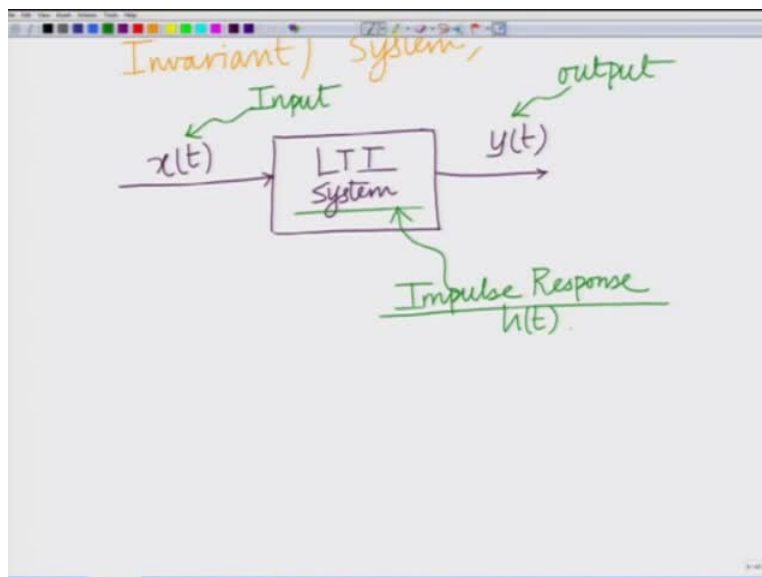
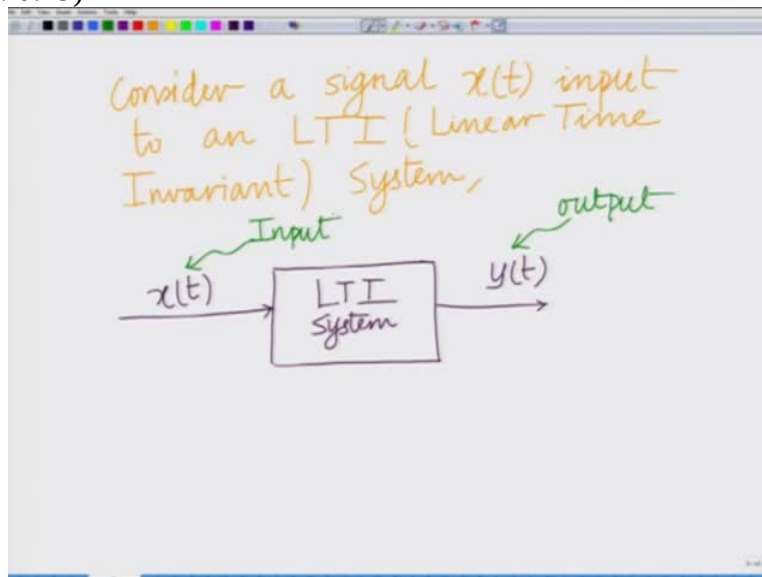


Principles of Communication- Part I
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Module No 2

**Lecture 07: Profession of Signal through Linear Time Invariant Systems (LTI)
and Cross-Correlation of Signals**

Hello welcome to another module in this massive open online course. So let us continue our discussion on linear time invariant systems and what happens when a signal $x(t)$ is input to a linear time invariant system.

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So consider a signal $x(t)$ input to an LTI that is your linear time invariant system, remember yesterday we elaborated on what did, what a linear, the properties is of a linear time invariant system that is linearity and time invariance. So let us consider a signal $x(t)$ that is given as an input to a linear time invariant system, okay. So we have our LTI system, okay. So $x(t)$ and let the corresponding input output be $y(t)$.

So this is our LTI system, so $y(t)$ is the output of the LTI system, $x(t)$ it is the input to the LTI system, correct? And any LTI system is characterized by an impulse response response $h(t)$. What is the impulse response $h(t)$? That is if the impulse that is your direct Delta function $\delta(t)$ is given as input to the LTI system, the output is represented by $h(t)$ which is the impulse which is the response to the direct Delta function $\delta(t)$ and this is known as the impulse response of the LTI system.

And this fundamentally characterizes the properties in the behavior of the LTI system. And the output corresponding to any signal any input signal $x(t)$ can be described in terms of the input $x(t)$ and the impulse response $h(t)$ of the LTI system as follows.

