

Measure and Integration

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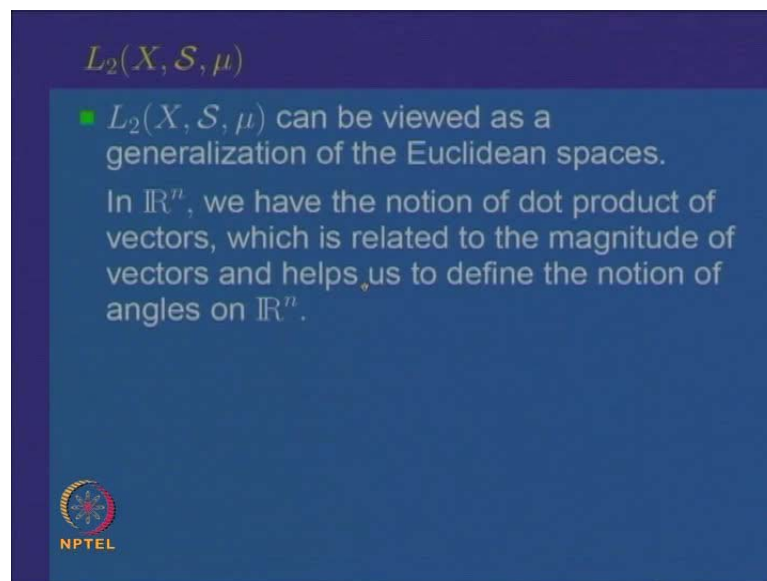
Module No. # 09

Lecture No. # 35

$L_2(X, S, \mu)$

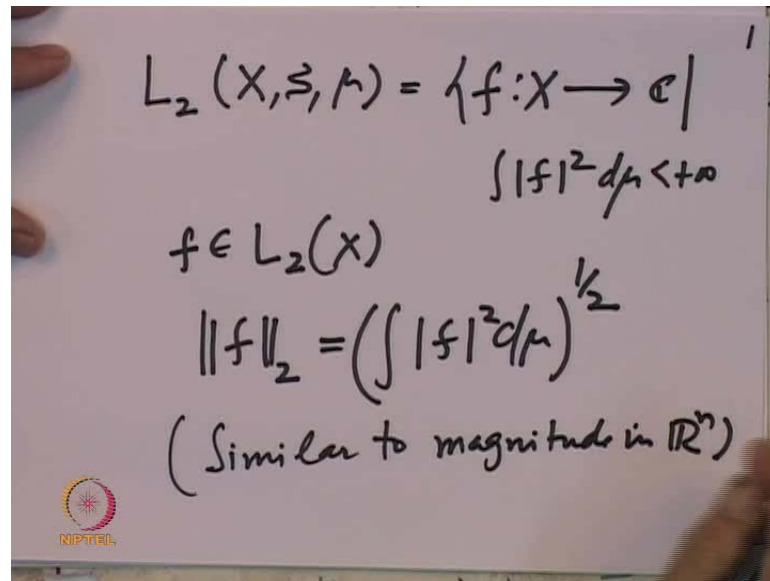
Welcome to lecture 35 on Measure and Integration. In the previous lectures, we had been looking at the p power integrable functions on a measure space X, S, μ and we have studied some general properties of this function spaces. Today, we will look slightly more in detail the special subspace namely, when p is equal to 2.

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So, the topic for today's discussion would be space $L_2(X, S, \mu)$ the space of complex valued square integrable functions. This is a special space in the sense that this can be viewed as a generalization of the Euclidean spaces. Recall in \mathbb{R}^n , the Euclidean space we have the notion of dot product of vectors which is related to the magnitude of the vectors and also help us to define the notion of angles on \mathbb{R}^n .

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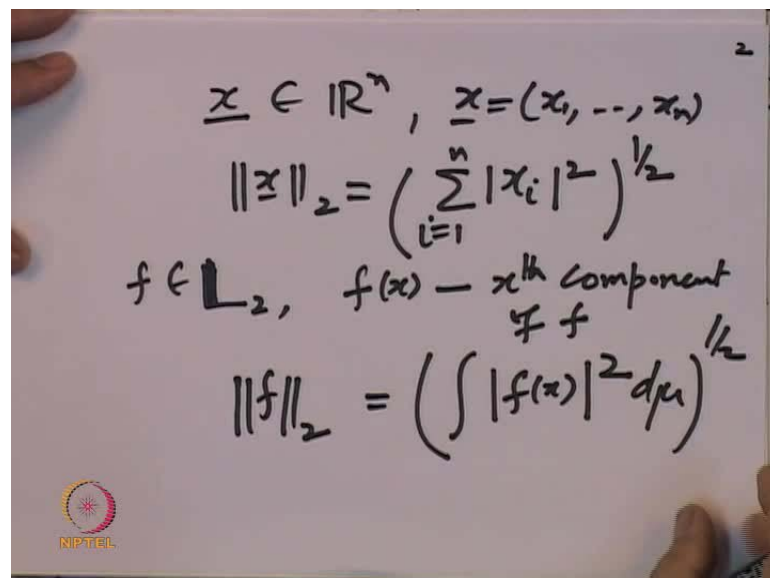
Handwritten mathematical definitions for the L_2 space on a whiteboard. The text includes the definition of $L_2(X, \mathcal{S}, \mu)$ as the set of functions $f: X \rightarrow \mathbb{C}$ such that $\int |f|^2 d\mu < +\infty$. It also states $f \in L_2(X)$ and provides the formula for the norm $\|f\|_2 = \left(\int |f|^2 d\mu \right)^{1/2}$, noting that this is similar to the magnitude in \mathbb{R}^n . The NPTEL logo is visible in the bottom left corner.

$$L_2(X, \mathcal{S}, \mu) = \{f: X \rightarrow \mathbb{C} \mid \int |f|^2 d\mu < +\infty\}$$
$$f \in L_2(X)$$
$$\|f\|_2 = \left(\int |f|^2 d\mu \right)^{1/2}$$

(Similar to magnitude in \mathbb{R}^n)

Let us see, what we can do as far as L_2 is concerned. So, we look at the space L_2 of X , \mathcal{S} , μ to be the space of all functions, which are defined on X which are complex valued, such that $\int |f|^2 d\mu$ is finite. For such spaces, for a function f in L_2 of X , we have defined the norm of this which is equal to integral of $|f|^2 d\mu$ raised to power $1/2$. We said that this is very much like the magnitude in \mathbb{R}^n , so this is similar to magnitude in \mathbb{R}^n .

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Handwritten mathematical definitions for the L_2 norm in \mathbb{R}^n on a whiteboard. It defines $\underline{x} \in \mathbb{R}^n$ as $\underline{x} = (x_1, \dots, x_n)$ and gives the formula for the norm $\|\underline{x}\|_2 = \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2}$. It also notes that $f \in L_2$ and $f(x)$ is the x^{th} component of f , leading to the formula $\|f\|_2 = \left(\int |f(x)|^2 d\mu \right)^{1/2}$. The NPTEL logo is visible in the bottom left corner.

$$\underline{x} \in \mathbb{R}^n, \underline{x} = (x_1, \dots, x_n)$$
$$\|\underline{x}\|_2 = \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2}$$

$f \in L_2$, $f(x)$ — x^{th} component of f

$$\|f\|_2 = \left(\int |f(x)|^2 d\mu \right)^{1/2}$$

Let us just briefly recall what the magnitude in \mathbb{R}^n was. So, for a vector say x in \mathbb{R}^n the Euclidean norm is defined as - look at the components - if x is having components x_1 to x_n then look at mod of x_i square sigma 1 to n and the square root of that and for a function f in L^2 ; let us regard f as a vector with $f(x)$ as the component, so this is the x th component of f .

How would we define the norm? Look at the x th component $f(x)^2$, so this is very much similar to what we have done for Euclidean look at the component square and now sum it up; but here, the summation is over x and any x belonging to X any indexing set. So, it is nothing but integral $d\mu$ and then the power $1/2$. In that sense this is a perfect generalization of the Euclidean norm to arbitrary spaces.

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$$\begin{aligned} \underline{x} \in \mathbb{R}^n, \quad \underline{x} &= (x_1, \dots, x_n) \\ \underline{y} &= (y_1, \dots, y_n) \\ \langle \underline{x}, \underline{y} \rangle &= \sum_{i=1}^n x_i y_i \\ \underline{x}, \underline{y} \in \mathbb{C}^n \\ \langle \underline{x}, \underline{y} \rangle &= \sum_{i=1}^n x_i \bar{y}_i \\ \|\underline{x}\|^2 &= \langle \underline{x}, \underline{x} \rangle \end{aligned}$$

Now on L^2 , for a vector x belonging to \mathbb{R}^n with components x_i , we have the notion of what is called the dot product; so dot product if x and y are two vectors with y components as y_i , then x comma y the dot product is defined as $x_i y_i$ and i equal to 1 to n .

