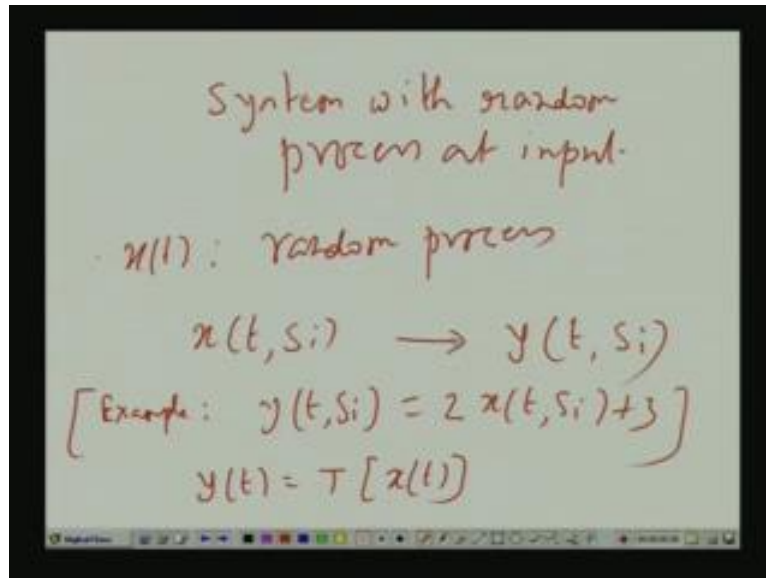


Probability and Random Variables
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Lecture - 32
System with Random Process at Input

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Today, we discuss system with random process input, suppose x_t is a random process; that means you conduct experiment and with a every time you have an experiment you know you observe 1 waveform. So, may be 1 particular waveform we denote as before like this; where s_i denotes the i 'th out come. So, in the i 'th out come you get a waveform as a function of time, so this is represented by x_t comma s_i . So, this actually a function of time, s_i just denotes the fact, s_i just stands for the fact that it is related to the i 'th out come a particular waveform. Now, suppose we have some kind of some rule, we develop or we have in which some kind of rule by which given the particular waveform we generate another waveform by a rule.

Some waveform we call it y and obviously that waveform was is associated with the i 'th outcome of the experiment. So, that which we can write it as $y_t s_i$; for example, you can have a sa very simple rule is nothing but say twice $x_t s_i$ say plus 3, which means whenever you observe a waveform say what the i 'th outcome you have got a particular waveform this is x_t comma s_i . Now, take that waveform multiplied by 2 at 3 to each at t

whatever waveform you get that is what in this case is $y(t, s_i)$. So, that means, for the same outcome s_i , we have a waveform for x and we did not have a waveform for y . So; that means, we can say that we can generate a random process $y(t)$ whose waveform for the s_i it's outcome is given by $y(t, s_i)$. In other words given, $x(t)$ we say that we generate a random process as T working on $x(t)$, where t stands for operator or that is the rule or the way of mapping they given waveform $x(t, s_i)$ we mapped it to another waveform by some rule. So, that mapping or that rule that is contained in this operator T ; So, T takes or each waveform of this random process $x(t)$ and maps it to another waveform $y(t)$. So, $y(t)$ there by you generate a random process $y(t)$; Obviously, because with each outcome we are not only getting the waveform for $x(t)$; I am also getting a waveform for $y(t)$ and therefore, you can also say that $y(t)$ is such a random process whose waveform for the s_i it's outcome is $y(t, s_i)$.

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$$y(t) = T[x(t)]$$

$$\begin{array}{ccc}
 x(t, s_i) & \left. \begin{array}{c} \lrcorner \\ \lrcorner \end{array} \right\} \text{Same} \Rightarrow & T[x(t, s_i)] \\
 x(t, s_i) & & = T[x(t, s_i)]
 \end{array}$$

\Rightarrow Deterministic Systems

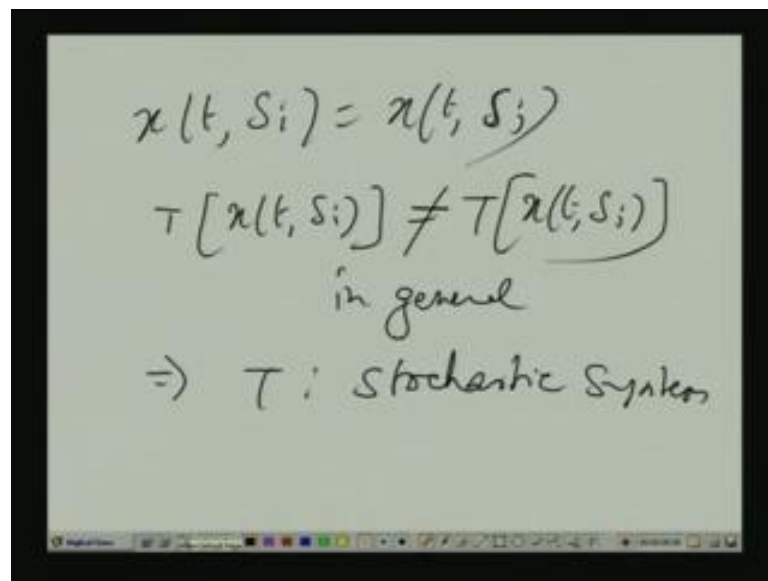
So, I repeat again we have this; so this T which is actually a rule by which you map each waveform of $x(t)$ into a waveform of $y(t)$, this rule actually in a bigger sense is called system. So, T stands for the system operator or the mapping rule, but we normally call we normally preferred to use the term system. So; that means, a system takes $x(t)$ and maps it to some waveform of $y(t)$. Now, if it, so happens that giving this waveform $x(t, s_i)$ when t works on it. If T that the system works only on this as a function of time and takes s_i as parameter, then we say it is a deterministic system. That is suppose s_i and s_j are 2 different outcomes, but the waveforms are same. If, so happens then 2 waveforms are same. In that case, if these 2 are same, implies T of $x(t, s_i)$ should

be same as $T x_t, s_j$. That is the system only you look at the time t ; it does not look at the experimental outcome, it does not look at the s values.

