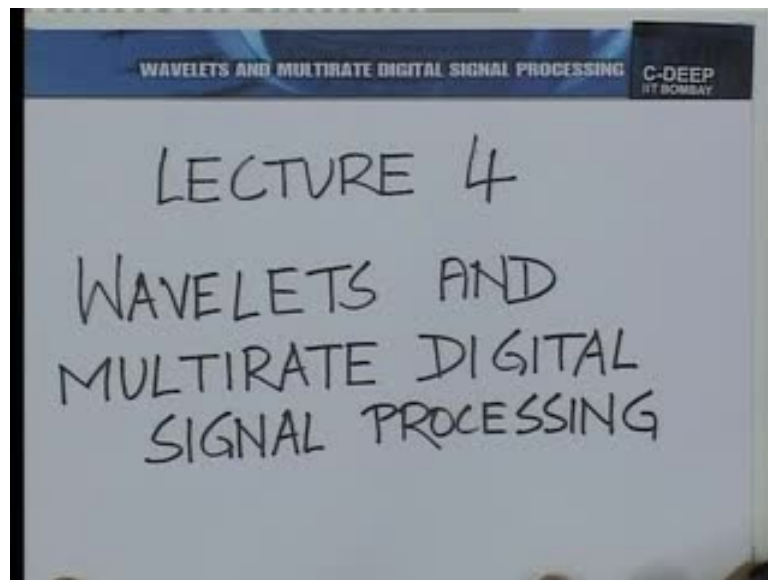


Advanced Digital Signal Processing – Wavelets and Multirate
Prof. V.M. Gadre
Department of Electrical Engineering
Indian Institute of Technology, Bombay

Lecture No. #04
Wavelets and Multirate Digital Signal Processing

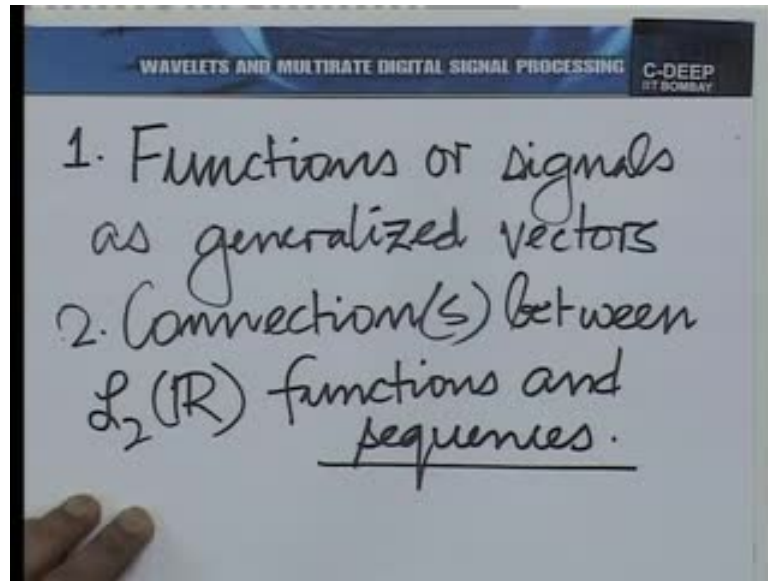
A warm welcome to the fourth lecture on the subject of wavelets and multirate digital signal processing in which we intend to build further, the connection between signals or functions in $L^2 \mathbb{R}$ and vectors, and therefore, we wish to build further, the idea of thinking of functions as belonging to linear spaces and characterizing them in a manner, slightly different from what we were doing in the previous lecture.

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So, just to put our discussion in perspective, this is the 4th lecture on the subject of wavelets and multirate digital signal processing and what we intend to discuss in this lecture is the following, let me put down the points, one by one.

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The 1st thing that I wish to talk about today is to think of functions as generalized vectors.

This idea is going to be useful to us in many different contexts in this course. So, we need to understand this connection between functions or signals and vectors in depth; we shall spend some time on it today.

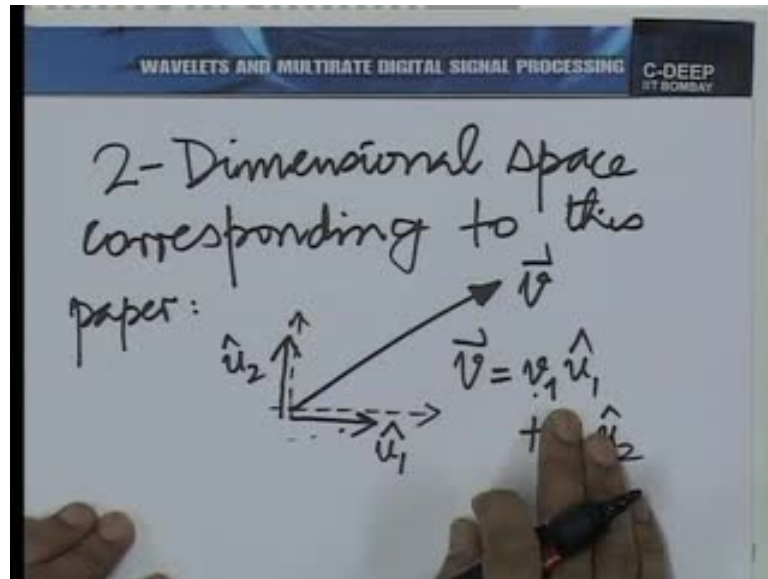
Secondly, the connection between $L_2(\mathbb{R})$ functions, connection or connections between $L_2(\mathbb{R})$ functions and sequences, we wish to understand this in greater depth.

So, what we are going to show in the later part of this lecture is that one can intimately relate processing of a function to processing of an equivalent sequence and whatever we are doing to try and gain information from or modify a function, can be done by equivalently processing or modifying, that sequence corresponding to the function.

Let us then, embark on the 1st of these 2 objectives now. You see, let us begin by asking what characterizes a vector after all? Let us take a minute and reflect.

What characterizes a two-dimensional vector, for example? A two-dimensional vector is essentially characterized by 2 coordinates, which are independent, we call them perpendicular coordinates. Actually, the idea of perpendicularity there is also intimately related to the idea of independence.

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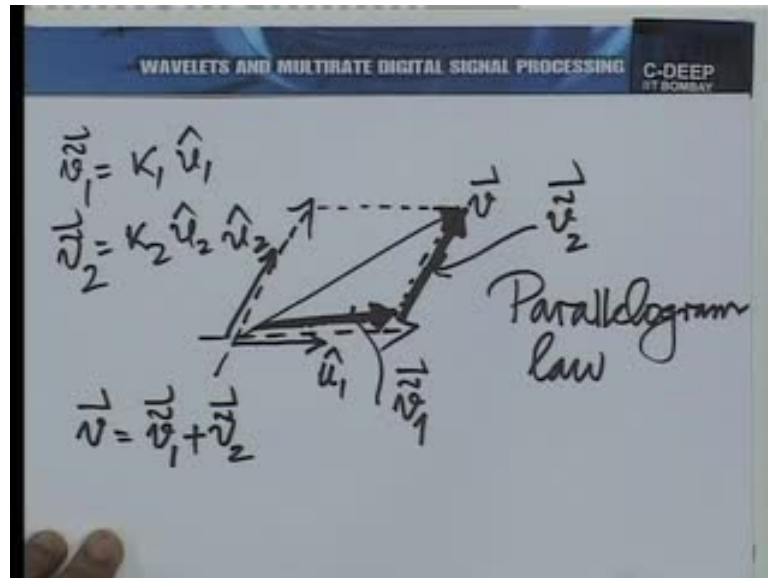
So, for example, let me treat the plane of the paper as a 2-dimensional space; the 2-dimensional space corresponding to this paper. Well, let us take any vector on this 2-dimensional space, so this vector be v , I am marking it as v .

There are many different ways to characterize this vector. In fact, notionally, an infinite number of ways and one of those ways is to choose the following 2, so called, perpendicular axis. So, we choose one axis like this and another axis like this and choose a unit vector along each of them. So, I have, say, unit vector, let me call it \hat{u}_1 along this axis and another unit vector \hat{u}_2 along this axis, and then I could write v or I could write this, sorry, just the vector v uniquely as, say, v_1 times \hat{u}_1 plus v_2 times \hat{u}_2 , whereby v_1 and v_2 characterizes vector v uniquely in this 2-dimensional space, with respect to the coordinates system generated by \hat{u}_1 and \hat{u}_2 , and there is an infinity of such coordinate systems.

In fact, one infinity of such coordinate systems can be generated simply by rotating this coordinate system of $\hat{u}_1 \hat{u}_2$. It is very easy to see, that if I take this structure $\hat{u}_1 \hat{u}_2$ and rotate it by any angle in this 2-dimensional plain, it would give me a new coordinate system. So, there is infinity of orthogonal coordinate systems in 2-dimensional space and in fact, there is also a relation between all these infinite orthogonal coordinate systems, simple enough. And orthogonal coordinate systems are not the only kinds of coordinate

system for a 2-dimensional vector. So, for example, the same two-dimensional space can be described by the following different coordinate systems.

