

Advanced Digital Signal Processing - Wavelets and Multirate

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Lecture No. # 15

Time and Frequency Joint Perspective

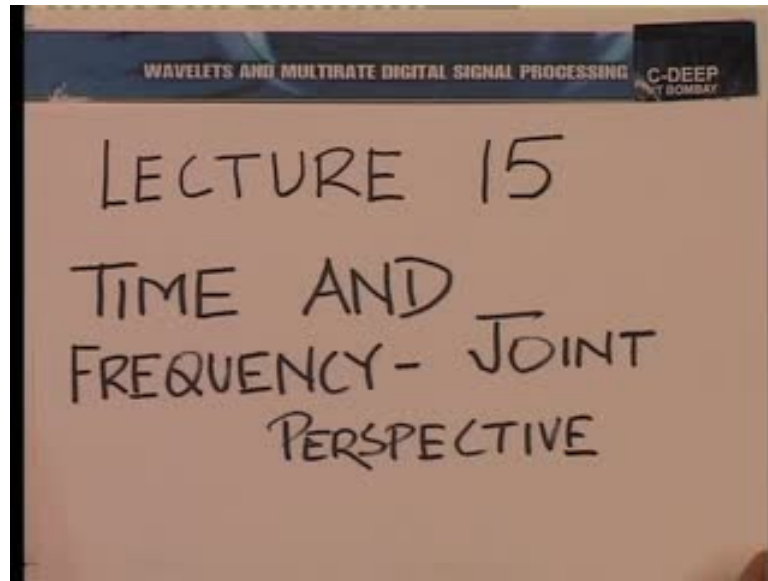
A warm welcome to the 15th lecture, on the subject of Wavelets and Multirate digital signal processing. We had raised a few questions in the previous lecture, which we set out to answer in this lecture. The questions pertain to the fact, that we were building more than one multiresolution analysis and specifically the question was, what are we looking for; when we go from one multiresolution analysis to the other? Why cannot we be content with Haar multi resolution analysis? We have been singing the phrases of the Haar multiresolution analysis so frequently, in some of the previous lectures.

We been saying that, it tells us most of the things that a multiresolution analysis does. Why then, should we look for others? What is inadequate in the Haar? Well, if you remember, when we began this course; we set out to do something which a basic course on signal theory or signals and systems does not. Namely look at two domains simultaneously. We have been trying to do this, for a while. So, when we took the Haar, we did at some point talk about the filtering aspect of it. We also talked about the ideal to which we strive in the filter bank and if you recall that ideal had to do with the frequency domain, not the time domain.

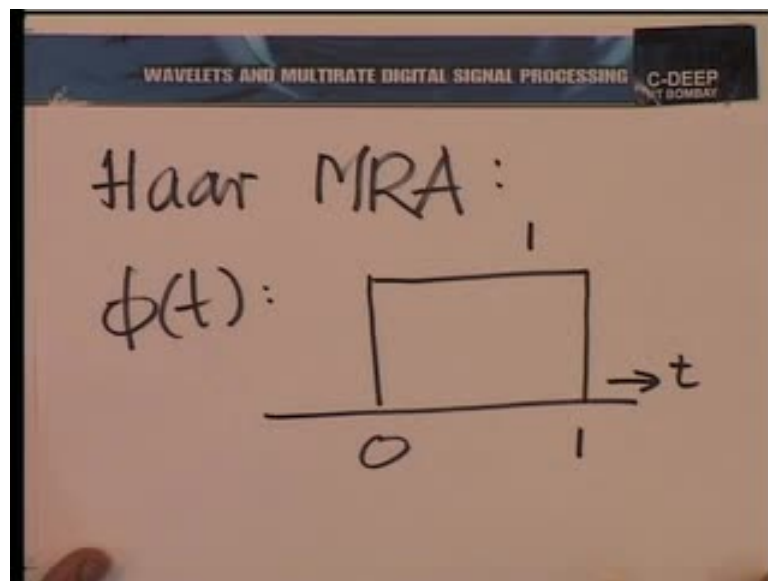
So somewhere, we have not been giving as much importance to the other domain so to say, as we should. We been working in the natural domain of time, but we have not quite being doing justice to the frequency domain. When we take a Haar multiresolution analysis, what does it do in the frequency domain? We have not quite understood this as yet and the first thing that we would like to do is to answer this question. What does the Haar multiresolution analysis do in the frequency domain? And then, how do these filter banks that comprise the Haar, build up to that frequency domain behavior?

So let us set out first, to answer a more basic question. What is the Fourier transform of the scaling function and the wavelet function in the Haar multiresolution analysis?

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So, we shall call today's lecture a joint perspective, time and frequency and we shall set out first, to look at the Fourier transform of $\phi(t)$ and $\psi(t)$ in Haar. The Haar MRA. Let us, begin with the $\phi(t)$ and let us look at this Fourier transform. Remember, how $\phi(t)$ looked, this was $\phi(t)$. Let us obtain its Fourier transform.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP BOMBAY

$$\hat{\phi}(\Omega) = \int_{0}^{1} \phi(t) e^{-j\Omega t} dt$$

analog radian (angular) frequency

$$= \int_{0}^{1} e^{-j\Omega t} dt$$

We shall use this notation for the Fourier transform, a cap on top and capital omega to denote the angular frequency, analog angular frequency, radian or angular frequency and this is easy to calculate. Well, I am saying 0 to 1; actually, I should write minus if you like, I can write minus infinity to plus infinity, but the non zero part is only between 0 and 1. So, it is also alright to write between 0 and 1 only for this specific function.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP BOMBAY

$$= \frac{e^{-j\Omega t}}{-j\Omega} \Big|_0^1$$
$$= \frac{1 - e^{-j\Omega}}{j\Omega}$$

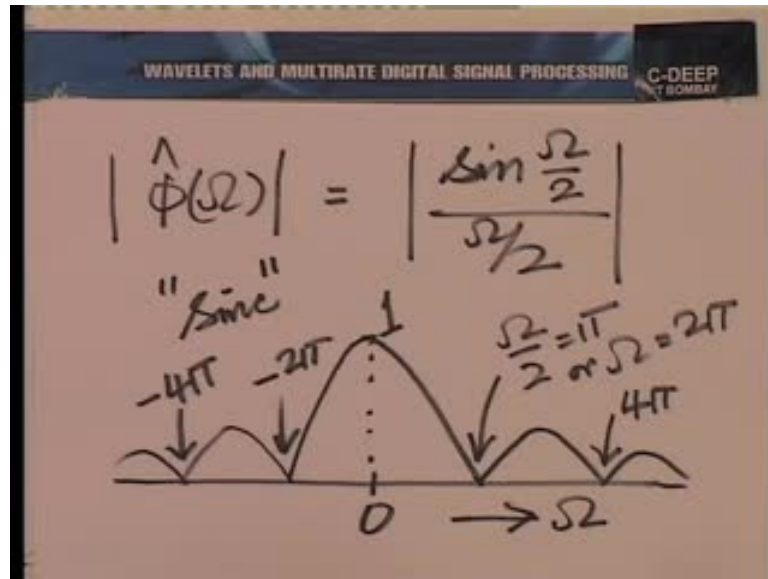
And this of course, equal to integral 0 to 1, e raise to the power minus j omega t dt, which evaluates to e raise the power minus j omega t, by minus j omega from 0 to 1 and that is 1 minus e raise the power minus j omega by j omega and we can simplify this.

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$$\begin{aligned}
 &= e^{-j\Omega/2} (e^{j\Omega/2} - e^{-j\Omega/2}) \\
 &= \frac{e^{-j\Omega/2} (e^{j\Omega/2} - e^{-j\Omega/2})}{2 \cdot j\Omega/2} \\
 &= e^{-j\Omega/2} \cdot \left(\frac{\sin \frac{\Omega}{2}}{\Omega/2} \right)
 \end{aligned}$$

We can take an e raise the power minus j omega by 2 common, in the numerator and that is easy to interpret. In fact if you wish, you can make; even make this by j omega alright 2, and then this becomes e raise the power minus j omega by 2. If I take the 2 j and this together, I get sin omega by 2 there and omega by 2 here. So, this is the Fourier transform.

