

Advanced Digital Signal Processing –Wavelets And Multirate.
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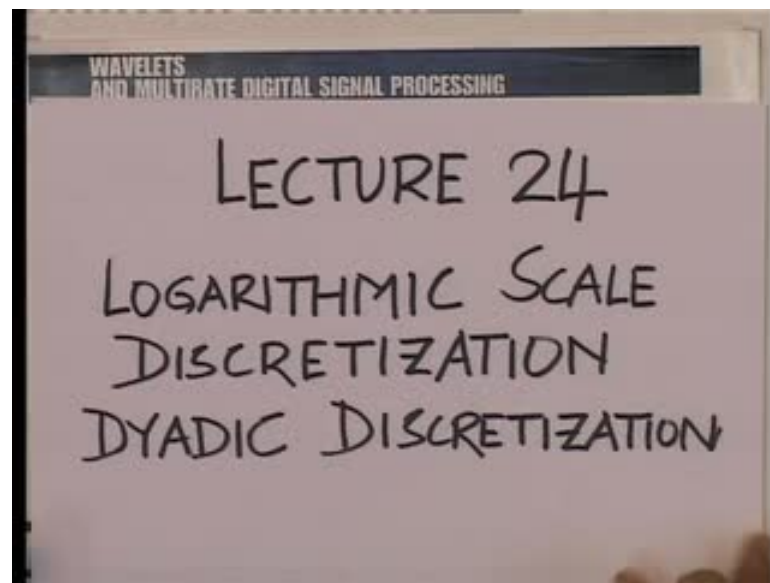
Lecture No. # 24
Logarithmic Scale Discretization Dyadic Discretization

A warm welcome to the twenty fourth lecture on the subject of wavelets and multirate digital signal processing. In the previous lecture, we had look that the admissibility condition for the continuous wavelet transform in depth. The admissibility condition was essentially a condition for reconstruct ability from the continuous wavelet transform, as we saw.

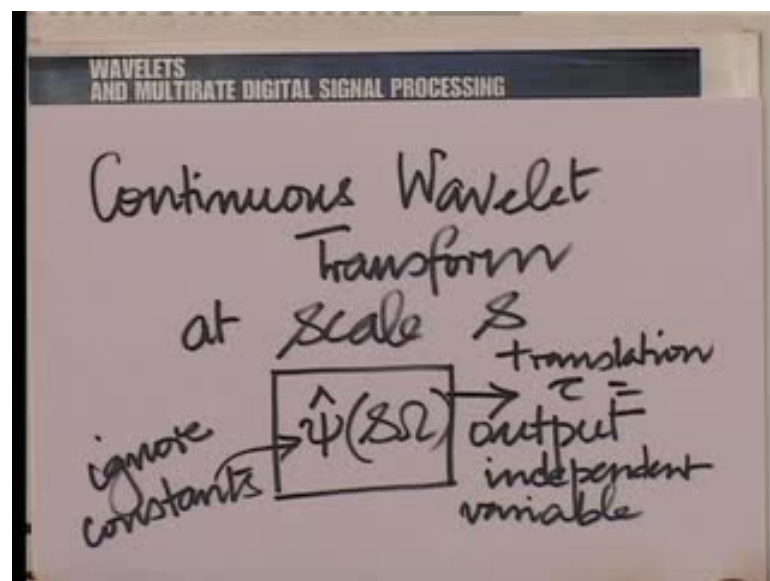
The continuous wavelet transform - we note it was extremely redundant to use a continuous scale and continuous translation, meaning essentially a two dimensional representation. When what we are dealing with is a one-dimensional entity is extremely redundant, very wasteful, and therefore, we began to ask the question of discretization, can we discretize the scale and can we discretize the translation parameter?

We first addressed the question of discretizing the scale, and in that context, we had noted what happens when we change the scale. We shall proceed in this lecture today to build on this discretization of scale in great a depth and to consider one particular kind of discretization of scale, with which, we have been dealing in the first half of the course, namely: dyadic discretization.

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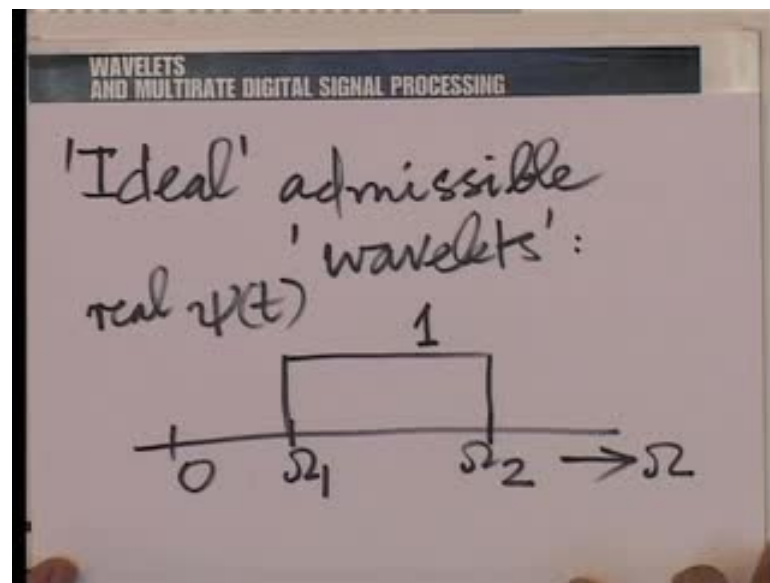


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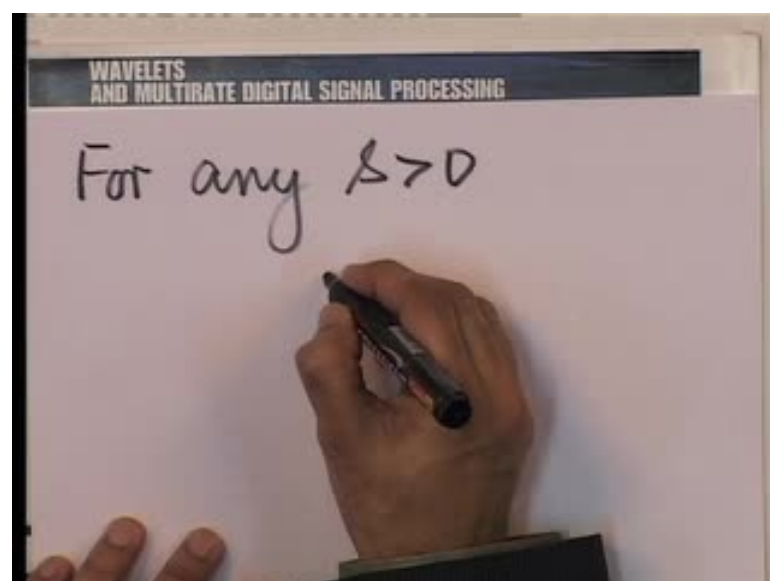
Therefore, I intend in the lecture today to talk about logarithmic scale discretization in general and dyadic discretization in particular. With that little background, let us look once again to what the continuous wavelet transform does. We had noted that a continuous wavelet transform essentially operates like a filter, both in synthesis and in analysis. So, we had noted that the continuous wavelet transform, at scale s is essentially a filtering operation.

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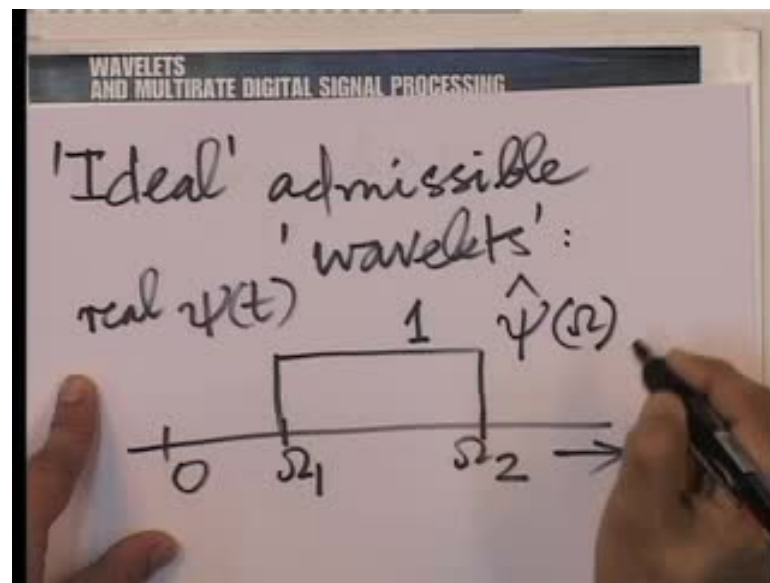


The filtering operation with the frequency response given by $\psi(\omega)$ with some constant of course, at the moment, let us simply ignore the constants. So, you see, the output independent variable here can be interpreted as the translation. The output independent variable is τ here, alright? Well, then, if we take an ideal filter or an ideal wavelets, how to speak? We are taken examples of ideal frequency response that were admissible and those were strictly band limited.

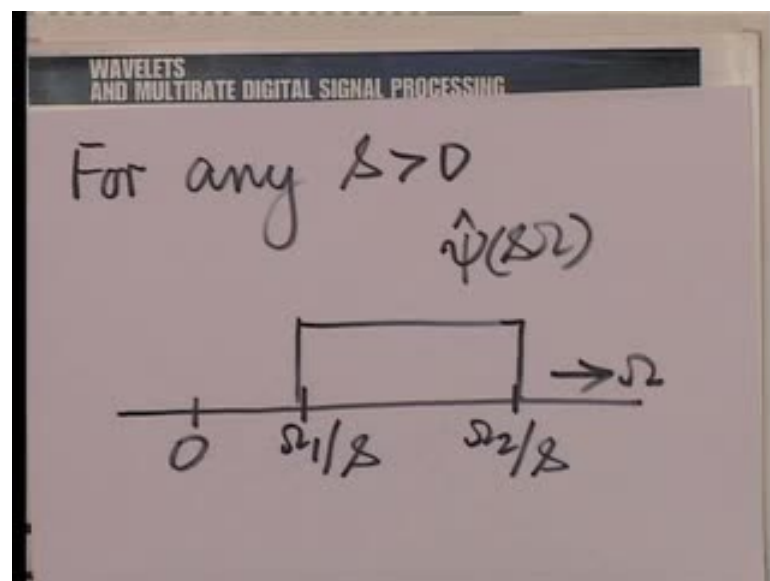
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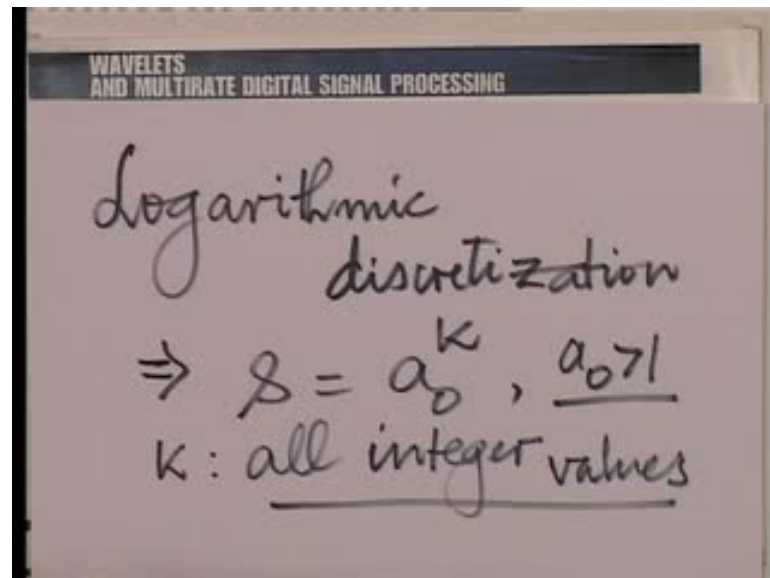
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They had a ψ cap, which look something like this. Of course, I am assuming that the wavelets are real, and, and looking only at the positive side of the frequency axis. So, we had note it, that, that the admissibility condition is essentially a statement of band pass character, and therefore, if we consider an ideal band pass function like this and take a dilation of the same, we again get a band pass function. So, for any s greater than zero as we should have it, ψ cap $s \omega$. This was of course the Fourier Transform - ψ cap ω . ψ cap $s \omega$ would look like this, and we once again note that there is a

contraction or expansion dilation in general, both of the band and the center frequency, and that is the reason why we are argued that logarithmic discretization make sense.