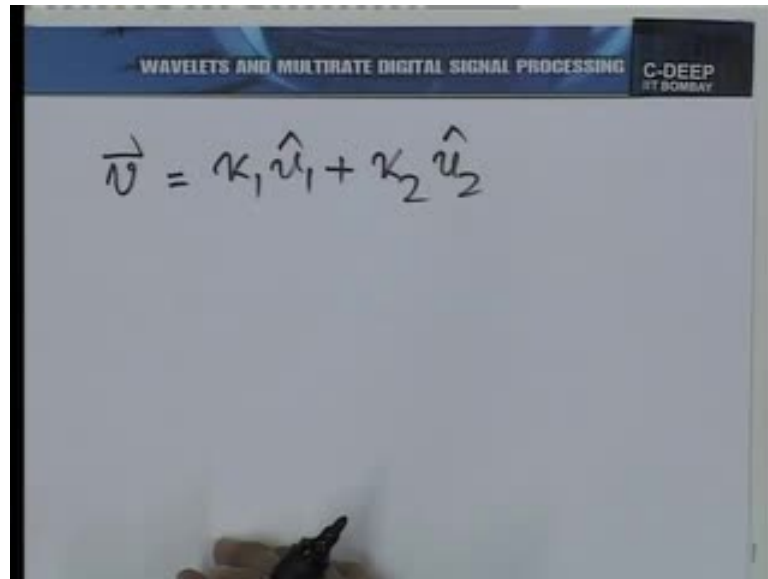
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So, I will draw the same vector v and it is perfectly alright to choose a coordinate system something like this. I could choose one coordinate like this, another coordinate like this and of course, I could again have the unit vectors in these 2 directions, \hat{u}_1 so to speak, \hat{u}_2 and I could express v in terms of \hat{u}_1 and \hat{u}_2 . Indeed, I could complete a parallelogram here, so using the parallelogram law, I could draw a line parallel from the tip of this vector to this \hat{u}_2 , another one parallel to \hat{u}_1 from the tip of the vector and it is very easy to see, that this dot dash vector here plus this dot dash vector here gives me v . Let me highlight, that dot dash vector.

This vector here plus this vector here gives me v . Let me call this \tilde{v}_1 and is a vector, and let me call this \tilde{v}_2 that is again, a vector. Of course, we have, v is \tilde{v}_1 plus \tilde{v}_2 and it is very easy to see, that \tilde{v}_1 as a vector is some multiple of \hat{u}_1 and similarly, \tilde{v}_2 as a vector is some multiple of \hat{u}_2 .

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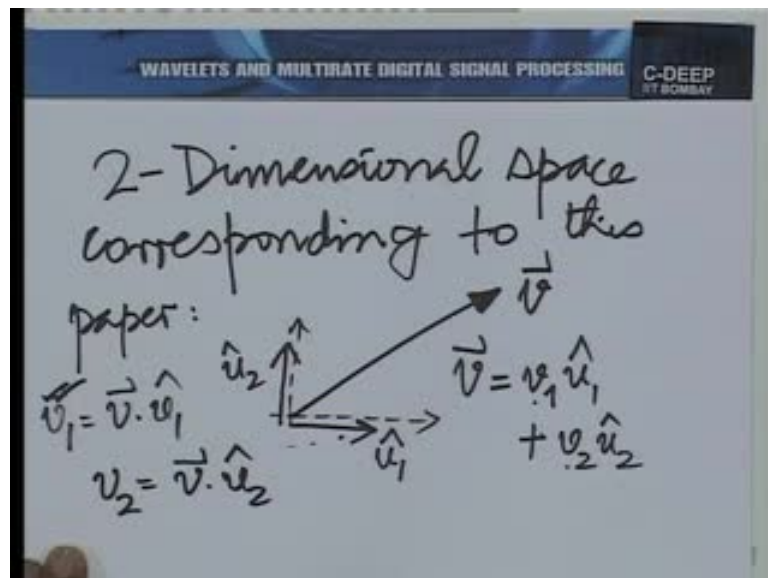


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
BT BOMBAY

$$\vec{v} = \kappa_1 \hat{u}_1 + \kappa_2 \hat{u}_2$$

Thereupon, I have, v is some multiple of u_1 plus some other multiple of u_2 , $\kappa_1 u_1$ plus $\kappa_2 u_2$. The only catch is determining κ_1 and κ_2 is a little more difficult than determining the constants in the previous representation. In fact, let me go back to that previous representation.

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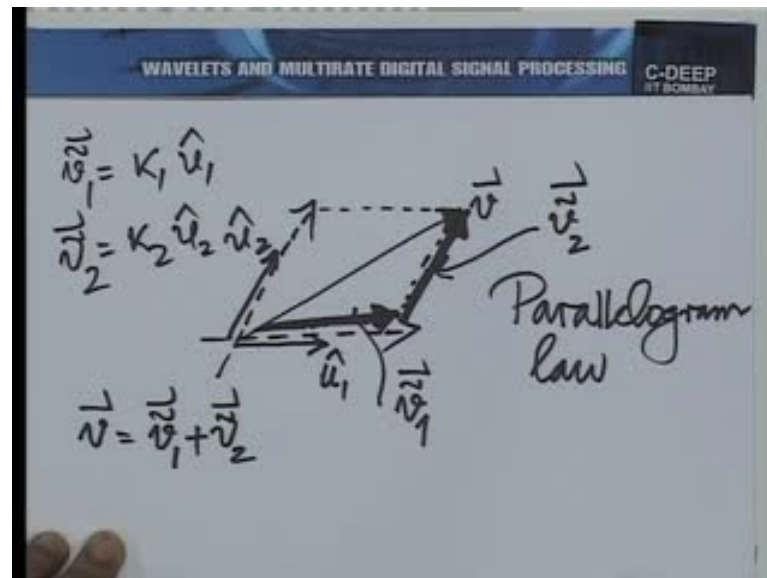
2-Dimensional Space corresponding to this paper:

$$\vec{v}_1 = \vec{v} \cdot \hat{u}_1$$
$$v_2 = \vec{v} \cdot \hat{u}_2$$
$$\vec{v} = v_1 \hat{u}_1 + v_2 \hat{u}_2$$

I had this representation previously, where v is $v_1 u_1$ cap plus $v_2 u_2$ cap and remember, v_1 and v_2 here, of course, are constants and very easy to obtain because I can simply obtain them by taking a dot product of v with u_1 cap and v with u_2 cap. So,

in fact, in the sense of dot products, v_1 is indeed $v \cdot u_1$ and v_2 , I mean, v_1 is a coordinate not as a vector, v_2 is a coordinate, is the dot product of v with u_2 cap, simple enough.

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Such a simple relationship does not exist in this context. While we are not hard put to describe the process by which we obtain k_1 and k_2 , it simply says, construct the parallelogram; expressing this analytically is a bit of work.

So, it is definitely very clear from this example, that an orthogonal or a perpendicular coordinate system has its advantages. It is always nice to have a perpendicular coordinate system in two-dimensional space to represent any two-dimensional vector. The same idea can, of course, be extended to three-dimensions too and then, one could also conceive of more than three-dimensions, four-dimensions, N -dimensions and then, in principle, an infinite number of dimensions too. Now, there again, when we talk about infinite dimensional situations, we have countably infinite and uncountably infinite finer points, but for the moment, infinite is difficult enough.

So, infinite dimensional vectors, in fact, lead us to the idea of functions. Now, it is a little difficult to understand infinite dimensional vectors all at once, so to progress towards infinite dimensional vectors, it is easier first to start from finite dimensional vectors of larger and larger dimension, and all that we need to do is to understand, that what

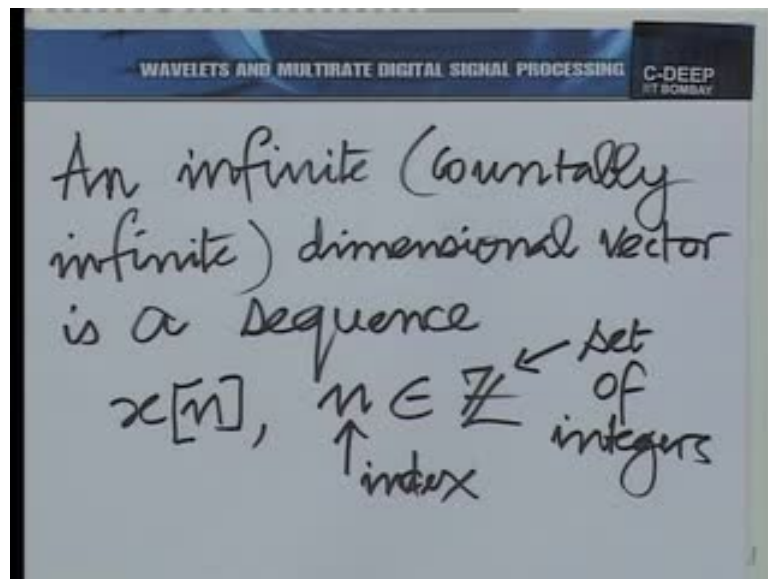
characterizes the dimension of a vector is really the number of independent coordinates, that it has.

For example, a three-dimensional vector has 3 independent coordinates, a four-dimensional vector would have 4 and N-dimensional vector N, and an accountably infinite dimensional vector would have accountably infinite number of dimensions or countably infinite number of coordinates.

By countable we mean, we can put the coordinates or the dimensions in one to one correspondence with the set of integers. So, we can talk about the 0th coordinate; we can talk about the 1th coordinate; we can talk about the minus 1th coordinate; the minus 2th coordinate, and so on, so forth.

What are we talking about here then, if we talk about an infinite dimensional vector? We are, in fact, talking about sequences; we build up the idea from there.

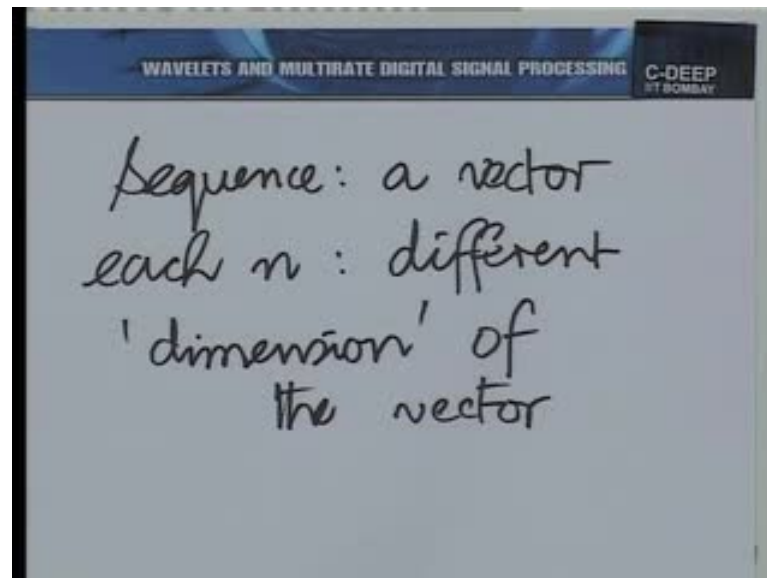
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So, here we are, let us make a note of this. An infinite dimensional vector or rather an infinite - countably infinite - dimensional vector is, essentially, a sequence. So, for example, we have a sequence x of n , where n belongs to set of integers over all the integers; recall, that **this script** \mathbb{Z} is the representation of the set of integers and this is called the index variable.