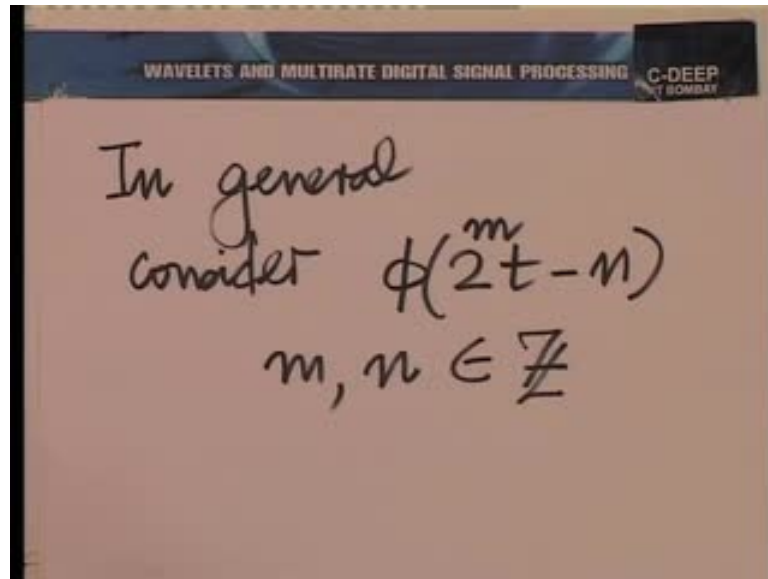
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Now, let us look at the magnitude of this Fourier transform. So, in fact I could straight away sketch this. It is essentially the magnitude of $\sin \omega$ by 2 by ω by 2 . A sketch would look like this. This is the very familiar function to most electrical engineers. We call it, the so called Sinc function, the Sinc pattern. People have different names for it; they call it the sampling function, the Sinc function and whatever other names. Anyway this is the point where ω by 2 is equal to π or ω is 2π . And this of course, is the point where ω is 4π and this where it is minus 2π , this where it is minus 4π and so on so forth.

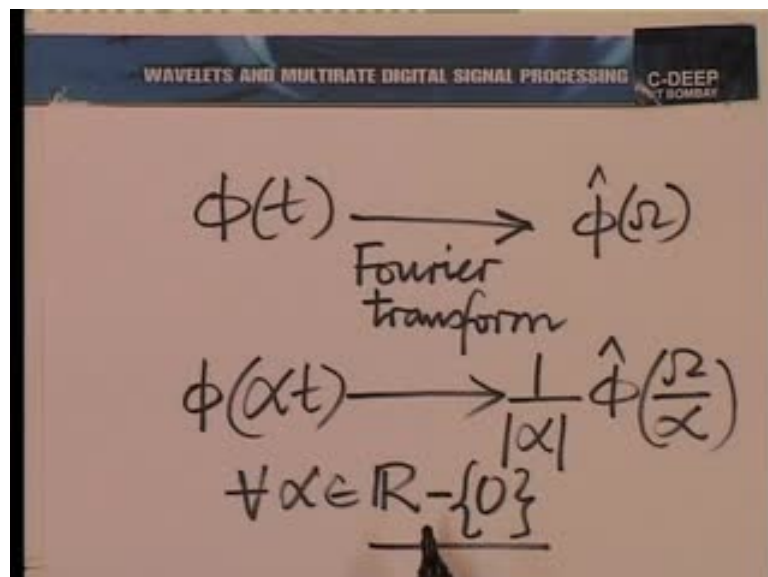
At this point the magnitude takes the value 1. In fact, we call that the magnitude of the Fourier; in fact, the value of the Fourier transform at ω equal to 0 is indicative of the area and ϕt . So, the integral of ϕt overall t .

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Anyway so much so for the magnitude and now, we know what to do when we compress and expand? Let us look in general, at the Fourier transform of phi 2t minus n, so we should interpret that. You know we are talking about dyadic dilates and translates. So, let us consider phi, in fact, 2 raise the power of m t minus n if you please, for m and n belonging to the set of integers.

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Now, you will recall that there is a very simple result in the Fourier transform which says, that if phi t has the Fourier transform phi cap omega as we do. In general, then phi

$\phi(t)$ has the Fourier transform, 1 by mod α $\hat{\phi}(\omega)$ by α , for all α belonging to the real numbers other than 0 . This notation says all real numbers except 0 . Of course, α cannot be 0 . So for example, even if α is negative we can use this. In particular for example, if α is minus 1 we have a reflection of the Fourier transform as well.

So, using this, we now take care of the two so called distortion or modification that we made in ϕ ; the translation and the dilation. In fact, the translation does not affect the magnitude. The translation only affects the phase or the angle of the Fourier transform. So, I can even forget about the translation, I need only look at the 2 raise the power of m term there.

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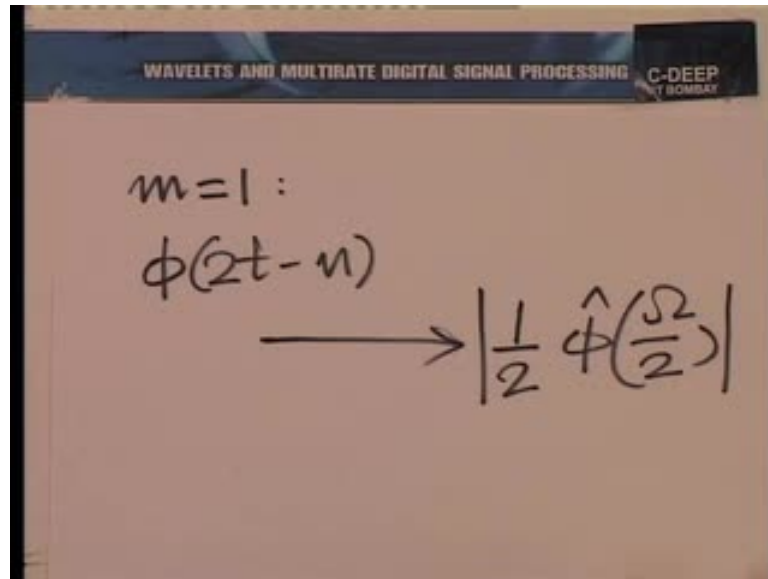
Magnitude only:

$$\phi(t) \xrightarrow{\text{Fourier Trans}} |\hat{\phi}(\omega)|$$

$$\phi(2t-n) \xrightarrow{\text{Fourier Trans}} \left| \frac{1}{2^m} \hat{\phi}\left(\frac{\omega}{2^m}\right) \right|$$

So, here, if I restrict myself to magnitude only then $\phi(t)$ has the Fourier transform, $\hat{\phi}(\omega)$ with the magnitude of mod of this. $\phi(2t-n)$ would then have a Fourier transform mod $\hat{\phi}(\omega)$ divide by 2 raise the power of m . Of course, with the constant, the same 2 raise the power of m , but in the denominator here.

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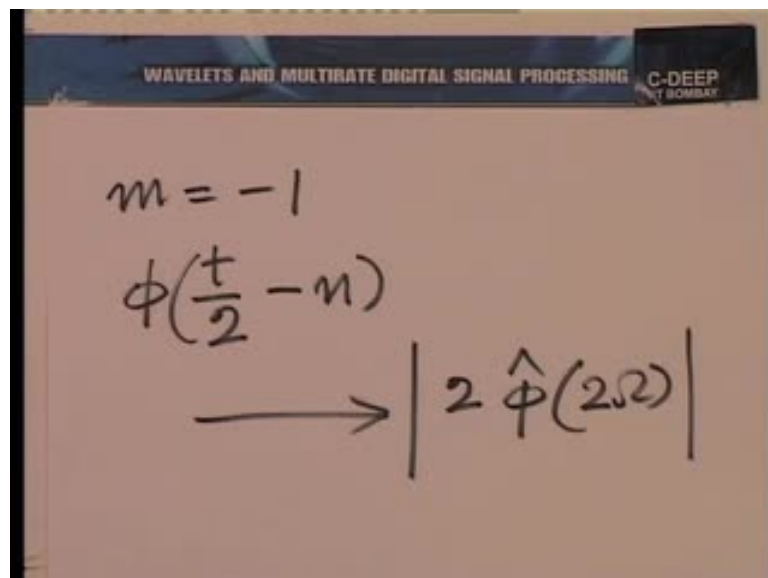


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP T BOMBAY

$$m=1:$$
$$\phi(2t-n) \longrightarrow \left| \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right) \right|$$

So, Of course, let us take an example, suppose m is equal to 1 and minus 1 to fix our ideas and n equal to minus 1.

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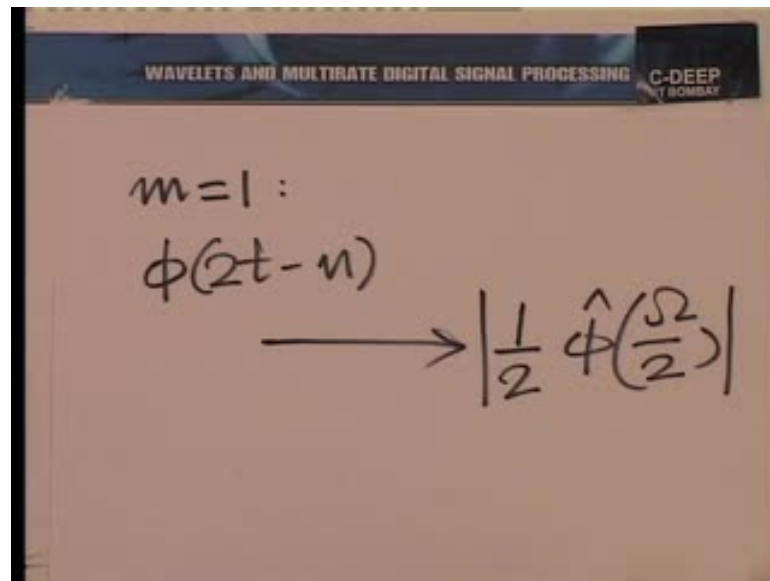


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP T BOMBAY

$$m=-1$$
$$\phi\left(\frac{t}{2}-n\right) \longrightarrow \left| 2 \hat{\phi}(2\Omega) \right|$$

Now, notice that the n is entirely absent here.