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Impulse Response $h(t)$.

$$y(t) = x(t) * h(t)$$

input signal CONVOLUTION OPERATOR

$$= \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$

The image shows a digital whiteboard with handwritten mathematical expressions. At the top, the word "signal" is written in green with a pink arrow pointing to the first integral. To its right, "CONVOLUTION OPERATOR" is written in green. The first integral is
$$= \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$
 with a pink arrow pointing to the $x(\tau)$ term. Below it, the second integral is
$$= \int_{-\infty}^{\infty} x(t-\tau) h(\tau) d\tau$$
 with a pink arrow pointing to the $x(t-\tau)$ term.

$y(t)$ that is the output corresponding to any signal $x(t)$ is $x(t)$ convolved with $h(t)$ where $x(t)$ is your input signal, this is the convolution operator $y(t)$ is $x(t)$ convolved with $h(t)$, therefore if I know the impulse response $h(t)$ of the LTI system that completely characterises the LTI system for any given in other arbitrary input signal $x(t)$ the output $y(t)$ can be derived by can convolving the input $x(t)$ with the impulse response $h(t)$.

And this convolution operation is represented as this is equal to basically integral - infinity to infinity $x(t - \tau) h(\tau) d\tau$ or that is your $x(\tau) h(t - \tau) d\tau$ which is also equal to integral - infinity to infinity $x(t - \tau) h(\tau) d\tau$. So the output $y(t)$ is the input $x(t)$ convolved with the impulse response $h(t)$ and this has a very interesting, the convolution has a very interesting has a convolution follows a very interesting property.

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Handwritten notes on a digital whiteboard. At the top, three Fourier transform pairs are listed, each with a double-headed arrow between the time and frequency domains:

$$\begin{aligned}x(t) &\longleftrightarrow X(F) \\ y(t) &\longleftrightarrow Y(F) \\ h(t) &\longleftrightarrow H(F)\end{aligned}$$

Below these, the convolution equation is written:

$$y(t) = x(t) * h(t)$$

A curved arrow points from the convolution equation to a boxed equation in the frequency domain:

$$Y(F) = X(F) \cdot H(F)$$

Handwritten notes on a digital whiteboard, identical to the previous slide. It shows the same Fourier transform pairs and the convolution equation. The boxed frequency domain equation is also present:

$$Y(F) = X(F) \cdot H(F)$$

Below the boxed equation, the text "Multiplication in Frequency Domain" is written in cursive, with an arrow pointing to the multiplication symbol in the equation.

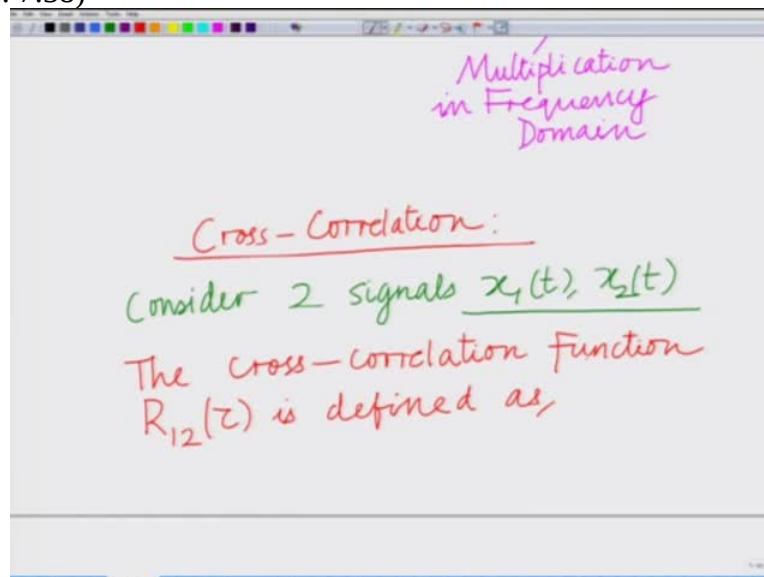
If I consider the Fourier transform of the input and output, let $x(t)$ have a Fourier transform pair with $X(F)$ the output $y(t)$ we Fourier transform pair with $Y(F)$ that is Fourier transform of $y(t)$ is given by $Y(F)$. The Fourier transform of the impulse response $h(t)$ be given by H of F then we have in the time domain $y(t)$ equals $x(t)$ convolved with $h(t)$. This implies in the frequency domain the Fourier transform $Y(F)$ equals the input Fourier transform $X(F)$ times that is the product $H(f)$. That is convolution in the time domain becomes multiplication in the frequency domain.

