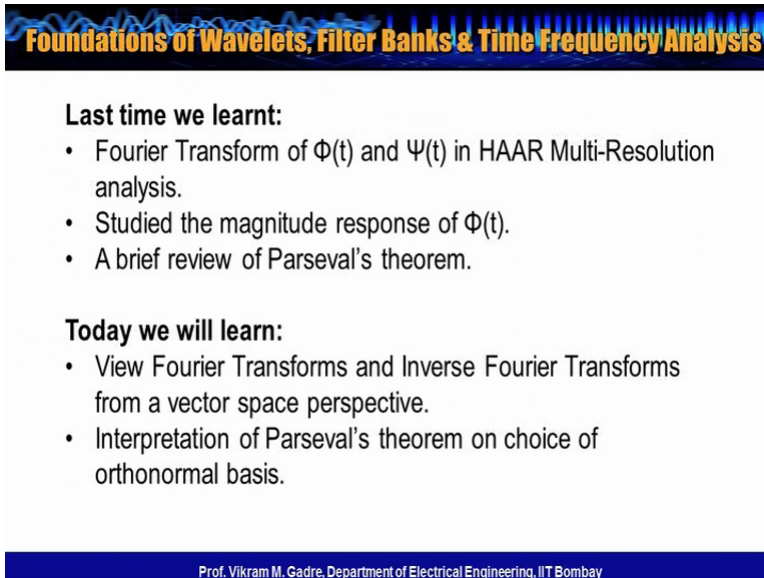


Foundations of Wavelets, Filter Banks and Time Frequency Analysis.
Professor Vikram M. Gadre.
Department Of Electrical Engineering.
Indian Institute of Technology Bombay.
Week-6.
Lecture -15.2.
Revisiting Fourier Transform and Parseval's Theorem.

(Refer Slide Time: 0:19)



Foundations of Wavelets, Filter Banks & Time Frequency Analysis

Last time we learnt:

- Fourier Transform of $\Phi(t)$ and $\Psi(t)$ in HAAR Multi-Resolution analysis.
- Studied the magnitude response of $\Phi(t)$.
- A brief review of Parseval's theorem.

Today we will learn:

- View Fourier Transforms and Inverse Fourier Transforms from a vector space perspective.
- Interpretation of Parseval's theorem on choice of orthonormal basis.

Prof. Vikram M. Gadre, Department of Electrical Engineering, IIT Bombay

You are representing a function with respect to the basis formed by the rotating complex exponentials or the phasors for different frequencies Ω . The Fourier transform is essentially a projection of a function, in this case a function in $L^2\mathbb{R}$ on a particular element of that basis, a particular rotating phase. The inverse Fourier transform reconstructs the original function from its components. It is worth recalling some of these points, finer points because it helps for them to be firmly embedded in our consciousness in a course like this. So even if it means a little bit of repetition, let us emphasise those points again.


(Refer Slide Time: 1:29)

WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
IIT BOMBAY

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

FOURIER TRANSFORM

"Component" of $x(t)$
"along" $j\omega t$




WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
IIT BOMBAY

Inverse Fourier Transform


$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \hat{x}(\omega) e^{j\omega t} d\omega$$

"Reconstruction from components"



WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
IIT BOMBAY

Parseval's Theorem

$$\langle x(t), y(t) \rangle = \frac{1}{2\pi} \langle \hat{x}(\omega), \hat{y}(\omega) \rangle$$


What we said was if I took the Fourier transform, so if I take $x(t)$ and its Fourier transform $X(\omega)$ so to speak, then $X(\omega)$ is essentially a projection $x(t)$ raised to the power $e^{-j\omega t}$. A component of $x(t)$ along $e^{-j\omega t}$. This is a Fourier transform and the inverse Fourier transform, this is, so let us write that down, this is the Fourier transform. And we have the inverse Fourier transform which reconstructs $x(t)$ from its components with a factor of 2π here. And you will recall the interpretation that we gave this.

We had said that essentially, this is the component along a particular ω and this is the so-called unit vector with the factor of 2π . So if I take this and this together, it is like a unit vector here. And what we are seeing here is the original function is essentially the component multiplied by the unit vector integrated over all the components. Reconstruction of a vector, reconstruction from components, that is the interpretation. Now, with this background what we have said in Parseval's theorem is the following. These are products of $x(t)$ and $Y^*(t)$, so, you know you are talking about 2 different domains is equal to the inner product of $X(\omega)$ and $Y(\omega)$ but with the factor of 1 by 2π .