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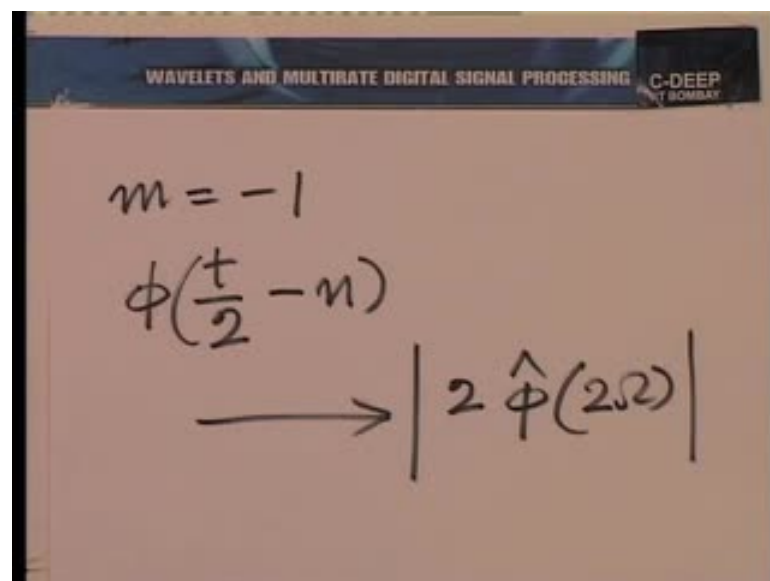


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP BOMBAY

$$m=1:$$
$$\phi(2t-n) \longrightarrow \left| \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right) \right|$$

Or here.

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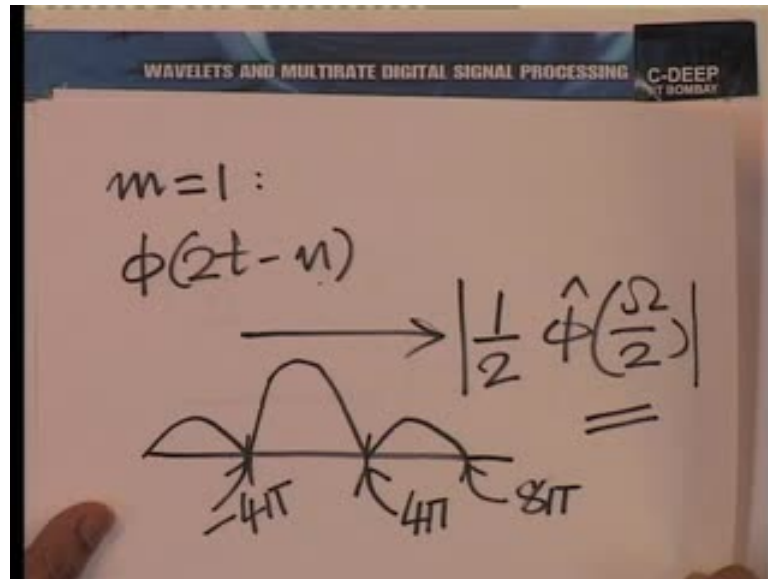


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP BOMBAY

$$m=-1$$
$$\phi\left(\frac{t}{2}-n\right) \longrightarrow \left| 2 \hat{\phi}(2\Omega) \right|$$

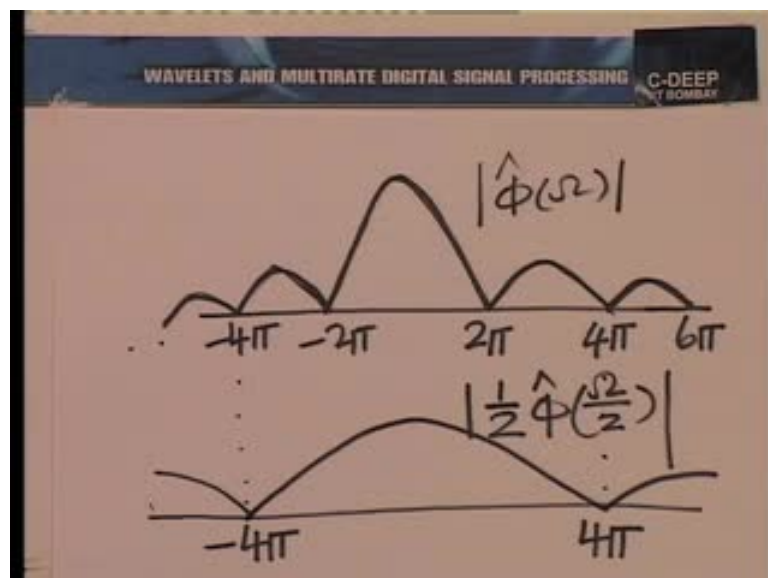
The n is irrelevant as far as the magnitude goes; the n only contributes to the phase. Let us sketch both of these, how would these look?

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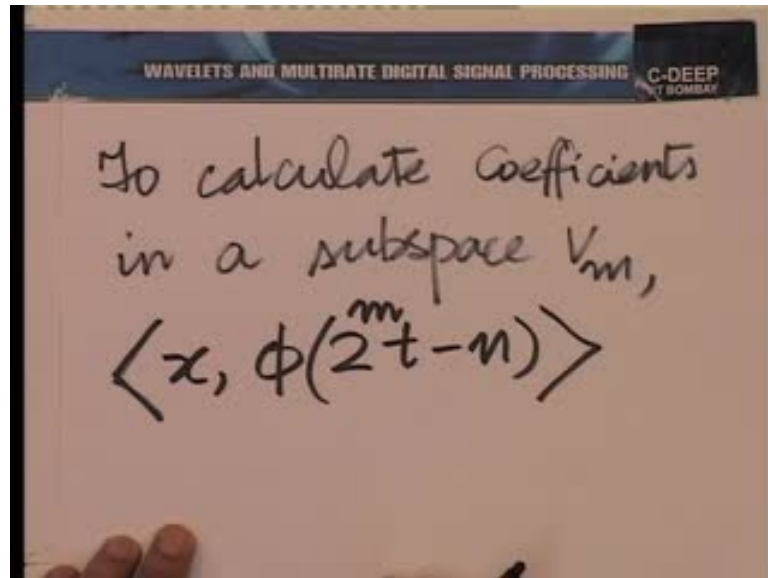
So, let me take the m equal to 1 case. Let me sketch this, so omega by 2. So, if I focus my attention on the main lobe and the principle side lobe that I have here. This was originally 2 pi, now it is become 4 pi. This was originally 4 pi, now it is become 8 pi and so to on this side. This is minus 4 pi and so on. See a stretched by a factor of 2. Let we put them together, the original Fourier transform and this Fourier transform with m equal to 1.

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and so on. This is I am just sketching $\text{mod } \phi \text{ cap } \omega$ and here I sketch, $\text{mod } \phi \text{ cap } \omega$ by 2 multiplied by half. Of course, this should be smooth here, all these are smooth.

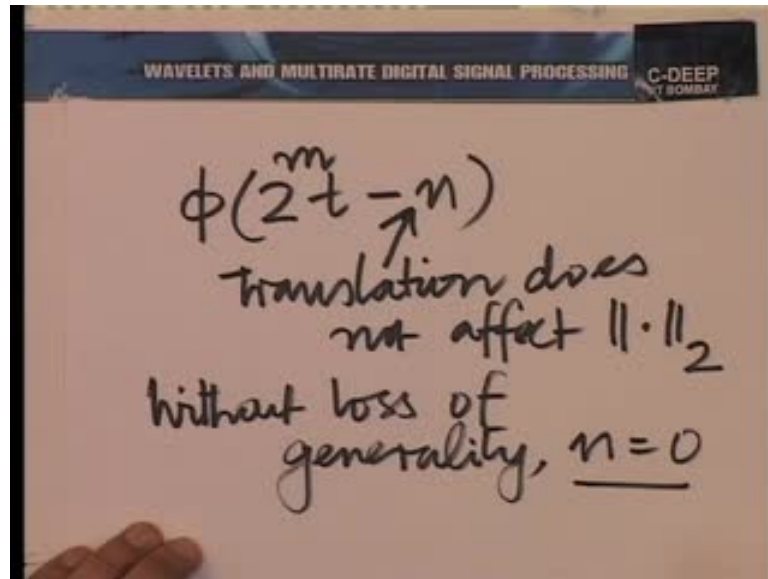
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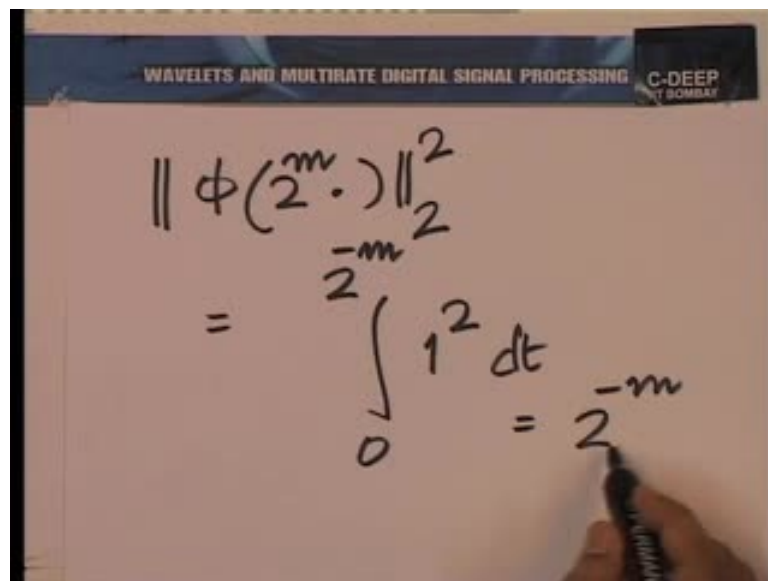
So, as expected when we squeeze in time, we have stretched in frequency and now, let us interpret what we are doing, when we take a dot product, to calculate the coefficients in a V_m . To calculate coefficients in a subspace V_m , what are we effectively doing? We are taking an inner product, namely, the inner product of an x with ϕ 2 raise the power of m t minus n , remember. Now, you know there is of course, a normalization here. So, if you want to work with an Orthonormal basis, then it should not quite be ϕ 2 raise the power of m t minus n . One must normalize it to make it unit norm.