If we are in complex plane, if x and y belong to \mathbb{C}^n then, the dot product $x \cdot y$ is defined as sigma $x_i y_i$ bar, i equal to 1 to n . This notion of dot product is related to the magnitude in the following way that in either \mathbb{R}^n or \mathbb{C}^n either one, the norm of x square is equal to the dot product of x with itself and we know that is the dot product in \mathbb{R}^n or \mathbb{C}^n gives the notion of the angle and orthogonality which helps us to do geometry in \mathbb{R}^n .

The basic idea of today's lecture would be on the space L^2 of x , we already have the notion of norm; the notion of distance. We will define the notion of inner product or the notion of dot product on L^2 of x and show how it is related to the notion of distance. So that helps us to define the notion of orthogonality for perpendicularity of two elements in L^2 . So, we can do geometry in L^2 of X, \mathcal{S}, μ .

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Inner product on L_2

- For $f, g \in L_2(X, \mathcal{S}, \mu)$, define

$$\langle f, g \rangle := \int f(x) \overline{g(x)} d\mu(x),$$

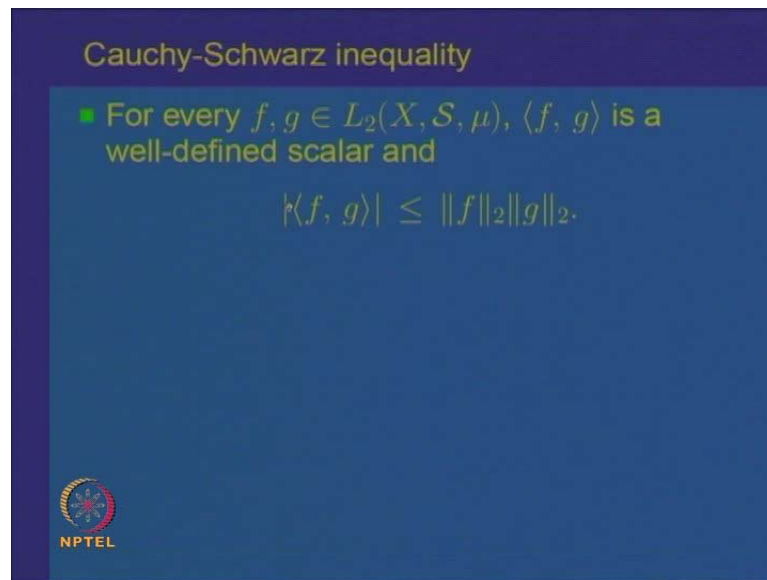
whenever it exists, where \overline{g} denote the complex conjugate function:

$$\overline{g}(x) := \overline{g(x)} \quad \forall x \in X.$$

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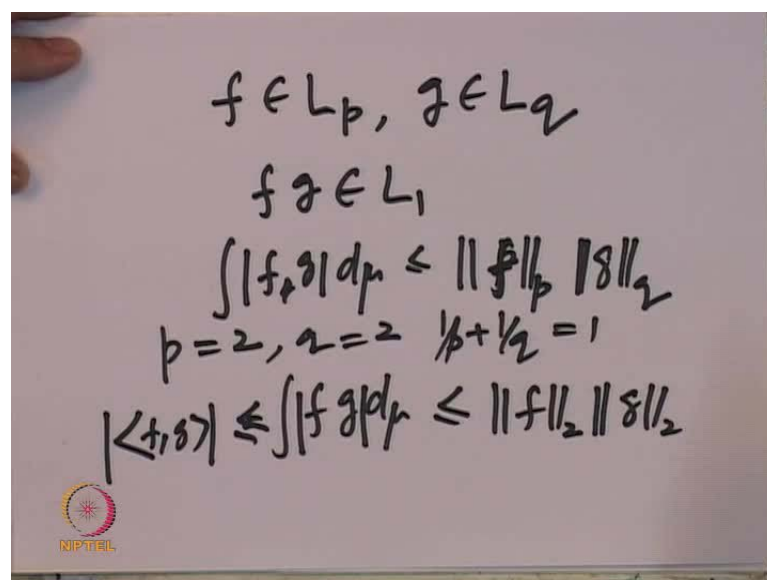
Let us define, what is the notion of the dot product in L^2 ? For functions f and g in L^2 , keep in mind our spaces are complex valued, so define the dot product of f with g are also called the inner product of f with g as integral of $f(x) \overline{g(x)} d\mu(x)$, where this \overline{g} is the complex conjugate of the function g , so $\overline{g(x)}$ is $\overline{g(x)}$. The inner product or the dot product between f and g ; for two functions f and g is defined as the integral of $f(x) \overline{g(x)} d\mu(x)$, which is perfectly similar to complex C to the power n $a + ib$ $\overline{a + ib}$.

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The first thing we want to say that this is well defined and that follows from the holders inequality that we had proved - recall holders inequality - it is said that if f is a function in L^p and g is a function in L^q then, $f \cdot g$ is function which is integrable and the integral of $f \cdot g$ is less than or equal to the L^2 norm. So, let me just say it once again, what we are meaning here - that should be integral.

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For holder inequality, we had that if f belongs to L^p and g belongs to L^q then, fg belongs to L^1 and integral of mod of f into g $d\mu$ is less than or equal to the p th norm

of f and the q th power of the norm of g . For p equal to 2, so that will give us that integral of $f \bar{g}$ $d\mu$ and p is equal to 2, q is equal to 2, so we have got that $1/p + 1/q$ is equal to 1. So, the Hölder's inequality will give us that this is less than or equal to L^2 norm of f into L^2 norm of g . This is bigger than or equal to $f \bar{g}$ inner product, is less than or equal to this, so that gives us the Cauchy-Schwarz inequality.

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Cauchy-Schwarz inequality

- For every $f, g \in L_2(X, \mathcal{S}, \mu)$, $\langle f, g \rangle$ is a well-defined scalar and

$$|\langle f, g \rangle| \leq \|f\|_2 \|g\|_2.$$



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Inner product on L_2

- For $f, g \in L_2(X, \mathcal{S}, \mu)$, define

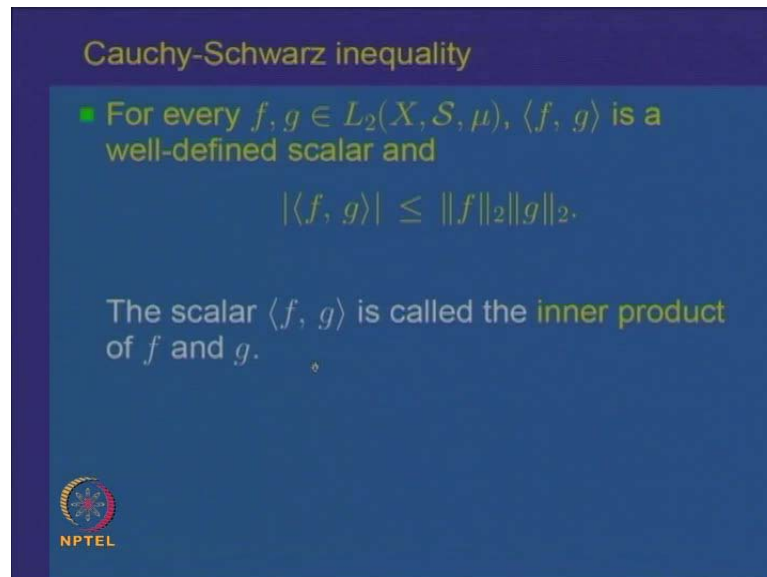
$$\langle f, g \rangle := \int f(x) \bar{g}(x) d\mu(x),$$

whenever it exists, where \bar{g} denote the complex conjugate function:

$$\bar{g}(x) := \overline{g(x)} \quad \forall x \in X.$$


So, this is also called the Cauchy-Schwarz inequality namely, the absolute value of the inner product between f and g is less than or equal to the norm of f into norm of g . So that says that this inner product is well-defined; it is a well-defined in quantity.


