\sqrt{\langle x, x \rangle}$  then the following should hold (also called the parallelogram law)

$$\|x\|^2 + \|y\|^2 = \langle x, x \rangle + \langle y, y \rangle$$

$$= \frac{1}{2} \left( \langle x, x \rangle + \langle x, y \rangle + \langle y, x \rangle + \langle y, y \rangle \right) \rightarrow \|x+y\|^2$$

$$+ \frac{1}{2} \left( \langle x, x \rangle - \langle x, y \rangle - \langle y, x \rangle + \langle y, y \rangle \right) \rightarrow \|x-y\|^2$$

$$= \frac{1}{2} \left( \|x+y\|^2 + \|x-y\|^2 \right)$$



Now: let  $x = [a, a, 0, \dots, 0]$   $y = [a, -a, 0, \dots, 0]$

Then:  $\|x+y\|_p = \left\| \begin{bmatrix} 2a \\ 0 \\ \vdots \\ 0 \end{bmatrix} \right\|_p = (|2a|^p)^{1/p} = 2|a|$

$$\|x-y\|_p = \left\| \begin{bmatrix} 0 \\ 2a \\ 0 \\ \vdots \\ 0 \end{bmatrix} \right\|_p = (|2a|^p)^{1/p} = 2|a|$$

$$\|x\|_p = \left\| \begin{bmatrix} a \\ a \\ 0 \\ \vdots \\ 0 \end{bmatrix} \right\|_p = (|a|^p + |a|^p)^{1/p} = 2^{1/p} |a|$$

$$\|y\|_p = \left\| \begin{bmatrix} a \\ -a \\ 0 \\ \vdots \\ 0 \end{bmatrix} \right\|_p = (|a|^p + |-a|^p)^{1/p} = 2^{1/p} |a|$$

For the parallelogram law to be satisfied

$$2 * 2^{2/p} |a|^2 = \frac{1}{2} * 2 * 2^2 |a|^2$$

$$\|x\|_p^2 + \|y\|_p^2 = \frac{1}{2} (\|x+y\|_p^2 + \|x-y\|_p^2)$$

i.e.  $2 * 2^{2/p} = 4$

i.e.  $2^{2/p} = 2$  which holds iff  $p=2$

Thus:  $\|x\|_p$  corresponds to an inner product  $\sqrt{\langle x, x \rangle}$

iff  $p=2$

iff since  $\exists \sqrt{\langle x, x \rangle} = \|x\|_2$   
 where  $\langle x, y \rangle = \sum_{i=1}^{\infty} x_i y_i$

## ⑥ Polynomial vector spaces

$P^n(x)$  = set of polynomials with coefficients in  $S$  with degree  $\leq n$  in a single variable  $x = \left\{ \sum_{i=0}^n s_i x^i \mid s_i \in S \right\}$

↳ Can be viewed as vector space  $S^{n+1}$  and dimension =  $n+1$

$P(x)$  = Set of polynomials with coefficients in  $S$  with no degree restriction but in a single variable  $x$

↳ Can be viewed as vector space of infinite sequences... All discussions (norm, inner product etc follow)

$P(x_1, \dots, x_k)$  = Set of polynomials in  $k$  variables

↳ Result similar to  $P(x) / P^n(x)$ ?

↳ How about finite/infinite sequences of matrices

$$P(x_1, \dots, x_k) = \begin{bmatrix} a_{11} \\ a_{12} \\ \vdots \end{bmatrix}$$



③ Function spaces:  $f: X \rightarrow V$  ( $V$  is vector space over  $S$ )

Basis  $= \{f_{ij}\}$   $\left\{ \begin{array}{l} \hookrightarrow \text{If } X \text{ is finite \& } V \text{ is finite dimensional,} \\ \text{then } f \text{ has dimension } |X| \dim(V) \end{array} \right.$

$\left. \begin{array}{l} f_{ij}(x_i) = v_j \\ f_{ij}(x_k) = 0 \\ \{v_1, \dots, v_t\} = \text{Basis}(V) \end{array} \right\} \hookrightarrow \text{If } X \text{ is finite \& } V \text{ is countably infinite dimensional then } \dim f \text{ is countably infinite (similarly construct basis)}$

$\hookrightarrow$  Else  $\dim f$  is uncountably infinite (Prove H/W)

$\hookrightarrow$  Normed spaces:

$$L^p = \left\{ f \mid f: X \rightarrow S, \|f\|_p = \left( \int_X |f(x)|^p dx \right)^{1/p} < \infty \right\}$$