So, now, we have a different interpretation for sequences. A sequence is like a vector and each n is a different dimension of that vector; I think that is important enough for us to write down explicitly.

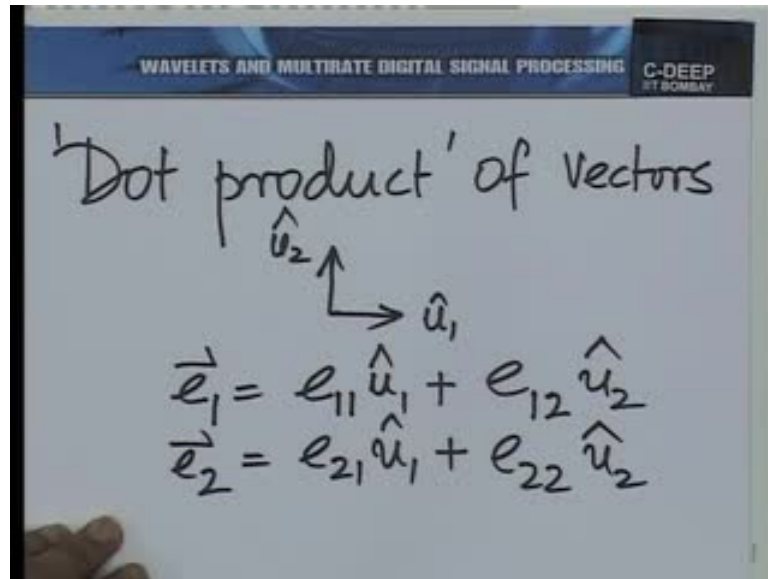
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So, a sequence is a vector. Each n is a different dimension of the vector and once we have this analogy, then extending other ideas of vectors to this context is not difficult at all. For example, adding 2 vectors, simple, add the sequences point by point; multiplying a vector by a constant, very simple, multiply each point of that sequence by that constant.

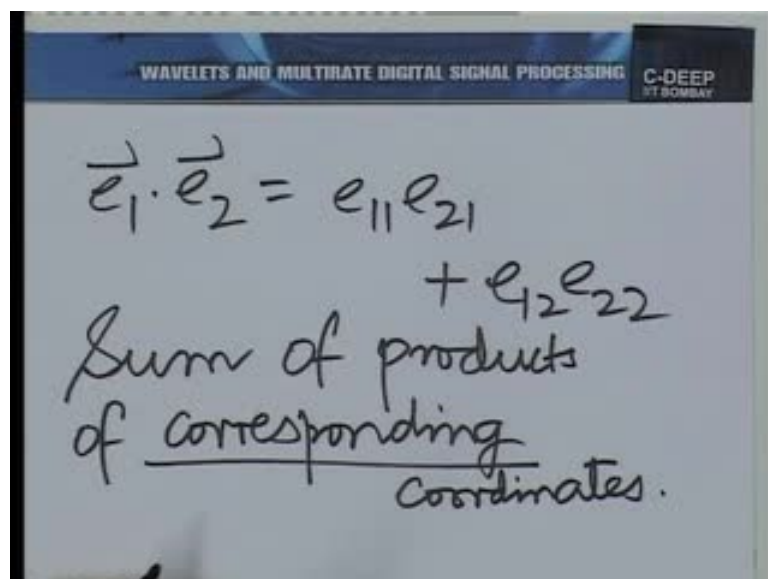
What we would like to do now is to extend some of the other ideas of vectors that we have. Some of the geometrical ideas to this, this context of infinite dimensional vectors and one of the very useful ideas that we have in the context of vectors, is the idea of a dot product. How do we take the dot product of 2 vectors in two-dimensional space? So, let us recall.

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So, suppose, for example, we choose a pair of orthogonal coordinates. So, we have \hat{u}_1 and \hat{u}_2 , as we did some time ago, orthogonal to one another, perpendicular to one another. And we have 2 vectors; let us call them \vec{e}_1 , which has the coordinates e_{11} and e_{12} . So, \vec{e}_1 is $e_{11}\hat{u}_1 + e_{12}\hat{u}_2$, and similarly, \vec{e}_2 has a vector has the coordinates $e_{21}\hat{u}_1 + e_{22}\hat{u}_2$.

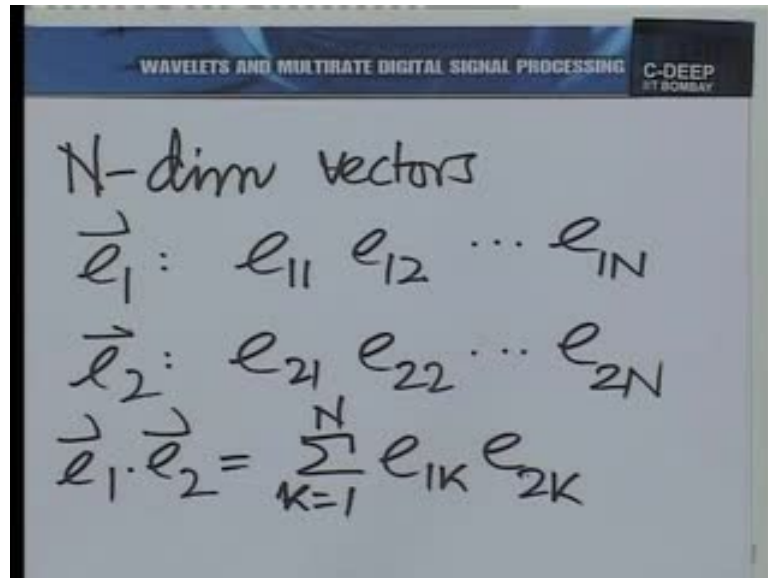
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Then, the dot product of \vec{e}_1 and \vec{e}_2 , $\vec{e}_1 \cdot \vec{e}_2$ as we write it, is essentially, $e_{11}e_{21} + e_{12}e_{22}$. So, it is the sum of products of corresponding coordinates; two-

dimensions, easy enough to understand; three-dimensions, easy to extend; in fact, N-dimensions, equally easy to extend.

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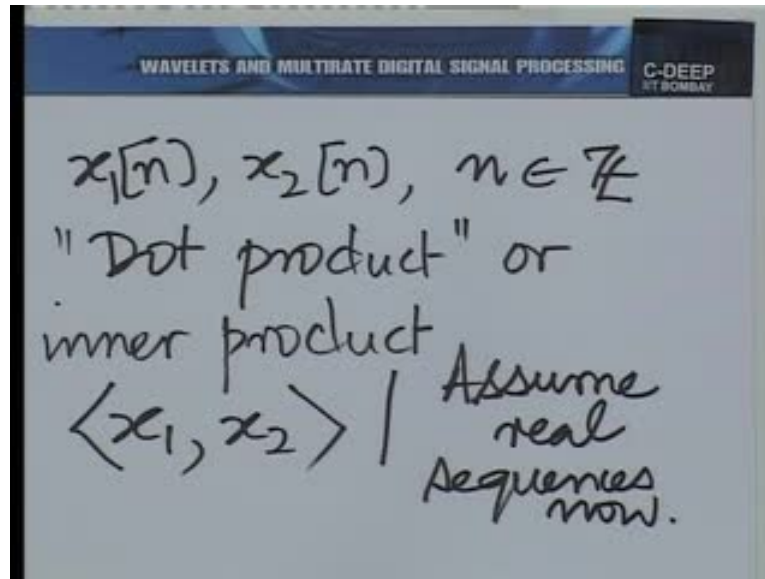
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
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N-dimn vectors
 $\vec{e}_1: e_{11} e_{12} \dots e_{1N}$
 $\vec{e}_2: e_{21} e_{22} \dots e_{2N}$
 $\vec{e}_1 \cdot \vec{e}_2 = \sum_{K=1}^N e_{1K} e_{2K}$

Suppose, we had 2 N-dimensional vectors characterized by coordinates, say e_{11} to e_{1N} . So, you have 2 N-dimensional vectors, e_1 characterized by coordinates e_{11} e_{12} up to e_{1N} and similarly, e_2 characterized by the coordinates e_{21} e_{22} up to e_{2N} .

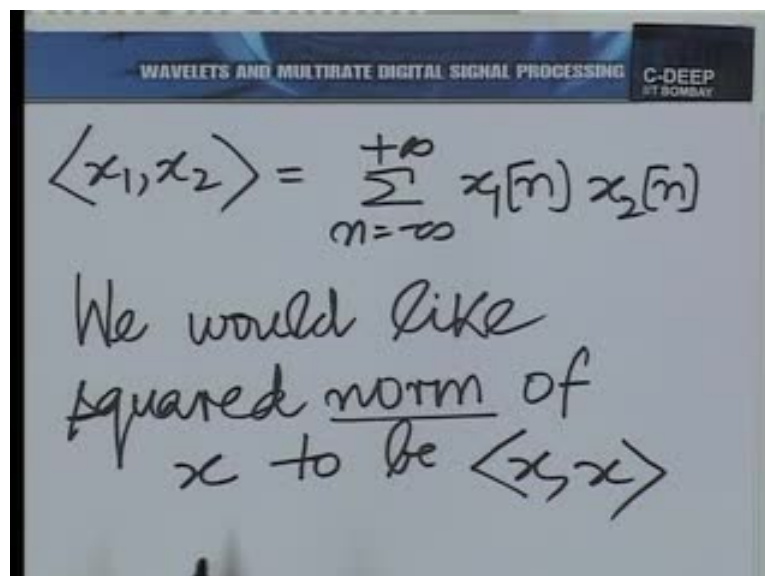
Then, of course, e_1 dot e_2 is easy to express if we generalize this. It is essentially, summation K from 1 to N ; e_{1K} times e_{2K} . So, dot product generalized to N-dimensions; of course, we assume these are orthogonal coordinates.