So, whether this time you observe a waveform or after some time for another outcome you observe the same waveform in both cases, your output will be same, because a system only looks at the function as a function of time. It does not look at the particular experimental out come with which is it associated. If that be, so then is called deterministic system, but if it, so happens that a system not only treats this as a function of time, but also looks at the particular experimental out come. That is; even if, after getting this waveform x_t comma s_i for another outcome x_j s_j you have the waveform.

Suppose if the system acts on that waveform differently that is if not all it changes its behavior from outcome. Then the system that is χ , now system is changing is property with time from outcome to outcome.

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$$x(t, s_i) = x(t, s_j)$$
$$T[x(t, s_i)] \neq T[x(t, s_j)]$$

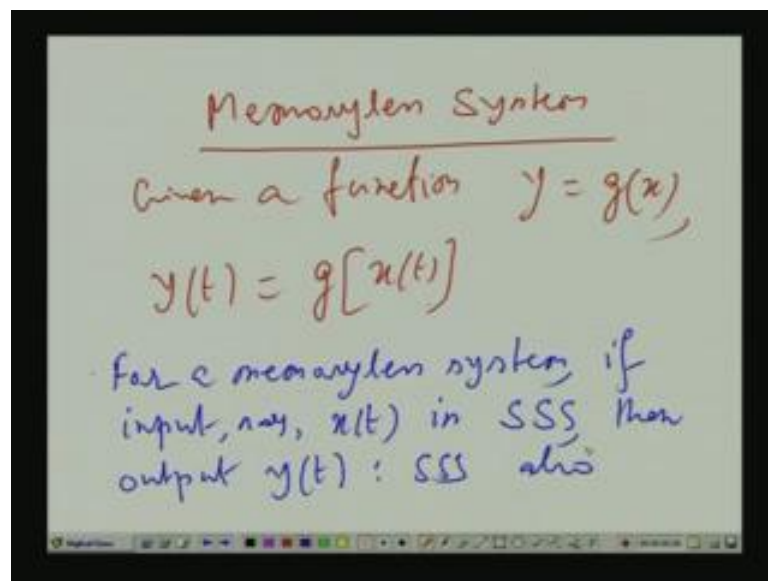
in general

$$\Rightarrow T: \text{Stochastic System}$$

So, in that case system is called stochastic system. If that is, suppose it is given that is for 2 outcomes s_i and s_j . The 2 wave forms stand out to be same, but if the system is stochastic it will not give the same output, because it is not only look it is not looking at this waveform just as function of time. It is also considering a particular out come s_o , for a are are out come s_i its behavior will be different from it is the action when the outcome is s_j . So, in general in general t of not equal to in general, this leads to T as stochastic system. That is, if T such a state whose individual parameters built in parameters.

So, internal parameters they vary randomly that is they vary they undergo change from experimental outcome to outcome; obviously, the action of the system will also change in that cases system is stochastic. So, its behavior on this $x(t)$ and $x(t)$ even if though if the waveforms are same will be different, because you know with experimental change in experimental outcome system parameters are change. So, system behavior has changed, but in this course we will be considering mostly system that is deterministic not stochastic. A particular class of a particular case of deterministic system is what is called memory less system.

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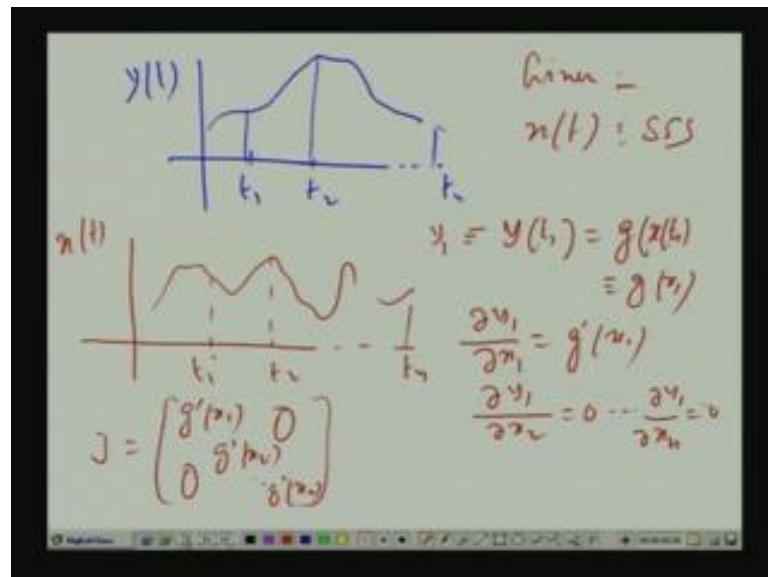


That is suppose, given a function y equal to $g(x)$, that is give any value for x we find out y . This memory less system will be given by $y(t)$ as g va of $x(t)$; That is $x(t)$ that is for each experimental outcome you got some waveform for $x(t)$. What g does at each t , it takes the value and evaluates this function g of x that is g of that value and gives you the corresponding y . So; obviously, here when g works on that value it does not look at the past or future of $x(t)$ as a result we call it memory less system. I repeat again, what it does we know this function that giving any value of x , how to work out $g(x)$ and what is y that we know. Then this system is constructed like this; that this random process $x(t)$ is the input g work on it and gives you $y(t)$.