This becomes multiplication. This is a way convolution satisfies a very interesting property that is if 2 signals are convolved in the time domain that is $y(t)$ is the convolution of $x(t)$ with $h(t)$ then the Fourier transform of $y(t)$ that is $Y(F)$ is obtained by the multiplication of the Fourier transform that is the Fourier transform of $x(t)$ that is $X(F)$ with the Fourier transform of $h(t)$ that is $H(F)$.

So $Y(F)$ equals $X(F)$ times $H(F)$ that is convolution in the time domain is multiplication is equivalent to multiplication of the 2 signals in a of the Fourier transform of that is 2 respective signals in the frequency domain and this is 1 of the fundamental properties of linear systems and it is also very applicable, 1 of the fundamental principles of, communication because we are going to look at several instances where input signal is passed through an LTI system and we derive an output signal and therefore it is very important to remember that the Fourier transform, right?

Or spectrum of the output signal is derived by the multiplication of the spectra of the Fourier transform. So the input signal and the impulse response alright. Now let us look at another concept that is the cross correlation between 2 signals.

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$$R_{12}(\tau) = \int_{-\infty}^{\infty} x_1(t) x_2^*(t - \tau) dt$$

Auto correlation
For lag τ

Characterizes the extent or degree of SIMILARITY between $x_1(t)$, $x_2(t)$ for a shift of τ .

$$R_{12}(\tau) = \int_{-\infty}^{\infty} x_1(t) x_2^*(-(\tau - t)) dt$$

a shift of τ .

$$R_{12}(\tau) = \int_{-\infty}^{\infty} x_1(t) \underbrace{x_2^*(-(\tau - t))}_{\tilde{x}_2(\tau - t)} dt$$

$$\tilde{x}_2(t) = x_2(-t)$$

$$\Rightarrow \tilde{x}_2(\tau - t) = x_2(t - \tau)$$

$$\Rightarrow x_2^*(\tau - t) = \tilde{x}_2^*(t - \tau)$$

Handwritten derivation on a digital whiteboard:

$$\Rightarrow \tilde{x}_2(\tau - t) = x_2(t - \tau)$$

$$\Rightarrow x_2^*(\tau - t) = x_2^*(t - \tau)$$

$$\Rightarrow \int_{-\infty}^{\infty} x_1(t) \tilde{x}_2^*(\tau - t) dt$$

CONVOLUTION of
 $x_1(t) * \tilde{x}_2^*(t)$

$$= x_1(t) * x_2(-t)$$

So we will start looking at another concept that is the cross correlation. The cross so consider for this purpose considered 2 signals $x(t)$ and $y(t)$ or 2 signals $x_1(t)$, $x_2(t)$, correct? The cross correlation function $R_{12}(\tau)$ the cross correlation function are $R_{12}(\tau)$ is defined as, well $R_{12}(\tau)$ equals integral - infinity to infinity $x_1(t) x_2^*(t - \tau) dt$.

$\int_{-\infty}^{\infty} x_1(t) x_2^*(t - \tau) dt$, this is $R_{12}(\tau)$, this is the auto correlation for a lag τ that is what we are doing is getting $x_1(t)$. Multiplying it by the shifted version of x_2 , that is $x_2^*(t - \tau)$, multiplying $x_1(t)$ by $x_2^*(t - \tau)$ and integrating - from - infinity to infinity, this basically this is a measure of the extent of similarity between $x_1(t)$ and $x_2(t)$ for the lag τ .

So this characterizes, this is important to know the meaning of the auto correlation characterizes the extent or degree of similarity between $x_1(t)$, $x_2(t)$ for a lag or a shift of τ . For a shift the extent of similarity between $x_1(t)$ and $x_2(t)$ for a shift of τ that is corresponding to the shift of τ . What is the extent of similarity between the signal $x_1(t)$ and $x_2(t - \tau)$?

Alright, now simplify this. That is we have $R_{12}(\tau)$ or $R_{12}(\tau)$ we have written this as $\int_{-\infty}^{\infty} x_1(t) x_2^*(t - \tau) dt$ I am going to write this as $\int_{-\infty}^{\infty} x_1(t) x_2^*(\tau - t) dt$. Now let us define $\tilde{x}_2(t)$ as $x_2(-t)$ this implies $x_2^*(\tau - t) = x_2^*(\tau - (-t)) = x_2^*(\tau + t)$ this implies $x_2^*(\tau - t) = x_2^*(\tau + t)$ therefore here instead of $x_2^*(t - \tau)$ I can substitute, here I can substitute $\tilde{x}_2^*(t)$.

