

Fundamentals of Wavelets, Filter Banks and Time Frequency Analysis.

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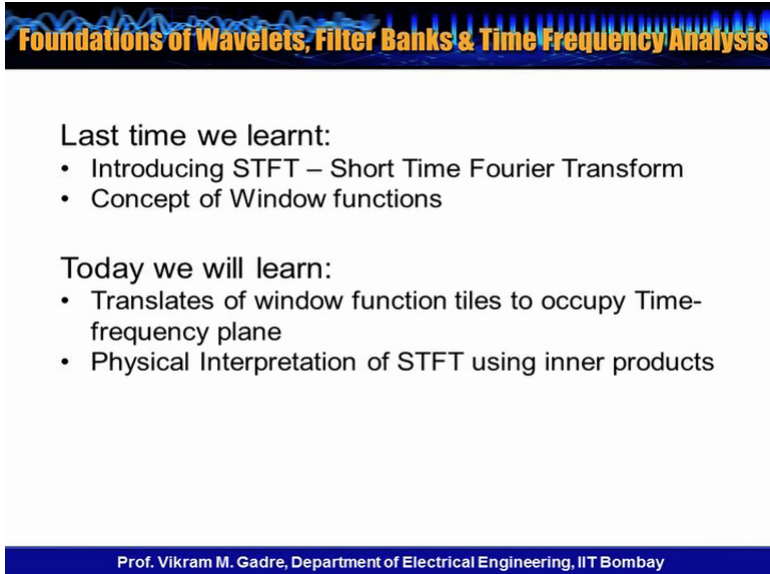
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Week-8.

Lecture-21.2.

STFT: Time domain and frequency domain formulations.

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Foundations of Wavelets, Filter Banks & Time Frequency Analysis

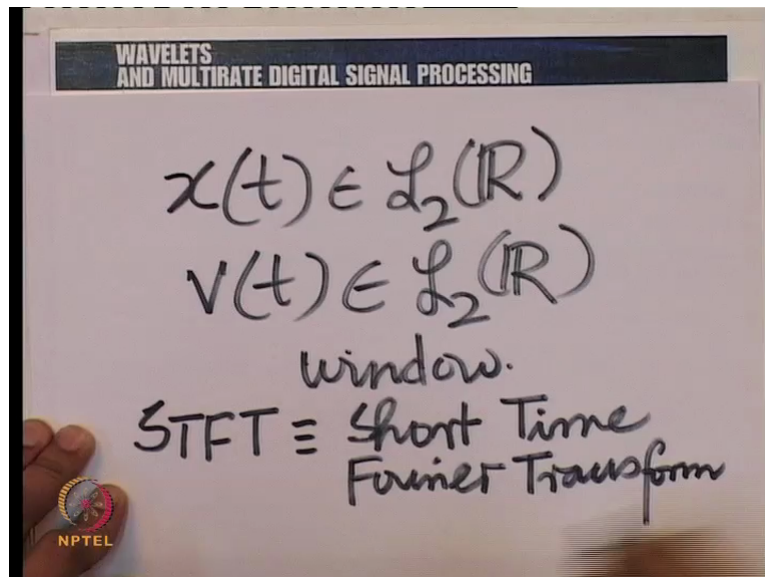
Last time we learnt:

- Introducing STFT – Short Time Fourier Transform
- Concept of Window functions

Today we will learn:

- Translates of window function tiles to occupy Time-frequency plane
- Physical Interpretation of STFT using inner products