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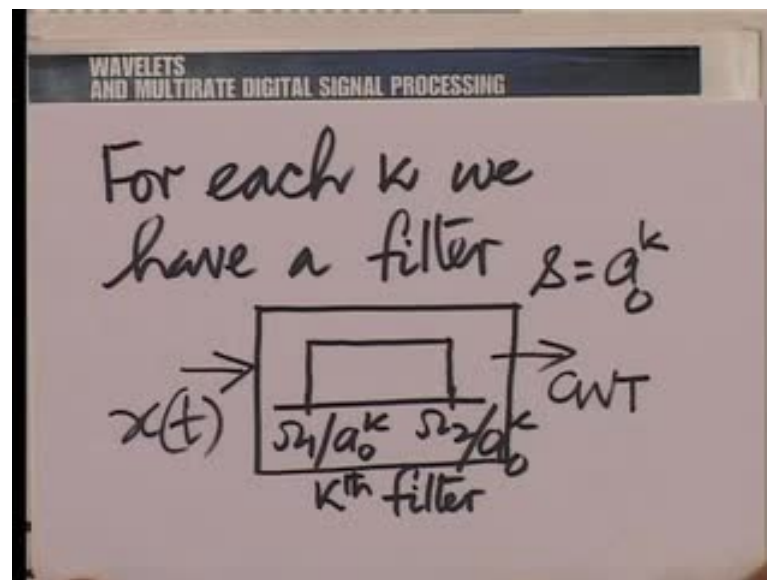


The image shows a handwritten note on a slide. The slide has a title bar that reads "WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING". The handwritten text on the slide is as follows:

$$\text{Logarithmic discretization}$$
$$\Rightarrow s = a_0^k, \underline{a_0 > 1}$$
$$k: \underline{\text{all integer values}}$$

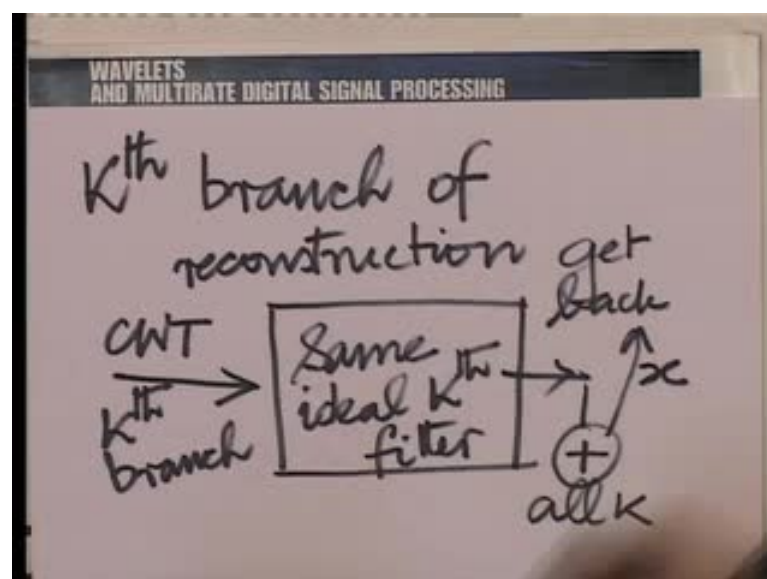
Logarithmic discretization means s should be discretized; s should be discretized according to a naught to the power of k , where k runs over all the integers, and we said: as long as a naught is positive and greater than one and k runs over all the integers. We do not need to take a naught less than one also as the possibility, because of after all, when k runs over all the integers, the less than one possibility is taken care of. I had given you the example of two, for example, two raise to the power of k over all integer k is the same thing as say half raise to the power of k over all integer k . With that little recapitulation of our discussion, let us go again in depth to analyze, what will happen when we discretize the scale? So, in other words, when we discretize the scale, what we are trying to ask is - what you do to each of these filters and how do you recombine what you have done to each of these filters to get back the original x of t .

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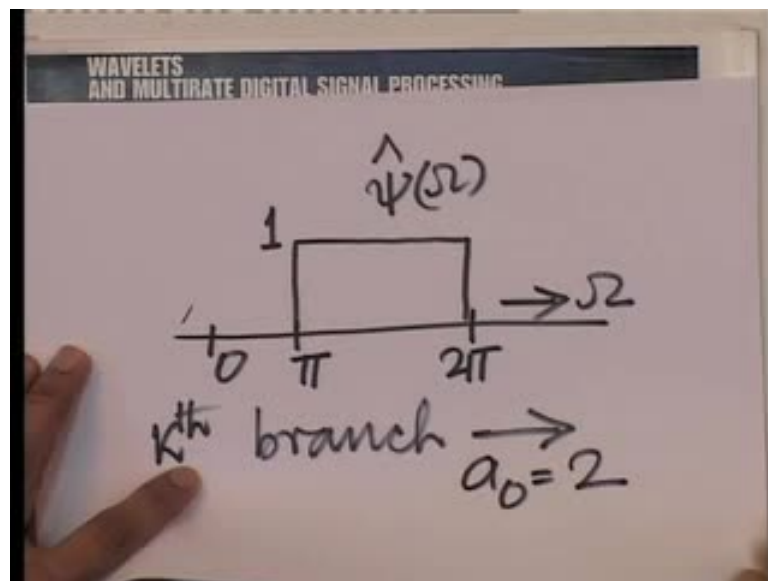
In other words, for each such k , we had a filter, and you see, how the frequency response of the filter looks; I mean let start with the ideal situation, and then, let's get to the practical situation eventually. In the ideal situation, the filter, the k th filter so to speak would have a frequency response like this. So, it would have a band going from ω_1 divided by a_0 to the power k to ω_2 divided by a_0 to the power of k , where s is, of course, a_0 to the power k .

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Now, it is obvious how we can reconstruct this part after all. If things are ideal, and in fact, if you do pass $x(t)$, the input through an ideal filter like this to get the continuous wavelet transform here. The reconstruction should also proceed by doing the same thing. So, in fact, the k th branch of reconstruction should essentially have the continuous wavelet transform coming in here, the k th branch; I mean the continuous wavelet transform being fed to the same ideal filter, and if we add this over all k , we should get back $x(t)$.

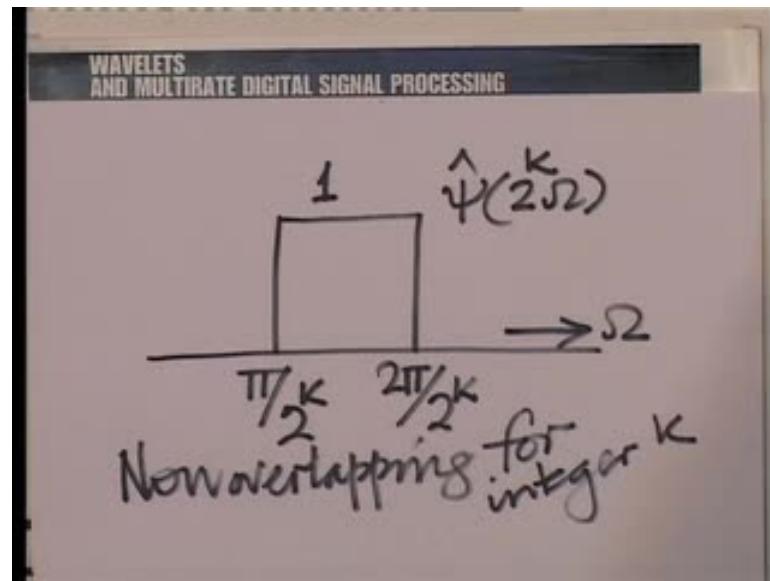
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So, in the ideal situation, the answer is very simple. What each branch, so to speak each k th branch in some sense of this continuous wavelet transform does is to extract a particular band on the frequency axis, and now, we can also see very easily, interpret very easily what each branch should do. Each branch should have a separate non overlapping band. Just to take an example, suppose you had a band pass filter of the following kind, so, if you had $\psi(\omega)$ looking something like this, suppose this was $\psi(\omega)$ one between π and 2π and zero everywhere else, and naturally, mirrored on the negative side of the ω .

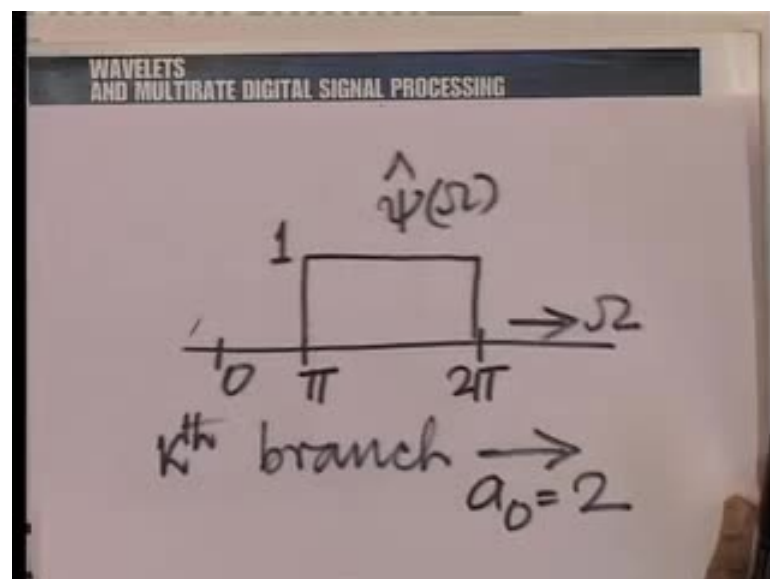
Then the k th branch would essentially do the following: the k th branch would cover the band between π divided by two to the power k if a_0 where equal to k . So, I must also specify what a_0 is - k th branch with a_0 equal to two.

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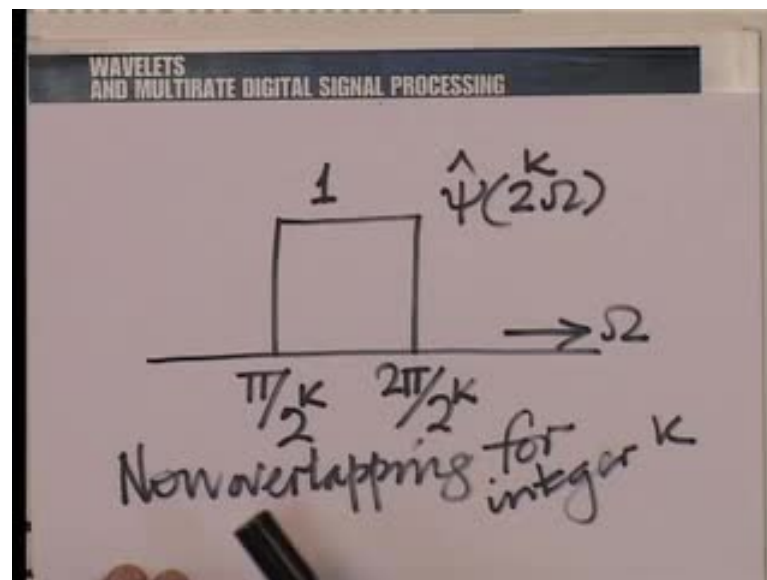


So, essentially, the k th branch would take the frequencies from π by two raise the power of k to two π by two raise the power k , and we can see that in this particular case for different integers k , these bands would be non overlapping. So, it is easy to fix our ideas. In fact, what I have done is to put before you the very idealization of the dyadic multi resolution analysis which we been looking at all this while.