So, I take an instance. Let us take any arbitrary m and let us look at the norm. So, if we consider ϕ 2 raise the power of m t and again, the minus n does not affect the norm. I am talking about the L_2 norm and therefore, we can as well take without loss of generality, we could take n equal to 0.

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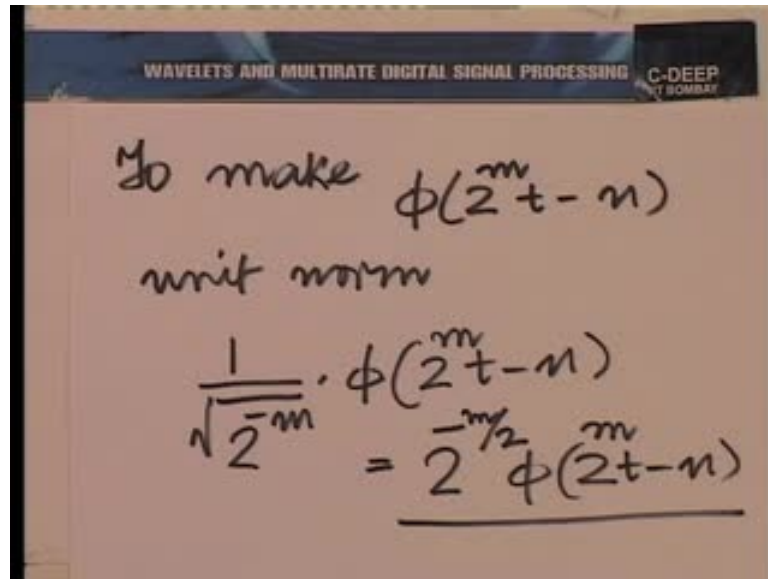


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Let us find out the norm therefore. So, I am putting 2 raise the power of m and then a dot, dot denotes the argument of the function, but we are treating the function as an entity. So, I do not use the explicit argument here. The norm of this is essentially, integral. Now, you know when you go over $\phi(2^m \cdot)$. You are talking about 0 to 2 raise the power minus m here, 1 square dt and that is obviously, 2 raise the power minus m . And therefore, if you want to make this unit norm, then you must divide by; this is of course, the square of the norm.

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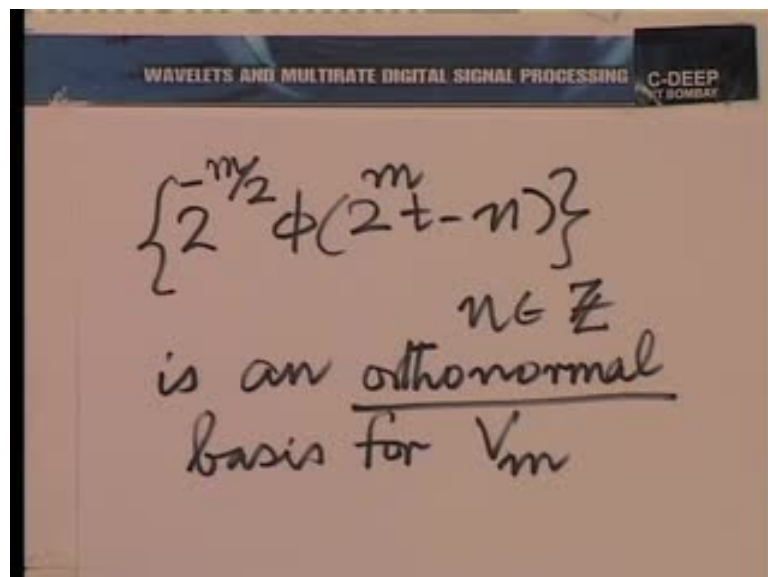


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
T BOMBAY

To make $\phi(2^m t - n)$
unit norm
$$\frac{1}{\sqrt{2^{-m}}} \cdot \phi(2^m t - n)$$
$$= \underline{2^{-m/2} \phi(2^m t - n)}$$

So, if you want to make this unit norm, you must divide the function by the square root of this. We must consider 1 by square root of 2 raise the power minus m times phi 2 raise the power of m t minus n and that is easily seen to be 2 raise the power minus m by 2, phi 2 raise the power m t minus n. So, this is the unit norm, this is now an orthonormal basis. So, let us make that note.

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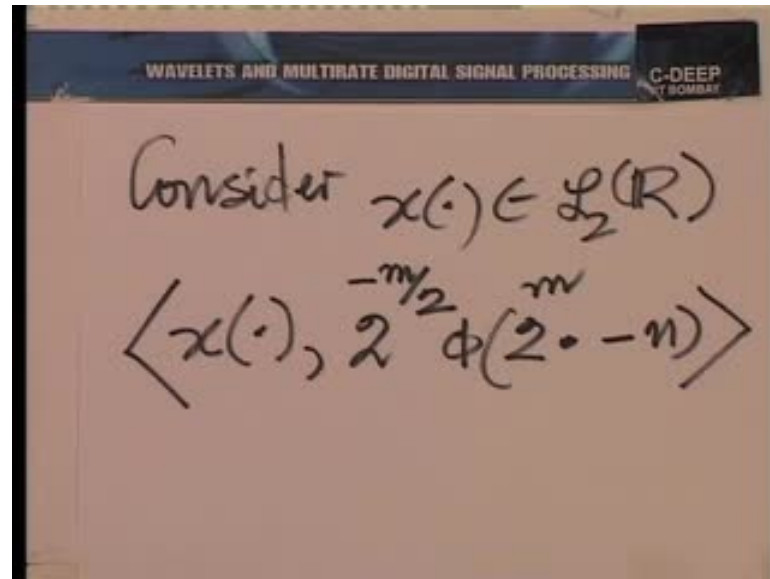
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
T BOMBAY

$\{2^{-m/2} \phi(2^m t - n)\}_{n \in \mathbb{Z}}$
is an orthonormal
basis for V_m

2 raise the power minus m by 2, phi 2 raise the power m t minus n, for all integer n, is an orthonormal basis for V_m . You know, what V_m means? V_m is the mth subspace in the

ladder of subspaces that leads to $L^2 \mathbb{R}$ as you go right wards and to the trivial subspace with only the zero element as you go left wards. So, it is that m th subspace in the ladder, and now we have an orthonormal basis for it. Anyway, now let us interpret what happens, when we take the dot product of any function in $L^2 \mathbb{R}$, with an element of this orthonormal basis.

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So, consider $x(t)$ belonging to $L^2 \mathbb{R}$ or if you please x with the argument belonging to $L^2 \mathbb{R}$ as is the correct way to write it. And then, consider the inner product of this x , with this orthonormal basis element $2^{-m/2} \phi(2^n \cdot -n)$. Now, we are going to invoke, the Parseval's theorem. You will remember that we had discussed the Parseval's theorem a while ago, in one of the earlier lectures. When I had talked about the relationship of functions and vectors, I had mentioned the significance of Parseval's theorem. There are different ways of stating it. Parseval's theorem on one hand says; that the inner product is preserved as we go from time to frequency.

Now, here if we use angular frequency a factor of 2π is needed. If we use hertz frequency, that 2π factor is not required. But since, we are working with angular frequency; it would be safer to retain that factor of 2π . But, that factor of 2π apart, what Parseval's theorem says is that, after all when you go from the function to its Fourier transform in effect, you are representing the same function in a different basis.

You are representing the function, with respect to the basis formed by the rotating complex exponentials or the phases 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 final points because, it helps for them to be firmly embedded in our consciousness in a course like this.