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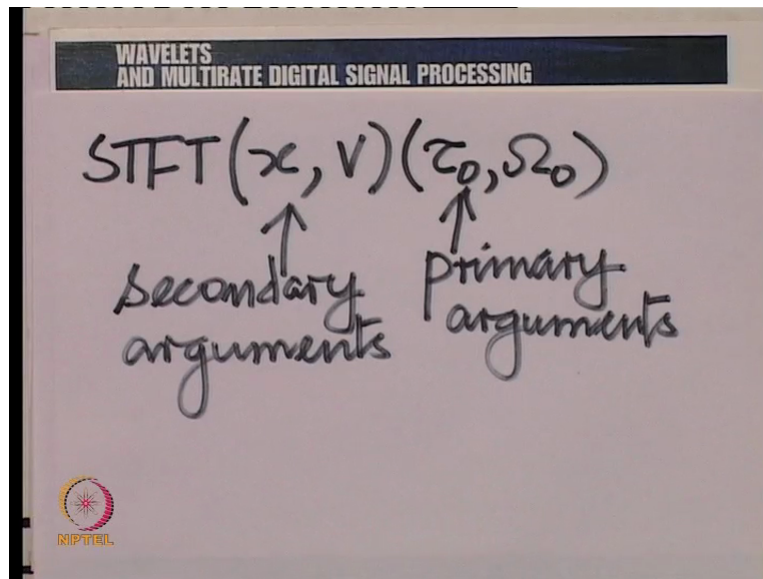
WAVELETS AND MULTIRATE DIGITAL SIGNAL PROCESSING

$$x(t) \in L_2(\mathbb{R})$$
$$v(t) \in L_2(\mathbb{R})$$

Window.

STFT \equiv Short Time Fourier Transform

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Anyway, with that background let us now use as tiles translates modulates of this window. So what we do is to construct a continuum of the following kinds of dot products. We have a function x_t belonging to $L^2\mathbb{R}$ and we have chosen a window v_t , the short time Fourier transform of x_t , so the short time various transform is abbreviated often by STFT. So let us use this abbreviation now onwards. The short time Fourier transform or STFT of x with respect to V as we shall call it.

STFT of x with respect to V , so we have 2 arguments, the function whose short time Fourier transform is being constructed and the window with respect to which it is being constructed. Now this is essentially the set of arguments which determine what short time Fourier transform we are constructing but there is also a pair of arguments for this transform which are the translation and the modulation arguments. So these are I would call the so-called primary arguments of the short time Fourier transform.

And these, one should call the secondary arguments. To explain this idea of primary and secondary arguments, let us take for example the Fourier transform. In the Fourier transform, the primary argument is the frequency, the angular frequency capital Ω . The secondary argument is the function whose Fourier transform is being calculated. So when you calculate the Fourier transform of x_t , the secondary argument is the function x , the transform is the whole operator.

It takes this function x as a secondary argument and the primary argument is the angular frequency. So with that little explanation let us go back to the short time Fourier transform of x with respect to the window V evaluated, this is how we would read it, the short time Fourier

transform of x with respect to the window V evaluated at the translation τ_0 and the modulation Ω_0 . Let us write that down. It would be essentially a dot product of the following.

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$$= \text{dot product of (inner product)} \\ x(t) \text{ with } v(t - \tau_0) e^{j\Omega_0 t}$$

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$$= \int_{-\infty}^{+\infty} x(t) \overline{v(t - \tau_0)} e^{j\Omega_0 t} dt$$