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Now, we can even take this to infinite dimensions. So, we can think of the dot product of 2 sequences, let us say, x_1 and x_2 . So, we have here, for example, 2 sequences $x_1[n]$ and $x_2[n]$, defined over the set of integers n , over all the integers. They are, so called, dot product or inner product, as the formal name is. So, we see, instead of dot product, now you would like to use a term inner product to generalize and we denote the inner product this way.

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For the moment, let us assume, these are real sequences for the moment. In that case, if we generalize, it is easy to see, that the inner product of x_1 and x_2 is simply summation on n , n running from all the way from minus to plus infinity, $x_1(n)$ times $x_2(n)$. And of course, it is clear, that the dot product or the inner product, as we are going to call it, in this generalized situation is commutative, that means, if I interchange the rows of x_1 and x_2 , the result does not change.

However, we would like this inner product or dot product notion to give us some of the powers and some of the conveniences that the dot product offers in the context of vectors. One so called convenience, so one such, so called, interpretation or meaning, that we derived from the dot product is the notion of magnitude. So, in fact, one could think of the notion of magnitude as induced from a dot product if one desires, or in other words, one could calculate the magnitude of a vector by using the notion of a dot product as one path towards the calculation of magnitude.

Incidentally, the word magnitude of vectors is used for small dimensional vectors, like 2 and 3 dimensional, but when we go to these generalized situations of N -dimensional vectors or countably infinite dimensional vectors, we replace the word magnitude by the word norm.

So, we say, that we would like the squared norm of x to be the dot product of x with x , as is the case with vectors. So, if you recall, $A \cdot A$, where A is a vector in two or three-dimensions, for that matter, is the magnitude squared of A . The same should hold good here. When we take the dot product of a sequence with itself, it should give us the squared norm of that sequence, where norm is a more general word for the magnitude.

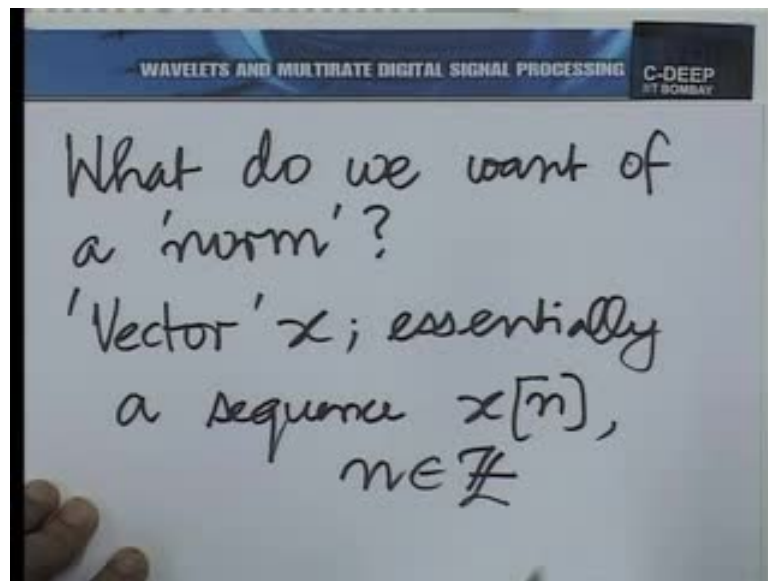
In fact, in $L^2 \mathbb{R}$, the norm is representative of the energy, but at this moment we are not talking about $L^2 \mathbb{R}$ because we have not yet come to the situation where we are dealing with functions of continuous variables. So, we will postpone that interpretation for a minute, not very far away from now, and once again come back to sequences.

Even for sequences, when we take the dot product of a real sequence with itself, we indeed get something that we likened to energy of the sequence. So, it is not uncommon to refer to the dot product of a real sequence, or for that matter, sequence with itself as the energy in that sequence. Anyway, I kept emphasizing real for a good reason. When we talk about the magnitude of a vector, or for that matter, there is more generalized

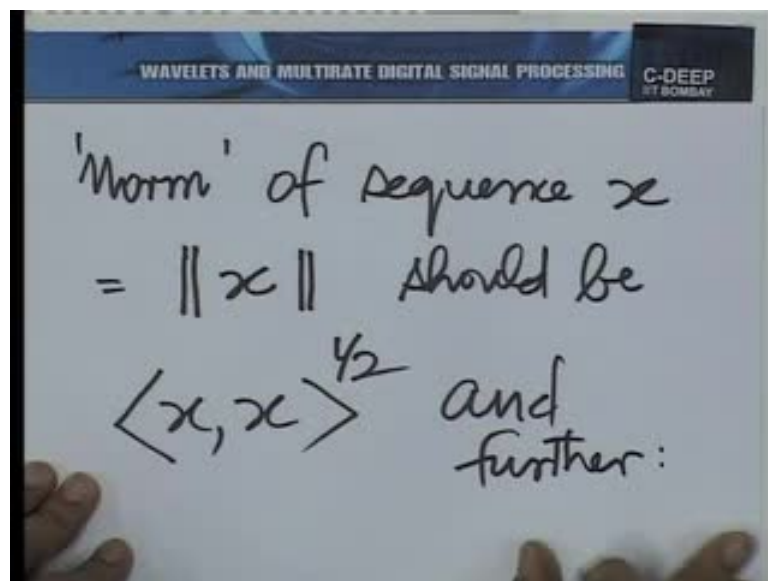
word norm, what is it that we expect of a magnitude? We want the magnitude or the norm to be a non-negative number and in fact, strictly positive if that vector is non-zero.

So, there are the following things that we demand of this concept of norm or magnitude, let us write them down, it is a useful and a powerful idea to have around us. So, what do we want of a norm?

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP AT BOMBAY

$$\|x\| \geq 0$$

and $\|x\| = 0$

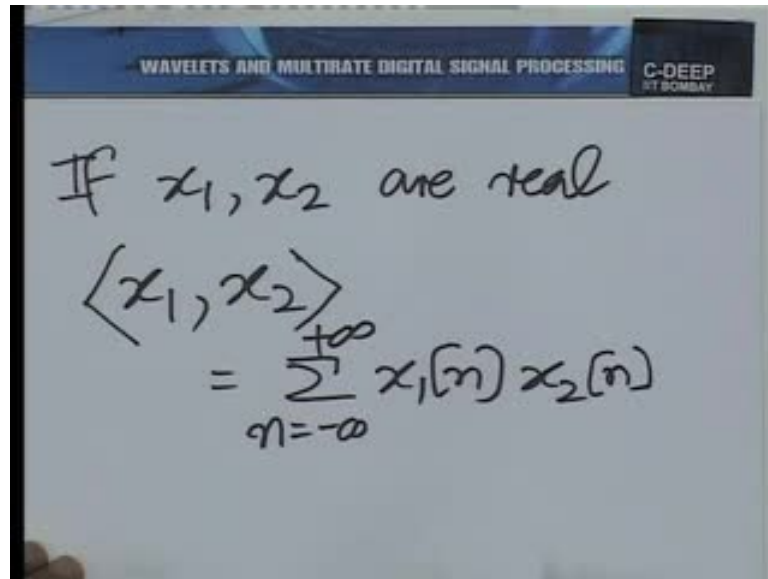
$$\Leftrightarrow x = 0.$$

i.e. $x[n] = 0 \quad \forall n \in \mathbb{Z}$

So, if I have a vector x , essentially a sequence $x[n]$, n over the set of integers, then it is norm, which we shall denote in the following way. We denote it like this, should be essentially, the dot product of x with x square root, and further, we would want norm of x to be non-negative. And if at all the norm of x is 0, that implies, and is implied by the sequence itself being 0 everywhere; that is, $x[n]$ is equal to 0 for all n belonging to set of integer. This is important, so we do not want that norm to be 0 unless the sequence itself is the 0 sequence.

A non-zero sequence, even if it is non-zero, at one point must have a non-zero norm and a 0 sequence must have a 0 norm.

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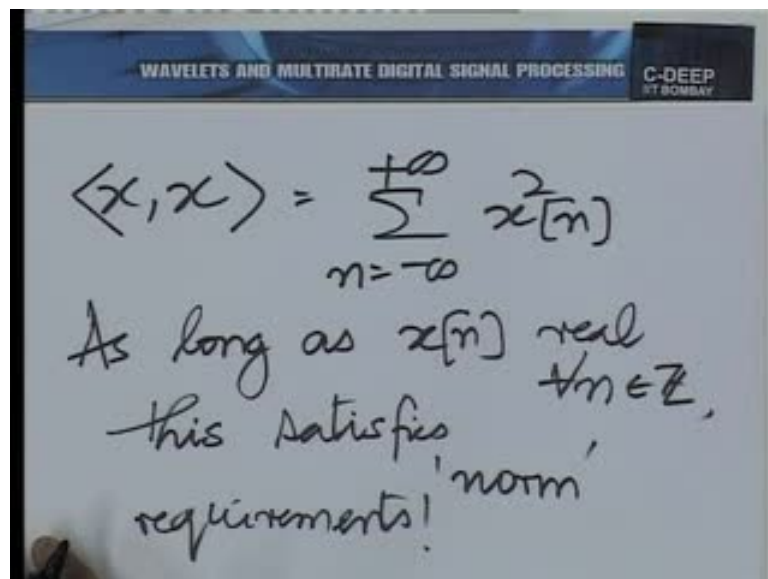
A handwritten slide from a presentation. At the top, there is a blue header with the text "WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING" and "C-DEEP FT BOMBAY" on the right. The main content is handwritten in black ink on a light background. It starts with "If x_1, x_2 are real", followed by the equation $\langle x_1, x_2 \rangle = \sum_{n=-\infty}^{+\infty} x_1[n] x_2[n]$.

WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP FT BOMBAY

If x_1, x_2 are real

$$\langle x_1, x_2 \rangle = \sum_{n=-\infty}^{+\infty} x_1[n] x_2[n]$$

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP FT BOMBAY

$$\langle x, x \rangle = \sum_{n=-\infty}^{+\infty} x^2[n]$$

As long as $x[n]$ real $\forall n \in \mathbb{Z}$,
this satisfies requirements! 'norm'

Does are dot product satisfy this? Well, for real sequences, it does. If $x[n]$ is real, rather, if x_1, x_2 are real and we take the following definitions, the dot product of x_1 and x_2 is, essentially, summation on n going from minus to plus infinity $x_1[n] x_2[n]$, then the dot product of x with x is essentially summation n running from minus to plus infinity $x^2[n]$, and as long as $x[n]$ is real for all n belonging to \mathbb{Z} , this satisfies the requirements of norm.

It is non-negative and it is 0 if and only if the sequence is identically 0, but what if this is complex. So, we have to allow complex sequences too. One of the coordinates could be complex and in fact, the situation could be such, that x squared n could be plus 1 for one of the coordinates and minus 1 for some other coordinate in that case, because when you square a complex number, nothing guarantees the output is going to be non-negative. In fact, nothing even guarantees the output is going to be real, where is the question of non-negative? So, this definition is not going to work when x_1 and x_2 are complex sequences in general and we need to tweak the definition a little.

Well, it is not that difficult after all. What we want is that for every coordinate you must get a non-negative quantity when you take point by point products. So, all that we need to do for that purpose is to complex conjugate the 2nd argument in that summation.

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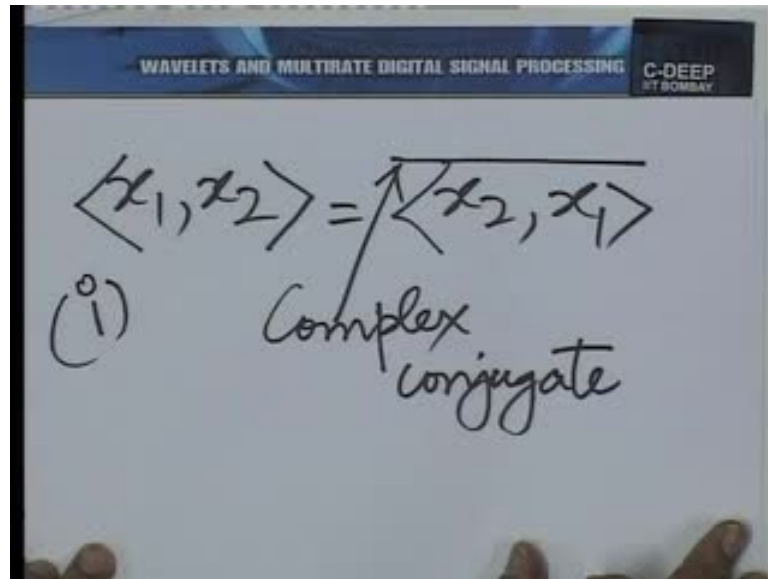
The following small change for complex sequences: $\langle x_1, x_2 \rangle$

$$= \sum_{n=-\infty}^{+\infty} x_1[n] \overline{x_2[n]}$$

↑
complex CONJUGATE

So, the small change for complex sequences, we will do our job. Dot product of x_1 with x_2 is summation overall n $x_1[n] \overline{x_2[n]}$, where bar denotes the complex conjugate.

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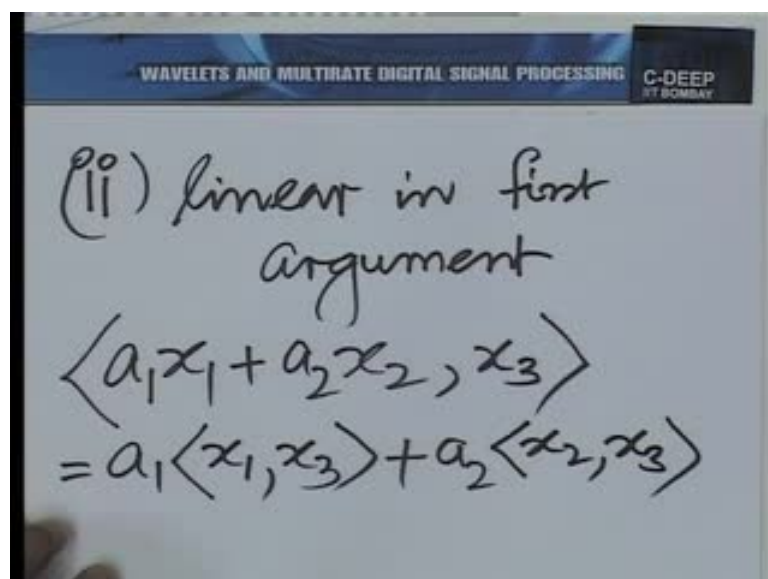
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$$\langle x_1, x_2 \rangle = \overline{\langle x_2, x_1 \rangle}$$

(i) Complex conjugate

Now, one point to note here when we make this little change is that, that commutativity property is lost. So, if I take the inner product x_1 with x_2 and then, if I take the inner product x_2 with x_1 , there is a complex conjugate relationship and this is the more general requirement of a dot product. In fact, this is the simplest way in which one can define a dot product between sequences; there are many other ways, again infinite number of ways, but at this moment we shall not go into the other ways, they will only confuse us. This is what is called the standard inner product, but one can have many other nonstandard inner products, which obey the following conditions.

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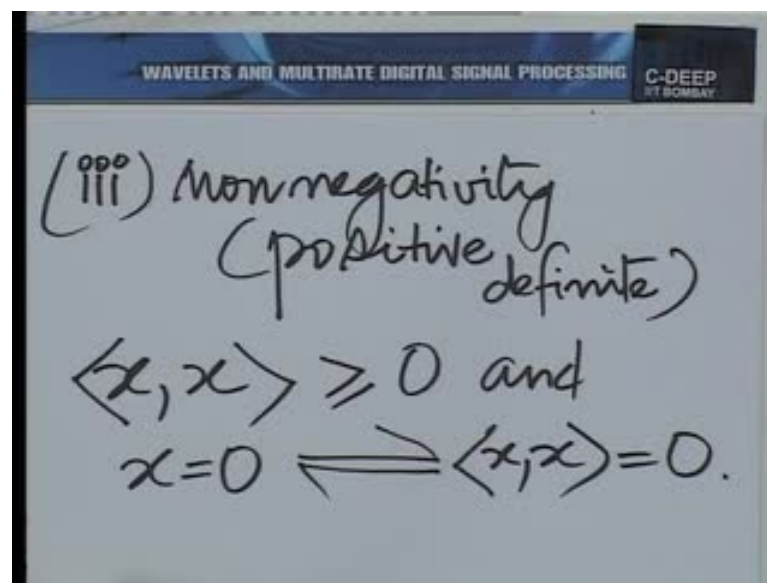
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(ii) linear in first argument

$$\langle a_1 x_1 + a_2 x_2, x_3 \rangle = a_1 \langle x_1, x_3 \rangle + a_2 \langle x_2, x_3 \rangle$$

The 1st condition is this that we write down here; the inner product of x_1 with x_2 is the complex conjugate of the inner product of x_2 with x_1 . Secondly, the inner product is linear in the 1st argument. In other words, if I take $a_1 x_1$ plus $a_2 x_2$, where in general a_1 and a_2 could be complex and take the inner product with x_3 , it is essentially a_1 times the inner product of x_1 with x_3 plus a_2 times the inner product of x_2 with x_3 ; this is the 2nd requirement of an inner product, linearity in the 1st argument.

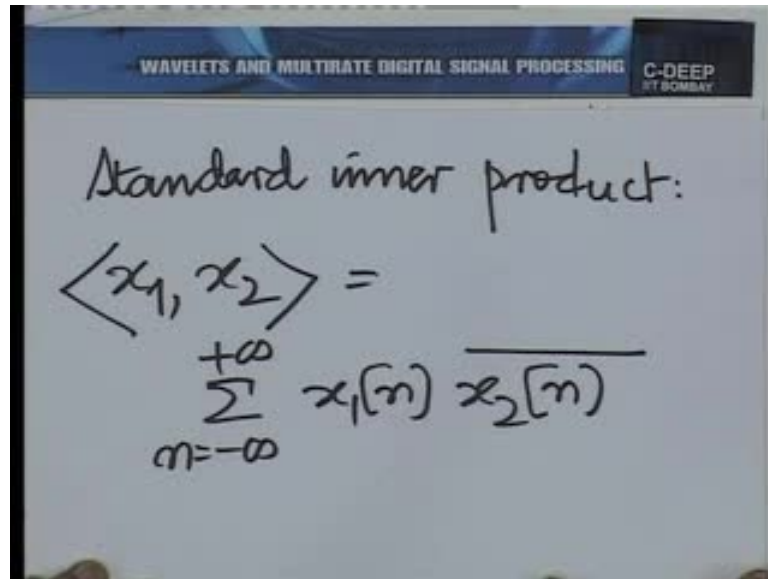
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The 3rd requirement of the inner product is what we have been building towards all this while, namely what is called the positivity or non-negativity. In fact, positivity is more appropriate, positive definiteness, namely the inner product of x with x is always greater than equal to 0, and x equal to 0 implies and is implied by the inner product of x with x being 0.

In fact, any operation between 2 sequences, x_1 and x_2 , which obeys these 3 conditions, is called an inner product and the standard inner product that we have just described is one such, which we shall use very frequently. So, in the discussion, henceforth, when we say inner product of sequences, we mean the standard inner product unless otherwise specified.