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The image shows a whiteboard with handwritten mathematical notes. At the top, there is a header that reads "WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING" and "C-DEEP BOMBAY". The main content of the whiteboard is as follows:

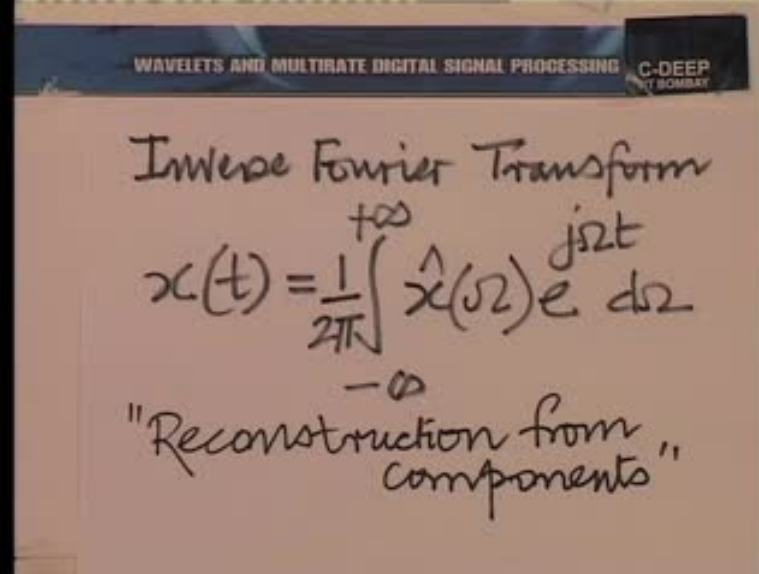
$$x(t) \rightarrow \hat{x}(\omega)$$

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

Below the integral, the words "FOURIER TRANSFORM" are written. Underneath that, there is a note: "Component of $x(t)$ along $e^{j\omega t}$ ".

So, even if it means a little bit of repetition. Let us emphasize those points again. What we said was, if I took the Fourier transform, so if I have $x(t)$ and its Fourier transform, $\hat{x}(\omega)$ so to speak and $\hat{x}(\omega)$ is essentially a projection $x(t) e^{-j\omega t}$ dt. A components of $x(t)$ along $e^{j\omega t}$. This is the Fourier transform and the inverse Fourier transform this is so, let us write this down, this is the Fourier transform.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
T BOMBAY

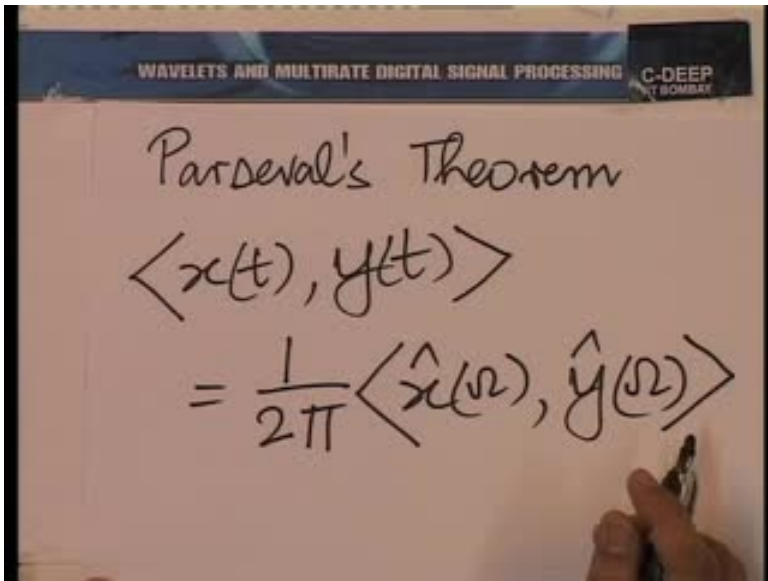
Inverse Fourier Transform

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

"Reconstruction from components"

And we have the inverse Fourier transform, which reconstructs $x(t)$ from its components with the 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 saying 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 such the interpretation.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP
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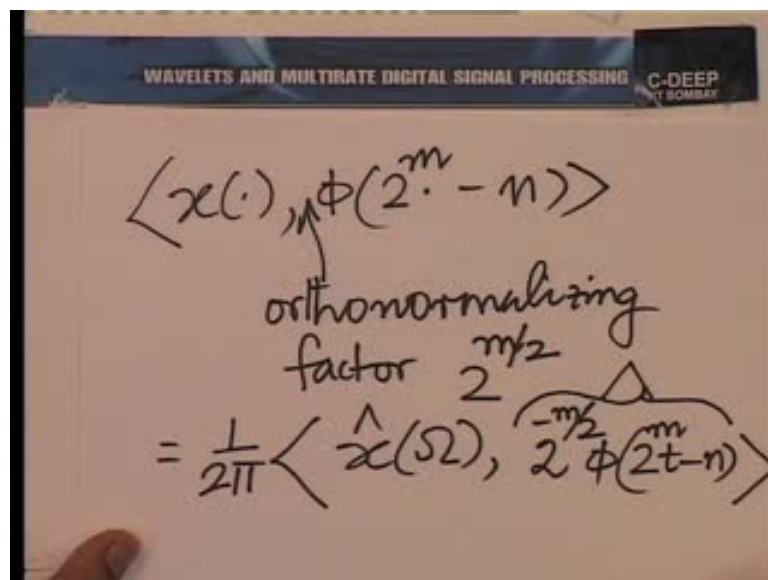
Parseval's Theorem

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

Now, with this spec round, what we have said in Parseval's theorem is the following. The inner product of $x(t)$ and $y(t)$; so, you know I am talking about two different domains is equal to the inner product of x cap omega and y cap 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. The inner product has nothing to do with the choice of basis. The inner product between two 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 within a factor of constant, this constant appears, because of the angular frequency radians, radians per second I mean. Otherwise, if you want to take hertz 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 on to V_m .

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

orthonormalizing
factor $2^{-m/2}$

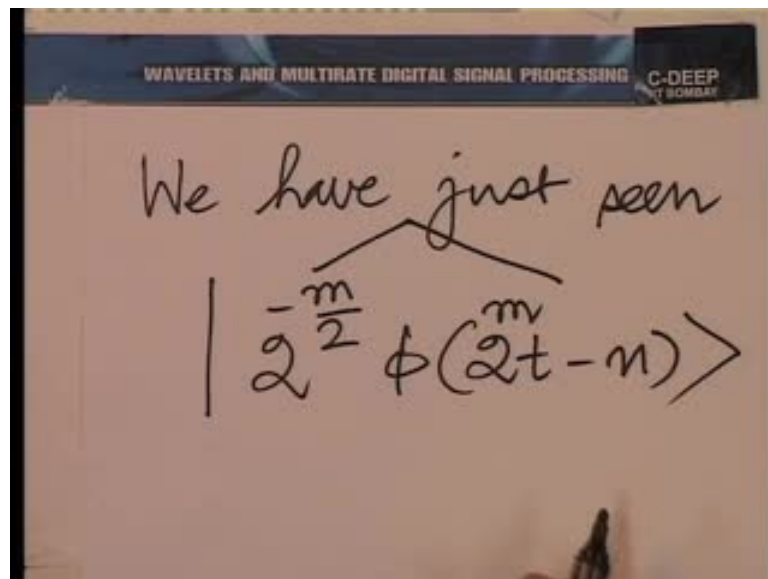
$$= \frac{1}{2\pi} \langle \hat{x}(\Omega), 2^{-m/2} \phi(2^m t - n) \rangle$$

So, when we consider, the inner product that we were doing of few minutes ago. The inner product of x , with $\phi(2^m t - n)$ and all if you wish to make it orthonormal, then introduce the orthonormalizing factor, $2^{-m/2}$. This

is going to be equal to 1 by 2π 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 transform of this whole things. So, please do not misunderstand what I am writing to be the expression itself, but understand it to be the Fourier transform of this.

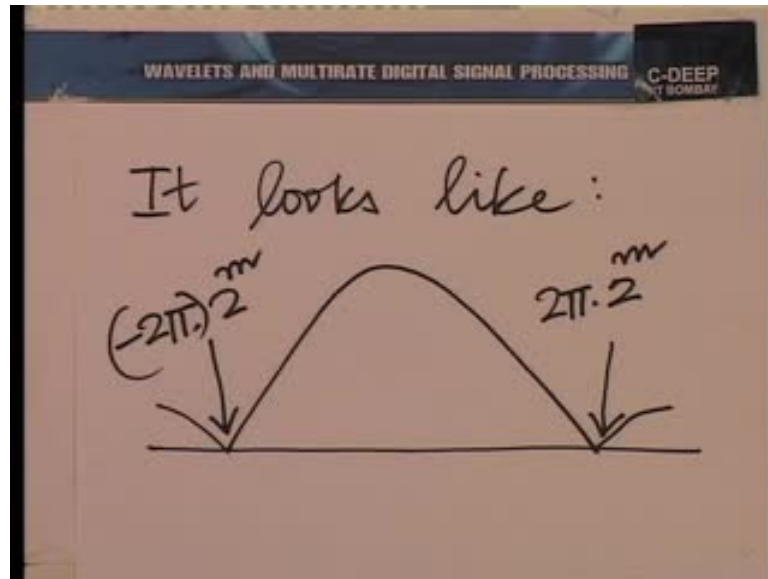
So, you have 2 raise the power of m t minus n , but take the Fourier transform. So, I am saying Fourier transform of this whole thing here. Now, I would like to interpret this graphically first. What are we going to do when we take the inner product in the Fourier domain; so, what is this going to look like?