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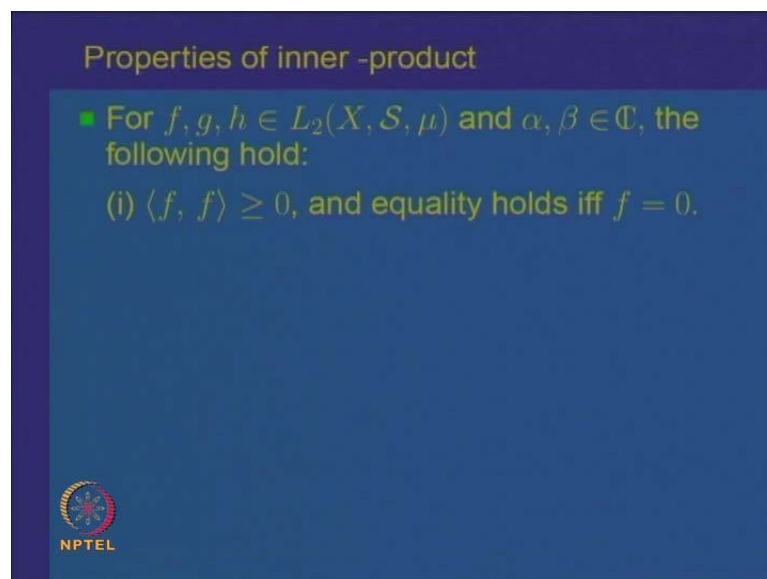
Cauchy-Schwarz inequality

- For every $f, g \in L_2(X, \mathcal{S}, \mu)$, $\langle f, g \rangle$ is a well-defined scalar and
$$|\langle f, g \rangle| \leq \|f\|_2 \|g\|_2.$$

The scalar $\langle f, g \rangle$ is called the **inner product** of f and g .




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Properties of inner -product

- For $f, g, h \in L_2(X, \mathcal{S}, \mu)$ and $\alpha, \beta \in \mathbb{C}$, the following hold:
 - (i) $\langle f, f \rangle \geq 0$, and equality holds iff $f = 0$.



For every f and g in L_2 , we have the notion of the inner product, so f comma g which is the inner product is well defined quantity. This behaves perfectly similar to that of the inner product for ordinary vectors in \mathbb{R}^n or \mathbb{C}^n that means, this is a function which is defined on L_2 cross L_2 . It has the following properties namely the inner product of f

with f is always bigger than or equal to 0 and the equality holds, if and only if, f is equal to 0.

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$$\begin{aligned} \langle f, f \rangle &= \int |f|^2 d\mu \geq 0 \\ &= 0 \Leftrightarrow |f| = 0 \text{ a.e.} \\ &\Leftrightarrow f \in L_2, f = 0. \\ \langle f, g \rangle &= \int f \bar{g} d\mu = \overline{\int \bar{f} g d\mu} \\ &= \overline{\langle \bar{f}, g \rangle} = \langle g, f \rangle \end{aligned}$$

Let us look at this property that how is this true. So, inner product of f with itself is nothing but integral of mod f square $d\mu$. This is always bigger than or equal to 0 and this will be equal to 0, if and only if, our function mod f is equal to 0 almost everywhere. So, if and only if, f is belonging to L_2 as I treated as element of L_2 f is equal to 0.

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Properties of inner -product

- For $f, g, h \in L_2(X, \mathcal{S}, \mu)$ and $\alpha, \beta \in \mathbb{C}$, the following hold:
 - (i) $\langle f, f \rangle \geq 0$, and equality holds iff $f = 0$.
 - (ii) $\langle f, g \rangle = \overline{\langle g, f \rangle}$.
 - (iii) $\langle \alpha f + \beta g, h \rangle = \alpha \langle f, h \rangle + \beta \langle g, h \rangle$.
 - (iv) $\langle f, \alpha g + \beta h \rangle = \bar{\alpha} \langle f, g \rangle + \bar{\beta} \langle f, h \rangle$.
 - (v) $\|f\|_2 = \langle f, f \rangle^{1/2}$.

The first property is obvious; namely, the dot product of f with itself is always bigger than or equal to 0. The second property says, the dot product or the inner product of f with g is same as the inner product of g with f , so we are interchanging f comma g and the complex conjugate of it.

The inner product of f with g is same as the complex conjugate of the inner product of g with f and that is quite simple to verify from the definition. So, if we have got the inner product of f with g ; so inner product of f with g is equal to $\int f \bar{g} d\mu$ and that is equal to $\int \bar{g} f d\mu$ and this is equal to $\int g \bar{f} d\mu$. So, that verifies the property namely, the inner product of f with g is equal to the inner product of g with f bar.

Similarly, it is easy to verify using that the integral is linear. It is easy to verify that the inner product is linear in the first variable that means, αf plus βg inner product with h is equal to α times the inner product of f with h plus β times the inner product of g with h .


Similarly, in the second variable is this conjugate linear because of the property of 2. So, f inner product with αg plus βh is same as α bar of f g inner product f g plus β bar of inner product of f h . In the second variable, it is not linear, it is a conjugate linear. Finally, this property is obvious namely, the L_2 norm of f is square root of the inner product of f with itself.

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Inner-product spaces

- An arbitrary vector space H over the field \mathbb{R} (or \mathbb{C}), with a map


$$\langle \cdot, \cdot \rangle : H \times H \longrightarrow \mathbb{R}$$
 (or \mathbb{C}), having the properties (i) to (iv) as for L_2 , is called an **inner product space**.



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Properties of inner-product


- For $f, g, h \in L_2(X, \mathcal{S}, \mu)$ and $\alpha, \beta \in \mathbb{C}$, the following hold:
 - (i) $\langle f, f \rangle \geq 0$, and equality holds iff $f = 0$.
 - (ii) $\langle f, g \rangle = \overline{\langle g, f \rangle}$.
 - (iii) $\langle \alpha f + \beta g, h \rangle = \alpha \langle f, h \rangle + \beta \langle g, h \rangle$.
 - (iv) $\langle f, \alpha g + \beta h \rangle = \overline{\alpha} \langle f, g \rangle + \overline{\beta} \langle f, h \rangle$.
 - (v) $\|f\|_2 = \langle f, f \rangle^{1/2}$.



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Inner-product spaces

- An arbitrary vector space H over the field \mathbb{R} (or \mathbb{C}), with a map
$$\langle \cdot, \cdot \rangle : H \times H \longrightarrow \mathbb{R}$$
(or \mathbb{C}), having the properties (i) to (iv) as for L_2 , is called an **inner product space**.
On every inner product space H , it is easy to show that
$$\|u\| := \langle u, u \rangle^{1/2}, u \in H,$$
is a norm on H , called the **norm induced** by the inner product.

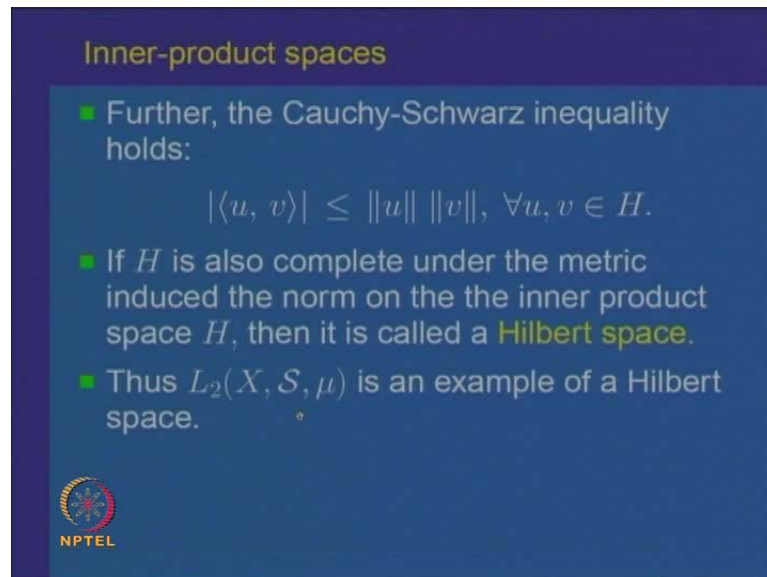


All the properties that we have for the dot product in \mathbb{R}^n or \mathbb{C}^n are defined for the inner product in L_2 . This is not very special for L_2 in fact, one can look at any vector space H over the field of real or complex numbers. If one has a function which is defined on H cross H taking values in the underlying field of real or complex having properties similar to that I, II, III, IV, V of L_2 , so one can define what is called an inner product space.