$$= \int_{-\infty}^{+\infty} x(t) \overline{v(t - \tau_0)} e^{-j\Omega_0 t} dt$$

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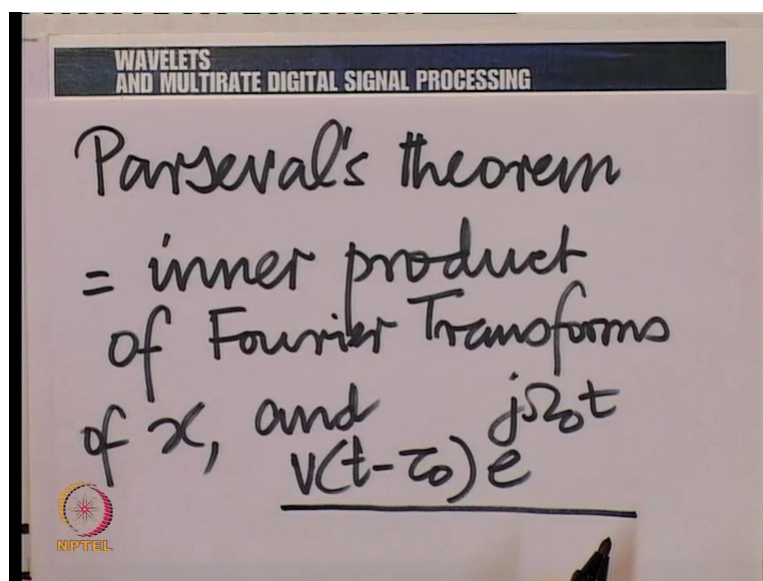
Dot product or the inner product of $x(t)$ with V translated by τ_0 and modulated with e raised to the power $j\Omega_0 t$. And how would dot product look, let us write it down. $x(t)$ multiplied by $\overline{v(t - \tau_0)} e^{j\Omega_0 t} dt$. And if you please, we could simplify this. Now a couple of words here. If $v(t)$ is the real function, as we have been considering all this while, whether it is a Gaussian or whether it is a raised cosine or the triangular function or if you like the extreme case of the rectangular window, this complex conjugation is redundant, it is not required.

If you look at it, what we are doing here in alternate sense is to 1st multiply $x(t)$ by a window appropriately translated, so t_0 is the location of the window. Essentially what we are trying is to extract information of $x(t)$ around the point t equal to t_0 . And then we are taking a Fourier transform, that is another way to interpret it. The short time Fourier transform is essentially a process of piecing or breaking the function into pieces followed by a Fourier transformation.

Here we are assuming a continuous t_0 and a continuous capital Ω_0 . And therefore we are talking about the continuous short time Fourier transform. Now, let us invoke Parseval's theorem to get different perspective on the same expression here. So let us go back to the expression here. We have taken a dot product of $x(t)$ with a translated version of the window and then a Fourier transformation on this product. So for example if $v(t)$ were to be a triangular window, essentially it would extract the information of $x(t)$ around the point t_0 with some weighting done by the rectangular, by the triangular function or the rectangular function.

In the rectangular function, there is no weighting, in the raised cosine or the triangular or the Gaussian function there is a weighting, different weights given to different points around t_0 , followed by a Fourier transformation, so looking at the frequency content in a certain region. Now if we invoke Parseval's theorem on this, what does it give us?


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Fourier Transform
of $v(t - \tau_0) e^{j\Omega_0 t}$

$$= \int_{-\infty}^{+\infty} v(t - \tau_0) e^{j\Omega_0 t} e^{-j\Omega t} dt$$




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Let $t - \tau_0 = \lambda$

$$= \int_{-\infty}^{+\infty} v(\lambda) e^{j(\Omega_0 - \Omega)(\lambda + \tau_0)} d\lambda$$

$$= e^{j(\Omega_0 - \Omega)\tau_0} \int_{-\infty}^{+\infty} v(\lambda) e^{j(\Omega_0 - \Omega)\lambda} d\lambda$$



Parseval's theorem says, this is also the inner product of the Fourier transforms of x and V translated by τ_0 and modulated by capital Ω_0 . Therefore we need to calculate the Fourier transform of this quantity, that is a task that we need to do. The Fourier transform of $v(t - \tau_0) e^{j\Omega_0 t}$ can easily be seen to be the following. So it is minus to plus infinity $v(t - \tau_0) e^{j\Omega_0 t}$ so the power minus, well $j\Omega_0 t$ multiplied by $e^{-j\Omega t}$ integrated with respect to t .

And let us simplify this, so of course we employ the standard process of replacement of argument. So let us replace the argument $t - \tau_0$. And noting that for a fixed τ_0 , when t runs from minus to plus infinity, λ also runs from minus to plus infinity, this would become integral from minus to plus infinity $V(\lambda) e^{j(\Omega_0 - \Omega)\lambda}$ now of

course collecting terms, this is Ω_0 minus Ω times t which is λ plus τ_0 d λ .