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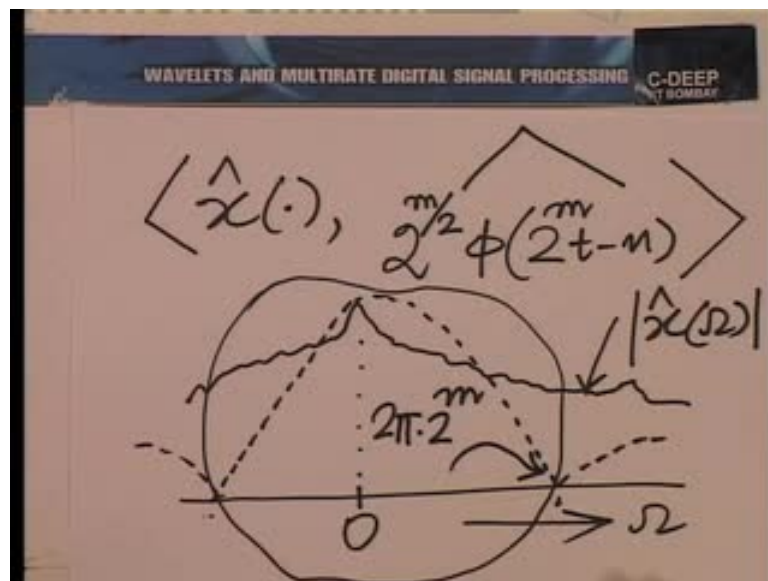


We have just seen the magnitude of this quantity. And that looks like this. This is the place where, you have 2π multiplied by 2 raise the power of m . Some effect you have expanded that main lobe and all the side lobes by a factor of 2 raise the power of m . In particular with m equal to plus 1 , you have expanded by 2 . If m is equal to minus 1 , you would have contracted it by a factor of 2 and so on.

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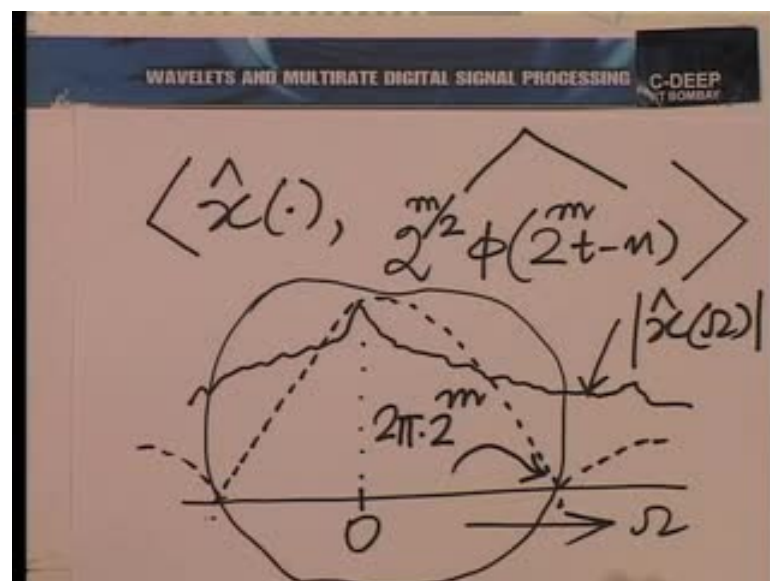


So, when we take the inner product, in the inner product x cap and the cap of this. What are we going to do? We have the Fourier transform, so let us understand it graphically as I said. Let us say, this is the Fourier transform of x whatever it be I mean, let us understand in the magnitude sense first. 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 emphasize most the area of the Fourier transform around the main lobe of ϕ . You see after all, it is the magnitude which plays the significant role here. When we take the dot product, we are going to multiply the two Fourier transforms and integrate overall 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 would emphasize, the corresponding part of the Fourier transform contained under the main lobe. What we are saying is essentially, that part of the Fourier transform of the original function x , which is contained in the main lobe is emphasized as against 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.

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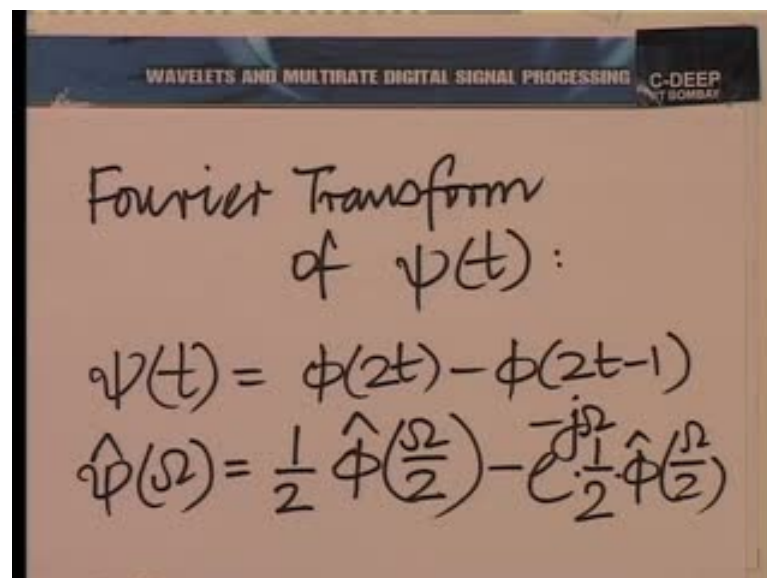


This point is 2π multiplied by 2 raised to the power of m . So, take for example, m equal to 0 . You are essentially this dotted line here, is the magnitude of the Fourier transform of the appropriate dilate of ϕ and of course, we do not care so much about the translate. The effect of the translation is only to change the angle and it does not reflect in the magnitude. So, for m equal to 0 we are essentially emphasizing, a region of the frequency axis broadly speaking, between minus 2π and plus 2π . When we take m equal to 1 , we are emphasizing a region between minus 4π and plus 4π . When we take

m equal to minus 1, we are emphasizing a region between minus pi and plus pi and so on so forth, add and fill it up.

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 zero 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 zero frequency. Now, what happens, when we consider psi t that is unequally interesting interpretation? Let us do that. In other words, here we have analyzed the implications of taking the projection on one of those subspaces in the ladder. What happens when we project on one of the incremental subspaces? That is also an interesting question.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP BOMBAY

Fourier Transform
of $\psi(t)$:

$$\psi(t) = \phi(2t) - \phi(2t-1)$$

$$\hat{\psi}(\Omega) = \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right) - e^{-j\Omega} \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right)$$

So, for that you must first consider the Fourier transform of psi t and that is easy to do. Psi t as you know is phi of 2 t minus phi of 2 t minus 1 and you can easily find its Fourier transform. Psi cap omega is therefore, going to be equal to; well, when you multiply by 2 here, you would be dividing by 2 in the other domain and of course, here we need to take care of the minus 1. So, e raise the power minus j omega times the same expression. We can evaluate this. So, you know it is I think maybe we should it would be better to come to this a little more systematically.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP T BOMBAY

$$\phi(2t) \rightarrow \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right)$$
$$\phi(t-n) \rightarrow e^{-j\Omega n} \hat{\phi}(\Omega)$$

So, $\phi(2t)$ would have the Fourier transform, $\hat{\phi}(\Omega)$ by 2 multiplied by half. Now, $\phi(t-n)$ would in general have the Fourier transform, $e^{-j\Omega n}$ times $\hat{\phi}(\Omega)$ and when we consider $\phi(2t-n)$ in general.