In general an inner product space is defined to be a vector space H on which there is a notion of inner product defined. What is an inner product? It is a function defined on H


cross H to \mathbb{R} with those properties. Once we have a notion of inner product that gives rise to notion of magnitude, by the property that the magnitude of a vector u in H is nothing but defined as the dot product of u with square root itself.

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Inner-product spaces

- Further, the Cauchy-Schwarz inequality holds:
$$|\langle u, v \rangle| \leq \|u\| \|v\|, \forall u, v \in H.$$
- If H is also complete under the metric induced the norm on the the inner product space H , then it is called a **Hilbert space**.
- Thus $L_2(X, \mathcal{S}, \mu)$ is an example of a Hilbert space.


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One verifies that Cauchy-Schwarz inequality holds for this kind of inner product and that means, this is a well-defined norm, so Cauchy-Schwarz inequality will say that is a norm defined on it. Once, you have the notion of norm that gives rise to a metric on the underline vector space and one can ask whether it is complete under that metric or not.

On a vector space inner part is defined, so it becomes an inner product space and inner product space gives rise to a norm and if the underlying metric induced by the norm is complete one says, H is a Hilbert space, that is a general definition of a Hilbert space.

So, our L_2 is an example of a Hilbert space, because $L_2(X, \mathcal{S}, \mu)$ is a vector space on which a notion of norm - the L_2 norm - is defined and that L_2 norm is related to the inner product, just now we have seen. We have already seen as Riesz-Fischer theorem, which said that $L_2(X, \mathcal{S}, \mu)$ is a complete metric space in the L_2 metric, so this is an example of a Hilbert space.

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Orthogonality

- Functions $f, g \in L_2(X, \mathcal{S}, \mu)$, are said to be **orthogonal** if
$$\langle f, g \rangle = 0.$$
and write $f \perp g$.
- For $f \in L_2(X, \mathcal{S}, \mu)$, and S a subset of $L_2(X, \mathcal{B}, \mu)$, we write
$$f \perp S \text{ if } \langle f, h \rangle = 0 \forall h \in S.$$
- **Pythagoras' Identity:** If $f, g \in L_2(X)$ and $f \perp g$, then $\|f + g\|_2^2 = \|f\|_2^2 + \|g\|_2^2$.

Once we have the notion of the inner product, one can define the notion of two elements in L_2 to be orthogonal or perpendicular to each other. So, we say two elements f and g in L_2 are orthogonal to each other, if the inner product between them is equal to 0 and that is what we have for vectors in \mathbb{R}^n also that the dot product is equal to 0.

We write this as f perpendicular to g ; so f perpendicular to g is defined as saying that the inner product of f with g is equal to 0. We can also define the inner product if element f orthogonal to a subset S , so writing it as f orthogonal to a subset S means that f is perpendicular to every element of S .

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A photograph of a whiteboard with handwritten mathematical equations. The equations are: $f, g \in L_2, f \perp g$; $\|f+g\|_2^2 = \langle f+g, f+g \rangle$; $= \langle f, f \rangle + \langle g, f \rangle + \langle f, g \rangle + \langle g, g \rangle$; $= \|f\|_2^2 + \|g\|_2^2$. In the bottom left corner of the whiteboard, there is a small circular logo with a star and the text 'NPTEL' below it.

$$\begin{aligned} f, g \in L_2, f \perp g \\ \|f+g\|_2^2 &= \langle f+g, f+g \rangle \\ &= \langle f, f \rangle + \langle g, f \rangle \\ &\quad + \langle f, g \rangle + \langle g, g \rangle \\ &= \|f\|_2^2 + \|g\|_2^2 \end{aligned}$$

So, f is perpendicular to S will mean that f comma h , the inner product is equal to 0 for every element h in S . Similarly, we can define orthogonality of two sets also. With this one can prove what is called the Pythagoras identity namely, if f and g are two functions in L_2 and f is orthogonal to g then, the norm square of f plus g is equal to norm f square plus norm g square.


Let us just quickly verify the Pythagoras identity namely, if f and g are two elements in L_2 and f is orthogonal to g then, we want to compute the L_2 norm of square of this. By definition, this is related to the inner product, so this is inner product of f plus g with itself. Now, using the property of linearity what we will get is, this is f comma f plus g comma f plus f comma g plus the inner product of g with itself.

So, that gives you norm of f square plus norm of - the last term will give you - norm of g square, but f is orthogonal to g that means, f comma g is equal to 0 and g comma f is also equal to 0. So these two terms, the inner products of g with f and f with g both are equal to 0. So, we get what is called the Pythagoras identity namely, the norm of f plus g square is equal to norm f square plus norm g square, whenever f is orthogonal to g .

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Closed subspaces

- Let S be a nonempty subset of $L_2(X, \mathcal{S}, \mu)$.
- (i) We say S is a **subspace** of $L_2(X, \mathcal{S}, \mu)$ if
$$\alpha f + \beta g \in S$$
for all $\alpha, \beta \in \mathbb{C}$ and $f, g \in L_2(X, \mathcal{B}, \mu)$.
- (ii) A subspace S is called a **closed subspace** if it is closed under the L_2 - metric, i.e., for every sequence $\{f_n\}_{n \geq 1}$ in S with
$$\lim_{n \rightarrow \infty} \|f_n - f\|_2 = 0$$
for some $f \in L_2(X, \mathcal{S}, \mu)$, we have $f \in S$.



Let us carry over this idea of orthogonality a bit further; let us take S any nonempty subset of L_2 . We call S a subspace of L_2 , those who have done a bit of linear algebra will recognize this definition - L_2 is a vector space, so we are looking at a vector subspace of L_2 . So, a set S is nonempty subset is called a subspace of L_2 , if for any f and g in S , so this should be in S not in L_2 and $\alpha\beta$ in complex number αf plus βg belongs to S , then we say S is a subspace, so this is not L_2 it is S .

So that means, for any two elements $\alpha\beta$ in S the linear combination αf plus βg should be in S , in that case it is called a subspace of S . A subspace of S is called a closed subspace, if it is closed under the metric on L_2 that is L_2 metric. So, it should be a closed set that means what? That means, whenever we are get a sequence f_n in S and f_n converges to a function f in L_2 norm, then the limit must also be inside S . So that is what is a definition of a closed over subspace. It is a subspace and it is a closed set under the L_2 metric, so that is the notion of a closed subspace.


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Orthogonal complement

Let S be any nonempty subset of $L_2(X, \mathcal{S}, \mu)$ and let

$$S^\perp := \{g \in L_2(X, \mathcal{S}, \mu) \mid \langle f, g \rangle = 0 \forall f \in S\}.$$

Then S^\perp is a closed subspace of $L_2(X, \mathcal{S}, \mu)$.



For given a set S in L_2 denoted by S upper suffix perpendicular - this is also called the orthogonal complement of S to be all elements in the space L_2 which are perpendicular - to all elements of S . Given a set S , we are looking at all elements in L_2 which are orthogonal to every element of S . So, that is called S perpendicular and this is called the orthogonal complement of S and the claim is that this is a closed subspace of L_2 .


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S a subset of L_2

$$S^\perp = \{f \in L_2 \mid f \perp h \forall h \in S\}$$

S^\perp is a subspace

$h, g \in S^\perp, \alpha, \beta \in \mathbb{C}, f \in S$

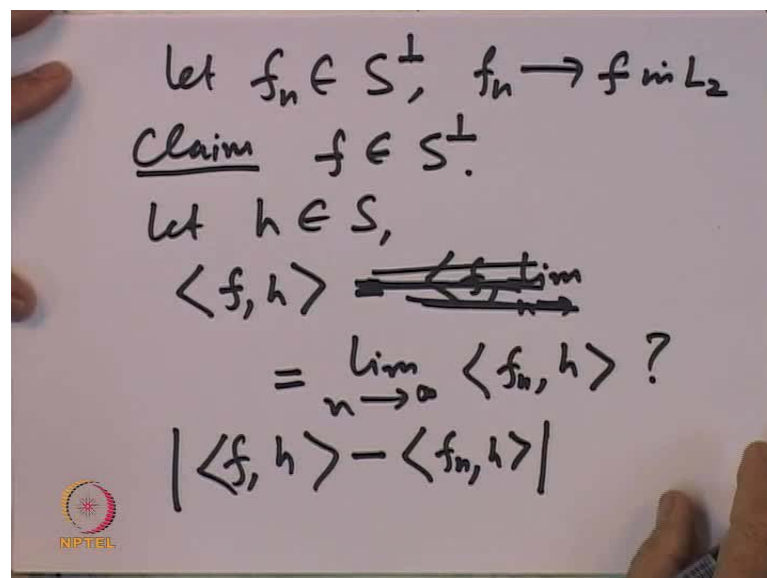
$$\langle \alpha h + \beta g, f \rangle$$
$$= \alpha \langle h, f \rangle + \beta \langle g, f \rangle$$
$$= 0$$


Let us verify this fact that S a subset of L_2 and S perpendicular is the set of all elements f in L_2 , such that f perpendicular to h for every h in S then, claim is that first of all S

perpendicular is a subspace. Let us take some h and g belonging to S perpendicular and α and β belonging to C . Then $\alpha h + \beta g$ - let us take an element f , for f in S this will be equal to α times $\langle f, h \rangle + \beta$ times $\langle f, g \rangle$ using the property of linearity in the first variable for the inner product.