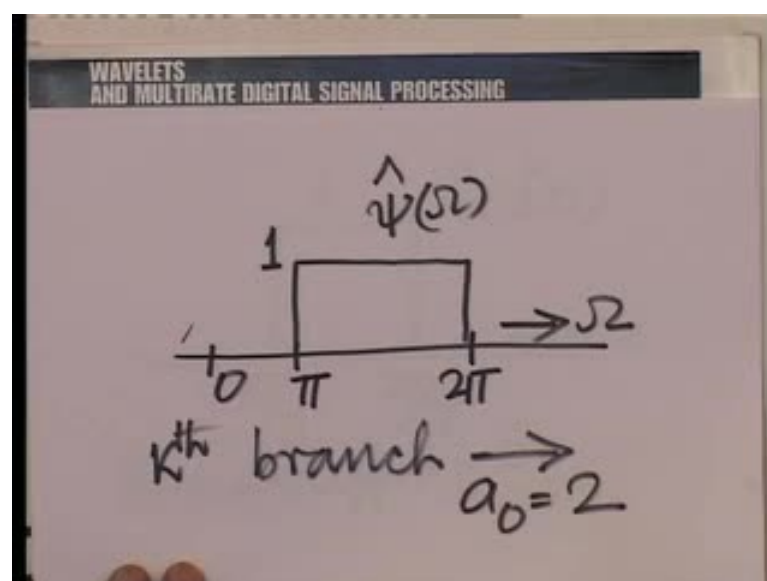
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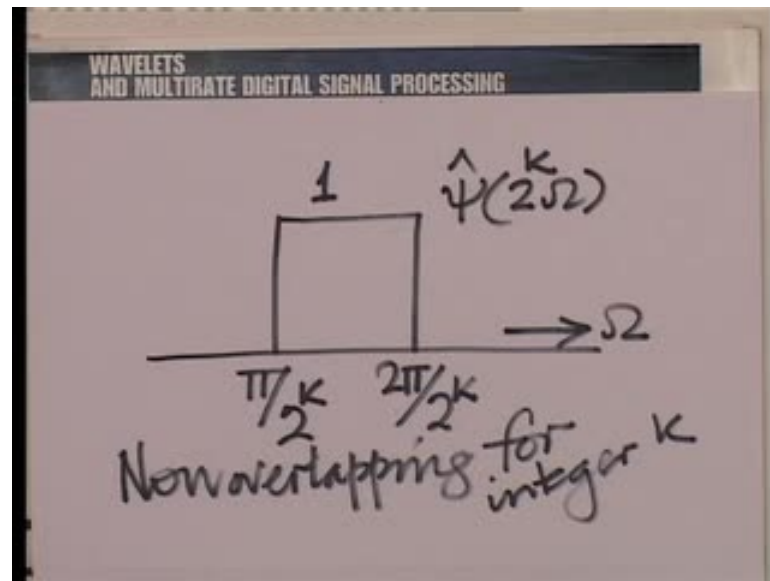
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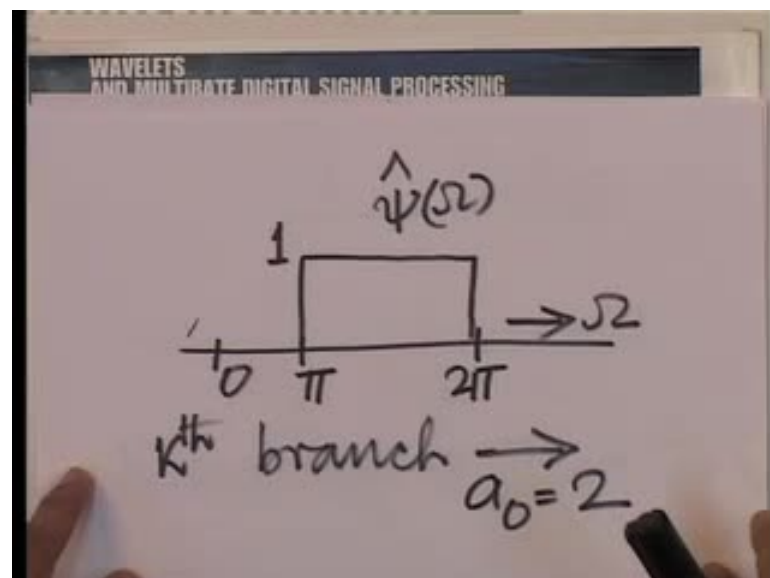


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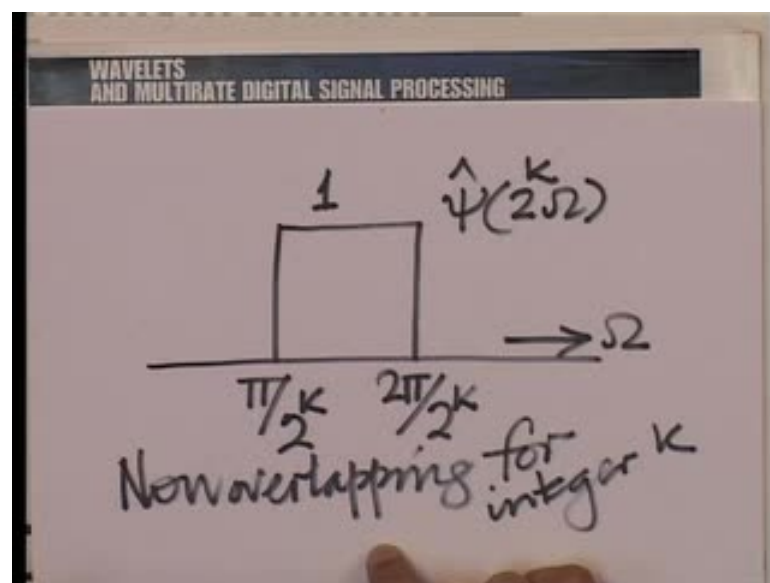
In fact, let us look back at this frequency domain here. You see, take for example, k equal to one k equal to one would take to over the band π by two to π , and if you look at the original band, its π to two π k equal to minus one would take you from two π to four π . Again non overlapping with π to two π , and you can continue in a similar way k equal to two k equal to minus two and you can proceed; k equal to zero needless to say, keeps you in the original band - π to two π .

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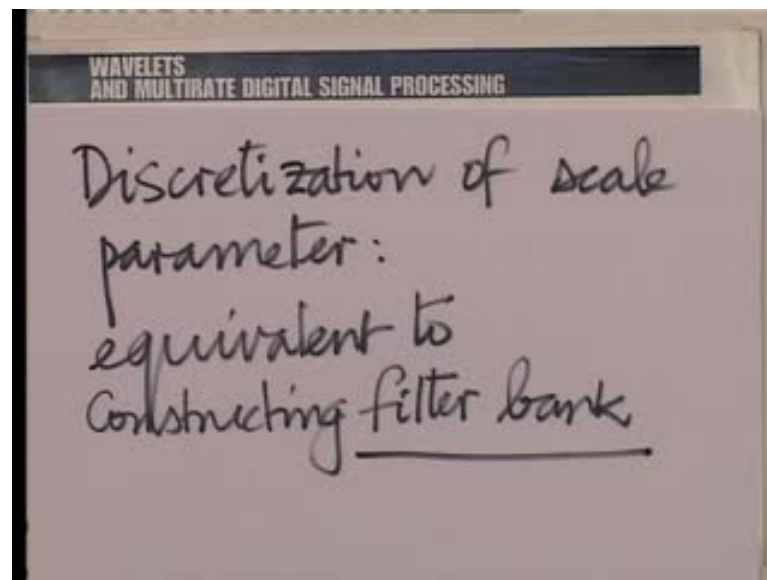
So, now, we have beginning to understand what the continuous wavelet transform does in an idealization sets. You see, in that case, how we really just talking about doing this only in the frequency domain? Well, we must not forget what we are trying to do and which we can never do because of the uncertainty principle is to do what I have just shown you on the frequency axis ideally, but do it with time limited functions, and alas if you look at the inverse Fourier Transform of this $\hat{\psi}(\omega)$ that we which just drew here, it is far from time limited. In fact, it decays as slowly as the reciprocal of time terribly, anyway.

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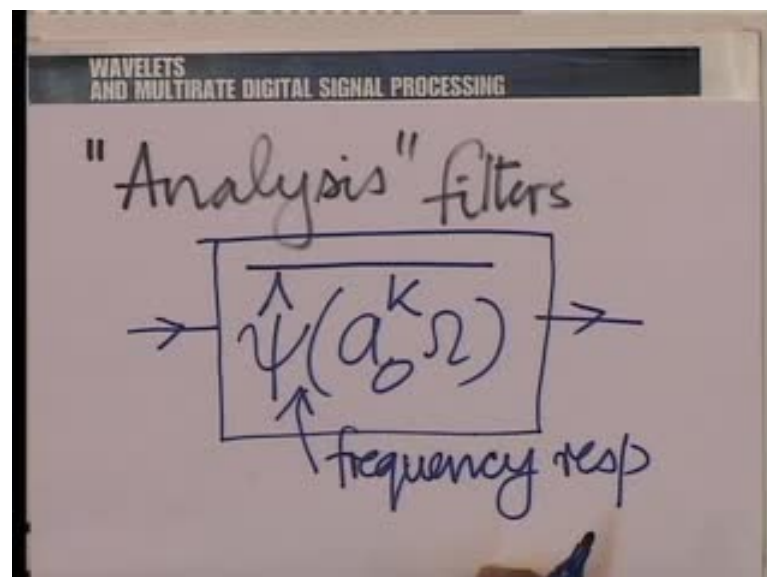
So, this is easy now to interpret and understand, and now, it is also easy to see that if you take the ideal situation, when you could simply put the same filter on the analysis side and the synthesis side. This also makes it clear how we get the notion of filter banks when we discretize.

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So, you see, the discretization of the scale parameter is equivalent to constructing a filter bank. Recall what a filter bank is. A filter bank is a collection of filters although with the common input or with all the outputs sum together to get a common output, and of course, more subtly the filters are interrelated.

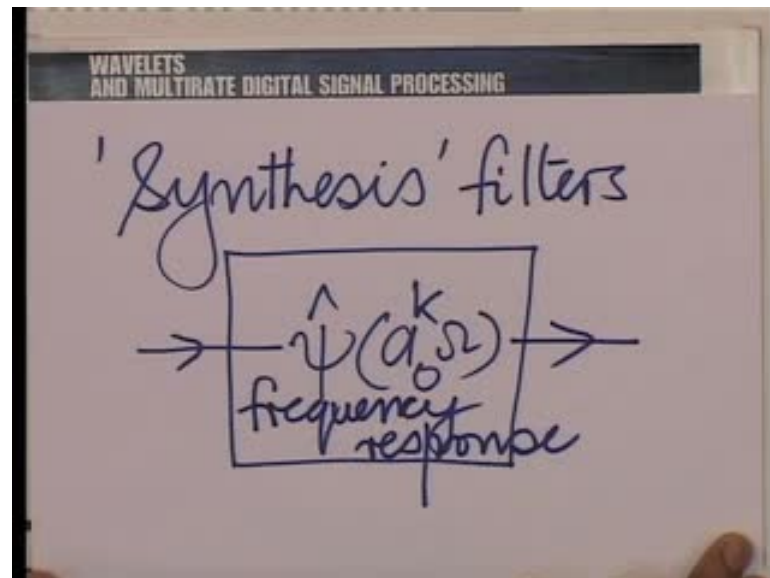
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Now, all these qualities are satisfied in the k th branches that we are talking about here. In fact, in the particular k th branches that we speak of here are analysis filter have the following pattern. The analysis filter is the ones that create the continuous wavelets

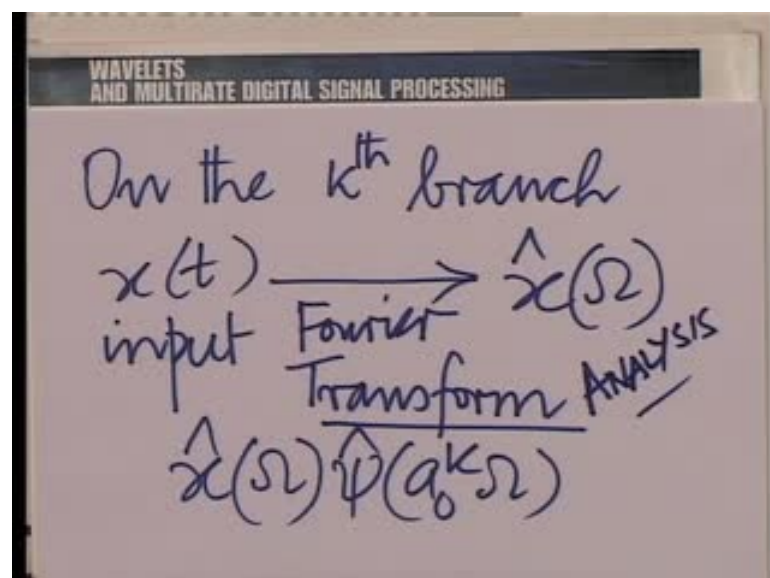
transform. Strictly speaking, we should put a complex conjugate here as was the case with the continuous wavelet transform.