Its meaning is take any experimental outcome say $x(t)$ that is i 'th outcome. Therefore each, so you get an waveform, in that waveform for each time t you have a value for that waveform that is amplitude or whatever g works on that s as for this given

function and gives you a value that value is assigned to $y(t)$. So, here as you see it will work on the instantaneous value of this waveform at a particular time it will not it will neither look at future not look at past. So, this will be called memory less system. A memory less system, if the system is memory less than in the input is stationary. If it a strict sense stationary SSS then output also SSS that we can see.

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Let us let me write down first, will prove it first this is $y(t)$ if I note time t_1, t_2, \dots, t_n . We will find the joint statistics joint probability density or distribution any 1 will do. For the samples of y at these values t_1, t_2 , and then if that joint density does not undergo any change if a constant shift is giving to all these time points that is instead of t_1 I go $t_1 + c, t_2 + c, t_n + c$. If you still get the same joint probably statistics then by the definition its strict sense stationary. We will check suppose it is giving that input is strict sense stationary, that is $x(t)$ if you take a waveform given $x(t)$ given is SSS. Now, you remember we did some study on this.

You know joint statistics of a group of random variables and that is given joint density for x that is $x(t_1), x(t_2), \dots, x(t_n)$. If I take those variables and if, I take those random variables take that joint density, how to obtain the joint density for y . That is how to obtain the joint density for $y(t_1), y(t_2)$ and $y(t_n), \dots, y(t_n)$. So, there is a formula that involved the joint density for x that is $x(t_1), x(t_2)$ up to $x(t_n)$ and the Jacobean what also Jacobean. Jacobean was in this that Jacobean you remember in this case for the Jacobean will turn out to be this excuse me is this. First this $y(t_1)$ it is a function of just $x(t_1)$.

It is not a function of the other variable x_2, x_3, \dots, x_n , that is y_1 ; obviously, $\frac{\partial y_1}{\partial x_1}$ is $g'(x_1)$; obviously $\frac{\partial y_1}{\partial x_2}$ is 0 dot dot dot $\frac{\partial y_1}{\partial x_n}$ is 0. What is x_2 , x_2 is x_2 ; what is x_n that is x_n it is a it is a short notation. Similarly, when you consider y_2 and call it y_2 then $\frac{\partial y_2}{\partial x_1}$ that is 0. Because y_2 , that is y_2 is just g of x_2 that is g of x_2 . So, we can under we will see the Jacobean becomes a diagonal matrix, $g'(x_1)$ $g'(x_2)$ dot dot dot $g'(x_n)$ and 0 here 0 here and that expression involved the determinant of this matrix. So, determinant is obviously, product of this mod of determinants, so mod of the product of the diagonal entries.

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$$p_y(y_1, y_2, \dots, y_n; t_1, t_2, \dots, t_n) = \frac{p_x(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n)}{|g'(x_1) g'(x_2) \dots g'(x_n)|}$$

So, if you do that what happens to the joint density that is; this y stands for the fact that you know, I am looking at the joint density for the y variables at say y_1, y_2, \dots, y_n at what values t_1, t_2, \dots, t_n that will be P_x by that formula divided by the Jacobean expression and Jacobean you have seen it will be mod of $g'(x_1) g'(x_2)$ dot dot dot $g'(x_n)$. Now, we are we are giving the fact that, input is strict sense stationary

So, you see is it true for output? For that if I give a constant shift if t_1 is changed to say $t_1 + c$, t_2 is changed to say $t_2 + c$, t_n is changed to say $t_n + c$.

Then here also, by this formula t_1 should be changed to $t_1 + c$ t_2 should be changed to $t_2 + c$ t_n should be changed to $t_n + c$, but since input is strict sense stationary this will not produce change. I will get the because the shift is constant, because of strict sense stationary, I will get the same expression. So, it does not make any change here and we see denominator this is independent of time. So, it does not get changed by this shift in t .

So; that means, there is no change in the lhs expression, so whether you give keep t_1 t_2 up to t_n or give each of the values a constant shift by c which; obviously, means that output was a strict sense stationary and if strict sense stationary it is WSS also. We now, consider a particular case of systems called linear system, which is not necessary memory less system.

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The image shows a whiteboard with the following handwritten text:

Linear Systems

$$y(t) = L[x(t)]$$

$$L[x(t, s_i)] = y(t, s_i)$$

$$L[x(t, s_j)] = y(t, s_j)$$

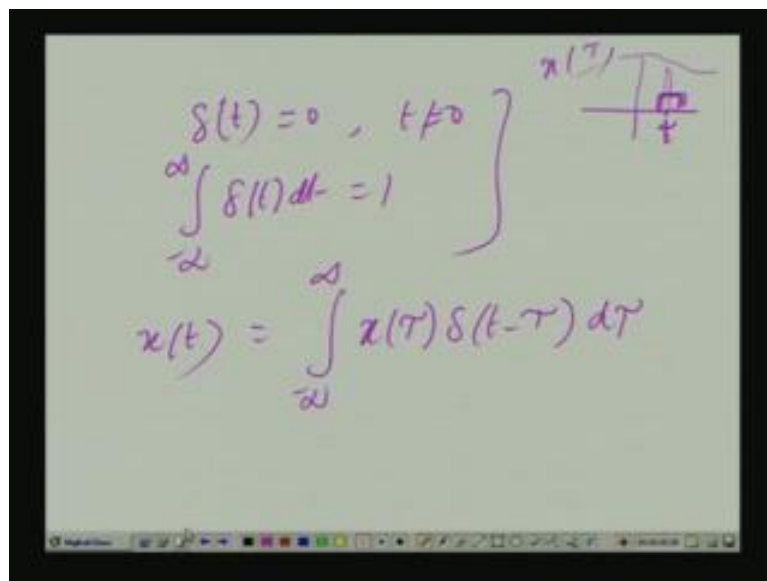
$$L[c_1 x(t, s_i) + c_2 x(t, s_j)] = c_1 y(t, s_i) + c_2 y(t, s_j)$$