Should be structure preserving between domain & range }  $S = \mathbb{C}$  or  $\mathbb{R}$   
i.e. measurable

$L^p$  is Banach for  $p \geq 1$

$L^p$  is Hilbert only for  $p=2$

$$\langle f, g \rangle = \int_X f(x) \overline{g(x)} dx$$

## ④ Vector space of $m \times n$ matrices

- Show that the following are vector spaces (assuming scalars come from a set  $S$ ), and then answer questions that follow for each of them: Set of all matrices on  $S$ , set of all polynomials on  $S$ , set of all sequences of elements of  $S$ . (HINT: You can refer to [this book](#) for answers to most questions in this homework.) How would you understand the concepts of independence, span, basis, dimension and null space (chapter 2 of [this book](#)), eigenvalues and eigenvectors (chapter 5), inner product and orthogonality (chapter 6)? EXTRA: Now how about set of all random variables and set of all functions.

Let us consider space of matrices:

$$\left\{ \begin{bmatrix} s_{11} & \dots & s_{1m} \\ \vdots & & \vdots \\ s_{n1} & \dots & s_{nm} \end{bmatrix} \mid s_{11} \dots s_{nm} \in S \right\} \text{ over scalars } S$$

So far we considered  $S = \mathbb{R}$

Obvious that this is a vector space (since multiplication etc are defined on  $S$ )  
For simplicity, let  $S = \mathbb{R}$  & let us consider a norms for matrices, induced by norms for vectors

Let  $N(x)$  be a vector norm satisfying the vector norm axioms:

(Define  $\|A\| = f(A, N(x))$  for any/all  $x \in \mathbb{R}^n$ )

Then we will define a **matrix norm**

$$\sup f(s) = \hat{f}$$

S&S if  $\hat{f}$  is minimum upper bound

$$M_N(A) = \sup_{x \neq 0} \frac{N(Ax)}{N(x)}$$

Can you prove that this is indeed a valid norm?

as the matrix norm induced by  $N(x)$

what, for example, will be

$M_N(I) \rightarrow \text{Ans: } 1$

irrespective of  $N(x)$ ?

examples

(a) If  $N(x) = \|x\|_1 = \sum_{i=1}^n |x_i|$

$\|x\|_p = \left( \sum_{i=1}^n |x_i|^p \right)^{1/p}$

$\|A\|_1 = ?$

Ans:  $\|Ax\|_1 = \sum_{i=1}^n \left| \sum_{j=1}^m a_{ij} x_j \right| \leq \sum_{i=1}^n \sum_{j=1}^m |a_{ij}| |x_j|$

abs value of sum  $\leq$  sum of abs values

Changing order of summation:

$\|Ax\|_1 \leq \sum_{j=1}^m |x_j| \sum_{i=1}^n |a_{ij}|$

Let  $C = \max_j \sum_{i=1}^n |a_{ij}|$

Then  $\|Ax\|_1 \leq C \|x\|_1$

$$\Rightarrow \|A\|_1 = \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} \leq C$$

But consider an  $x = [0 \dots 0 \cdot 1 \cdot 0 \dots 0]$

$k^{\text{th}}$  position, where  $k$  is column index  $j$  for which

$$C = \sum_{i=1}^n |a_{ik}|$$

Then  $\|x\|_1 = 1$  &  $\|Ax\|_1 = C$  (Show this)

$$\Rightarrow \|A\|_1 = \max_j \sum_{i=1}^n |a_{ij}| \quad \text{i.e. if } N(x) = \|x\|_1 \text{ then } M_N(A) = \max_j \sum_{i=1}^n |a_{ij}|$$

(b) Similarly,

$$\text{if } N(x) = \|x\|_2 = \left[ \sum_{i=1}^n |x_i|^2 \right]^{1/2}$$

$$\|A\|_2 = \left[ \text{dominant eigenvalue of } A^T A \right]^{1/2}$$

(c) If  $N(x) = \|x\|_\infty = \max_i |x_i|$

$$\|A\|_\infty = \max_i \sum_{j=1}^m |a_{ij}|$$

$$\left\{ \lim_{p \rightarrow \infty} \left( \sum_{i=1}^n |x_i|^p \right)^{1/p} \right.$$

Other matrix norms:

$$\|A\|_F = \sqrt{\sum_{i,j} a_{ij}^2}$$

Frobenius norm =  $\sqrt{\text{Trace}(A^T A)}$

→ special case

- ① If  $A$  is symmetric:  $\|A\|_F^2 = \lambda_1(A)^2 + \lambda_2(A)^2 + \dots + \lambda_n(A)^2$
- ② For general  $A$ :  $\|A\|_F^2 = \sigma_1(A)^2 + \dots + \sigma_k(A)^2$   
rank( $A$ ) =  $k$ .