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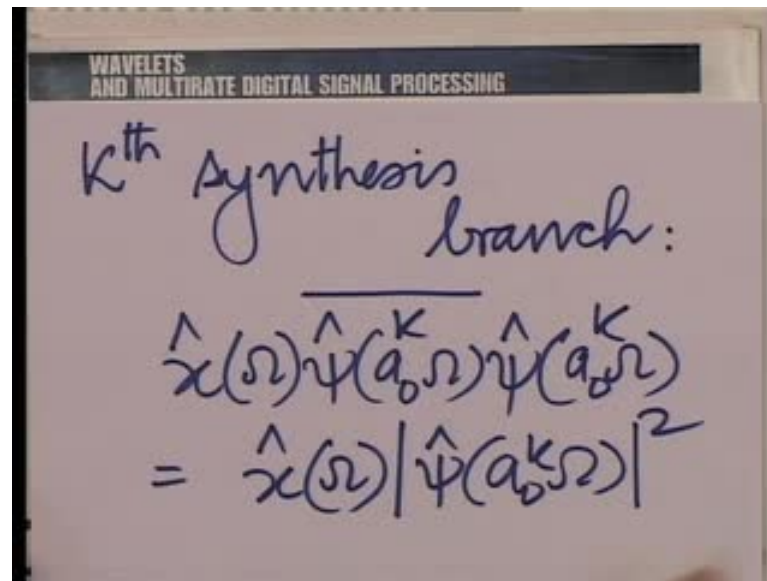


So, these are the analysis filters, and this is the frequency response if you ignore the constants the syntheses filters, again if you ignore the constants, have a frequency response that looks like this. This is the k th branch, and therefore, if you analyze what you are doing all over the, k , k th branches for k going from minus to plus infinity, you are doing the following.

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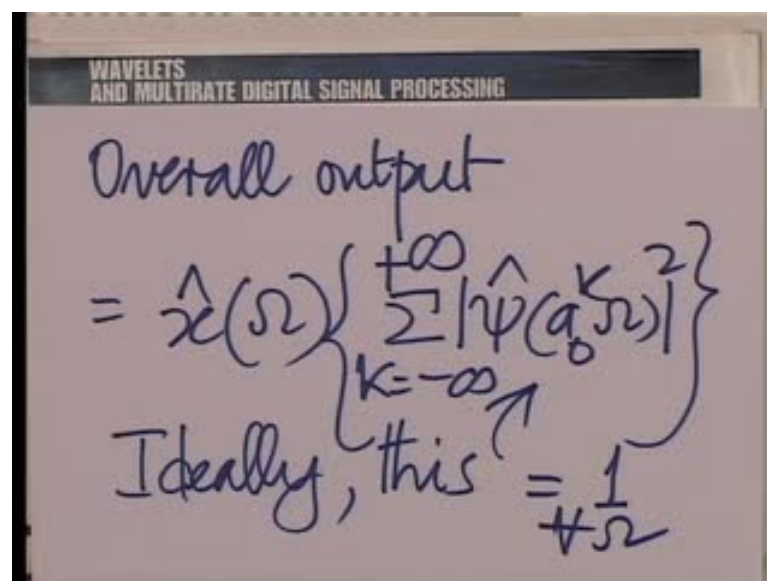
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

k^{th} Synthesis branch:

$$\hat{x}(\omega) \hat{\psi}(a_0^k \omega) \hat{\psi}(a_0^k \omega) = \hat{x}(\omega) |\hat{\psi}(a_0^k \omega)|^2$$

On the k th branch, if the input signal x of t has the Fourier Transform $\hat{x}$ of ω , then we are actually constructing $\hat{x}$ of ω times $\hat{\psi}$ of a_0 to the power k of ω complex conjugates, and therefore, this is on the analysis side I mean, and on the synthesis side, the k th branch multiplies this again by $\hat{\psi}$ of a_0 raised to the power of k of ω , which essentially amounts to multiplying the original Fourier Transform by a modulus squared of $\hat{\psi}$ of a_0 raised to the power of k of ω , and on the output side, one is going to sum up all these outputs of the branches.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

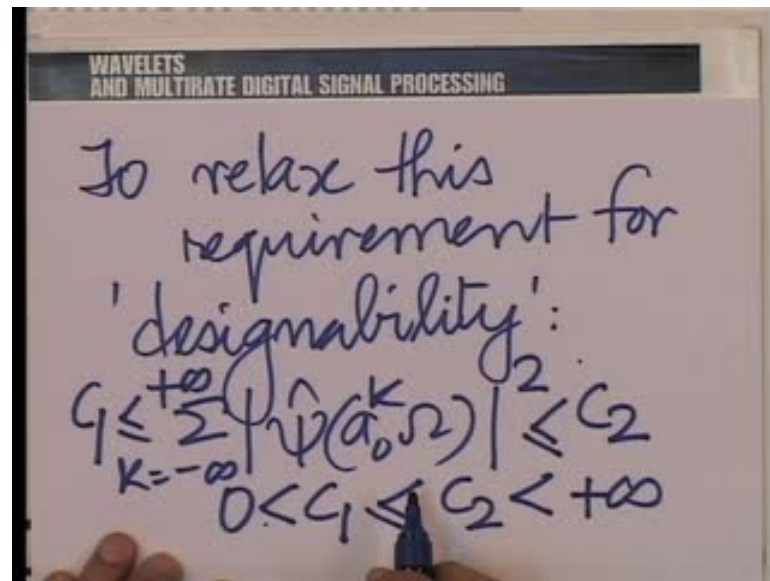
Overall output

$$= \hat{x}(\omega) \left\{ \sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \omega)|^2 \right\}$$

Ideally, this $\uparrow$ $= \frac{1}{4\omega}$

So, the final output, overall output is going to be x cap ω multiplied by a summation on all k ψ cap a naught raise the power of k ω modulus the whole squared, and a minute ago, we saw what this was for the ideal situation. Ideally, this should be one; one for all ω . Of course, that is where the challenge is. We wanted to be one for all ω with time limited functions, and that is the whole challenge.

(Refer Slide Time: 21:36)



WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

To relax this requirement for 'designability':

$$C_1 \leq \sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \omega)|^2 \leq C_2$$

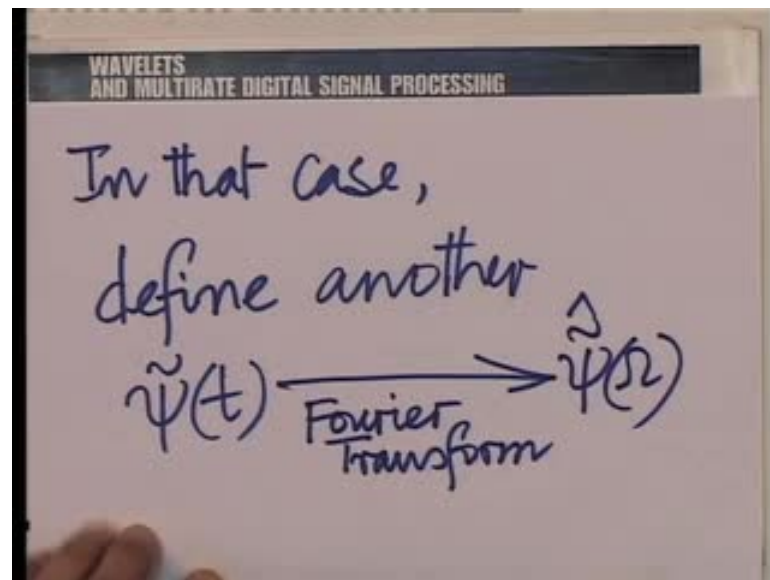
$$0 < C_1 \leq C_2 < +\infty$$

Anyway, how do we relax in the frequency domain if we want to meet this challenge? Well, relaxation means do not quite insist, it must be a constant. Allow it to be between two positive constants. Now, I shall insist on strictly positive constants and we will see why. What I am saying is to relax this requirement, for design, for designability. Let this quantity ψ cap a naught to the power k ω mod squared summed over all k all integer k be between two constants. So, the greater than or equal to a constant, let say c one, and less than equal to a constant, let say c two, and we are told that c one is of course, less than equal to c two; c two is less than infinity and c one is greater than zero, not this is greater than, not greater than or equal to, and of course, this is strictly less than, not less than, whereas, this cannot go to infinity.

So, here, there is the possibility of equality; here there is not. These are strict in equality; that means c one is strictly positive, c two is strictly finite, and of course, it is not difficult to see that this is the nonnegative quantity. So, it cannot possibly become negative.

So, all that we are saying is instead of insisting that, this be a strict constant, allow it to be between two constants, and this is the case, then can be make a small change to the synthesis filter and reconstruct? That is the first question that we shall **answer** That is very easy to say to see.

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You see, in that case, let us define another function psi tilde from psi. Now, psi tilde shall be defined in the frequency domain. So, let psi tilde t have the Fourier Transform psi tilde cap omega. So, we will show that, you know, even if the wavelet psi t does not quite obey this strict constancy of the sum, we can still build a reconstruction approach. That is what we are trying to show.

(Refer Slide Time: 24:41)

WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\hat{\tilde{\psi}}(\Omega) = \frac{\hat{\psi}(\Omega)}{\sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \Omega)|^2}$$

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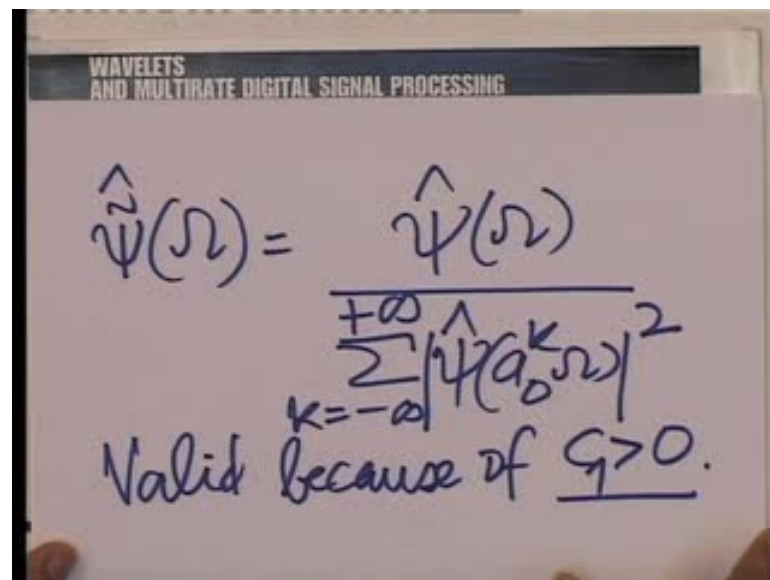
WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

To relax this requirement for 'designability':

$$C_1 \leq \sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \Omega)|^2 \leq C_2$$

$0 < C_1 \leq C_2 < +\infty$

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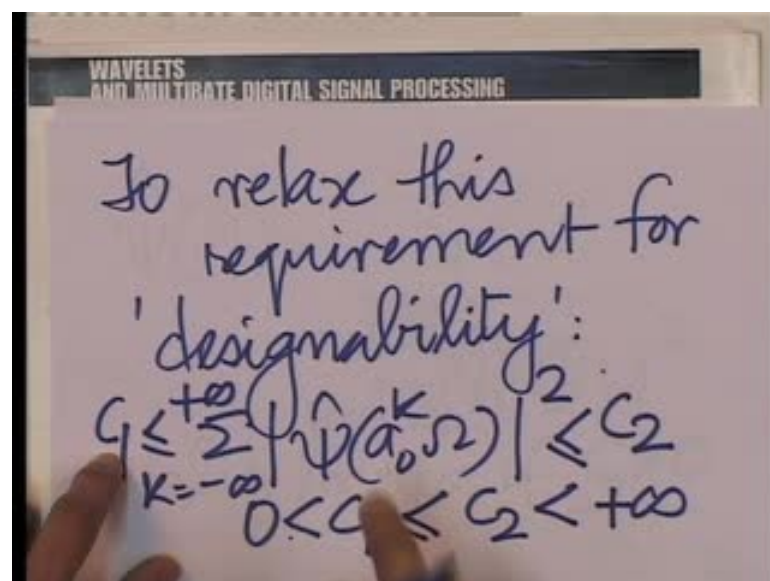
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\hat{\tilde{\psi}}(\omega) = \frac{\hat{\psi}(\omega)}{\sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \omega)|^2}$$

Valid because of $c_1 > 0$.