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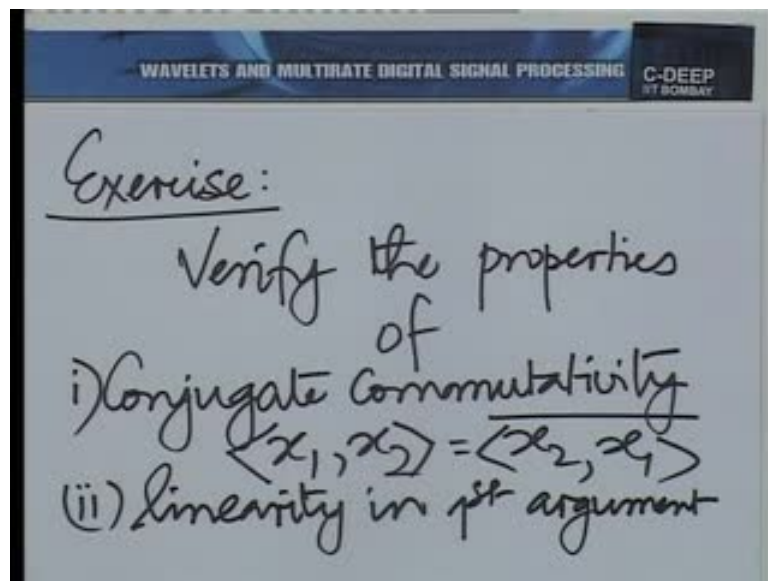


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Standard inner product:

$$\langle x_1, x_2 \rangle = \sum_{n=-\infty}^{+\infty} x_1[n] \overline{x_2[n]}$$

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Exercise:

Verify the properties of

i) Conjugate Commutativity
 $\langle x_1, x_2 \rangle = \overline{\langle x_2, x_1 \rangle}$

(ii) Linearity in 1st argument

So, let us just verify this for completeness; let us verify this for the standard inner product. The inner product of 2 sequences, x_1 x 2 is essentially, the sum n going from minus to plus infinity $x_1[n] \overline{x_2[n]}$ definition. The 1st property, as we said, is complex conjugate easy to verify. So, in fact, I leave it to you as an exercise; verify the properties of what is called conjugate commutativity, the first property and linearity, linearity in the first argument. I leave it as an exercise, easy enough to do.

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(iii) Positive definiteness:

$$\langle x, x \rangle = \sum_{n=-\infty}^{+\infty} x[n] \overline{x[n]}$$
$$= \sum_{n=-\infty}^{+\infty} |x[n]|^2 = 0$$

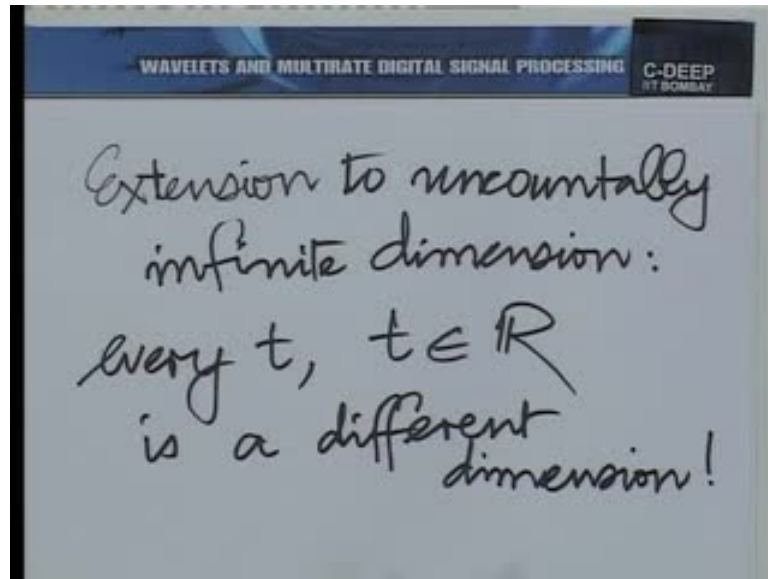
iff $x[n] = 0 \forall n$

But we shall, because it is so important, verify the 3rd property, the positive definiteness. Indeed, if we take the dot product of x with x , it is summation n going from minus to plus infinity $x[n] \overline{x[n]}$. We should, summation n going from minus to plus infinity $|x[n]|^2$ and it is very easy to see, that this is equal to 0 if and only if $x[n] = 0$ for all n .

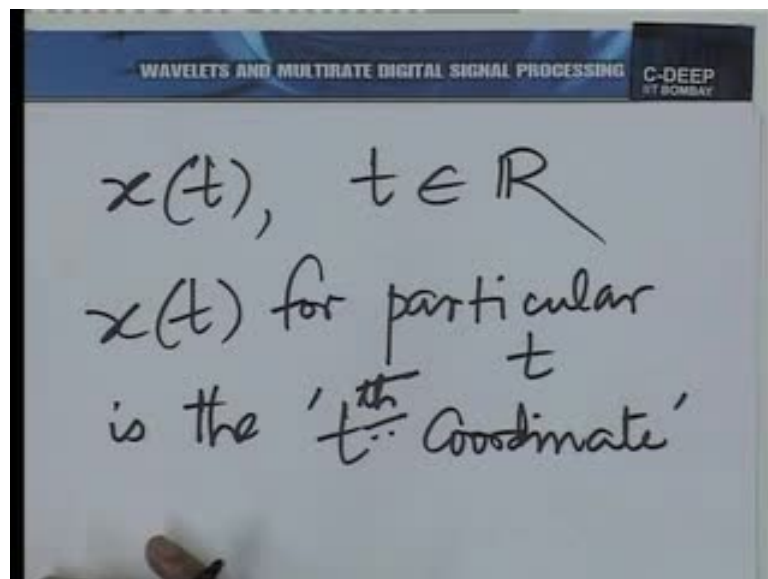
Even if one of the coordinates is non-zero, that particular $|x[n]|^2$ is going to be non-zero and it is going to contribute a positive term and of course, it is very easy to see, that each term for every n , I mean, is strictly positive if $x[n]$ is non-zero, so far so good. So, now, we have build up the idea of inner product or dot product between 2 sequences, which is going to be useful to us.

So, we move from two-dimensional to three-dimensional to N -dimensional, n is finite and then, to countably infinite dimensional.

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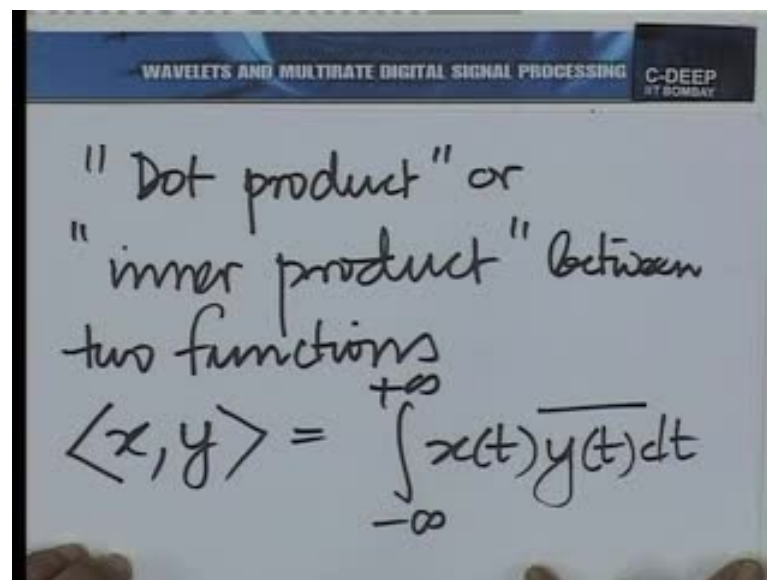


Now, let us move to uncountably infinite dimensional. So, suppose I take a function of the continuous variable t , how can I extend these notions? So, extension to uncountably infinite dimensions, well this is going to be very difficult in general, but very easy in particular. If we simply accept, that every t , for real t is a different dimension, simple. So, if you have a function x of t , t over the real numbers; x of t for a particular t is the t th coordinate, so to speak, and there is an uncountably infinite number of such coordinates indexed by the real numbers.

So, in principle, in a given function you have complete liberty to put down the value of x at every different point t , the only catch is we have agreed, that we would like to make the functions square integrable. So, that, that puts some restriction on x t , but not a very serious one, even so.

Now, you know, dealing with infinite dimensional spaces, if we wish to do it very rigorously and very, very carefully and you know, to satisfy the fastidious mathematician is a difficult job and we do not really intend to do that, all the way, in this course. If some of us do wish to take that puritanical perspective, one of course, would benefit from it in some ways and one could look up a book of, on function analysis, but what we wish to do is, rather to give intuitive understanding of some of the concepts at different places.

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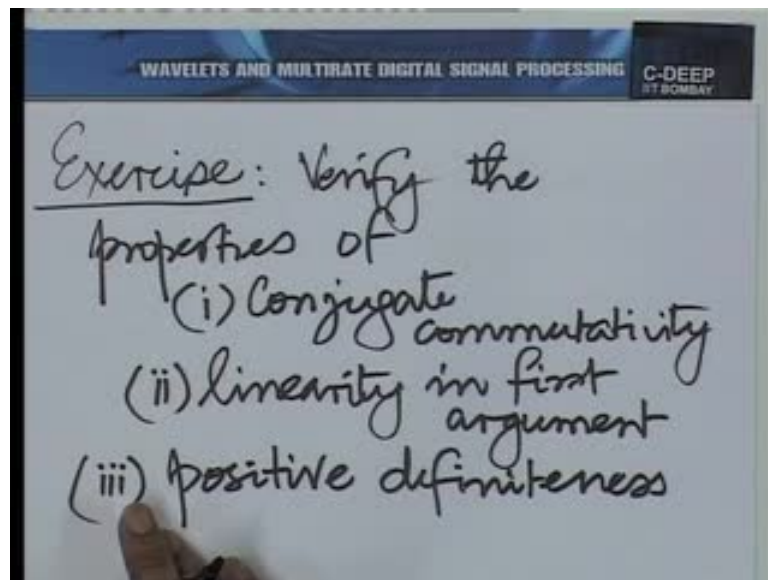
"Dot product" or
"inner product" between
two functions

$$\langle x, y \rangle = \int_{-\infty}^{+\infty} x(t) \overline{y(t)} dt$$

The intuitive understanding will not be different from a more rigorous understanding for those specific situations, but it may not quite be complete. Even so we would not suffer too much in our study of wavelets, in our applications of wavelets if we take this intuitive path to some extent, not all the time, I mean, to some extent in the context of dealing with infinite dimensional spaces. So, with that little prelude, let us come back to this uncountably infinite dimensional space of functions on the real line, in which case we can generalize. So, we can generalize the notion of a dot product or inner product between 2 functions.