So therefore this becomes a cross correlation becomes this is integral - infinity to infinity $x_1(t) x_2^*(t - \tau) dt$ and notice this is nothing but convolution of $x_1(t)$ and $x_2^*(t)$. This basically you will see realize that from all definition this is convolution of, it follows from the definition of convolution that this is a convolution of $x_1(t)$ with $x_2^*(t)$ but realize $x_2^*(t)$ is $x_2^*(t)$, I am sorry this is $x_2^*(t)$ conjugate $t - \tau$.

This is a convolution of $x_1(t)$ with $x_2^*(t)$ but $x_2^*(t)$ is equal to $x_2^*(t)$ conjugate $-t$. So this is a convolution of $x_1(t)$ with $x_2^*(-t)$, okay.

(Refer Slide Time: 14:01)

The image shows a handwritten derivation on a whiteboard. At the top, the integral $\int_{-\infty}^{\infty} x_1(t) \tilde{x}_2^*(t - \tau) dt$ is written. Below it, a bracket indicates this is the "CONVOLUTION of $x_1(z) * \tilde{x}_2^*(z)$ ". This is then equated to $(x_1 * \tilde{x}_2^*)(z)$. Finally, the result is boxed as $R_{12}(z) = x_1(z) * \tilde{x}_2^*(z)$. A red arrow points from the text "Auto correlation of $x_1(t), x_2(t)$ " to the boxed equation.

So R_{12} or you can also replace this by τ that is R_{12} or you can also replace this by τ . Since t here is the variable of integration, so this is τ this is $-\tau$, so I have $R_{12}(\tau) = x_1(\tau) * x_2^*(-\tau)$, remember this is the auto correlation of $x_1(t)$ and $x_2(t)$ corresponding to a lag of τ . This is the auto correlation function of $x_1(t)$, 2 different signals $x_1(t)$ and $x_2(t)$ corresponding to a lag of τ .

We have shown that this is basically the convolution between $x_1(\tau)$ and $x_2^*(-\tau)$. This is convolution between x_1 and the conjugate version of x_2 . There is conjugate x_2 and flip it, $x_2^*(-\tau)$ $x_2^*(-\tau)$ is nothing but flip it about the origin and take the complex conjugate, alright. And this auto correlation is an important measure because it is it is a measure of the degree of similarity between x_1 and x_2 for a lag of τ , alright.

And this has a lot of uses as we are going to see later in our study of communication system. This measure of out of this measure that is the auto correlation between 2 different signals $x_1(t)$ and $x_2(t)$, okay.

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$$R_{12}(z) = x_1(z) * x_2^*(-z)$$

Auto correlation of $x_1(t), x_2(t)$. Lag or shift

$$S_{12}(F) = \int_{-\infty}^{\infty} R_{12}(z) e^{-j2\pi Fz} dz$$

Now let us try to look at this in the now let us look at this in the frequency domain naturally convolution and the time is that is convolution in the tau domain, of course now this is not even time this is tau, remember this is the lag or shift now if we take to the Fourier transform of this that is denote this by $S_{12}(F)$ equals integral - infinity to infinity $R_{12}(\tau) e^{-j2\pi F\tau} d\tau$ then we know that this convolution in the Tau domain is multiplication in the frequency domain.

(Refer Slide Time: 16:43)

$$R_{12}(\tau) = x_1(\tau) * x_2^*(-\tau)$$

Auto correlation of $x_1(t), x_2(t)$. Lag or shift

$$S_{12}(f) = \int_{-\infty}^{\infty} R_{12}(\tau) e^{-j2\pi f\tau} d\tau$$
$$S_{12}(f) = X_1(f) \cdot \frac{\text{FT}(x_2^*(-t))}{?}$$