Now, because f belongs to S and h, g are in S perpendicular, so this quantity inner product is 0 and the second inner product is 0, so the sum this inner product is equal to 0 (Refer Slide Time: 21:42). So that says, if h and g belong to S perpendicular and α and β are on the C , then $\alpha h + \beta g$ is always orthogonal to every element of S . Hence, it belongs to S perpendicular, so S perpendicular is a subspace.

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Next, let us prove that this is a closed subspace; we want to check that there is a closed subspace. So, let f_n belong to S perpendicular and f_n converge to f in L_2 , we want to check it. So, claim that f belongs to S perpendicular, for that let us take any element h belonging to S and we want to compute $\langle f, h \rangle$ and the claim is this $\langle f, h \rangle$ this is equal to $\langle f, h \rangle = \lim_{n \rightarrow \infty} \langle f_n, h \rangle$ - so what is h ? What is sorry not this is not true.

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$$\begin{aligned} &= |\langle f - f_n, h \rangle| \\ &\leq \|f - f_n\|_2 \|h\| \\ &\quad \downarrow 0 \\ \therefore \langle f, h \rangle &= \lim_{n \rightarrow \infty} \langle f_n, h \rangle \\ &= 0 \quad \forall h \in S \\ \Rightarrow f &\in S^\perp \end{aligned}$$

The whiteboard image shows a handwritten derivation. At the top, it states $= |\langle f - f_n, h \rangle|$. Below this is the inequality $\leq \|f - f_n\|_2 \|h\|$. A downward arrow with a '0' below it indicates the limit. The next line is $\therefore \langle f, h \rangle = \lim_{n \rightarrow \infty} \langle f_n, h \rangle$. Below that, it says $= 0 \quad \forall h \in S$. The final conclusion is $\Rightarrow f \in S^\perp$. In the bottom left corner of the whiteboard, there is a small circular logo with a star and the text 'NPTEL' below it.

Let h belongs to S , now the claim is that since, f_n converges to f in L^2 , so this is equal to limit n going to infinity of f_n comma h . So, this is a very simple thing to verify because, if we look at the difference of the two f and h , so why is this true? This is true because, we look this inner product of f with h and inner product of f_n with h and look at the absolute value of this then, we can write this as this; so this quantity is equal to absolute value of f minus f_n inner product with h . By Cauchy-Schwarz inequality, this is less than or equal L^2 norm of f minus f_n and L^2 norm of h and this goes to 0.

Therefore, we get f with h inner product is equal to limit n going to infinity inner product of f_n with h . Since, each f_n belongs to S perpendicular h is in S , so that implies that each term is equal to 0, so this is equal to 0 for every h belonging to S . So that implies that f belongs to S perpendicular. This proves that S perpendicular for any set S , if we look at its orthogonal complement then that is a closed subspace of h , so this is also called the orthogonal complement of a set.

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Best approximation

We next state an important result which seems geometrically obvious.


Theorem:
Let $f \in L_2(X, \mathcal{S}, \mu)$, and let S be a closed subspace of $L_2(X, \mathcal{S}, \mu)$ and

$$\alpha := \inf \{ \|f - g\|_2 \mid g \in S \}.$$

Then there exists a unique function $f_0 \in S$ such that

$$\alpha = \|f - f_0\|_2.$$

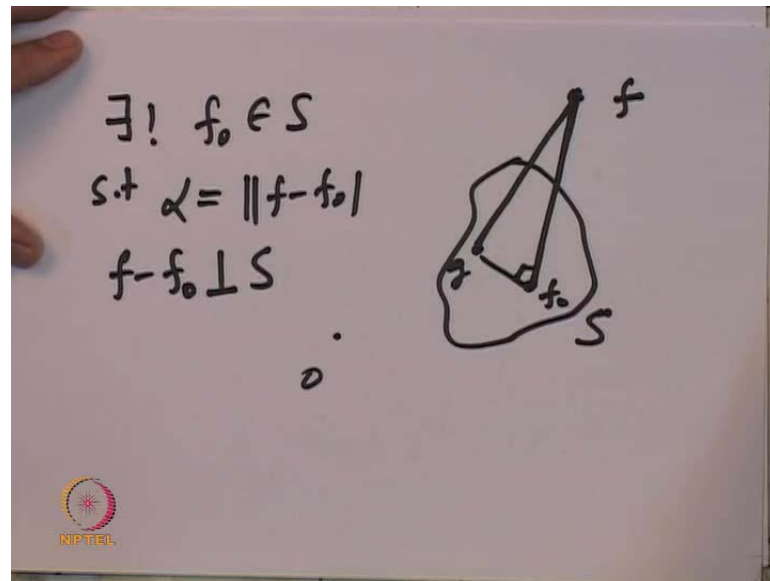
Further, if $f \notin S$ then $0 \neq (f - f_0) \perp S$.



We next state an important result which seems geometrically obvious, which can be proved for any Hilbert space, so we will just look at it for our space L_2 , will not prove this result, we will just assume this result the proof can be referred to the book.

The result says that if f is a L_2 function and S is a closed subspace of L_2 then, look at the number α which is the infimum of all the L_2 distances of f from g where g is any element in S . Then, the theorem says that this infimum is attained at some point in S , so that means there exists not only it is attained then there is a unique function f_0 belonging to S such that this infimum α is equal to norm of f minus f_0 . Further, if f does not belong to S then, look at the difference of f minus f_0 that is always going to be perpendicular to S , so that is the claim of the theorem.

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We will not prove it, we will just geometrically analyze this result a bit. So, look at the closed subspace S of L^2 , this is a closed subspace of L^2 and we have got a function f , which is outside it, what we are going to do is? We are going to look at any point inside S point g and look at the L^2 distance of this, so it says that there is a value called f_0 , such that L^2 distance of f from it is the minimum. If this is 0 says, if I look at f minus f_0 so that is going to be orthogonal to it, so that is the theorem.

There exist a unique point f_0 belonging to S . Such that α the infimum is equal to the distance of f minus f_0 and f minus f_0 is perpendicular to S . Look at this vector f minus f_0 that is always orthogonal to this S . Geometrically in a sense given a point and given a subspace it is kind of the projection gives you the minimum. So, this is a generalization of the projection theorem for finite dimensional spaces.

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Best approximation

We next state an important result which seems geometrically obvious.


Theorem:
Let $f \in L_2(X, \mathcal{S}, \mu)$, and let S be a closed subspace of $L_2(X, \mathcal{S}, \mu)$ and

$$\alpha := \inf \{ \|f - g\|_2 \mid g \in S \}.$$

Then there exists a unique function $f_0 \in S$ such that

$$\alpha = \|f - f_0\|_2.$$

Further, if $f \notin S$ then, $0 \neq (f - f_0) \perp S$.



This is also called the best approximation theorem for Hilbert spaces. Let us look at once again, it says that if given a closed subspace of L_2 - a function f in L_2 - look at α the minimum of the distances between f and elements of S says, there is value there is a function in S , where this value is attained.


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Consequences

- Let S be a proper closed subspace of $L_2(X, \mathcal{B}, \mu)$. Then $S^\perp \neq \{0\}$.
- Let S_1, S_2 be subsets of $L_2(X, \mathcal{S}, \mu)$. Then the following hold:
 - (i) S_1^\perp is a closed subspace of $L_2(X, \mathcal{S}, \mu)$ and

$$S_1 \cap S_1^\perp \subseteq \{0\}.$$

If S_1 is also a subspace, then $S_1 \cap S_1^\perp = \{0\}$.



As a consequence of this, it also says that if S is proper closed subspace of L_2 then, its perpendicular cannot be 0, because if it is proper then there is an element f minus f_0 which is not 0, it is orthogonal to it.