And now of course we have a simple answer that. We can rewrite this little bit, taking out common terms. So you will notice that when we expand this product, we have that e raised to the power $j\Omega_0$ minus, rather Ω_0 minus Ω times τ_0 . And e raised to the power $j\Omega_0$ minus Ω times τ_0 is independent of λ , so I could bring it outside the integral. So I am saying this becomes e raised to the power $j\Omega_0$ minus Ω times τ_0 and rest of it inside.

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$$\int_{-\infty}^{+\infty} v(\lambda) e^{-j(\Omega - \Omega_0)\lambda} d\lambda = \hat{v}(\Omega - \Omega_0)$$

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Fourier Transform
of $v(t - \tau_0) e^{j\Omega_0 t}$

$$= \{ \hat{v}(\Omega - \Omega_0) \} \{ e^{j(\Omega_0 - \Omega)\tau_0} \}$$

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$$STFT(x, v)(\tau_0, \Omega_0)$$

$$= \frac{1}{2\pi} \int_{-\infty}^{+\infty} \hat{x}(\Omega) \cdot \hat{v}(\Omega - \Omega_0) e^{j(\Omega - \Omega_0)\tau_0} d\Omega$$


And now let me see what remains inside, what remains inside is $V \text{ Lambda } E_0$ raised to the power $J \text{ Omega } 0 \text{ minus } \text{Omega times Lambda d Lambda}$. So let me write that part down. So now here I should write only the integral, I have left out the other terms. The integral is essentially $V \text{ lambda } e$ raised to the power, I will rewrite it, $\text{minus } J \text{ Omega minus } \text{Omega } 0 \text{ times Lambda d Lambda}$ which is easily seen to be the Fourier transform of V but evaluated at $\text{Omega minus } \text{Omega } 0$ instead of at Omega .

So all in all we have the following. We have the Fourier transform of $v t \text{ minus } \tau_0$, e raised to the power $J \text{ Omega } 0 t$ is the Fourier transform of V evaluated at $\text{Omega minus } \text{Omega } 0$ multiplied by e raised to the power $J \text{ Omega } 0 \text{ minus } \text{Omega times } \tau_0$. And we shall substitute this in the Parseval's theorem expression. So we have the short time Fourier transform, secondary argument of x with respect to the window v evaluated at τ_0 and capital $\text{Omega } 0$ is essentially the following inner product with a factor of $1 \text{ by } 2 \text{ pie}$ remember.

So it is $x \text{ cap } \text{Omega } v \text{ cap } \text{Omega minus } \text{Omega not } e$ raised to the power $J \text{ Omega } 0 \text{ minus } \text{Omega times } \tau_0 \text{ d } \text{Omega}$ with this whole complex conjugate. And let us simplify that and once again let us note that when we expand this, we get 2 terms e raised to the power $J \text{ Omega } 0 \tau_0$ and e raised to the power $\text{minus } J \text{ Omega } \tau_0$, out of which only the 2nd depends on capital Omega , the 1st does not, the 1st can be brought outside the integral.

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$$= e^{-j\Omega_0 \tau_0} \dots$$

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{x}(\Omega) \hat{v}(\Omega - \Omega_0) e^{j\Omega \tau_0} d\Omega$$

NIPTEIL

So what we have in effect is e raised to the power minus $J \Omega_0 \tau_0$ times the following. $\frac{1}{2\pi}$, just the constant \times cap Ω we cap Ω minus Ω_0 e raised to the power $J \Omega_0 \tau_0$ $D \Omega$. This is complex conjugated here, here the complex conjugate has been taken care of as it is. And now we have a very beautiful interpretation for this, this looks very much like an inverse Fourier transform if you think about it. It is the inverse Fourier transform evaluated at the point τ_0 .

Inverse Fourier transform of what, of the Fourier transform of x multiplied by the Fourier transform of the window shifted to lie around Ω_0 . So now this makes a lot of sense. The short time Fourier transform as we expected has an interpretation, a very similar interpretation both in time and frequency.