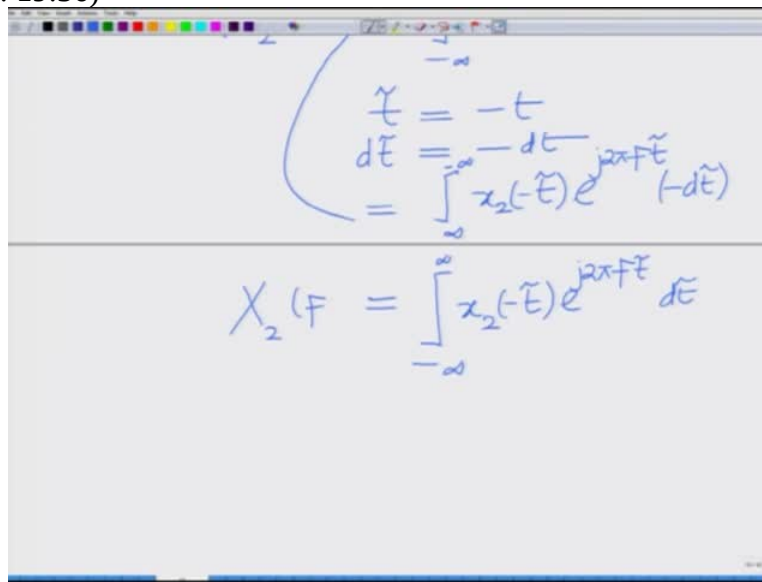
$$S_{12}(f) = X_1(f) \cdot \frac{\text{FT}(x_2^*(-t))}{?}$$
$$x_2(t) \longleftrightarrow X_2(f)$$
$$x_2(t) = \int_{-\infty}^{\infty} X_2(f) e^{j2\pi ft} df$$

$$\begin{aligned}
 X_2(F) &= \int_{-\infty}^{\infty} x_2(t) e^{-j2\pi F t} dt \\
 \tau &= -t \\
 d\tau &= -dt \\
 &= \int_{\infty}^{-\infty} x_2(-\tau) e^{j2\pi F \tau} (-d\tau)
 \end{aligned}$$

Therefore we can write from the property of linear system or from the property of convolution $S_{12} F$ is $x_1 F$ times the Fourier transform of x_2 tilde or x_2 conjugate Fourier transform of x_2 conjugate $-t$. Now what we have to do is we have to derive the Fourier transform of F 2 conjugate $-t$ that is what is the Fourier transform of F 2 conjugate $-t$? Now we know that if $x_2 t$ forms a Fourier transform pair with $x_2 F$ then $x_2 t$ is given by the inverse Fourier transform that is $x_2 F e$ to the power of $j 2 \pi F t d F$ consider the conjugate of this, well if I replace this by $-t$ this becomes over here this becomes $-t$.

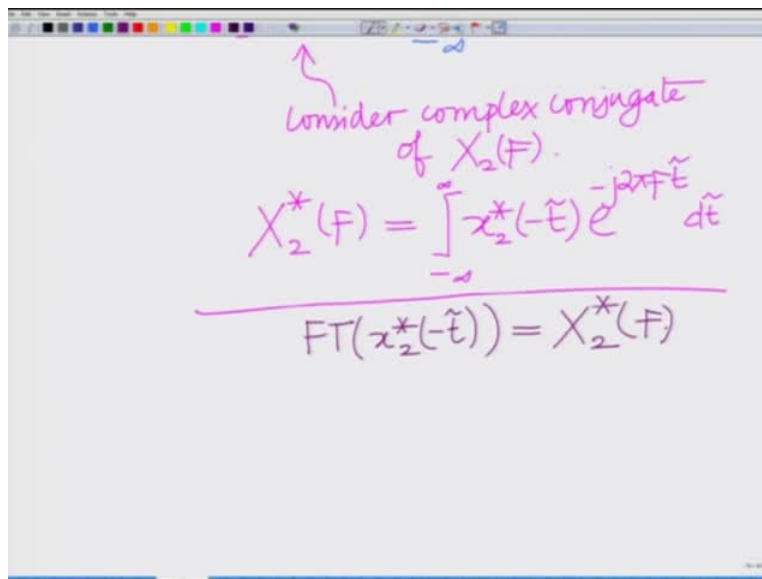
And now if I consider the conjugate, correct? Now if I for instance let us look at this rather than the inverse Fourier transform, let us look at the Fourier transform that is x_2 of F equals $-\infty$ to ∞ x_2 of $t e$ to the power of $-j 2 \pi F t$, correct? Now if I replace dt , correct? Now if I replace t by $-t$ so let us now replace or let us say t tilde equals $-t$ then what we have is dt tilde equals $-dt$. So this integral becomes t tilde equals $-t$, so this integral becomes $-\infty$ to ∞ x_2 of $-t$ tilde e to the power of $j 2 \pi F t$ tilde $-d$ tilde. I can remove this $-$ sign by interchanging the order of integration.