This linear system that is another important system linear system here, we use the notation L to denote for that stand for that operator system. Because as L stands for linearity system is linear if suppose you give 1 waveform for a particular outcome x . Suppose $x(t, s_i)$ L of that is some waveform $y(t, s_i)$, again $x(t, s_j)$ another waveform $y(t, s_j)$; now suppose you have an experimental outcome where the waveform is actually a linear combination of these 2 waveforms. That is suppose, you have got a situation where $x(t, s_k)$ is nothing but, linear combination of these waveforms. Whatever you had for this c_1 times that plus c_2 times $x(t, s_j)$.

Now, if it is linear then L working on this guy, it would generate; obviously, $y(t, s_k)$ the waveform for y for the same outcome x_k . If it is linear, if these become same as the response c_1 time, the response due to this first waveform that is $x(t, s_i)$ plus c_2 times the response due to the second waveform $x(t, s_j)$. That is, if you it is transferred to be c_1 times; $y(t, s_i)$ this particular waveform plus c_2 times $x(t, s_j)$; which means principle of super position should be valued. That is if you have 1 waveform $x(t, s_i)$ you $y(t, s_i)$. Similarly, another waveform $x(t, s_j)$ get $y(t, s_j)$.

comma sj; now, if you linearly combine them as $c_1 x(t) + c_2 x(t)$ waveforms are linearly combined.

The corresponding response should be simply c_1 times, the response due to $x(t)$ that is $y(t) + c_2$ times, the response due to $x(t)$; that is, c_2 times $y(t)$. So, the principle of superposition should be valued. If that be, so then it is called linear system. A particular class of linear system is linear time invariant system, I will come to that later, but before that let us do some let us review recall something from system theory that is how, to describe a continuous a continuous time system. Now, you all are familiar with this thing this direct delta function or unit impulse function delta (Refer Slide Time: 20:19)



What is delta? Delta t such that is 0, if t not equal to 0 and integral of delta $t dt$ all of this is direct delta function or unit impulse function. You all know this, that we start with box function whose area is 1 start compressing the box, but keeping the area 1. So, the height increases final in the limiting case it becomes that impales area remains same as 1, but since outside originate is 0; at originate putting up to infinity that kind of function. Then since it is not a class on system theory, I cannot go into those details, but any waveform any functions say $x(t)$.

For the time being for convenience I am dropping this second parameter that is earlier I was writing $x(t, s_i)$ or $x(t, s_j)$. For time been, I will just write it has $x(t)$ meaning waveform that is a function of t , but actually since there are coming from random process, I should also mention s_i or s_j indicate the particular outcome with which 0 waveform is associated, but for the timing just for covariance I am dropping that a

notation, I mean dropping that symbol. So, now given any $x(t)$; this there in systems theory you can verify also you can write it as this.

You obviously, get $x(t)$, because again this Δt as far as this $\Delta t - \tau$ $\Delta t - \tau$ is a impulse shifted at τ . That is you are doing an approximation of the impulse first by taking a new box of any area 1, you have shifted into a τ earlier it was it sent at origin and now you start compressing it here. And finally, it in the limiting it becomes an impulse at τ . Now, what is happening is you are not only taking the box multiplying the box by a value $x(\tau)$. So, area is not 1, but $x(\tau)$ and now, start from compressing it. So, area will still given $x(\tau)$, actually we are multiplying with your multiplying Δt first your approximating $\Delta t - \tau$ by x to box of area 1 and there is a function $x(\tau)$, this x is τ you are multiplying it by τ .

You have shifted in to t actually, because this all these are function of τ t is constant t is your choice from outside. So, t is fixed, so there is a box approximate in the Δt there is function of τ . It is center is t remembers $\Delta t - \tau$ or $\Delta \tau - t$ they are same. Δt such a function whether t you replace by $t - t$ or keep t it does not make any change. So, actually $\Delta t - \tau$ means $\Delta \tau$ shifted by t . So, centre is at t ; you are multiplying it by $x(\tau)$, when you multiply the 2; these are only in this area only where the box is located you get a multiplication otherwise everywhere else is this function $x(\tau)$ is multiply by 0.

Now, you start compressing the box keeping the area 1. So, in the limiting case what happens is this, you get a value $x(t)$, because if the box is sufficiently narrow, then what the area is of box not just 1, but multiplied by the value of the function. Now, if the box is sufficiently the narrow that function its infinitely small width then that function as actually approximately value equal to extreme that is the value at the centre point that is the $x(t)$ time in the box area that is $x(t)$. You still compress it further, so area will be you will get finally, and you're integrating. So, you get actually that is $x(t)$ details have given in system stage book I just this is just to kind of recap, so that you know we do not get really confused.