Essentially, if I take 2 functions, x and y , both on the variable t , that dot product is not going to be a summation any more, but integral. So, $\int x(t) \bar{y}(t) dt$, taking that idea further of multiplying corresponding coordinates and instead of summing, you now integrate. So, the integral replaces the operation of summation here.

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Now, of course, it is easy to verify and I leave that as an exercise to you the properties of linearity and the commutativity and so on. So, I leave it to you as an exercise here; verify the properties of conjugate commutativity. In other words, if I interchange the order of the arguments, there is a complex conjugation involved, 2nd of linearity in the 1st argument.

So, if I take a linear combination of 2 vectors or 2 functions in the 1st argument, then the corresponding inner products are also similarly, linearly combined and 3rd, positive definiteness. So, I leave this to you as an exercise, but what I wish to emphasize at this point is the **Parseval's** theorem of which we are aware in the context of the Fourier transform. So, let me recapitulate, that very important theorem in the context of the Fourier transform.

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Parseval's theorem

$$x(t) \longrightarrow \hat{x}(\nu)$$

Hertz frequency
Variable

$$\hat{x}(\nu) = \int_{-\infty}^{+\infty} x(t) e^{-j2\pi\nu t} dt$$

in Hz

And let us also give an interpretation to it. You see, the Parseval's theorem, as we know it for continuous functions says, that if $x(t)$ has the Fourier transform, now I am going to use the frequency, hertz frequency variable, so this is the hertz frequency variable, ν . In other words, what I mean by that is that the Fourier transform of $x(t)$ is, essentially, integral $x(t) e^{j2\pi\nu t} dt$ integrated overall time t . So, this is the hertz frequency variable in hertz.

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$$\hat{x}(\Omega) = \int_{-\infty}^{+\infty} x(t) e^{-j\Omega t} dt$$

angular frequency
variable for $\Omega = 2\pi\nu$
continuous time

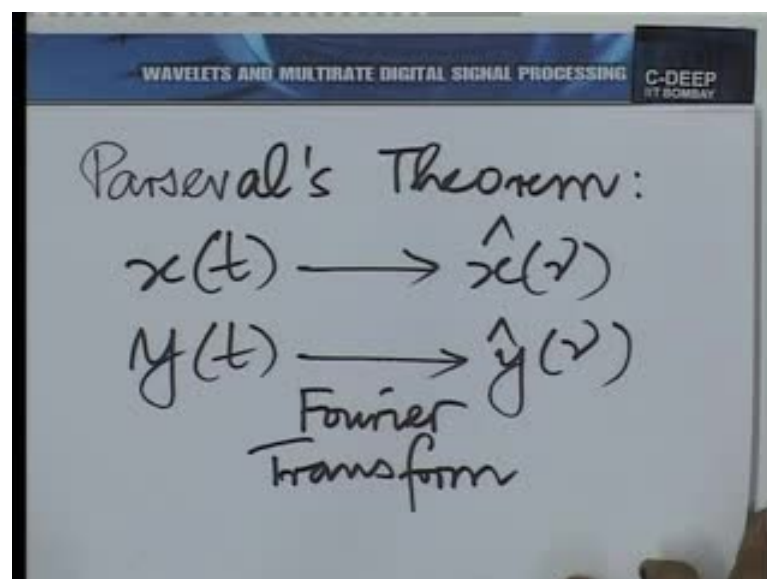
Recall, that you can also have an angular frequency variable. So, for example, you could write x cap of ω now and use this capital ω . When we are talking about continuous time we are going to follow some, **notions of**, different notation for continuous time and discrete time.

So, we use this as the angular frequency variable for continuous time in which case x cap ω is x of $t e^{j\omega t}$. And there are simple relation between capital ω and ν , ω is $2\pi\nu$ angular frequency in hertz frequency. Well, simple things, but we should put down all our cards in the beginning, so we do not get confused later.

Now, again this is a little bit of abuse of notation because I am using x cap of capital ω here and I am using x cap of ν there, and depending on the context, I must interpret either hertz frequency in the argument or angular frequency in radians per second in the argument.

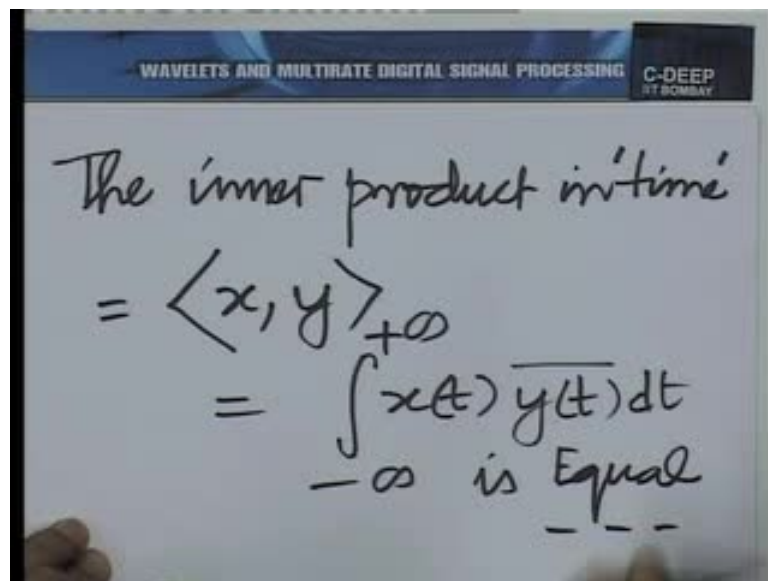
Normally, from the context, it shall be clear and if there is some confusion likely, we will make it clear by expressive statement, but remember that from the context we should be clear, whether we are dealing with hertz frequency or angular frequency radian per second. Anyway, with these details, let us come back to the Parseval's theorem.

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What does Parseval's theorem say? The Parseval's theorem says the following - if you have the Fourier transforms of x and y , so if $x(t)$ has the Fourier transform $X(\omega)$ and $y(t)$ has the Fourier transform $Y(\omega)$, this arrow denotes the Fourier transform, then there is an equivalence of the Fourier transform inner product and the time inner product. That is what the Parseval theorem says in our language, now.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP IIT BOMBAY

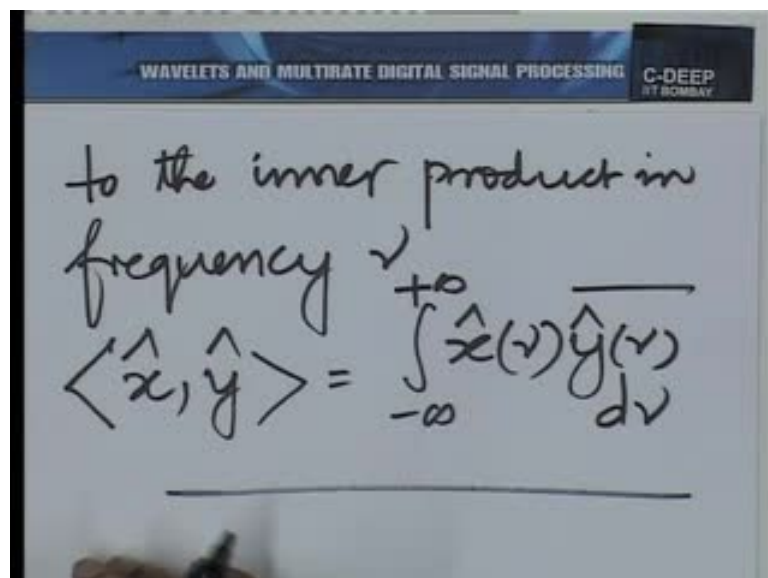
The inner product in time

$$= \langle x, y \rangle_{+\infty}$$

$$= \int_{-\infty}^{+\infty} x(t) \overline{y(t)} dt$$

--- is Equal ---

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP IIT BOMBAY

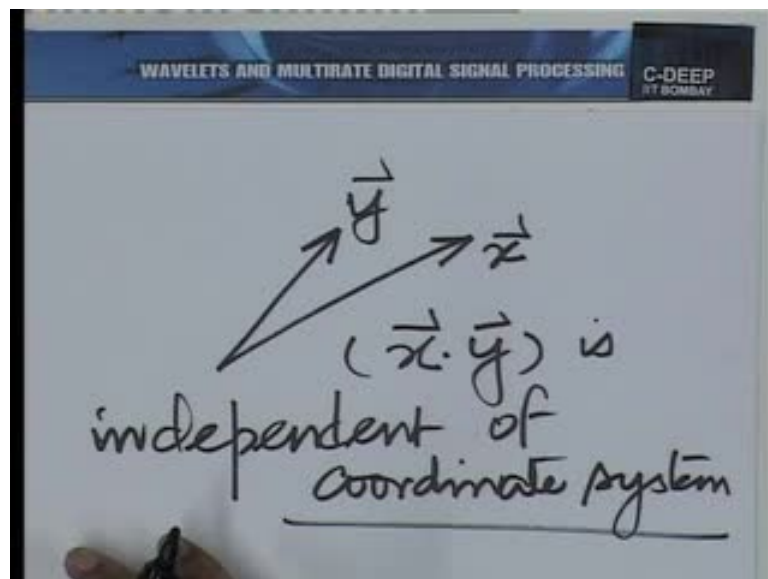
to the inner product in frequency

$$\langle \hat{x}, \hat{y} \rangle = \int_{-\infty}^{+\infty} \hat{x}(\omega) \overline{\hat{y}(\omega)} d\omega$$

So, the inner product in time, so to speak, is equal to the inner product in frequency. In other words, if you take x cap and y cap and construct their inner product in the same way, treating the frequency as the independent **variable or the argument...**

Now, this is very beautiful and a very powerful interpretation of the Parseval's theorem. When we talk about the inner product perspective, we were, we have a very different way of looking at Parseval's theorem and in fact, if we really think of it a little more deeply, Parseval's theorem become so much more intuitive when we talk in terms of inner products. And let me take a minute now to show you why it is so intuitive.