In the language of components, what is the interpretation? The interpretation is that in calculating the inner product it does not matter whether one is using one orthonormal basis or another, the result is the same. So the inner product is not a function of which orthonormal basis you use, you could be using a time basis, you could be using a frequency basis, you could be using impulses in time, you could be using the complex exponentials. The inner product has nothing to do with the choice of basis. The inner product between 2 vectors remains the same whatever basis we choose to express the vectors, that is the statement being made in Parseval's theorem here.

We are saying represent the functions in the natural basis of impulses or represent them in its Fourier basis, the inner product is the same. Of course to the within a factor of constant, this constant appears because of the angular frequency radians, radians per second I mean, otherwise if you were to take heard frequency, this factor would be absent as well. Anyway, this was an important result that we had seen when we looked at functions from a perspective of vectors. And now we shall use this Parseval's theorem to interpret this idea of projection onto VM.

(Refer Slide Time: 6:39)

WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
IIT BOMBAY

$$\langle \chi(\cdot), \phi(2^m \cdot - n) \rangle$$

orthonormalizing
factor $2^{m/2}$

$$= \frac{1}{2\pi} \langle \hat{\chi}(\omega), 2^{-m/2} \phi(2t - n) \rangle$$

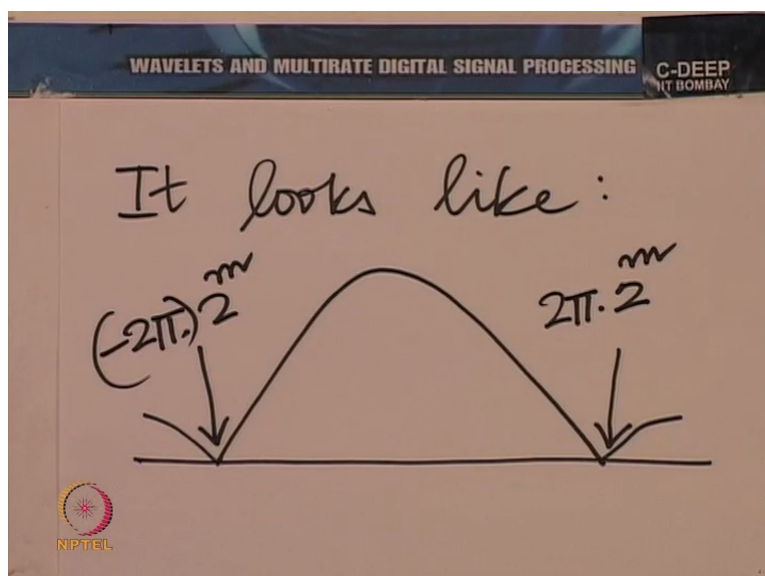
NPTTEL

WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
IIT BOMBAY

We have just seen

$$\left| 2^{-m/2} \phi(2t - n) \right\rangle$$

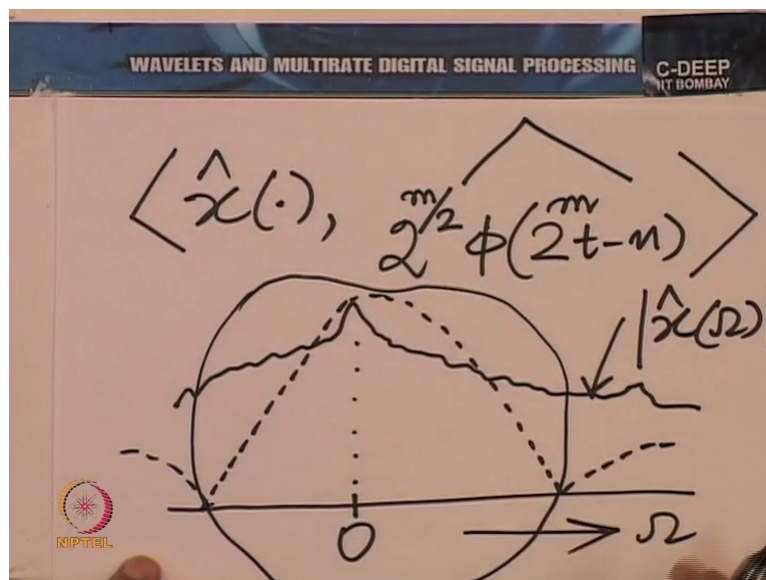
NPTTEL



So when we consider the inner product that we were doing a few minutes ago, the inner product of X with Φ_2 raised to the power m $t - n$ and now if you wish to make it orthonormal, then introduce the orthonormalizing factor, 2 raised to the power m by 2 , this is going to be equal to 1 by 2 pie times the Fourier transform of each of them. So the Fourier transform of X , of course with its argument, the argument remember is going to be ω here. And I shall write here the Fourier terms of this whole thing, so please do not misunderstand what I am writing to be the expression itself but understand it to be the Fourier transform of this.