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$$\begin{aligned} \delta(t) &\xrightarrow{L} h(t) = L[\delta(t)] \\ \delta(t-\tau) &\xrightarrow{L} h(t, \tau) = L[\delta(t-\tau)] \\ y(t) &= L[x(t)] = L\left[\int_{-\infty}^{\infty} x(\tau) \delta(t-\tau) d\tau\right] \\ &= \int_{-\infty}^{\infty} L[x(\tau) \delta(t-\tau)] d\tau \\ &= \int_{-\infty}^{\infty} x(\tau) h(t, \tau) d\tau \end{aligned}$$

Now, if such a system is pass through a linear system what happens; in general delta t that is an impulse at origin we denoted it by this delta t. If delta t is passed through a linear system output is called the impulse response h_t as L of delta t. If instead of delta t, I give delta t minus tau this is more general case. So, in the general case the output here will depend on what, the 2 indices if the function of t I agree, but it also depend on that shift parameter tau by which that delta has been shifted. If the system is shift invariant or time invariant then it should only depend, it should not depend on it should only depend; I mean the waveform that you get here will be same as h_t just h_t will be delayed by same amount tau.

That is without any shift or without any delay in delta t you got h_t ; now, you delay the delta t by some tau. So, corresponding response also should be h_t minus tau. That is, its previous response delayed by the same amount, but in general, if that happens when the system is time invariant. That is, there was a gap of tau in giving the input delta t that is why, I am giving now, input delta t minus tau within that time period tau the system has not change its property system is time invariant and therefore, the response now also will be in nature same as the previous response h_t , but if you plot it on the same scale It is now again delayed by the same amount tau.

So, on the same scale if you plot it, it will be not h_t , but h_t minus tau that happens in the time in variant case, but in the more general case, system may change its property within the period tau. So, output for delta t minus tau will not only will depend on 2 things

It will be a function of t , but it will also be a function of τ , if instead of τ you delay the impulse further by say τ' . The corresponding impulse response might be different, because the system is undergoing change in property with time. That is why, in the general case, it is a function of 2 things $h(t, \tau)$.

Only when it is time invariant it is not $h(t, \tau)$, but it is just $t - \tau$. That is the same $h(t)$ delayed by the same amount τ , because input is delayed by same amount, but in general if it is not system time invariant that is if property also changes with time. Then impulse as a function of time should also depend on τ , how much this parameter τ by which the input is shifted. If it is shifted further maybe I will get some other kind of waveform, not only just the delayed version of $h(t)$, but even the nature will change. So, let us take this; so in this case $y(t)$, we have already expressed $x(t)$ in terms of like this in previous slide and L now, it is a superposition, because of integral and system is linear, integral only means summation, but if the system is linear superposition principle works.

So, if you apply the system after the summation you should get the same thing; if you apply the system before that is L over this integral is same as integral over L working on these inputs. Remember system response over a sum of inputs is same as sum of the individual input responses. That is, because of linearity you should have L working on this integral is only a summation that is individual responses summed. Earlier I was taking the response of $y(t)$ by first summing the various inputs by the integral then applying to the input, because of linearity that would be same as first applying individual input $x(t - \tau)$ getting the corresponding response and then summing the individual responses by an integral and remember here t is function of time on Δx is kind of amplitude.

So, L will work only on this, so this is $x(t - \tau)$ and I know L working on $t - \tau$, as I said earlier it will be function of 2 variables in the general case $h(t, \tau)$ remember this result. Now, using this result we will prove some other results.

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$$\begin{aligned}y(t) &= \int_{-\infty}^{\infty} x(\tau) h(t, \tau) d\tau \\ \mu_y(t) &= E[y(t)] = \int_{-\infty}^{\infty} E[x(\tau)] h(t, \tau) d\tau \\ &= \int_{-\infty}^{\infty} \mu_x(\tau) h(t, \tau) d\tau \\ &= L[\mu_x(t)]\end{aligned}$$

So, I write again first if I take the mean of this $\mu_y(t)$ at particular time t that is E of $y(t)$ that is E of the right hand side, but right hand side is a summation and remember E was, E that expectation operator is linear, that is E expected value of sum of random variable is same as some of the expected value of the individual response random variable. Only thing is here you have got a continuous sum in the form of integral, but still the linearity property of E operator prevails. So; that means, this is same as you can push E inside the integral and h is not random.

So, E will work only on x that is $\mu_x(\tau) h(t, \tau) d\tau$ and that is same as L working on this function. Remember when L was working on any function $x(\tau) h(t, \tau) d\tau$, but instead of $x(\tau)$ now other function $\mu_x(\tau)$. That is why $\mu_x(\tau) h(t, \tau) d\tau$; so as though the mean function that is if you take that mean or expected value of x you again get a waveform. It is not a random waveform, but waveform if you pass it through the system whatever output you get that is what you get, for the mean waveform of the output this is number 1 result.

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Special case: LTI

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

$$= \int_{-\infty}^{\infty} x(t-\tau') h(\tau') d\tau'$$

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

In the special case, when it is LTI linear and time invariant LTI. In that case, we know that $y(t)$ is not t comma τ , but it is actually really a delayed percent by τ . If giving delta as input you got $h(t)$ as impulses response now, delta d shifted by τ . So, $h(t)$ also will be shifted by the same amount τ . That happens where when it is not only linear it is linear an time invariant that a special case. And they if you replace t by minus τ y τ prime So, τ is t minus τ prime and then you see $d \tau$ is minus of $d \tau$ prime and when τ is minus infinity τ prime is plus infinity. So, plus infinity here minus infinity here $d \tau$ means minus $d \tau$ prime prior minus has come you reverses the 2 limits again.

So, minus infinity another minus comes. So, it becomes plus and you get back $x \tau$ prime h you get back this $x(t-\tau')$ $h(\tau')$ $d \tau'$. I repeat again if you replace t minus τ by τ prime. So, $d \tau$ prime is minus $d \tau$, so $d \tau$ is minus $d \tau$ primes. So, minus is coming you look at the limits τ becoming minus infinity means; τ prime plus infinity and here, minus infinity. So, plus infinity to minus infinity Reverse the limits minus infinity to plus infinity. So, another minus sign, so 2 minuses cancel each other and you get back this relation. And no point in carrying out τ prime you can again bring back your good old τ .