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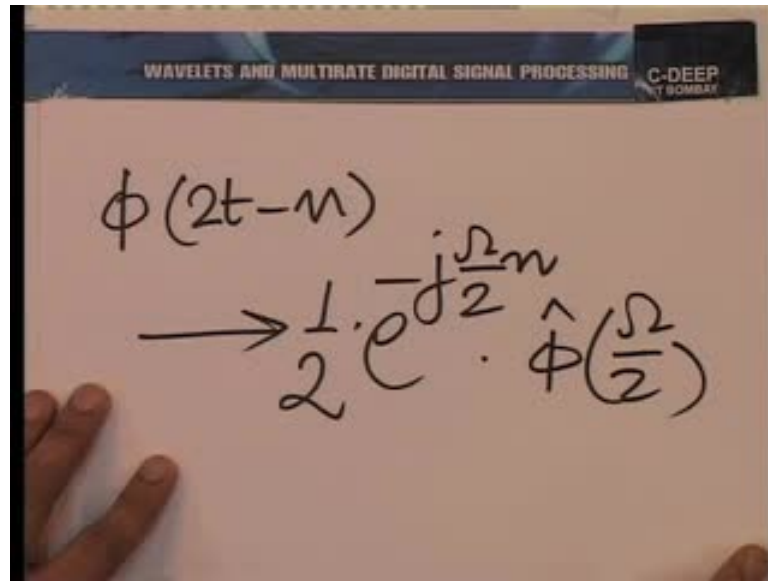
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP T BOMBAY

$$\phi(2t-n) \rightarrow \frac{1}{2} \left\{ e^{-j\Omega n} \cdot \hat{\phi}(\Omega) \right\}$$

$\Omega \leftarrow \frac{\Omega}{2}$

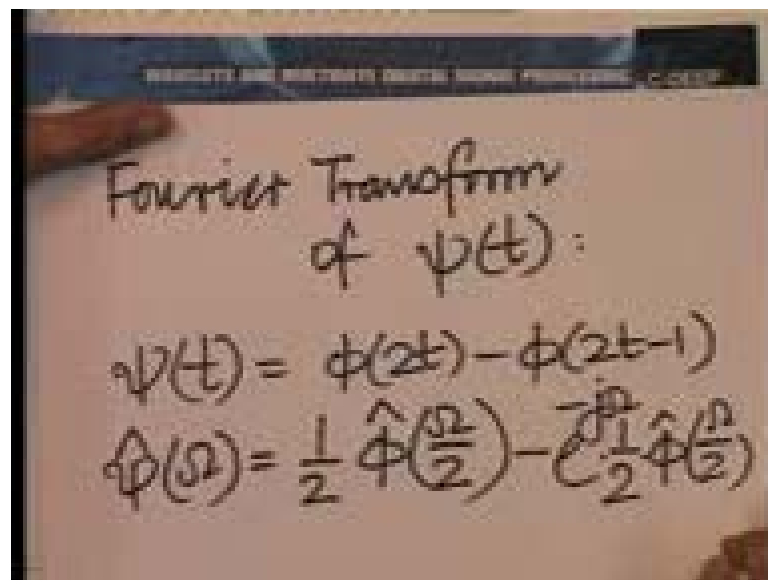
So, here we are going to replace t by $2t$. We should take this replacement Ω by $\frac{\Omega}{2}$ and then multiply by half. So, this replacement must be in the whole thing.

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$$\phi(2t-n) \rightarrow \frac{1}{2} \cdot e^{-j\frac{\Omega n}{2}} \cdot \hat{\phi}\left(\frac{\Omega}{2}\right)$$

And therefore, phi of 2 t minus n would have the Fourier transform, half e raise the power minus j omega by 2 times n, phi cap omega by 2 and therefore, we need to make a little correction here.

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Fourier Transform
of $\psi(t)$:

$$\psi(t) = \phi(2t) - \phi(2t-1)$$
$$\hat{\phi}(\Omega) = \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right) - e^{-j\frac{\Omega}{2}} \frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right)$$

The correction is we need to replace omega by omega by 2 here. So, that we can rewrite that part of the expression. It is more convenient.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP T BOMBAY

$$\psi(t) = \phi(2t) - \phi(2t-1)$$

↓ Fourier Trans

$$\frac{1}{2} \hat{\phi}\left(\frac{\Omega}{2}\right) - \frac{1}{2} e^{-j\frac{\Omega}{2}} \hat{\phi}\left(\frac{\Omega}{2}\right)$$

Psi t which is phi 2 t minus phi 2 t minus 1 has the Fourier transform, phi cap omega by 2 minus e raise the power minus j omega by 2 times phi cap omega by 2.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP T BOMBAY

$$= \frac{1}{2} (1 - e^{-j\frac{\Omega}{2}}) \hat{\phi}\left(\frac{\Omega}{2}\right)$$

$$= \left\{ \frac{1}{2} e^{-j\frac{\Omega}{4}} \cdot 2j \sin\left(\frac{\Omega}{4}\right) \frac{1}{j} e^{j\frac{\Omega}{4}} \dots \right\}$$

And if we aggregate terms, we have phi cap omega by 2 and this is easy to write. This is half; well we can play the same trick. We can extract an e raise the power minus j omega by 4 common from here and get 2 j. I am skipping the couple of steps; sin omega by 4 and you know the expression for this. This is essentially, 1 by 2 e raise the power minus j

omega by 4 times; well, I continue on the next page. It is really complicated, but not too difficult.

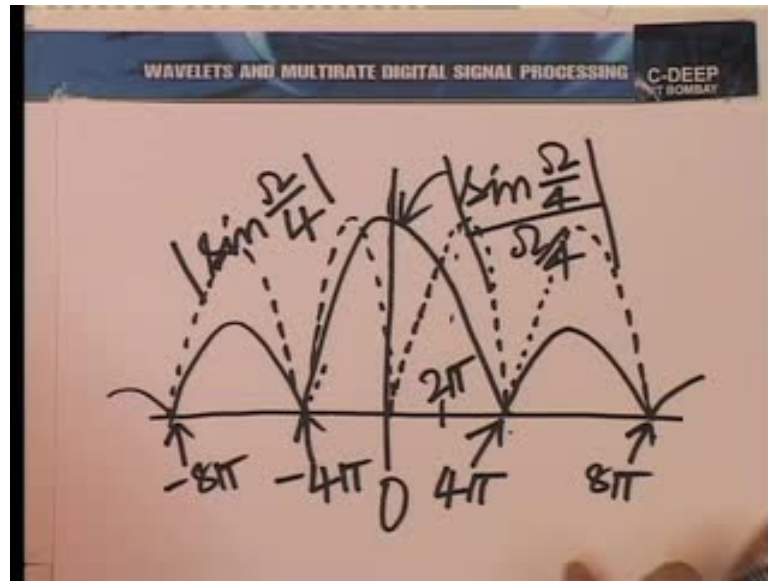
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... $\frac{\sin \Omega/4}{\Omega/4}$ Important

$= 2j e^{-j\Omega/2} \frac{\sin \frac{2\Omega}{4}}{\Omega/4}$

Sin omega by 4 divided by omega by 4. In fact, let us multiply all this and put it together. So, I have e raise the power minus j omega by 2 coming together j there, sin squared, omega by 4 divided by omega by 4 and a factor of 2 remains outside. So, what we need to focus on here is essentially this part. This is important, the rest of it is not because, it is this that really affects the magnitude seriously, rest of it is essentially a constant magnitude.

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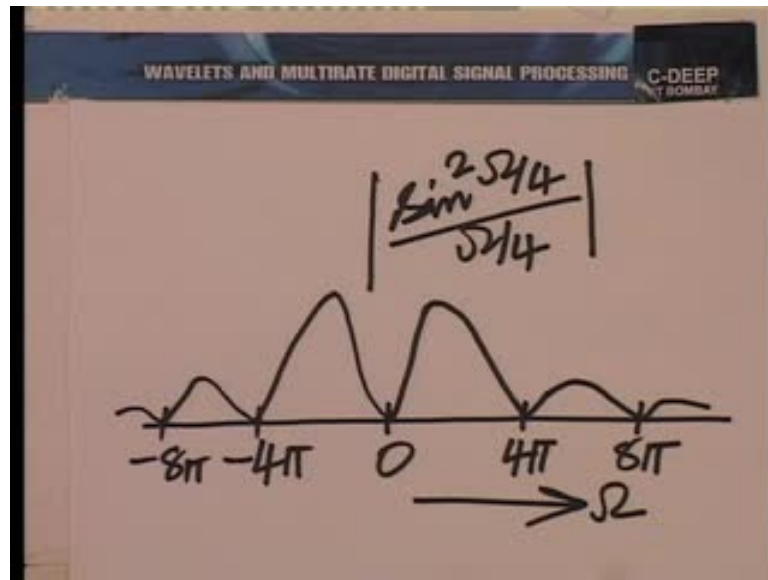
So, let us look at the magnitude of $\sin^2 \frac{\omega}{4}$. First, let us look at $\sin \frac{\omega}{4}$. That of course, would look like this, so this is the point 4π , minus 4π and so on and then, on the same graph I will draw this is $\sin^2 \frac{\omega}{4}$. And on the same graph, I will show in dotted $\sin \frac{\omega}{4}$ once again. So, that is going to look like this. It is going to have a maximum at π by 2 and then it is going to see ω by 4 . So, one cycle will be completed when ω by 4 equal to π that is ω equal to 4π .

So, one cycle been completed here, \sin^2 is going to look like this. So, in each span of 4π you have one half cycle being completed. So, this is one half cycle. Now, in the next span of 4π again, you have one more half cycle being completed, like this. This is the situation. Now, look at the situation, this was the zero frequency here. The solid line has a maximum at zero and it tapers off up to 4π on both sides. The dotted line reaches a maximum in between at 2π , when ω by 4 is equal to π by 2 alright.

So, ω is 2π , this point, the maximum here occurs at 2π and therefore, when you multiply this decreasing function by increasing function here. There is going to be some point of maximum in magnitude somewhere in between 0 and 4π and it is going to taper off to 0 at 4π again. So, let me focus, using the other side lobes. It is easier to understand. For example, between 4π and 8π , it is very easy to understand. You are multiplying essentially, two similar looking functions and so, you can see this is going to

be a maximum, somewhere it is going to taper off. And it is also clear that the maxima in the other side lobes are going to be weaker in the than the maxima in the main lobe here. All this, when you multiply the dotted function and the solid function, you are going to get a pattern something like this which I now sketch.