So, as an immediate consequence that if S is proper closed subspace of L^2 then its orthogonal complement cannot be 0 it has to be something else. Also means that if the orthogonal complement of something is 0 then S must be equal to L^2 , another way of stating the same thing is this. As I said, we will not be proving this theorem but, we will give some applications of this today. So, let us look at some properties of orthogonal complement before we go and to prove some general facts.

Let us take S_1 and S_2 be subsets of L^2 then the following properties hold namely, S_1^\perp is a closed subspace that we have already shown and S_1 they intersect only at the most at 0 , they are just sets.

(Refer Slide Time: 30:14)

$$\begin{aligned} f \in S_1 \cap S_1^\perp \\ \Rightarrow \langle f, f \rangle &= 0 \\ \Rightarrow \|f\| &= 0 \\ \Rightarrow f &= 0. \\ S_1 \cap S_1^\perp &\subseteq \{0\} \end{aligned}$$


The image shows a whiteboard with handwritten mathematical text. At the bottom left, there is a small circular logo with the text 'NIPTEL' below it.

S_1^\perp is a subspace because S_1 may not be a subspace. So, it says that if $S_1 \cap S_1^\perp$ is always inside 0 and that is obvious, because if f belongs to $S_1 \cap S_1^\perp$ then that means, the inner product of f with itself because f belongs to S_1 and it also belongs to S_1^\perp that must be equal to 0 . So, that implies norm of f is equal to 0 and that implies f must be equal to 0 .

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Consequences

- Let S be a proper closed subspace of $L_2(X, \mathcal{B}, \mu)$. Then $S^\perp \neq \{0\}$.
- Let S_1, S_2 be subsets of $L_2(X, \mathcal{S}, \mu)$. Then the following hold:
 - (i) S_1^\perp is a closed subspace of $L_2(X, \mathcal{S}, \mu)$ and
$$S_1 \cap S_1^\perp \subseteq \{0\}.$$
If S_1 is also a subspace, then $S_1 \cap S_1^\perp = \{0\}$.




Hence, $S_1 \cap S_1^\perp$ is inside 0. As an obvious consequence, if S_1 is also a subspace then $S_1 \cap S_1^\perp = \{0\}$, they do not have anything common other than the vector.

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Consequences

- (ii) If $S_2 \subseteq S_1$, then $S_1^\perp \subseteq S_2^\perp$
- (iii) $S_1 \subseteq (S_1^\perp)^\perp$, and
$$S_1 = (S_1^\perp)^\perp \text{ iff } S_1 \text{ is a closed subspace.}$$



The second property says that if S_2 is a subset of S_1 then, S_1^\perp is a subset of S_2^\perp and that is obvious, because if we take any element say h in S_1^\perp then, the inner product of h with every element of S_1 is equal to 0 and in particular with S_2 is equal to 0 that also belongs to S_2^\perp , so this property is obvious.

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$$\begin{aligned} f \in S_1, \quad h \in S_1^\perp \\ \Rightarrow \langle h, f \rangle = 0 \\ \Rightarrow f \in (S_1^\perp)^\perp \\ S_1 \subseteq (S_1^\perp)^\perp \end{aligned}$$

The image shows a whiteboard with handwritten mathematical equations. At the bottom left, there is a small circular logo with the text 'NPTEL' below it.

The third property says that S_1 is a subset of S_1 perpendicular perpendicular. So, orthogonal component of the orthogonal complement always includes S_1 . So that property is again obvious, because if we take f belonging to S_1 and h belonging to S_1 perpendicular then that implies h comma f the dot product is equal to 0, because f belongs S_1 and h belongs to S_1 perpendicular and that is equal to 0. That means, h is perpendicular to f ; this means that for every h in S_1 perpendicular f comma h or h comma f is equal to 0 that means, f is belonging to S_1 perpendicular perpendicular. So, S_1 is always a subset of S_1 perpendicular perpendicular.

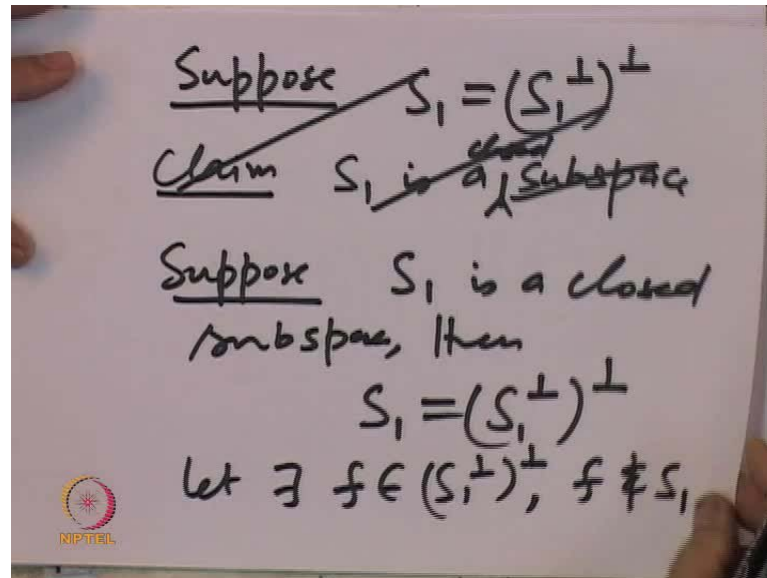
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$$\begin{aligned} \Rightarrow \langle h, f \rangle = 0 \\ \Rightarrow f \in (S_1^\perp)^\perp \\ S_1 \subseteq (S_1^\perp)^\perp \\ \text{In case } (S_1^\perp)^\perp = S_1 \\ \Rightarrow S_1 \text{ is a closed subspace.} \end{aligned}$$

The image shows a whiteboard with handwritten mathematical equations. At the bottom left, there is a small circular logo with the text 'NPTEL' below it.

We want to show that in case S_1 perpendicular perpendicular equal to S_1 , these two are equal then, the left hand side is an orthogonal complement of a subspace; so this is the closed subspace - this implies S_1 is a closed subspace.

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Let us prove the converse part namely, the converse is also true. Suppose, S_1 is equal to S_1 perpendicular perpendicular then the claim that S_1 is a closed subspace, so that we have already shown. We want to prove the way round, so this is not what we want to do is. Suppose S_1 is a closed subspace, so converse is if S_1 is a closed subspace then we want to show S_1 is also equal to S_1 perpendicular perpendicular.

To prove this, let us take there exist f belonging to - this is subset of this anyway - let us assume there is a S_1 perpendicular perpendicular f not in S_1 . In that case, let us apply our best approximation theorem, so implies by the theorem just now we stated which we did not prove, that there exist an element f_{naught} belonging to S_1 such that f minus f_{naught} is perpendicular to S_1 . So that means, f minus f_{naught} belongs to S_1 perpendicular.

So there is an element, f is not in S_1 ; so there is an element in f_{naught} in S_1 such that the difference is perpendicular to S_1 . Now, let us observe that this element f_{naught} belongs to S_1 which is contained in S_1 perpendicular perpendicular.

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The image shows a whiteboard with handwritten mathematical text. A hand is visible on the left side, pointing to the text. The text is as follows:

$$\begin{aligned} \text{Then} & \Rightarrow \exists f_0 \in S_1 \text{ such} \\ \text{that} & f - f_0 \perp S_1 \\ \Rightarrow & (f - f_0) \in S_1^\perp \text{ ---}^* \\ \text{As } & f_0 \in S_1 \subseteq (S_1^\perp)^\perp \\ \Rightarrow & f - f_0 \in (S_1^\perp)^\perp \text{ ---}^* \\ \Rightarrow & f - f_0 = 0 \Rightarrow f = f_0 \\ & \Rightarrow f \in S_1 \end{aligned}$$

A small logo for NIPTEL is visible in the bottom left corner of the whiteboard image.

So that implies so we have got f belonging to S_1 perpendicular perpendicular and f naught also belonging that means f minus f naught belongs to S_1 perpendicular perpendicular and f naught is also in the same thing and this being a subspace the difference must also belong to S_1 perpendicular perpendicular.

Now, the element f minus f naught belongs to S_1 perpendicular perpendicular and it also belongs to S_1 perpendicular. From here and here, it belongs to a subspace and orthogonal complement of it that means f minus f naught must be equal to 0, implying f is equal to f naught (Refer Slide Time: 36:33).