So you have 2 raised to the power $mt - n$ but take the Fourier transform, so I am saying Fourier transform of this whole thing here. Now I would like to interpret this graphically 1^{st} . What are we going to do when we take the inner product in the Fourier domain? So what is this going to look like? We have just seen the magnitude, we have just seen the magnitude of this quantity and that looks like this. This is the place where you have 2 pie multiplied by 2 raised to the power of m . So in effect you have expanded that main lobe and all the side lobes by a factor of 2 raised to the power of m , in particular which m equal to $+1$ you have expanded by 2 , if m is equal to -1 you would have contracted it by factor of 2 and so on.

(Refer Slide Time: 9:42)

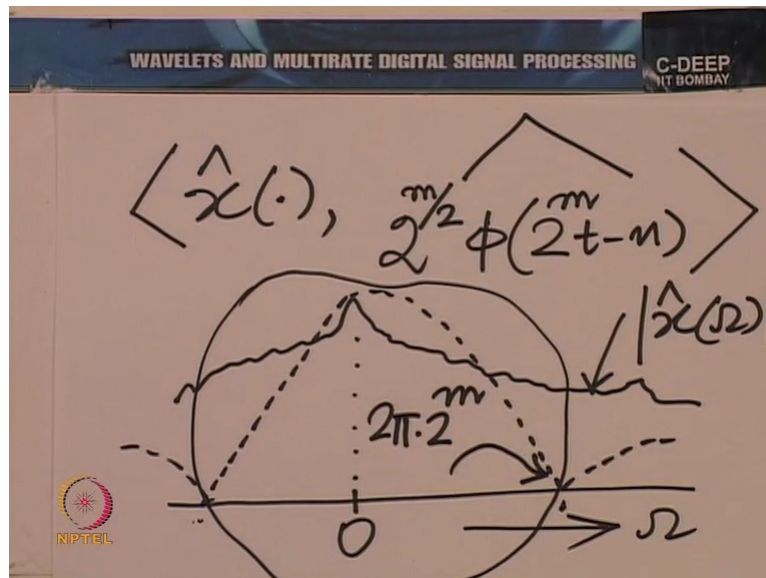


So when we take the inner product, in the inner product X cap and the cap of this, what are we going to do, we have the Fourier transform, so let us understand it graphically as I said. Let us say this is the Fourier transform of X , whatever it be, I mean let us understand in the magnitude sensors. So let this be the magnitude of the Fourier transform of X , this is 0 let us say here and this is the magnitude of the Fourier transform of the other argument, I am

showing only the main lobe and a part of the other side lobes. When you multiply them, their magnitudes are going to get multiplied and therefore essentially you are going to extract this band so to speak.

In a notional sense we are going to emphasise most the area of the Fourier transform around the main lobe of Phi. You see after all it is the magnitude which plays a significant role here. When we take the dot product we are going to multiply the 2 Fourier transforms and integrate over all frequencies. Where the magnitude is larger, the contribution will be larger, where the magnitude is smaller, the contribution will be smaller. So the side lobes would kind of suppress that part of the Fourier transform and the main lobe will emphasise the corresponding part of the Fourier transform contained under the main lobe.

(Refer Slide Time: 12:18)



What we are saying is that part of the Fourier transform of the original function X which is contained in the main lobe is emphasised as again all the rest in calculating the area. Now this also gives us an interpretation of what happens when we increase or decrease m in frequency. In fact let us look at this drawing once again. This point is $2\pi \cdot 2^m$. So take for example m equal to 0, you are essentially, this dotted line here is the magnitude of the Fourier transform of the appropriate dilate of Φ and of course we do not care so much about the translate, the effect of the translation is only to change the angle and it does not reflect in the magnitude.

So for m equal to 0 we are essentially emphasising a region of the frequency axis broadly speaking between -2π and $+2\pi$. When we take m equal to 1, we are emphasising region

between -4π and $+4\pi$. When we take m equal to -1 , we are emphasising a region between $-\pi$ and $+\pi$ and so on so forth ad infinitum. What are we saying? When we increase m , $m=1, 2, 3$ and so on, we are effectively keeping more and more information around the 0 frequency, we are broadening it. Of course we are narrowing in time but we are broadening in frequency, so we are keeping a larger band of frequencies but all around the 0 frequency. Now, what happens when we consider $\psi(t)$ that is an equally interesting interpretation, let us do that.