So, you see, suppose this is a definition of its Fourier Transform, define psi tilde cap omega to be psi cap omega divided by the sum here, and please note that since we have this condition here, this sum lies between two positive constants, and what is important here is the fact that it does not go to zero. The c_1 constant is important here, because the c_1 constant, this denominator never goes to zero, and therefore, this is valid because of c_1 being greater than zero. Otherwise, this definition would be meaningless.

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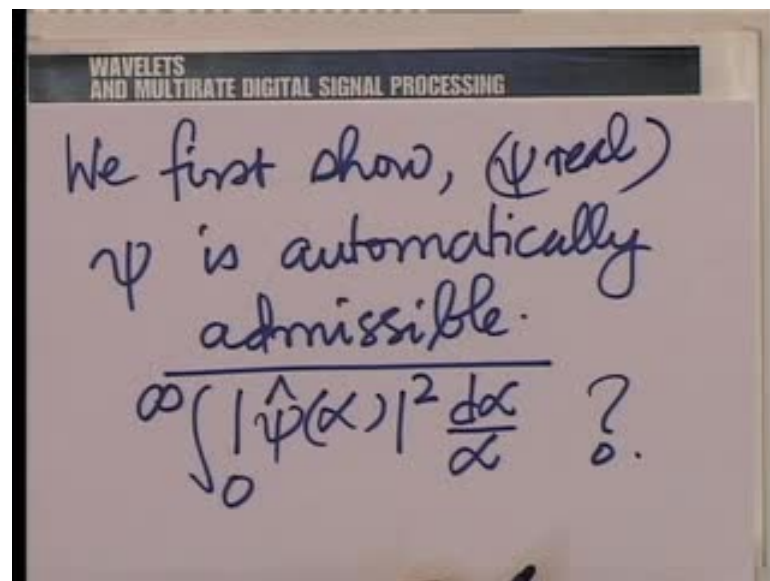
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

To relax this requirement for 'designability':

$$c_1 \leq \sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \omega)|^2 \leq c_2$$
$$0 < c_1 \leq c_2 < +\infty$$

So, the c one is required for this reason to be able to build this psi tilt to cap. Now, we will show something interesting. You see, we will show that we could use psi on the analysis side and psi tilde on the synthesis side, and we will first show that if psi obeys this requirement, then psi is admissible. So, we will show that now we have use c one to define psi tilde. We will use c two to prove admissibility. So, once we ensured this, we also have admissibility. Let us prove that.

(Refer Slide Time: 26:24)



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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\int_{-\infty}^{\infty} |\hat{\psi}(\alpha)|^2 \frac{d\alpha}{\alpha}$$

$$= \sum_{k=-\infty}^{\infty} a_0^{kH} \int_{a_0^k}^{a_0^{k+1}} |\hat{\psi}(\alpha)|^2 \frac{d\alpha}{\alpha}$$

Put $\alpha = a_0^k \beta$

By admissible, I mean we need to consider this integral, and we have of course, agree that $\psi(t)$ is real. So, let us make the remark here ψ is real; I mean it is a real wavelet. So, you see, let us break this integral into partial, but finite integrals of the following form summation k running from minus to plus infinity integral from a_0 to the power k to a_0 to the power of k plus one.

Now, let me take an example suppose for example, a_0 is equal to two, then the integral from zero to infinity is the same as the integral from two to the power k to two to the power k plus one with k running over all the integers. As k takes negative integer values, you are covering the range below one; as k takes positive integer values, we are covering the range from one and above one. That is the interpretation here.

So, for example, k equal to zero would cover the range, one to a_0 in general or one to two of particular a_0 is equal to two, and k equal to two will cover k equal to one will cover the range from two to four and so on and so forth, and k equal to minus one will cover the range from half to one and so on so forth in particular with a_0 equal to two. Anyway, with that, we keep the integral as it is, but as usual, we make a substitution on the integral. So, put α equal to a_0 raise to the power of k times β and make a change of variable in the integration.

(Refer Slide Time: 29:06)

WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\int_{a_0^{k+1}}^{a_0^k} |\hat{\psi}(\alpha)|^2 \frac{d\alpha}{\alpha} \quad d\alpha = a_0^k d\beta \quad \alpha = a_0^k \beta$$

$$= \int_1^{a_0} |\hat{\psi}(a_0^k \beta)|^2 \frac{d\beta}{\beta}$$

Now, when alpha is equal to a naught raise to the power of k times beta, this limit simply becomes one and this limit becomes a naught. (Refer Slide Time: 29:25) So, we have a naught raise to the power of k beta here, and of course, interestingly, d alpha is also a naught raise to the power of k d beta, and therefore, one can easily see that d alpha by alpha is equal to d beta by beta. So, in place of d alpha by alpha, I can simply write down d beta by beta.

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$$\int_0^{\infty} |\hat{\psi}(\alpha)|^2 \frac{d\alpha}{\alpha} = \int_1^{a_0} \left\{ \int_{-\infty}^{+\infty} |\hat{\psi}(a_0 \beta)|^2 \frac{d\beta}{\beta} \right\} \frac{d\beta}{\beta}$$

So, now, we have a beautiful summation coming in. What it means really is that the integral from zero to infinity psi cap alpha mod squared d alpha by alpha is equal to the integral from one to a naught summation if I interchange the order of the summation, and the integral which I can do simply, because this is a finite integral here, summation from k running from minus to plus infinity. In fact, I could interchange this order, because of I am guaranteed that I have a finite integral here and I am guaranteed that this is convergent for all beta; it lies between two positive constants. So, it is meaningful to do this, and in fact, now we have a condition on this. This is a nonnegative integrand here, and this is of course, you know, integral over a nonnegative quantity one by beta between one and a naught. If I can pull an upper bound on this, the upper bound on this is known to us - its c two.

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WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \beta)|^2$$

is upper bounded
by C_2

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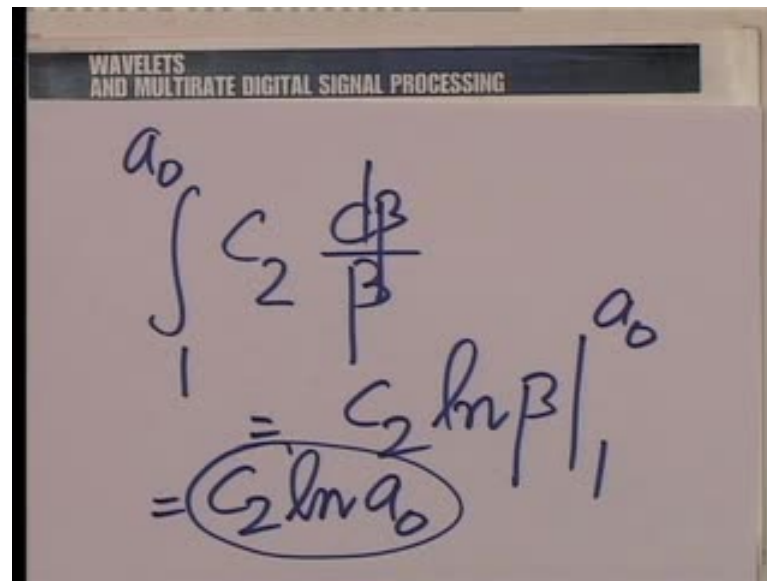
WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

'Admissibility
integral' $\int_0^{\infty} |\hat{\psi}(\alpha)|^2 \frac{d\alpha}{\alpha}$

is upper bounded:
by

So, the summation k running from minus to plus infinity $|\hat{\psi}(a_0^k \beta)|^2$ is upper bounded by C_2 , and therefore, the so called admissibility integral as we call it. We shall now introduce this term.

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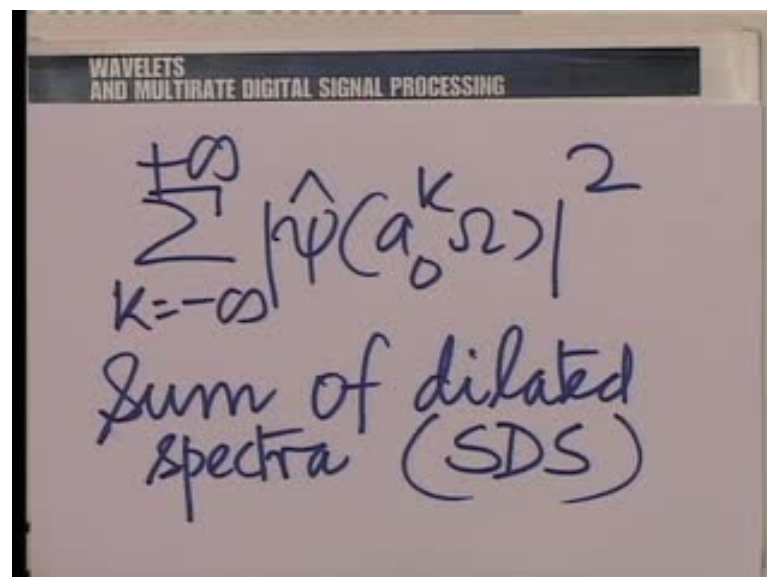


WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\int_1^{a_0} C_2 \frac{d\beta}{\beta} = C_2 \ln \beta \Big|_1^{a_0} = \boxed{C_2 \ln a_0}$$

The admissibility integral is upper bounded also by integral from one to a_0 of C_2 over β by $b\beta$, and this is a very easy integral to evaluate. It is essentially C_2 integrated log natural β from one to a_0 and that is $C_2 \log$ natural of a_0 , a neat and elegant result.

(Refer Slide Time: 33:25)



WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \Omega)|^2$$

Sum of dilated spectra (SDS)

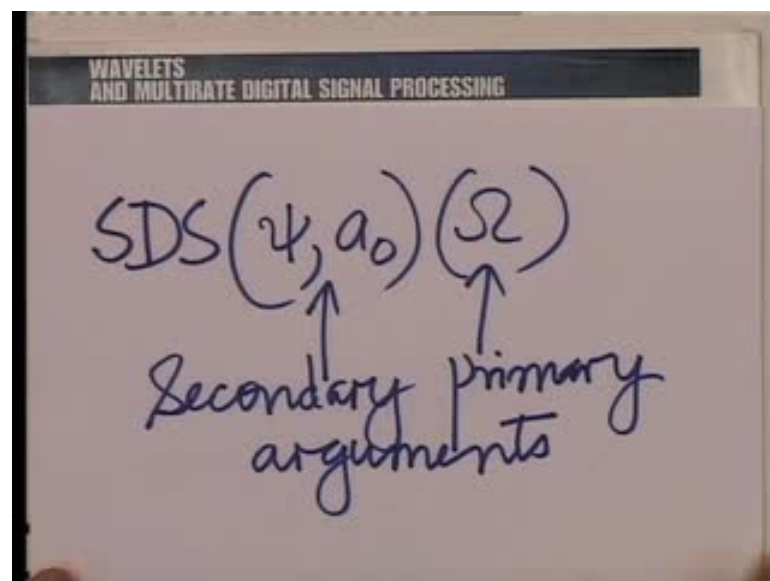
So, once I have the upper bound on the summation, I am guaranteed admissibility. So, we have shown that if ψ obeys this requirement, and now, we will introduce some terminology. We shall call this quantity which we are talking about all this while the

summation. We shall call this quantity a sum of dilated spectra and we shall abbreviate it by $s_d s$, and just as in the case of continuous wavelet transform, we have primary and secondary arguments for the sum of dilated spectra.