In such case $y(t)$ is what and this is call convolution between 2 sequences $x(t)$ and $h(t)$ this is the formula. Now, in this case what is $y(t)$, $y(t)$ is apply E put the E inside E works on $h(t-\tau)$ that is $y(t) = x(t-\tau) * h(\tau)$. That is again

convolution between these 2 functions $x(t)$ and $h(t)$ this a formula then let us find out this that.

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The image shows a handwritten derivation on a whiteboard. It starts with the convolution integral: $x(t) \xrightarrow{L} y(t) = \int_{-\infty}^{\infty} x(\tau) h(t, \tau) d\tau$. Below this, it calculates the expected value of the output: $R_{xy}(t_1, t_2) = E[x(t_1)y(t_2)]$. This is then expanded to $E\left[\int_{-\infty}^{\infty} x(t_1) x(\tau) h(t_1, \tau) d\tau\right]$. The expected value is moved inside the integral, resulting in $\int_{-\infty}^{\infty} R_{xx}(t_1, \tau) h(t_1, \tau) d\tau$. Finally, it is noted that this is the Laplace transform of $R_{xx}(t_1, \tau)$: $= \mathcal{L}\{R_{xx}(t_1, \tau)\}$.

There is system input is $x(t)$ there is system linear system not this a time invariance output is $y(t)$. We know $y(t)$ equal to remember only when it was time invariant it was it becomes a function of t minus τ and there I could replace t minus τ by τ prime and all that. I could write it in the reverse way, that instead of writing $x(\tau) h(t, \tau)$ could also write $h(\tau, t) x(t)$, but when this is not time invariant, I cannot proceed any further from here. So, I have to stick to this to most general I still keep the linearity only I do not bring in time invariance unless a necessary.

So, for the linear case; suppose, you want to find out 1 function on cross correlation function R_{xy} say t_1 t_2 that is expected value of $x(t_1) y(t_2)$, what is it? Now, you t_2 you know, you put in the formula that is $x(\tau) h(t_2, \tau)$ $d\tau$. And $x(t_1)$ multiplying that; so we bring that $x(t_1)$ within the integral. So, $x(t_1)$ brought within the integral $y(t_2)$ for that $x(\tau)$. Now, expected value can be brought inside, because expected values of summation are summation of the expected values h is deterministic. So, E will work only on this $x(t_1)$ into $x(\tau)$. So, what you get is $R_{xx}(t_1, \tau) h(t_2, \tau) d\tau$.

So; that means, what is it, that if you take it as a function of $R_{xx}(t_1, \tau)$ you just forget t_1 for the time being take it as a function of τ then isn't it. That it is the same

thing. If I will get, if you give the function of tau, forget t1 is 1 function of tau to the system that is input. Now, output; obvious, will be this function multiplied by ht coma tau d tau that is why I am getting here just to indicate that here L means, the integral will be with respect to tau, I denote tau here, I put tau here; that means, you take an integral multiply Rxx t comma tau into ht coma tau integral with respect to tau. Remember if the system works on any function on any quantity.

So, function of tau here that is why this tau is indicated L tau tau is a function. then what does is same do you multiply this by ht coma tau, integrate with respect to tau. So, that will give you the output at t. Here it is t 1, I will it will be better if I general if I simplified further for the LTI case that will be better, because actually this is this should be y x t1 y t2 this will t2, because this should be t2 . So, when I say L tau means you give this input or Rxx t1 in case my handwriting is not perfectly legible, you give this function Rxx t1 comma tau as a function of tau to the input to the system input were system evaluates the output at a time t2.

So; that means, this functions should be multiplied by h t2 comma tau integral with a tau that will give you the output at t2, that output will be function of t2 and also this t1 will carry, because t1 is given as a parameter from the beginning. So, it will be a function of t 1 comma t. In the special case, where it is LTI its time invariant also let us see what happens.

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Special case \rightarrow LTI

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

$$R_{xy}(t_1, t_2) = E[x(t_1) y(t_2)]$$

$$= \int_{-\infty}^{\infty} E[x(t_1) x(t_2 - \tau)] h(\tau) d\tau$$

$$= \int_{-\infty}^{\infty} R_{xx}(t_1, t_2 - \tau) h(\tau) d\tau$$

Special case LTI in such case, I know that $y(t)$ is, so what is $R_{xy}(t_1, t_2)$ that is E of $x(t_1)y(t_2)$. So, for $y(t_2)$ put t equal to t_2 ; here, multiply I mean push this parameter $x(t_1)$ from outside the integral t_2 inside integral and then bring in push this E operator inside the integral again. So, what you get there is this E working on $x(t_1)$, that has come from outside the integral then x say its $y(t_2)$ it should be $x(t_2 - \tau)$ h remains separately, because it is not sophisticated and what is this function this is R_{xx} .

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$$\Rightarrow L [R_{xy}(t_1, t_2)]$$

$$R_{xy}(t_1, t_2) = E [x(t_1) y(t_2)]$$

$$= \int_{-\infty}^{\infty} E [x(t_1) x(t_2 - \tau)] h(\tau) d\tau$$

$$= \int_{-\infty}^{\infty} R_{xx}(t_1, t_2 - \tau) h(\tau) d\tau$$

t_1 t_2 minus τ that means, we can write now this is equal to $L R_{xx}(t_2)$ and I write t_2 here. What does it mean this will be treated as a function of t_2 only; if the just to indicate that I put t_2 here. If that will be this function is given as a function of t_2 as a input to the system. So, what is the output and the LTI case. Simply, this will be replaced by t , I mean here instead of t_2 ; I will have t_2 minus τ that is what is happening here multiplied by $h(\tau)$ integral. These are cross correlation between input and output and what is the auto correlation in output.