(Refer Slide Time: 19:36)



Handwritten derivation showing the substitution $\tilde{t} = -t$ and $d\tilde{t} = -dt$ into the Fourier transform integral. The integral is then simplified to the standard form for $X_2(F)$.

$$\begin{aligned} \tilde{t} &= -t \\ d\tilde{t} &= -dt \\ &= \int_{-\infty}^{\infty} x_2(-\tilde{t}) e^{j2\pi F \tilde{t}} (-d\tilde{t}) \\ X_2(F) &= \int_{-\infty}^{\infty} x_2(-\tilde{t}) e^{j2\pi F \tilde{t}} d\tilde{t} \end{aligned}$$



Handwritten derivation showing the complex conjugate of $X_2(F)$. It includes a note to "consider complex conjugate of $X_2(F)$ ". The integral is then simplified to the standard form for $X_2^*(F)$.

$$\begin{aligned} &\text{consider complex conjugate of } X_2(F). \\ X_2^*(F) &= \int_{-\infty}^{\infty} x_2^*(-\tilde{t}) e^{-j2\pi F \tilde{t}} d\tilde{t} \\ \hline \text{FT}(x_2^*(-\tilde{t})) &= X_2^*(F) \end{aligned}$$

So this is if I flip is from infinity to - infinity equals - infinity to infinity therefore x_2 of $-t$ tilde e to the power of $j 2 \pi F d$ tilde, the $-$ sign in d tilde goes because I change the other of integration. Now if we consider the complex conjugate. Now consider complex conjugate if we consider the complex conjugate of $x_2 F$ I have x_2 conjugate of F equals - infinity to infinity x_2 conjugate of $-t$ tilde e to the power of $- j 2 \pi F t$ tilde d tilde.

And therefore now, what you can see from this? What you can see from this is that the Fourier transform of x_2 conjugate - t tilde the Fourier transform x_2 conjugate - t tilde is x_2 conjugate F . That is FT of x_2 or x_2 conjugate - t tilde equals x_2 conjugate of F .

(Refer Slide Time: 21:20)

Handwritten derivation on a whiteboard:

$$S_{12}(F) = X_1(F) \cdot \frac{FT(x_2^*(-t))}{?}$$

$$\underline{x_2(t)} \longleftrightarrow \underline{X_2(F)}$$

$$X_2(F) = \int_{-\infty}^{\infty} x_2(t) e^{-j2\pi Ft} dt$$

$$\begin{aligned} \tilde{t} &= -t \\ d\tilde{t} &= -dt \\ &= \int_{-\infty}^{\infty} x_2(-\tilde{t}) e^{j2\pi F\tilde{t}} (-d\tilde{t}) \end{aligned}$$

Handwritten derivation on a whiteboard:

$$\frac{\int_{-\infty}^{\infty} x_2(t) e^{-j2\pi Ft} dt}{FT(x_2^*(-\tilde{t}))} = X_2^*(F)$$

$$S_{12}(F) = X_1(F) \cdot FT(x_2^*(-t))$$

$$\uparrow$$

$$= X_1(F)$$

Fourier Transform of $R_{12}(z)$

$$R_{12}(z) = x_1(z) * x_2^*(-z)$$

Auto correlation of $x_1(t), x_2(t)$. Lag or shift

$$S_{12}(F) = \int_{-\infty}^{\infty} R_{12}(z) e^{-j2\pi Fz} dz$$

$$S_{12}(F) = X_1(F) \cdot \text{FT}(x_2^*(-z))$$

$$x_2(t) \longleftrightarrow X_2(F)$$

And therefore what we have over here that is if x_2 has Fourier transform $X_2(F)$ then $x_2^*(t)$ has Fourier transform $X_2^*(F)$. Therefore now we can write the Fourier transform $S_{12}(F)$, remember $S_{12}(F)$ is Fourier transform of your auto correlation function $R_{12}(z)$, that is $x_1(F)$ times Fourier transform $x_2^*(t)$ or $x_2^*(F)$ which is basically $x_1(F)$ times $X_2^*(F)$. This is the Fourier transform $x_1(F)$ times $X_2^*(F)$. Product of the Fourier that is $x_1(F)$ times $X_2^*(F)$.