Q: What abt inner products: (By virtue of trace)

Note: Not all normed spaces are inner prod spaces.

Eg:  $\|x\|_p = \left(\sum_i |x_i|^p\right)^{1/p}$  for  $p=2$   
 $\langle x, y \rangle = \sum_i x_i y_i$

For  $p=1$  or  $\infty$ ,  
No corresp. inner products

Read more on

[http://www.math.ucsd.edu/~njw/Teaching/Math271C/Lecture\\_04.pdf](http://www.math.ucsd.edu/~njw/Teaching/Math271C/Lecture_04.pdf)

Eg of Frobenius inner product:

$$\langle A, B \rangle = \sum_i \sum_j a_{ij} b_{ij}$$

Weighted inner product

$$\langle A, B \rangle_w = \sum_i \sum_j a_{ij} b_{ij} w_{ij} \text{ for } w_{ij} > 0$$

Hint: First prove (2)

You might use the orthonormal eigenvectors of  $A^T A$  as basis for column space of  $A$  and use this trick like in a previous lecture

Singular values & Eigenvalues of  $A$

$$Au = \sigma v$$

$$A^*v = \sigma u$$

$\Downarrow$

$$A^*Au = \sigma^2 u$$

i.e.  $\sigma^2$  is an eigenvalue  
of  $A^*A$

$$Au = \lambda u$$

$$\text{if } A^T = A \text{ or } A^* = A$$

$$A^*Au = \lambda^2 u \\ = \sigma^2 u$$

Basis for vector space of matrices ( $m \times n$ )

$$\left\{ \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & \dots & 0 & \end{bmatrix} \begin{bmatrix} 0 & 1 & \dots & 0 \\ 0 & 0 & \dots & \\ 0 & & & a \end{bmatrix} \dots \begin{bmatrix} 0 & \dots & 0 \\ \vdots & & \\ 0 & \dots & 1 \end{bmatrix} \right\}$$

$E_{11}$                        $E_{12}$                        $E_{mn}$

$m \times n$  linearly independent elements  
that span the space of all matrices  
of size  $m \times n$

$$B = \sum_{i,j} a_{ij} E_{ij} \equiv \begin{bmatrix} a_{11} \\ a_{12} \\ \vdots \\ a_{mn} \end{bmatrix}$$

This vector  $\in \mathbb{R}^{\frac{mn}{5}}$   
is a canonical  
representation of  $B$



$l_p(S)$  = set of sequences with scalar field  $S$   
that have finite  $p$ -norm

$$\|x\|_p = \left( \sum_{i=1}^{\infty} x_i^p \right)^{1/p}$$

$L_p(S)$  = set of measurable functions

$f: X \rightarrow S$  with finite  $p$ -norm

$$\|f\|_p = \left( \int_{x \in X} f(x)^p dx \right)^{1/p}$$

## RECALL

A linear map/linear operator  $T$  between two vector spaces  $X$  &  $Y$  is  $T: X \rightarrow Y$  s.t

$$T(\lambda x + \mu x') = \lambda T x + \mu T x'$$

$$\forall \lambda, \mu \in S$$

$$\forall x, x' \in X$$

If  $T$  is 1-1 & onto then  $T$  is called invertible.  $T^{-1}$  is defined s.t

$$T^{-1}: Y \rightarrow X \quad \text{s.t.} \quad T^{-1}y = x \quad \text{iff} \quad Tx = y$$

(a) if  $X$  &  $Y$  are normed spaces

$$\|T\| = \sup_{\substack{x \neq 0 \\ x \in X}} \frac{\|Tx\|}{\|x\|} \quad \text{is called operator norm}$$

$T$  is called bounded if:

$$\exists N \text{ s.t. } \|T\| \leq N \iff \exists M \text{ s.t. } \|Tx\| \leq M\|x\|$$

$T$  is bounded iff it is continuous  $\forall x \in X$

Proof: Recall that  $T$  is called continuous if given any  $\epsilon > 0$ ,  
 $\exists$  a  $\delta > 0$  s.t. whenever

$$\|x - x'\| \leq \delta \quad \|Tx - Tx'\| \leq \epsilon \quad x, x' \in X$$

(a) Suppose  $T: X \rightarrow Y$  is bounded. Then

$\forall x, x' \in X$  we have

$$\|Tx - Tx'\| = \|T(x - x')\| \leq M\|x - x'\|$$