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$$\begin{aligned}
 R_{yy}(t_1, t_2) &= E[y(t_1)y(t_2)] \\
 &= E\left[\int x(\tau_1)h(t_1, \tau_1)d\tau_1 \int x(\tau_2)h(t_2, \tau_2)d\tau_2\right] \\
 &= \iint R_{xx}(\tau_1, \tau_2)h(t_1, \tau_1)h(t_2, \tau_2)d\tau_1d\tau_2
 \end{aligned}$$

That is what is your R_{yy} t_1, t_2 ; this more interesting here simply replace by those expressions y t_1 was in the again in the just linear case not necessary time invariant. So, y t_1 for that we will have x τ_1 and then here x now there variable τ_2 ; take the E inside the integral and on x into x others are not stochastics. So, you get R_{xx} here it will be better if I take the time invariant case.

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$$\begin{aligned}
 &L_{t_2}\left[L_{t_1}\left[R_{xx}(t_1, t_2)\right]\right] \\
 R_{yy}(t_1, t_2) &= \iint E\left[x(t_1 - \tau_1)x(t_2 - \tau_2)\right]h(\tau_1)h(\tau_2)d\tau_1d\tau_2 \\
 &= \iint R_{xx}(t_1 - \tau_1, t_2 - \tau_2)h(\tau_1)h(\tau_2)d\tau_1d\tau_2 \\
 &\quad (t_1 - \tau_1, t_2 - \tau_2)
 \end{aligned}$$

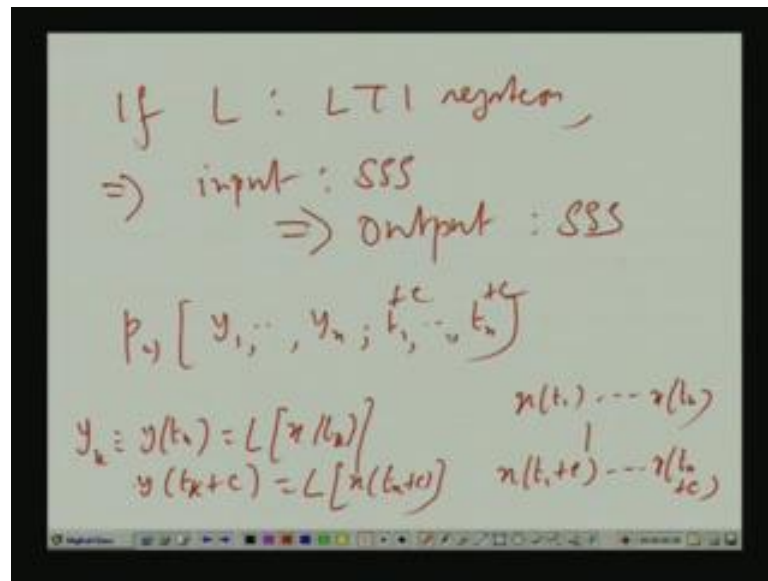
If it is time invariant there instead of Special case we consider now, LTI here. So, R_{yy} t_1 t_2 will be what E now, I am doing first R_y means E y t_1 into y t_2 . So, E of x t_1 minus τ_1 x t_2 minus τ_2 h τ_1 h τ_2 ; that is, R_{xx} what is the physical interpretation of

this very interesting this means what. If we just consider this inner integral $R_{xx}(t_1 - \tau_1, t_2 - \tau_2)$ multiplied by say $h(t_1 - \tau_1)$. So, that times it is the function of just t_1 . That means, first you are giving this guy suppose I do this; first $t_1 - \tau_1$ and then I will let t_1 first what does it mean; it will give rise to an output, what output $R_{xx}(t_1 - \tau_1, t_2 - \tau_2) h(t_1 - \tau_1) dt_1$ integral.

Apply here, so again it is it was a function of t_2 . So, now instead of t_2 ; it will become $t_2 - \tau_2$ and $h(t_2 - \tau_2)$ will come again another integral. So; that means, this input autocorrelation function is pass through the linear system, where it first exceed as a function of t_1 . Then the corresponding output which is the function of t_2 is again pass through the same system this time it takes this intermediate output as a function of t_2 . A particular thing you see if input is the wide sense stationary; If x is WSS then; obviously, $R_{xx}(t_1 - \tau_1, t_2 - \tau_2)$ it will be the function of just $t_1 - \tau_1$ plus $t_2 - \tau_2$; τ_1 and τ_2 their integration variables. So, once you carry out the integration there will be no trace of τ_1 and τ_2 there will be.

So, entire thing will be function of just $t_1 - t_2$; so that means, $R_{yy}(t_1, t_2)$ will be just function of $t_1 - t_2$. So, it is not individual t_1 or t_2 will matter, but the gap which means output also will be WSS. I mean in term of correlation in term of mean also it will be WS we have seen earlier. How, because you have seen that if input is a if input mean function you take $\mu_x(t)$ then output is $\mu_x(t)$ convolute with $h(t)$. So, if $\mu_x(t)$, because of stationarity it is a constant thing. Then the convolution will also yield a constant function 1 thing does not need any proof, we just make a statement we make a statement that if a system is linear.