(Refer Slide Time: 22:44)

$$S_{12}(F) = X_1(F) \cdot X_2^*(F)$$

Fourier Transform of $R_{12}(z)$

$$R_{12}(z) \longleftrightarrow S_{12}(F) = X_1(F) X_2^*(F)$$

Property of Fourier Transform of cross-correlation

Therefore what we have is R_{12} of τ which has Fourier transform S_{12} of F which is equal to X_1 of F times X_2 conjugate of F . This is the property of the Fourier transform of the cross correlation property of the, sorry this has to be the cross correlation therefore this is the Fourier transform of the cross correlation between 2 functions x_1 , this is the Fourier transform of the cross correlation that is we are considering 2 different signals $x_1(t)$ and $x_2(t)$.

So this is the property of the Fourier transform of the cross correlation that is $R_{12}(\tau)$ denotes the cross correlation between $x_1(t)$ and $x_2(t)$ for a lag of τ , okay. Now what we are going to do is we are going to consider a special case of the cross correlation that is the auto correlation, okay.

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Property of Fourier Transform of cross-correlation

Auto Correlation:

$$R_{11}(\tau) = \int_{-\infty}^{\infty} x_1(t) x_1^*(t-\tau) dt$$

Degree of self similarity of $x_1(t)$ for a Lag of τ .

$$S_{11}(F) = X_1(F) \cdot \text{FT}(x_1^*(-t))$$

$$= X_1(F) \cdot X_1^*(F)$$

$$S_{11}(F) = |X_1(F)|^2$$

Fourier-Transform of auto-correlation of $x_1(t)$ is $|X_1(F)|^2 \sim R_{11}(t)$

So now we are going to consider the auto correlation which is basically now the that is if I replace $x_2(t)$ by $x_1(t)$ itself that is the degree of self similarity of the signal that is characterize the degree of self similarity of the signal $x_1(t)$ for a lag of τ it becomes the auto correlation. So the auto correlation is $R_{11}(\tau)$ of τ is the degree of self similarity of $x_1(t)$ $x_1^*(t - \tau)$. What is this? This measures the degree of self similarity.

That is this is the degree of self similarity of $x_1(t)$ for a for a lag of τ s So therefore $R_{11}(t)$ from the property that we have deal shown before simply, we have simply replaced $x_1(t)$ by $x_2(t)$ by $x_1(t)$. So this is the convolution of $x_1(\tau)$ with $x_1^*(-\tau)$ that is your $R_{11}(\tau)$.

That is the auto correlation, this is the convolution operator and further if you look at the Fourier transforms that is $S_{11}(F)$ that is Fourier transform the auto correlation is Fourier transform of $X_1(\tau)$ which is $X_1(F)$ times the product in the time domain Fourier transform of $x_1^*(-\tau)$ which is nothing but as we have shown previously Fourier transform of $X_1^*(-\tau)$ is $X_1^*(F)$.

So we have $X_1(F)$ into $X_1^*(F)$, so this is magnitude $|X_1(F)|^2$. So $S_{11}(F)$ that is the Fourier transform of the auto correlation of $x_1(t)$. This is the Fourier transform and this is a very interesting property this is the Fourier transform of auto correlation of $x_1(t)$ is magnitude $|X_1(F)|^2$. That is this auto correlation which is $R_{11}(\tau)$ is magnitude $|X_1(F)|^2$ where $X_1(F)$ is the Fourier transform of $x_1(t)$, right? So $x_1(t)$ has Fourier transform $X_1(F)$, alright.

Okay, so this is important to keep in mind. So what we have shown is a very interesting property that is we have looked at the auto correlation function, alright. That is $R_{11}(\tau)$ which characterises the self similarity of the (si) signal $x_1(t)$ for a lag of τ its Fourier transform is nothing but the Fourier transform of magnitude square magnitude $|X_1(f)|^2$ where $X_1(f)$ is a Fourier transform of $x_1(t)$, alright.

So this is a very interesting property. Again we are going to rely on these properties several times that is the cross correlation between 2 signals x_1 and $x_2(t)$, the Fourier transform the cross correlation between x_1 and x_2 , alright. $X_1(f)$ and $X_2(f)$ and also the auto correlation of the signal $x_1(t)$ and the Fourier transform of the auto correlation of $x_1(t)$, alright. So these so we have looked at these concepts today studied several properties and we are going to rely on these properties in analysis of different communications (sys) systems, alright. So let us stop this module here and we will continue with other aspects, thank you.