The secondary argument is ψ , because it is for ψ that we are constructing a dilated, dilated or sum of dilated spectra, and of course, the primary arguments are a_0 and ω . Well, actually, a_0 is a gray area; you could treat it as secondary or primary. You know, in the continuous wavelet transform, for example, we distinguish primary and secondary by calling those arguments primary which did not change in the discussion in a particular context, and secondary for those that did or rather the other way around primary for those which did change, which were important in a particular context and secondary for those that did not.

So, for example, in the continuous wavelet transform, the primary arguments were the translation and the scale, those which were important in a given context, they changed in that context, and we were looking at the variation of those in a context, and the secondary arguments were the essentially the wavelet with which we are constructing the cwt or the continuous wavelet transform and the function on whom the continuous wavelet transform is being constructed.

(Refer Slide Time: 35:32)



(Refer Slide Time: 36:05)

WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\cancel{SDS(a_0)}$$

$$C_1 \leq SDS(\psi, a_0)(\Omega) \leq C_2$$

C_2 guarantees ADMISSIBILITY.

So, here to, we shall talk about $s d s$ with secondary arguments of ψ and a_0 and primary argument Ω , but of course, we could shift this, and what you are essentially saying is, $s d s$, $s d s \psi a_0$ as a function of Ω needs to be upper bounded and lower bounded, and the upper bound guarantees admissibility.

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WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

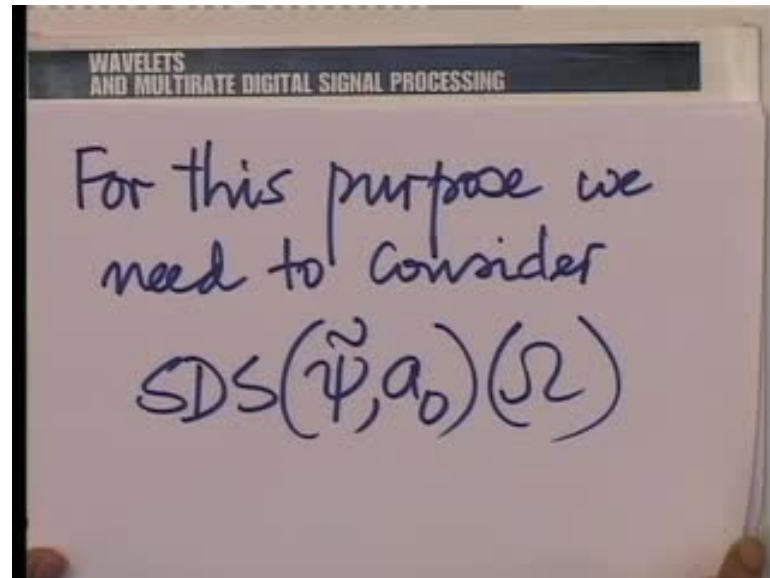
$$\hat{\tilde{\psi}}(\Omega) = \frac{\hat{\psi}(\Omega)}{SDS(\psi, a_0)(\Omega)}$$

We show that $\hat{\tilde{\psi}}$ is ADMISSIBLE

Now, we soon see in a minute that C_1 guarantees reconstruction, and in fact, we are going in that direction. We have already constructed a $\tilde{\psi}$, where $\tilde{\psi}(\Omega)$ is essentially $\hat{\psi}(\Omega)$ divided by $s d s \psi a_0$.

naught omega, and it is very easy to see that if $s_d s$ obeys the upper lower bound condition, ψ tilde is also going to be admissible.

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WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

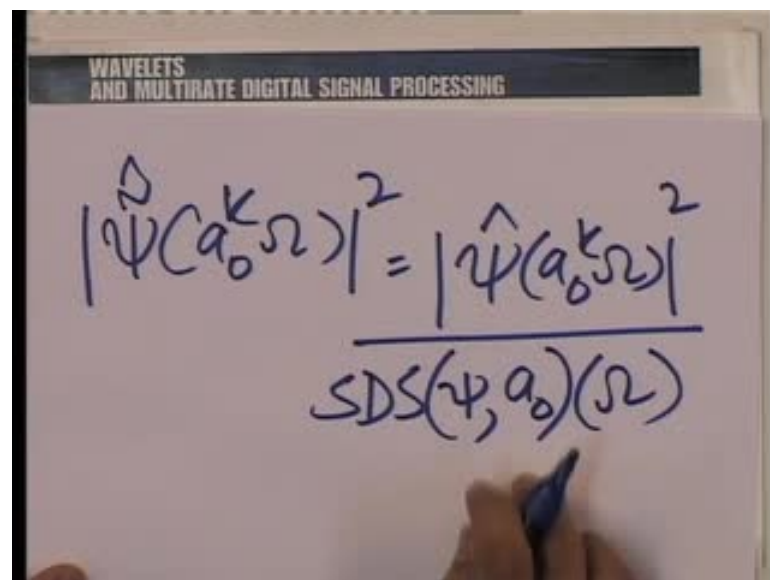
$$|\hat{\tilde{\psi}}(a_0^k \Omega)|^2 = \frac{|\hat{\psi}(a_0^k \Omega)|^2}{\sum_{l=-\infty}^{+\infty} |\hat{\psi}(a_0^l a_0^k \Omega)|^2}$$

So, we show that ψ tilde is admissible, and to prove that, all that we need to do is to try and look for bounds on the sum of dilated spectra of ψ tilde. So, we need to consider for this purpose - the sum of dilated spectra ψ tilde naught as a function of omega, and this is not at all difficult to compute. Indeed, let us in particular construct ψ tilde cap a naught raise to the power k omega, which is the typical term required or rather the

modulus of this squared, and the modulus of this squared is very easily seen to be the modulus of ψ cap a naught raise the power of k ω the whole squared divided by, that is the interesting thing.

Let me expand the denominator. I first try to summation on l to distinguish it from this k . Summation on l going from minus to plus infinity ψ cap a naught raise to the power of l . Now, in place of ω , I need to write a naught raise to the power of k ω , so I will write a naught raise to the power of k ω there, and now, this is easy to evaluate. In fact, you can see that this becomes a naught raise to the power of l plus k , and please remember the summation is on l and k is fixed. So, for a fixed k when l runs over all the integers as it does here, l plus k also runs over all the integers.

(Refer Slide Time: 39:43)



$$|\hat{\Psi}(a_0^k \Omega)|^2 = \frac{|\hat{\Psi}(a_0^k \Omega)|^2}{\text{SDS}(\Psi, a_0)(\Omega)}$$

So, what we have in the denominator is once again the sum of dilated spectra of ψ a naught evaluated at ω . That makes my life very easy. What I am saying in effect is ψ tilde cap a naught raise to the power of k ω mod squared is simply ψ cap a naught to the power of k ω mod squared divided once again by $s d s$ ψ a naught evaluated ω .

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WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

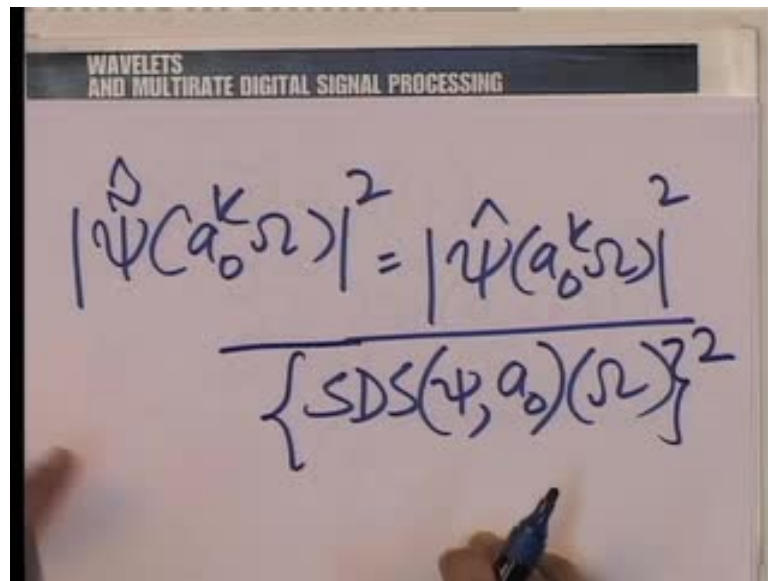
$$\frac{SDS(\tilde{\psi}, a_0)(\Omega)}{SDS(\psi)}$$
$$= \frac{\sum_{k=-\infty}^{+\infty} |\hat{\tilde{\psi}}(a_0^k \Omega)|^2}{SDS(\psi)}$$

(Refer Slide Time: 40:41)

WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\frac{|\hat{\tilde{\psi}}(a_0^k \Omega)|^2}{\left\{ \sum_{l=-\infty}^{+\infty} |\hat{\tilde{\psi}}(a_0^l a_0^k \Omega)|^2 \right\}^{1/2}}$$

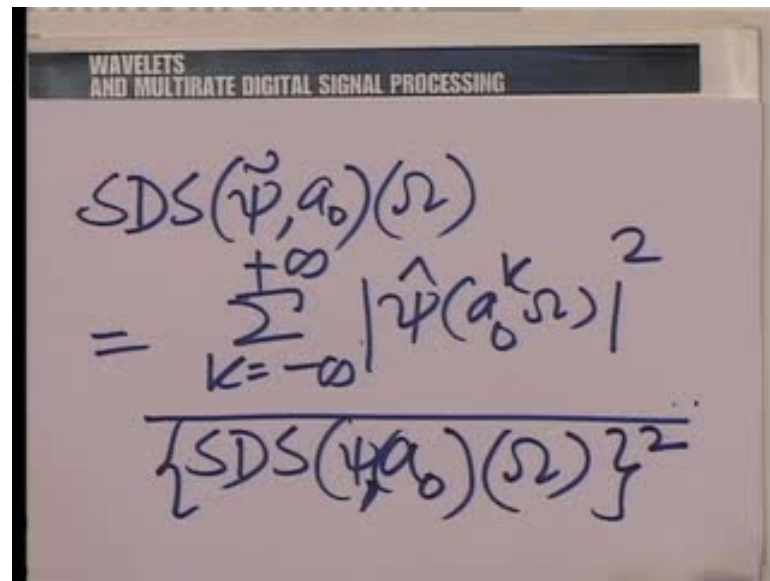
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$$\frac{|\hat{\psi}(a_0^k \Omega)|^2}{\{SDS(\psi, a_0)(\Omega)\}^2}$$

It makes my summation very easy to do. So, what I am saying in effect is that the sum of dilated spectra of ψ with parameter a_0 evaluated at Ω . In another words, summation only on the numerator. You see, the denominator is independent of k . In fact, a little correction which we need make here, you see, here, I need to put a square. So, although I had discuss this in the context when I took the modulus squared, I should also put a square here. I do not need to put a modulus, because this the anywhere nonnegative. So, with that little correction, what I have in the denominator, in fact, here also I need to make the correction, so I have $s d s \psi a_0 \Omega$ squared here; a little correction is required.