So that means what? That means f belongs to f naught f and where is f naught; f naught is in S_1 , so this f also belongs so we would **certify** the f in S_1 perpendicular perpendicular and we are getting that f is equal to f naught, where f naught is an element in S_1 , so that implies that f belongs to S_1 .


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Consequences

(ii) If $S_2 \subseteq S_1$, then $S_1^\perp \subseteq S_2^\perp$

(iii) $S_1 \subseteq (S_1^\perp)^\perp$,
and
 $S_1 = (S_1^\perp)^\perp$ iff S_1 is a closed subspace.

(iv) If S_1 and S_2 are closed subspaces and $f \perp g \forall f \in S_1$ and $\forall g \in S_2$, then
 $S_1 + S_2 := \{f + g \mid f \in S_1, g \in S_2\}$
is also a closed subspace.




What we are shown is, whenever f belongs to S_1 perpendicular perpendicular it also belongs to S_1 , so these two are equal. That proves the fact that if S_1 is equal to S_1 perpendicular perpendicular then S_1 is a closed subspace of it. Next, let us observe, the fact that if S_1 and S_2 are two closed subspaces and S_1 is perpendicular to S_2 then S_1 plus S_2 is also a closed subspace.

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S_1, S_2 - closed subspaces
 $S_1 \perp S_2$

$S_1 + S_2$: $f_1 + g_1 \in S_1 + S_2$
 $f_2 + g_2 \in S_1 + S_2$

$\Rightarrow \alpha(f_1 + g_1) + \beta(f_2 + g_2)$
 $= \underbrace{(\alpha f_1 + \beta f_2)}_{\in S_1} + \underbrace{(\alpha g_1 + \beta g_2)}_{\in S_2}$
 $\Rightarrow \in S_1 + S_2$

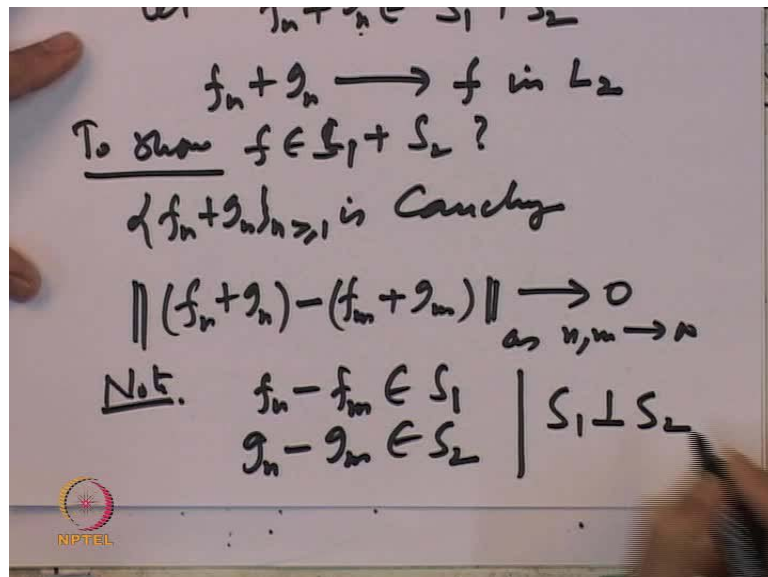


Let us observe that S_1, S_2 is closed subspaces and S_1 is perpendicular to S_2 . Let us take two elements; so let us look at S_1 plus S_2 , we want to show it is a subspace. Let us

take an element say $f + g$ in $S_1 + S_2$, where f belongs to S_1 and g belongs to S_2 .

Let us take another element, $f_2 + g_2$ also belonging to $S_1 + S_2$, where f_2 belongs to S_1 and g_2 belongs to S_2 . Then, for every α and β $\alpha f_1 + \beta f_2 + \alpha g_1 + \beta g_2$ is equal to $\alpha f_1 + \beta f_2$ plus $\alpha g_1 + \beta g_2$. Now, because S_1 is a subspace (Refer Time: 39:00) $\alpha f_1 + \beta f_2$ belongs to S_1 and $\alpha g_1 + \beta g_2$ belongs to S_2 . So, implies that this element belongs to $S_1 + S_2$.

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So that proves $S_1 + S_2$ is a subspace. To prove it is a closed subspace let us observe, let $f_n + g_n$ belong to $S_1 + S_2$, where f_n belongs to S_1 and g_n belongs to S_2 and $f_n + g_n$ converge to f in L_2 . To show $S_1 + S_2$ is closed that means we have to show that f belongs to $S_1 + S_2$, so that is what we have to show.

Now, let us observe that $f_n + g_n$ being convergent is Cauchy; $f_n + g_n$ is a Cauchy sequence. Let us look at $f_n + g_n - f_m - g_m$, so this norm goes to 0 as n and m go to infinity.

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Pyth agam

$$\| (f_n + g_n) - (f_m + g_m) \|^2$$

$$= \| (f_n - f_m) \|^2 + \| (g_n - g_m) \|^2$$

$$= \| (f_n + g_n) - (f_m + g_m) \|^2$$

$$\Rightarrow \| f_n - f_m \| \rightarrow 0, \| g_n - g_m \| \rightarrow 0$$

Now, note that $f_n - f_m$ belongs to S_1 and $g_n - g_m$ belong to S_2 , so this implies by Pythagoras theorem that the norm of $f_n - f_m$ square plus norm of $g_n - g_m$ square, this is equal to norm of $f_n + g_n - f_m - g_m$ square.

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$$\Rightarrow \{f_n\}_n \text{ is Cauchy}$$

$$\Rightarrow f_n \rightarrow h \in S_1$$

$$\| \wedge \quad g_n \rightarrow g \in S_2$$

$$\Rightarrow f_n + g_n \rightarrow h + g$$

$$\downarrow$$

$$\Rightarrow f = h + g \in S_1 + S_2$$

So that is by Pythagoras theorem and this goes to 0, so that implies that norm of $f_n - f_m$ goes to 0 and norm of $g_n - g_m$ goes to 0. So, meaning what? This says that f_n itself is Cauchy and g_n itself is Cauchy. So that implies that f_n is a Cauchy implying that f_n must converge to some h .

Similarly, g_n must converge to some g , all in L^2 . Similarly that implies that $f_n + g_n$ converges to $h + g$ and we know that this converges to f , so this implies that f is equal to $h + g$. Now, note that f_n is a sequence in S_1 and S_1 is closed, so this h belongs to S_1 and g belongs to S_2 , so this belongs to $S_1 + S_2$ (Refer Slide Time: 43:00).


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Consequences

(ii) If $S_2 \subseteq S_1$, then $S_1^\perp \subseteq S_2^\perp$

(iii) $S_1 \subseteq (S_1^\perp)^\perp$,
and
 $S_1 = (S_1^\perp)^\perp$ iff S_1 is a closed subspace.

(iv) If S_1 and S_2 are closed subspaces and $f \perp g \forall f \in S_1$ and $\forall g \in S_2$, then
 $S_1 + S_2 := \{f + g \mid f \in S_1, g \in S_2\}$
is also a closed subspace.




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Consequences

(v) If S_1 is a closed subspace, then
 $S_1 \cap S_1^\perp = \{0\}$ and
 $L_2(X, \mathcal{S}, \mu) = S_1 + S_1^\perp$.

Thus every $f \in L_2(X, \mathcal{S}, \mu)$ can be uniquely expressed as $f = g + h$, where $g \in S_1$ and $h \in S_1^\perp$.

This is also expressed as
 $L_2(X, \mathcal{S}, \mu) = S_1 \oplus S_1^\perp$,
and is called the **projection theorem**



This completes the proof that if S_1 and S_2 are closed subspaces and S_1 is perpendicular to S_2 , then $S_1 + S_2$ also is a closed subspace. Finally, we can claim that S_1 is a closed subspace and then we know that $S_1 \cap S_1^\perp$ is


0. In that case, L^2 is equal to S^1 plus S^1 perpendicular and the reason for that is because this intersection is 0, so there cannot be S^1 plus S^1 perpendicular is a closed subspace if it is not whole then there must be an element outside, which is not true. So that means for every closed subspace S^1 of L^2 ; L^2 can be expressed as S^1 plus S^1 perpendicular. That means, every element of L^2 can be represented as an element in S^1 plus an element in S^1 perpendicular and this decomposition will be unique because S^1 intersection S^2 ; S^1 is a subspace so there is nothing common between them.

This is also called sometimes the projection theorem that means, for every closed subspace S^1 of L^2 ; L^2 can be represented as S^1 , one writes as a direct sum of S^1 perpendicular namely, these two are equal and the intersection of these two subspaces is equal to 0.