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If L denotes an LTI system then input SSS means, output also SSS. Now, start with difficult output say you are tried to find out this y_1 that is y of t_1 dot dot dot say y_n , t_1 dot dot dot t_n . What is y_1 ? That y at t_1 . What is y at t_1 ? That is L of x t_1 . Remember I just say y_k is what is equivalent to y of t_k that is equal to L of x t_k ; suppose, you have somehow value of this joint density. Now, instead of t_1 or t_n suppose I give a shift plus c plus c ; that means, that time what will be y_k it will L of x t_k plus c . Now, you remember you understand that input this; if in the earlier case when there was no such c added the joint density will become actually function of the input statistics, because after all y_1 y_2 y_n they are coming from input; y_1 is L of t_1 L of x t_1 y_2 is L of x t_2 dot dot dot.

So; obviously, the joint density will depend on input statistics that is input joint density. Now, if I want to find out the same joint density not a t_1 , but t_1 plus c t_n plus 2 plus c t_n plus c . It should depend on input device statistics for what samples x of t_1 plus c x of 2 plus c dot dot dot x of t_n plus c , but you see input is stationary strict sense stationary. So, whether you have the variables x of t_1 dot dot dot up to x of t_n or you have got statistics wouldn't change, because input is given to strict sense stationary. What the LTI thing comes, because we said that, if y_1 is L of x t_1 that is a y of t_1 y_1 is y of t_1 . If y of t_1 is L of x t_1 , then because of time in variance y of t_1 plus c is L of nothing but x of t_1 plus c . That is, y of then t_k plus c , because of time invariance it is...

So, when you have $t_1 + c, \dots, t_n + c$ the corresponding joint density again will depend on the joint density of x of $t_1 + c, \dots, x$ of $t_n + c$, but, because of strict sense stationary of the input, that will that joint density will be that statistics will not be different from the earlier 1. When this constant shift c was not applied; obviously in the output case also 2 density will be same and you get strict sense stationary. As an example of this linearity you consider a differentiator.

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$$\begin{aligned}
 L[x(t)] &= x'(t) = y(t) \\
 \mu_y(t) &= \mu_{x'}(t) = L[\mu_x(t)] \\
 &= \mu_x'(t) \\
 R_{xy}(t_1, t_2) &= R_{x'x'}(t_1, t_2) \\
 &= L_{t_2}[R_{xx}(t_1, t_2)] = \frac{\partial R_{xx}(t_1, t_2)}{\partial t_2}
 \end{aligned}$$

A system which takes $x(t)$ and gives you first order derivative; it is obviously linear, because we know the derivation is linear derivation is a linear operator. So, what is $x'(t)$ actually is stands for $y(t)$ here; so what is $\mu_y(t)$ also you can denote as $\mu_{x'}(t)$. That is by our definition that will be like L working on $\mu_x(t)$ and L working on this means what you will have to differentiate it with respect to t . Then $R_{xy}(t_1, t_2)$ which is also you can write as $R_{x'x'}(t_1, t_2)$; that we have seen L work on t_2 ; we have seen earlier of what this input auto correlation function. Let us take it as a function of t_2 , and then let the system work on it, but system work on it means then it will be differentiate it again with respect to what t_2 . So, it will be $\frac{\partial}{\partial t_2}$ and just a couple of minutes and I will go ending up.

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$$\begin{aligned}
 R_{yy}(t_1, t_2) &\equiv R_{x'x'}(t_1, t_2) \\
 &= \frac{\partial^2 R_{xx}(t_1, t_2)}{\partial t_1 \partial t_2} \\
 &= \mu_{xx}'(t) \\
 R_{yy}(t_1, t_2) &= R_{x'x'}(t_1, t_2) \\
 &= L_{t_2} [R_{xx}(t_1, t_2)] = \frac{\partial R_{xx}(t_1, t_2)}{\partial t_2}
 \end{aligned}$$

The output R_{yy} which you can also denote as, that will be what; obviously, you have to apply this R_{xx} t_1, t_2 first with respect to say t_1 , I mean you to say R_{xx} t_1, t_2 will it pass into the system passed. As function of say t_2 then output will you past through the system, but that time is function of t_1 or vice versa that you have seen. So, if you do that that will be; obviously give rise to the double derivative first with respect to t_1 then with respect to t_2 or it does not matter it can be first ∂t and ∂t_1 and then $\partial \partial t_2$ or vice versa.

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$$\begin{aligned}
 R_{yy}(t_1, t_2) &\equiv R_{x'x'}(t_1, t_2) \\
 &= \frac{\partial^2 R_{xx}(t_1, t_2)}{\partial t_1 \partial t_2} \\
 x(t): \text{ WSS} & \quad \mu_{xx}(t) = \text{const} \\
 & \quad \mu_{xx}'(t) = 0 \\
 R_{xx}(t_1, t_2) &= R_{xx}(t_1 - t_2) \\
 R_{xx}'(\tau) &= -R_{xx}'(\tau); \quad R_{x'x'}(\tau) = -R_{xx}''(\tau)
 \end{aligned}$$

Suppose $x(t)$ WSS; that means, μ_{xx} is constant then μ_{xx}' which is derivative of this is 0. So, again in stationary mean wise you can easily have it verified here, that is for

the output also μ is constant, then how about auto correlation that is how about this $R_{yy}(t_1, t_2)$. Now, here we have got $R_{xx}(t_1, t_2)$ is a function of actually $R_{xx}(t_1 - t_2; t_1 - t_2)$ equal to say τ , so its function of τ . So, here, but I know its 1 minute here, if you take this derivative $\frac{d}{d\tau} R_{xx}(t_1 - t_2; t_1 - t_2)$ is τ we differentiate with respect to t_2 . First and t_1 first I limit as an exercise to you; you can easily verify these 2 things. $R_{xx}(\tau)$, $R_{xx}'(\tau)$ you can easily verify will be $R_{xx}(\tau)$ first order derivative and $R_{xx}''(\tau)$ will be second order derivative. So, that is all for today.

Thank you very much.