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Indeed, what Parseval's theorem says, in the language of inner product is this and let us do the same in 2 dimensions and then, it will be absolutely, amply clear.

So, I have 2 vectors, let us call them x and y . Now, what Parseval's theorem says is x dot y is independent of the coordinate system, simple enough. What coordinate system we choose to represent x and y does not affect the inner product, that is what Parseval's theorem says, in a way. **And to strengthen...**

See, it may not be obvious to you, why Parseval's theorem relates to this statement, it is obvious for two-dimensional vectors that the inner product is or the dot product is independent of the coordinate system. What is not obvious is, why is this related to the

Parseval's theorem? Well, towards that we need to go back to what x cap nu really is in a way and that will become clear if we write down the inverse Fourier transform.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP IIT BOMBAY

$$x(t) = \int_{-\infty}^{+\infty} \hat{x}(\nu) \cdot e^{j\nu t} d\nu$$

Reconstruct $x(t)$ from 'components' $\hat{x}(\nu)$.

So, we can write down $x(t)$ in terms of the inverse Fourier transform as x cap nu $e^{j\nu t}$ $d\nu$, ν is the hertz frequency variable again. So, in a way, what we are saying is, we are reconstructing $x(t)$ from its components. Each of the x cap nu for different values of ν is a component here and this is the way we have reconstructed $x(t)$ from its components, and in reconstruction we have used these vectors. Each of these $e^{j\nu t}$ is like a vector, is a function of the real axis.

The only catch is $e^{j\nu t}$ is not in L^2 function, so we have to deviate little bit there, from our discussion. But if we choose to ignore that fact, we have essentially taken these coordinates, multiplied them by the corresponding, so called, functions along each of the coordinates ν and added them to get the function $x(t)$.

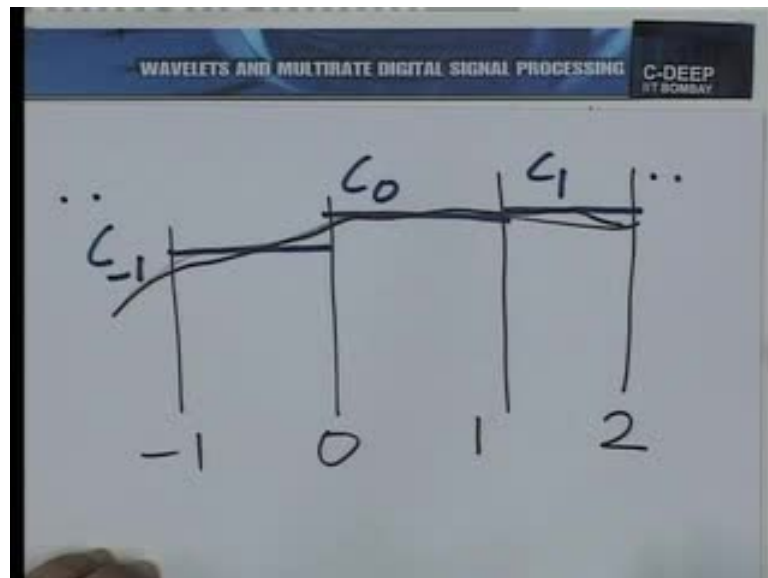
So, each of the x cap nu is like a different expression of the same vector x in a different coordinate system. So, what we are saying in Parseval's theorem is that the dot product is independent of the coordinate system. Whether we choose to use the standard coordinate system of time to represent the function or the slightly less obvious coordinate system of frequency to represent the same function, the dot product remains the same.

So, these and some other such interpretations are what are offered when we represent functions in terms of vectors or when we think of functions as generalizations of the ideas of vectors.

And now, for the last remark in this lecture, which we shall build on even greater in depth in the next, namely, what is a connection between functions and sequences, continuous functions and sequences.

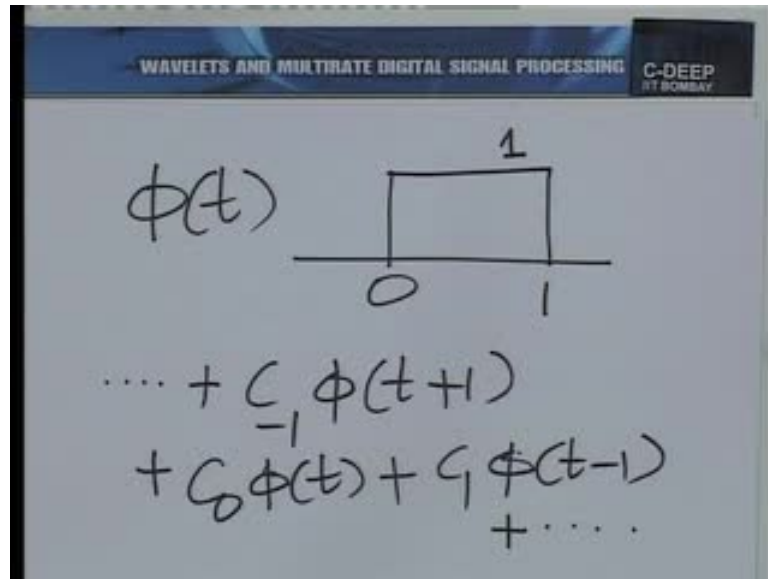
Just to initiate the discussion here, without completing it, completing it or rather taking it further, we shall do it in the next lecture, but just to initiate the discussion let us go back to the idea of piecewise constant approximation.

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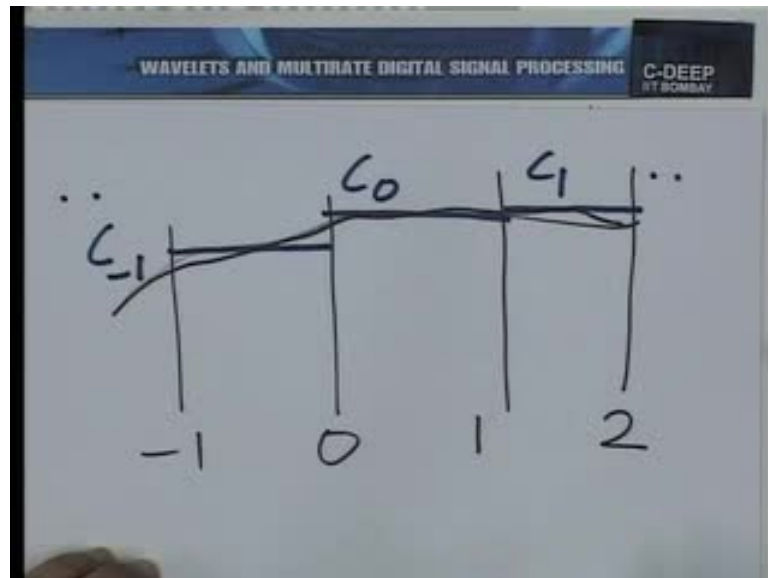
So, suppose we have this piecewise constant approximation of this function on intervals of length 1. So, I take the standard unit intervals and I make a piecewise constant representation of a function. So, I have this, so let the values be, let us say, C_{-1} here, C_0 there, C_1 there and so on, so forth.

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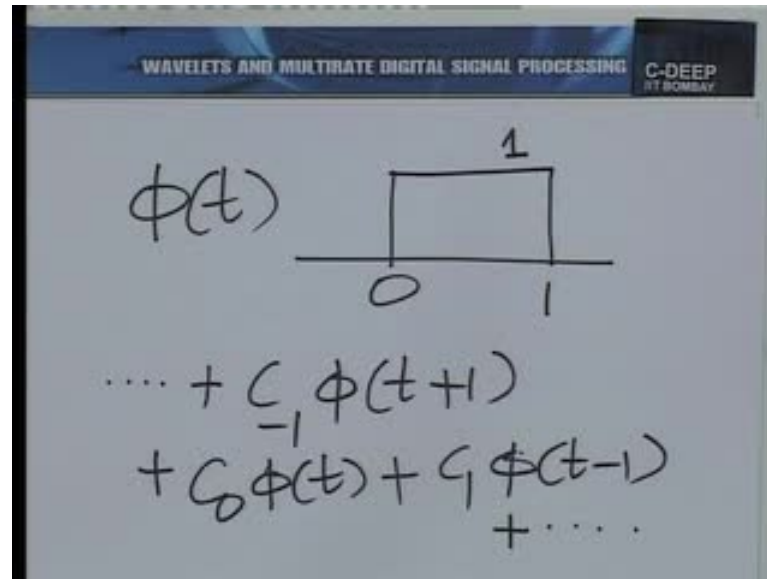


Now, it is very easy to see, that if I take the basic function $\phi(t)$, describe this way, 1 between 0 and 1 and 0 elsewhere, then this piecewise constant representation can be written as c_{-1} among other terms $\phi(t+1)$ plus $c_0 \phi(t)$ plus $c_1 \phi(t-1)$ and what have you afterwards.

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So, to conclude just this introduction of this correspondence, we can note, that equivalent to this piecewise constant representation, that I have here, this function in v_0 , that we talked about last time, equivalent to that function is the set of values: C_{-1} , C_0 , C_1 , and so on.

So, the sequence C_n , n overall the integers is equivalent to that piecewise constant function in v_0 . Any of them can be constructed from the other. From that piecewise constant function we can construct the sequence, from the sequence we can construct the piecewise constant function, given $\phi(t)$.

Now, this equivalence is what we shall take further and delve into deeper in the next lecture and in the next lecture, we should also build further these ideas of vector functions and sequences.

Thank you.