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Bounded linear functionals

- A map $T : L_2(X, \mathcal{S}, \mu) \longrightarrow \mathbb{C}$ is called a **bounded linear functional** if it has the following properties:
 - For every $f, g \in L_2(X, \mathcal{S}, \mu)$ and $\alpha, \beta \in \mathbb{C}$,
$$T(\alpha f + \beta g) = \alpha T(f) + \beta T(g).$$
 - There exists a real number M such that
$$|T(f)| \leq M \|f\|_2, \forall f \in L_2(X, \mathcal{S}, \mu).$$

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Next, let us come to analyzing maps on the space L^2 which is a vector space. As on any vector space you can analyze linear maps on the vector space taking values in the underlying field, here our vector space is L^2 actually it is a Hilbert space; so look at a map, which is a linear map T from L^2 to \mathbb{C} . We say it is a bounded linear functional if it has to following properties first of all, it should be linear; so T is a linear map as a vector space L^2 to \mathbb{C} . Secondly, we want that it is bounded in the sense that if there is a constant M such that norm of $T f$ is less than or equal to M times the norm of f^2 .

So, this called the boundedness of the linear map T . We say T is a bounded linear functional if T is linear on L^2 and an absolute value of $T f$ is less than or equal to a constant M times $\|f\|$ where M is a constant fix and this happens for every f in L^2 . It is quite to this condition, boundedness actually implies that T is also continuous, because if f_n converges to f then $\|T f_n - T f\|$ absolute value is less than or equal to M times norm of $f_n - f$ and that will go to 0.

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Bounded linear functionals

- It is easy to show that a linear functional $T : L_2(X, \mathcal{S}, \mu) \rightarrow \mathbb{C}$ is bounded, if and only if T is continuous at $0 \in L_2(X, \mathcal{S}, \mu)$.

Example:
Let $g \in L_2(X, \mathcal{S}, \mu)$ be fixed. Define the map

$$T_g : L_2(X, \mathcal{S}, \mu) \rightarrow \mathbb{C}$$

as follows:

$$T_g(f) = \langle f, g \rangle, \forall f \in L_2(X, \mathcal{S}, \mu).$$


So, it is easy to verify that T if a linear map is bounded, if and only if it is continuous and because you on vector space continuity at 0 are enough. So, one can verify easily that every bounded linear map is continuous at 0 is equivalent to it. One way of defining a continuous linear map is the following, fix any g in L^2 and look at the map T lower g defined on L^2 to be $T g$ at f is equal to f comma g for every f .

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Bounded linear functionals


It is easy to see that T_g is linear, i.e.,

$$T_g(\alpha f_1 + \beta f_2) = \alpha T_g(f_1) + \beta T_g(f_2)$$

$\forall \alpha, \beta \in \mathbb{C}$ and $f_1, f_2 \in L_2(X, \mathcal{S}, \mu)$,
and by the Cauchy-Schwarz inequality,

$$|T_g(f)| = |\langle f, g \rangle| \leq \|g\|_2 \|f\|_2.$$

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So that means, the value of T_g at f is defined as f g inner product of f with g for every f . It is easy to see that this is a linear map because of the inner product is linear in the first variable, that will give it is a linear map and it is bounded because of the Cauchy-Schwarz inequality. So, this is linear and by the Cauchy-Schwarz inequality it is a bounded linear map.

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Bounded linear functionals

■ It is easy to show that a linear functional


$$T : L_2(X, \mathcal{S}, \mu) \longrightarrow \mathbb{C}$$

is bounded,
if and only if
 T is continuous at $0 \in L_2(X, \mathcal{S}, \mu)$.

Example:
Let $g \in L_2(X, \mathcal{S}, \mu)$ be fixed. Define the map

$$T_g : L_2(X, \mathcal{S}, \mu) \longrightarrow \mathbb{C}$$

as follows:


$$T_g(f) = \langle f, g \rangle, \forall f \in L_2(X, \mathcal{S}, \mu).$$


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Bounded linear functionals

- As another application of the "best approximation" theorem, one proves the following:
- **Riesz representation theorem:**
Let $T : L_2(X, \mathcal{B}, \mu) \rightarrow \mathbb{C}$ be a bounded linear functional. Then there is a unique $g_0 \in L_2(X, \mathcal{S}, \mu)$ such that
$$T(f) = \langle f, g_0 \rangle \quad \forall f \in L_2(X, \mathcal{S}, \mu).$$



One way of constructing bounded linear functionals on L^2 is by taking the inner product of any element f with a fixed element g . This is an important theorem called the Riesz representation theorem, which says that this is the only way of constructing bounded linear functionals on L^2 . It says that if T is any bounded linear functional then, there is a unique g_0 belonging to L^2 such that Tf is equal to $\langle f, g_0 \rangle$. That means, every linear function T on L^2 arises via inner product of f with a fixed element g_0 and this g_0 is also unique.

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Riesz representation theorem

Outline of the proof

- We note that there does exist g_0 with the required claim, then $g_0 \in (\text{Ker}(T_g))^\perp$, where
$$\text{Ker}(T_g) := \{f \in L_2(X, \mathcal{B}, \mu) \mid T_g(f) = 0\}.$$

Note that $\text{Ker}(T)$ is a closed subspace. $\text{Ker}(T_g)$ is called the **kernel** of T_g .

- In case $\text{Ker}(T) = \{0\}$, then g_0 is the required function.



Let us just outline the proof of this. First of all, let us observe that there are two cases look at suppose that there exists any $g \neq 0$ with the required claim, then what will happen? Then, $g \neq 0$ must belong to what is called the Kernel of T of g that means what? A Kernel of g is all elements such that which are map to 0 and this is a closed subspace of Kernel of a bounded linear functional is a closed subspace.

So, if there is no g then this will be so; that means that our required claim will hold with T equal to 0. So, that is essentially saying the Kernel of T of g is a closed subspace of g and if one possibility is Kernel of T of g is equal to the whole space then it is equal to 0, because if Kernel of T of g equal to the whole space then T of g will be identically 0, so any $g \neq 0$ equal to 0 will satisfy.

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Riesz representation theorem

- In case $\text{Ker}(T)$ is a proper subspace, by "best approximation" theorem, there exists $g_1 \in \text{Ker}(T)^\perp$, $g_1 \neq 0$.
- One verifies that

$$g_0 = \left(\frac{T(g_1)}{\langle g_1, g_1 \rangle} \right) g_1$$

is the required (unique) function such that

$$T(f) = \langle f, g_0 \rangle, \forall f \in L_2.$$

 NPTEL

Let us assume that the Kernel of T is a proper closed subspace of it. Then by the best approximation theorem g_1 ; there is a g_1 in kernel of T perpendicular that is the consequence of the best approximation theorem and thus, g_1 will not be equal to 0. In that case, one verifies that if we take g_0 to be equal to T of g_1 divided by g_1 comma g_1 into g_1 , if this selection of g_0 is the required unique function such that T of f is equal to f g_0 for every f in L_2 .

Essentially, one implies the best approximation theorem to get an element g_1 in Kernel of T perpendicular. Why one is looking at Kernel of T perpendicular is because, if the required condition is to hold then that function g_0 has to belong to Kernel of T

perpendicular, because if it is $g \perp f$ and that means for f in kernel that must be 0 so the required $g \perp 0$ has to be from here. Let us pick up any element and then modify it and show that is required. So, this is what is called the Riesz representation theorem; so this is the Riesz representation theorem.

Today, what we have looked at is, the space L^2 is a perfect generalization of the space of the \mathbb{R}^n or the space \mathbb{C}^n that means there is a notion of an inner product defined on it, which is related to the norm and which gives the notion of perpendicularity.

We have proved; we stated one important theorem namely, if S is any closed subspace of L^2 and you take an element f in L^2 . There is a best approximation then there is an element $g \perp 0$ in the closed subspace, which best approximates within minimum distance from f . As a consequence of this; one consequence is the projection theorem namely every closed, if S is any closed subspace of L^2 then L^2 is a direct sum of S perpendicular.

The second consequence is characterizing all bounded linear functionals on the Hilbert space L^2 namely, the only way bounded linear functionals can be constructed on L^2 is via the inner product. That means T of f ; if T is a bounded linear functional and T of f must be equal to the inner product of f with an element $g \perp 0$, for some element $g \perp 0$ with the inner product. So that is characterization of bounded linear functions. Thank you.