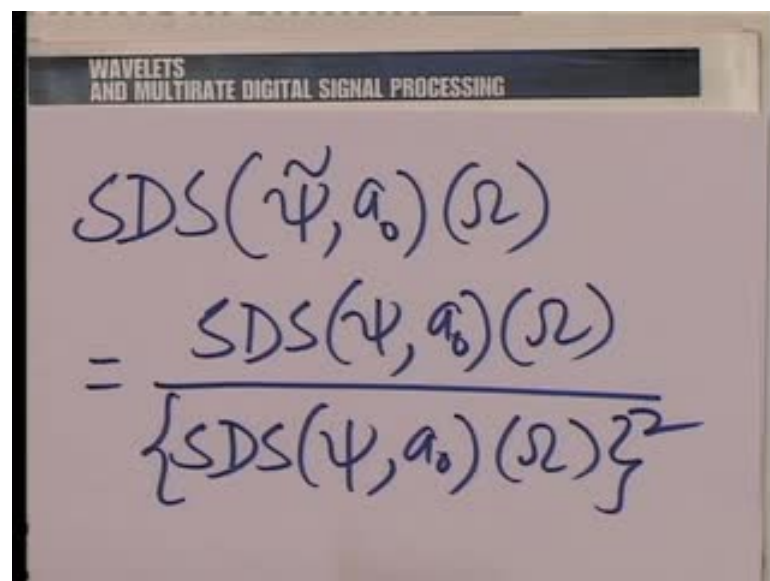
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The slide shows a handwritten equation for the squared magnitude of the Short-Time Fourier Transform (STFT) of a wavelet $\tilde{\psi}$ at scale a_0 . The equation is:

$$\begin{aligned} \text{SDS}(\tilde{\psi}, a_0)(\Omega) &= \sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \Omega)|^2 \\ &= \frac{\sum_{k=-\infty}^{+\infty} |\hat{\psi}(a_0^k \Omega)|^2}{\{\text{SDS}(\psi, a_0)(\Omega)\}^2} \end{aligned}$$

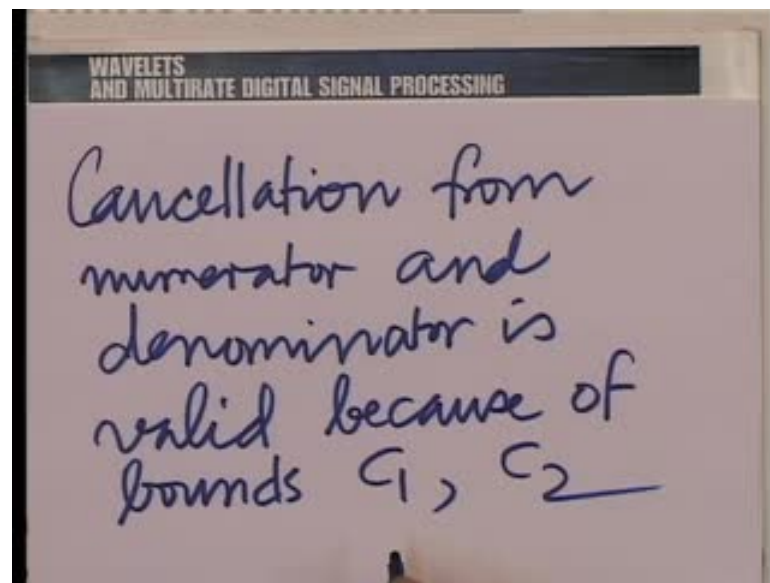
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The slide shows a handwritten equation for the Short-Time Fourier Transform (STFT) of a wavelet $\tilde{\psi}$ at scale a_0 . The equation is:

$$\begin{aligned} \text{SDS}(\tilde{\psi}, a_0)(\Omega) &= \frac{\text{SDS}(\psi, a_0)(\Omega)}{\{\text{SDS}(\psi, a_0)(\Omega)\}^2} \end{aligned}$$

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WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\frac{S DS(\tilde{\psi}, a_0)(\Omega)}{S DS(\psi, a_0)(\Omega)} = 1$$

So, here we have ψ a naught Ω squared, and now, this is familiar; this is essentially $S DS \psi$ a naught once again. So, what I have in effect is that $S DS \tilde{\psi}$ a naught evaluated at Ω is just $S DS \psi$ a naught evaluated Ω divided by $S DS$ squared, and this is where the lower bound becomes important. Since this is lower bounded by c_1 , one never goes to zero, and it is also upper bounded by c_2 , so never goes to infinity. It is valid to cancel. Cancellation from the numerator and denominator is valid because of the bounds, and on cancellation, we shall get $S DS$ the sum of dilated spectra of $\tilde{\psi}$ evaluated Ω with parameter a naught is just the reciprocal, simple, and in fact, now,

I can go a step further. I can use the bounds on the sum of dilated spectra for ψ to come up with bounds on the sum of dilated spectra of $\tilde{\psi}$. Let us do that.

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$$0 < c_1 \leq \text{SDS}(\psi, a_0)(\Omega) \leq c_2 < \infty$$

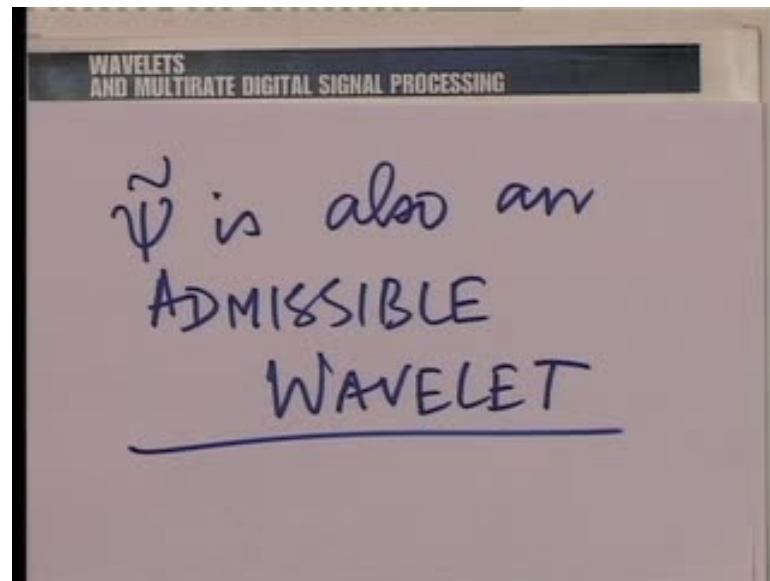
$$\infty > \frac{1}{c_1} \geq \frac{1}{\text{SDS}(\psi, a_0)(\Omega)} \geq \frac{1}{c_2} > 0$$

$$\text{SDS}(\tilde{\psi}, a_0)(\Omega)$$

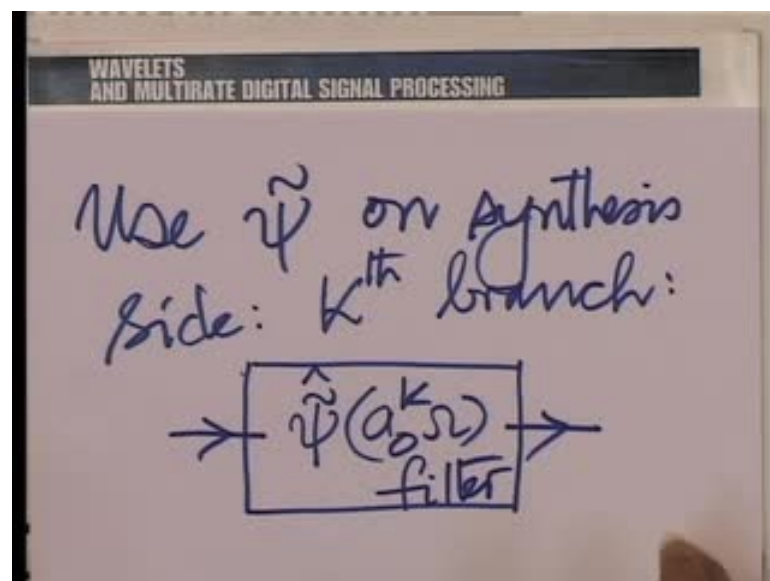
So, let us come back to this expression. You see, all that we need to do is to note that if $\text{SDS}(\psi, a_0)(\Omega)$ is bounded between c_2 and c_1 , this one greater than zero and this one less than infinity. Then, I can take the reciprocal on all side, because these are all nonnegative quantities and I can get infinity is greater than one by c_1 is greater than or equal to one by $\text{SDS}(\psi, a_0)(\Omega)$ is greater than equal to one by c_2 which is in term greater than zero, and this of course, is the sum of dilated spectra that we are looking for here. This is nothing but $\text{SDS}(\tilde{\psi}, a_0)(\Omega)$.

So, we have a bound; the bounds on the sum of dilated spectra of ψ give you the bounds on the sum of dilated spectra of $\tilde{\psi}$, and the same arguments that we used to show that ψ is admissible can be used to show that $\tilde{\psi}$ is also admissible.

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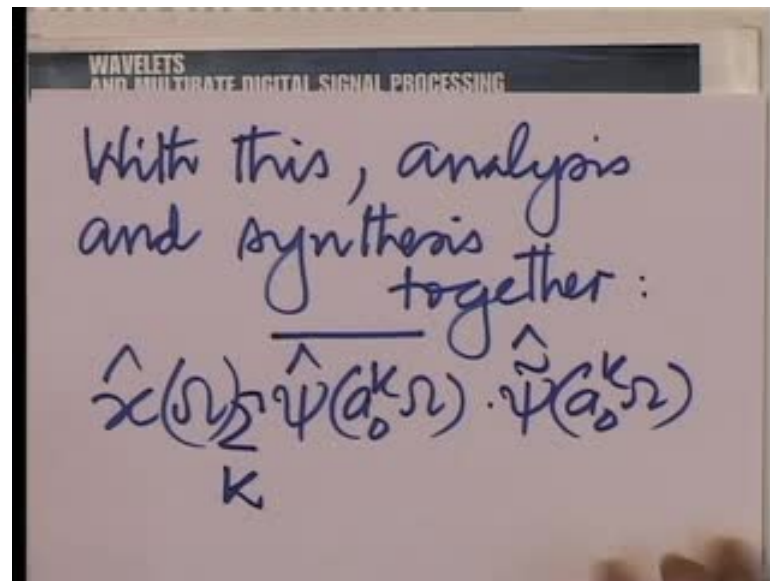


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So, $\tilde{\psi}$ is also an admissible wavelet, and therefore, I could not construct an asymmetric analysis and synthesis pyramid. On the analysis side, I would use the wavelet ψ ; on the synthesis side, I would use the wavelet $\tilde{\psi}$.

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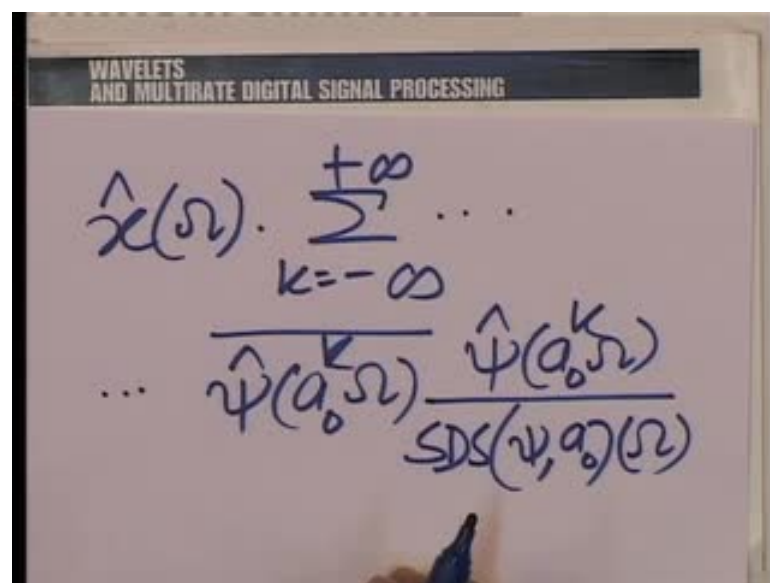
WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

With this, analysis
and synthesis
together:

$$\hat{x}(\omega) = \sum_k \hat{\psi}(a_0^k \omega) \cdot \hat{\psi}(a_0^k \omega)$$

So, all that I am saying is use psi tilde on the synthesis side; that means the kth branch is as follows: it has a filter with frequency response given by psi tilde cap a naught raise to the power of k omega, and now, we can see what happens to x when it goes to the analysis and synthesis side. With this, analysis and synthesis together would produce x cap omega times psi cap a naught raise to the power of k omega complex conjugate multiply by psi tilde cap a naught raise to the power of k omega, and this is easily seen to be, of course, this is summed over all k.