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Something like this this is going to of course, be mirror in this side and then, this way, this way and so on. So, this is mod sin squared omega by 4 by omega by 4, as a function of capital omega. Let us take a minute to look at this.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING C-DEEP BOMBAY

$\langle x(t), \psi(t-n) \rangle$

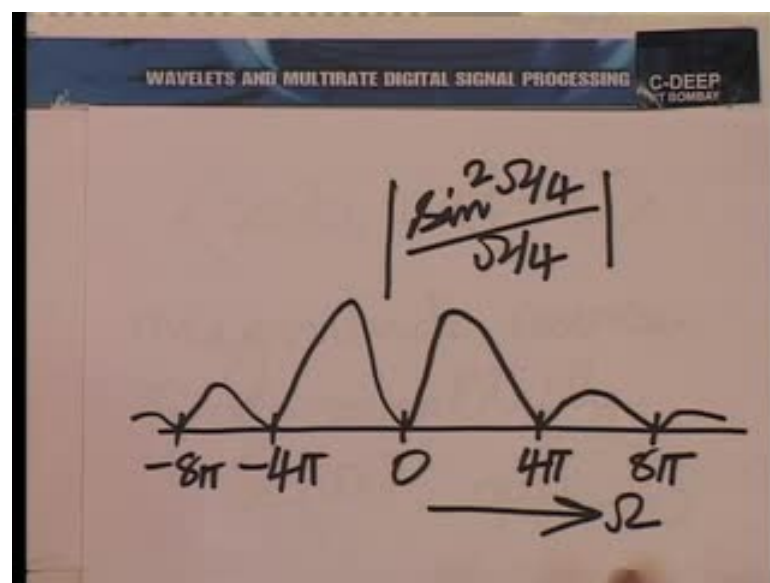
magnitude sense:

First multiply

$\hat{x}(\Omega), \psi(t-n)$

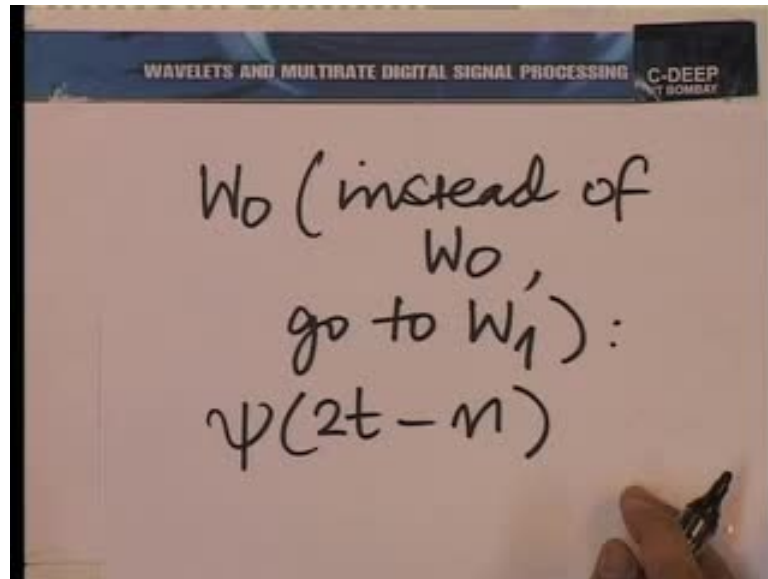
When we take an inner product, an inner product of a function $x(t)$. Now, here I am writing t , to explain that inner product is in time, with of which is ψ of t minus n . In the magnitude sense, what are we doing? We are first multiplying the Fourier transform of $x(t)$, with the Fourier transform of ψ of t minus n . This should be understood to mean the Fourier transform of ψ of t minus n , a cap on the whole function. So, when you multiply them and then, integrate that part of the Fourier transform of ψ of t minus n , which is significant in magnitude is going to be extracted out of the Fourier transform of the original function x and what is that significant part?

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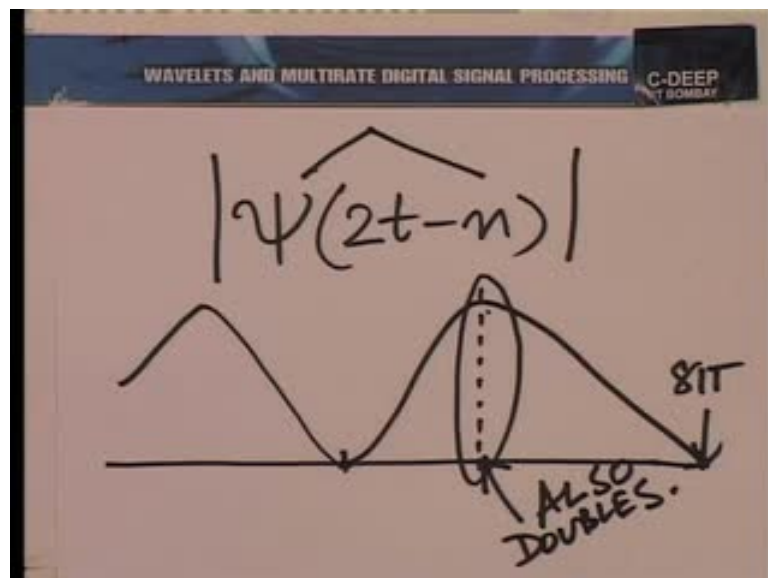
Let me put it back here for you; it is therefore, going to extract give emphasis to a band now. Not a band around zero, a band around some other frequency here. Of course, just symmetry, is always symmetry in frequencies. So, you could focus on the positive side of the frequency axis, but what is going to be done is to emphasis a band here and instead of ψ of t , if you take ψ of $2t$, for example.

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So, if you take, say instead of W 0. If you go to instead of W 0, go to W 1, in which case you would have psi of 2 t minus n and how would the Fourier transform of psi of 2 t minus n look?

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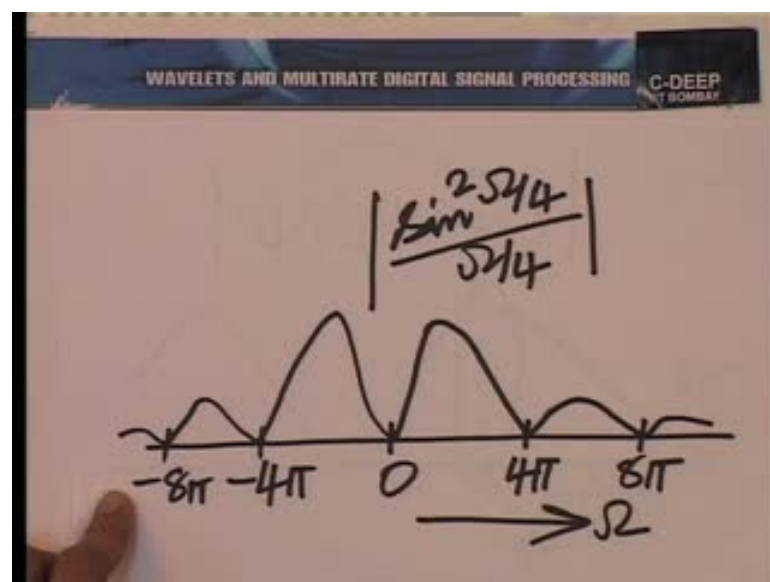
So, if we take the magnitude, if we take the Fourier transform of psi of 2 t minus n and plot its magnitude, its appearance would be something like this, it will be stretched. So, instead of going from 0 to 4 pi now, you would have a band between 0 and 8 pi, of course, symmetrically on the negative side. Now, there are two things. You see you must

keep in mind that, unlike the case of phi, in psi we have two changes taking places. One is that the band expands, the main lobe expands and each of the side lobes expand, but the second is that the center frequency, the point where, this was a maximum here.

You see this center frequency also doubles both the band and the center frequency double. That means now, you are emphasizing the different band, the center frequency is different, each time you go to a different W , you take W_0 you are emphasizing one band. When you take W_1 , you are doubling the center frequency. So, you are emphasizing the different band and of course, the band itself has doubled also. When you take W_2 , you are again doubling the center frequency, so you are emphasizing a different band and again the band is doubled. So, each time you are doubling the band and you are doubling the center frequency.

So, in a certain sense; that idea of complementarity. You know each time you put one increment layer. You are putting one more band and the band size is doubled here. Now, where in all this is our discontent with the Haar, why are we discontent? The discontent is because, even if we say we are emphasizing a band, it is only true to a certain extent.

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Let me put back this Fourier transform, and illustrate. We might say to an approximation that we are emphasizing this band, but that is only approximate. We are also, to some extent keeping this band and this and there of course, the negative corresponding pieces and that is where we are not content with the Haar. We want to keep a certain band,

focus our attention on it. We do not want interferences from the other side lobes that are there and in going to other multiresolution analysis; we are essentially trying to reduce that unwanted presence of the side lobes as much as we can. We have given a feel of, what our discontent is like. We shall build on this further in subsequent lectures. Thank you.