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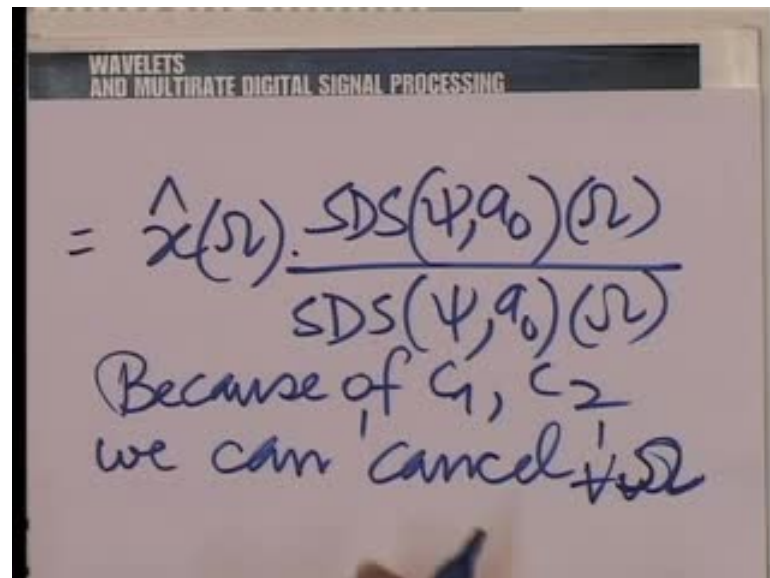
WAVELETS
AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$\hat{x}(\omega) \cdot \sum_{k=-\infty}^{+\infty} \dots$$

$$\dots \frac{\hat{\psi}(a_0^k \omega)}{SDS(\psi, a_0)(\omega)}$$

This is easily seen to be $\sum_k \hat{x}(\omega) \psi(a^k \omega) \psi(a^k \omega)^*$ for going from minus to plus infinity. $\psi(a^k \omega)$ raised to the power of k ω complex conjugate $\psi(a^k \omega)$ raised to the power of k ω divided by the sum of dilated spectra $\psi(a^k \omega)$ evaluated at ω , and once again, you can see what all this is leading to. This is independent of k .

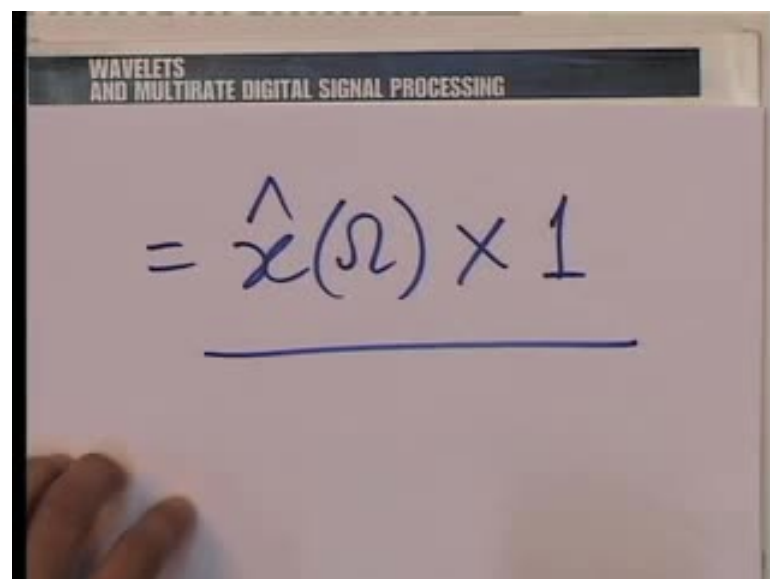
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$$= \hat{x}(\omega) \cdot \frac{SDS(\psi, a_0)(\omega)}{SDS(\psi, a_0)(\omega)}$$

Because of c_1, c_2
we can cancel ω

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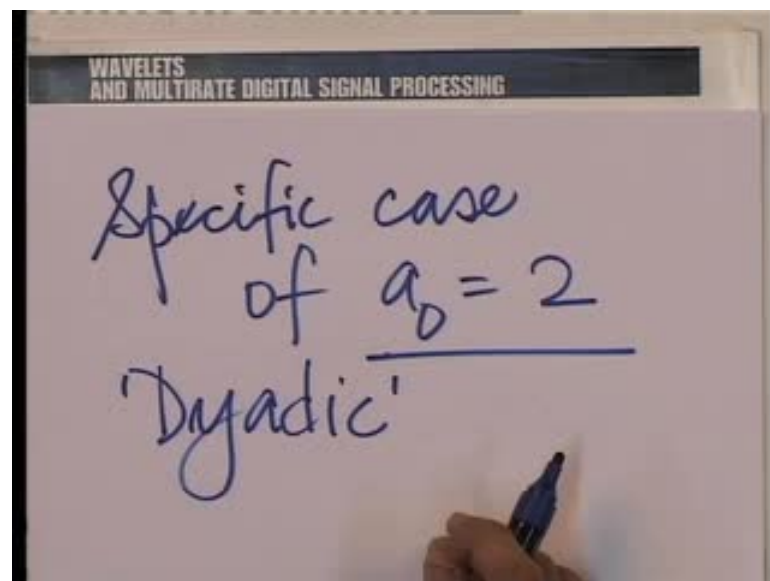
$$= \hat{x}(\omega) \times 1$$

This simply gives you $\hat{x}(\omega)$ into sum of dilated spectra $\psi(a\omega)$ divided by the sum of dilated spectra $\psi(a\omega)$, and once again, because of c_1 and c_2 , we can cancel for all ω and that gives us just $\hat{x}(\omega)$.

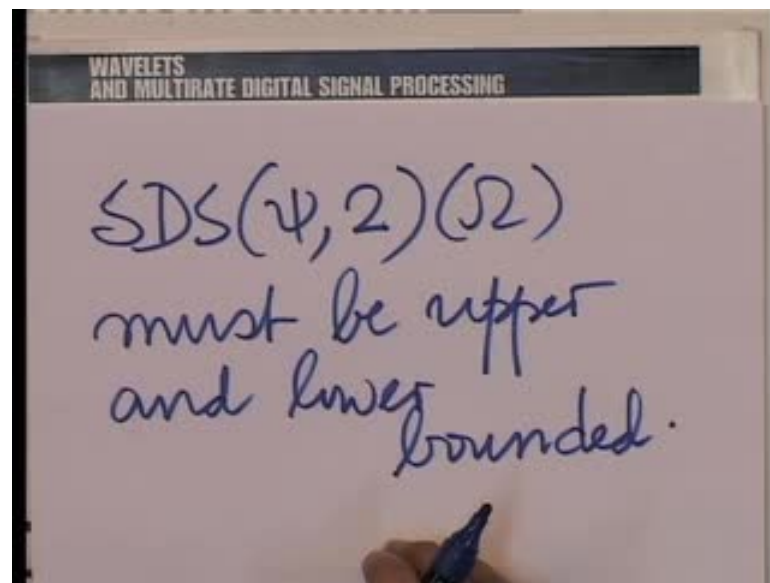
So, we reconstruct. Now, we have also brought in a new notion here. If we have made this relaxation, namely that the sum of the dilated spectra can lie between two positive constants, we also need to generalize our notion of analysis and synthesis a little bit.

We need to generalize it by allowing a different wavelet on the analysis side and on the synthesis side. You have ψ on the analysis side; $\tilde{\psi}$ on the synthesis side. So, we are slowly leading to a different paradigm in the context of filter banks. All this, while the wavelet on the analysis and the synthesis side has been the same.

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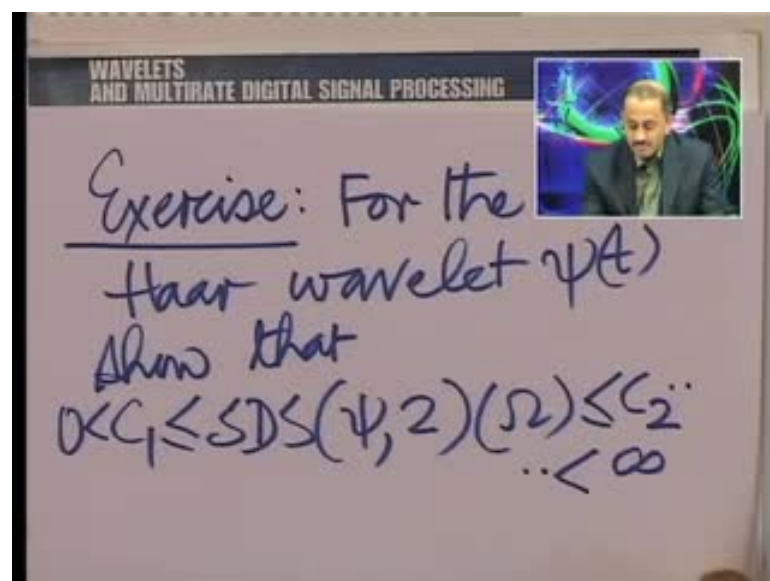


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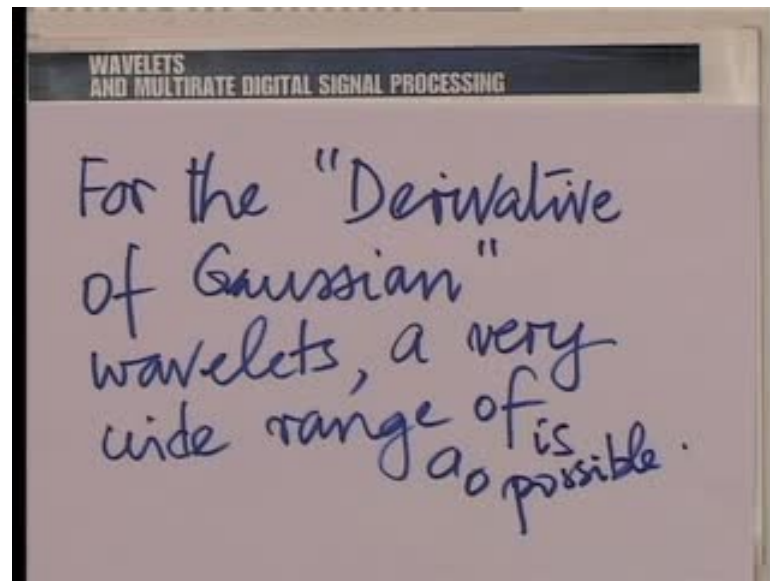
Now, we are allowing them to be different if we want to relax the context of design. Anyway, that remark apart. Let us come back to one specific case. The specific case a naught equal to two. In the specific case of a naught equal to two which is also called the dyadic case; dyadic refers to this - a naught equal to two. What we are asking for is that the sum of dilated spectra of psi two omega must be upper and lower bounded. Now, in particular, if one considers the haar wavelet, one can show that this is true. I do not wish to go through the details, but rather to leave it to you as an exercise.

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So, exercise for the haar wavelet, for example, show that the sum of dilated spectra as we have constructed lies between two positive bounds. Now, you know, to give your hint, it is clear that the upper bound essentially gives you admissibility, and it is not very difficult to show the haar wavelet is admissible. So, first, if you can show the simply haar wavelet is admissible, then this is kind of done.

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The lower bound is a little more tricky and that I leave as an exercise for the class to do, and of course, this also holds for the Gaussian wavelet, for example, so, I have also put down the following exercise. For the Gaussian wavelet of the derivative Gaussian, one can once again show a very wide range of a naught is possible. So, one can show that the sum of dilated spectra in the context of the derivative of Gaussian kind of wavelets is always between two bounds for a very wide range of a naught, and I encourage as an exercise to find out what that range of a naught is.

Now, one more remark - incase the sum of dilated spectra is one is constant for all omega. So, in other words, c one becomes equal to c two. Then we have the situation where psi and psi tilde are the same, and that is essentially the situation of orthogonal filter banks, the kind of filter banks that we have been talking about all this while.

So, in other words, the specific situation where analysis and reconstruction proceeds from the same wavelet is the case where the sum of dilated spectra. Well, it, you know,

please remember, here we are talking about a continuous translation parameter. So, we have not yet discretize the translation parameter.

So, it does not follow that the haar wavelet has the sum of dilated spectra equal to a constant, because we have discretize the translation parameter there, but if you are not to discretize it, then those wavelets where c_1 is equal to c_2 give you an orthogonal filter bank; that means the wavelet on the analysis and the synthesis side is the same, and in particular, when a naught equal to two, it gives you a dyadic analysis synthesis. **syst** With this thing, we conclude the lecture today and we shall build further on the dyadic case in the next lecture. Thank you.