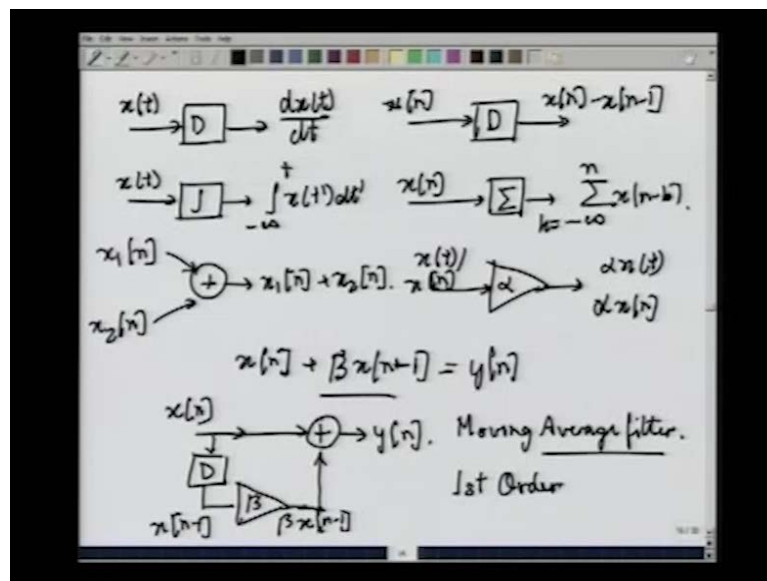


Signals and System
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Lecture - 22
Filters

We now, know that we have a means of differencing a discrete sequence, we have a means of carrying out a running form of a discrete sequence, we can differentiate a continuous time function, we can running integrate a continuous time function, we can do all these things. Suppose, without regard or prejudice to the manner in which these circuits are constructed which can do these calculations. Suppose, we nearly represent them by a block.

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Suppose, we say that we can have a system, which takes a continuous time signal $x(t)$ and yields at the output $\frac{dx(t)}{dt}$ or we add a similar discrete system, which took in $x[n]$ as the input and generated at the output is first difference $x[n] - x[n-1]$.

We would represent both these by D and D the context would make it very clear, whether we are concerned with a discrete system or a continuous time system. Similarly, we could have a running integrator or a running summer $x(t)$ goes inside and what you get at the output is integral minus infinity to t of $x(t)$ or $\sum_{k=-\infty}^n x[k]$, fine here you would have this. We would denote by this symbol and this we would denote by this symbol over

here, it would be the running summer $x[n]$ would yield summation minus infinity k equals minus infinity to n $x[k]$ minus k .

Suppose we have these blocks let us construct some more blocks, let us have a summing block, this block takes two inputs let us say $x_1[n]$ or $x_1[t]$ it does not matter, the symbol assume is the same $x_2[n]$ and the output it produces is $x_1[n]$ plus $x_2[n]$ this is what we would call a summing block.

We could also have a special block for a system, which nearly multiplies by a scalar constant we call that an amplifier in conventional electrical engineering. So, we could have an amplifier, which takes a sequence $x[t]$ or $x[n]$ as the input this depending upon whether it is discrete or continuous. And produces us the output let us say, if this has a gain of α you would get $\alpha x[t]$ or $\alpha x[n]$ as the output.

So, these are the four kinds of blocks that will constitute the objects of our next discussion. Our next discussion will be to actually build systems, which can solve differential equations or difference equations. Suppose we want to solve the difference equation that we have been playing around with till recently suppose, we have let us say even simpler system. Suppose we just had $x[n]$ plus $\beta x[n-1]$ equals $y[n]$ can be build an interconnection of these elementary blocks that we have over here to get $y[n]$, if we have the sequence $x[n]$ supply to us, that is not very hard.

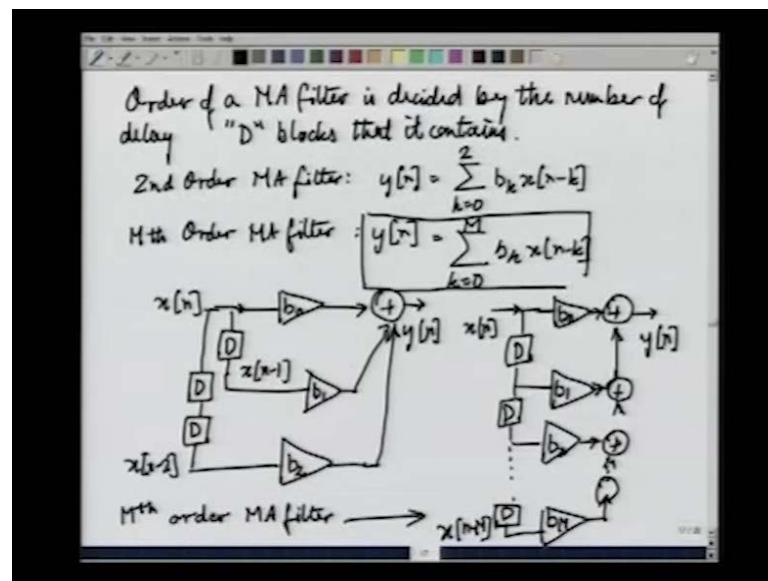
Let $x[n]$ be available over here we need to add $x[n]$ to something, namely to this term let us do that by constructing a summing block. Now, what is the other thing that we need to add to it is β times $x[n-1]$, but we do not have $x[n-1]$ available right now, we will have to make it by taking $x[n]$ out of here putting it in a D block. So, that what you get here is $x[n-1]$. Now, this is only $x[n-1]$ we want $\beta x[n-1]$. So, what we do is to append another amplifier block with a gain of β and this now is $\beta x[n-1]$.

So, you have $x[n]$ coming from this side $\beta x[n-1]$ coming from this side and we now, add the 2 and what you get here, would actually the $y[n]$. This is a very simple kind of filter, which just takes the present value of $x[n]$ and a certain scaled value of the previous $x[n]$ that is $x[n-1]$ and adds them to yield the current value of $y[n]$ in short $y[n]$ is said to be the moving average of $x[n]$.

This is called a moving average filter, it essentially creates a gradually moving average at every instant of time, and the window over which the input sequence is being averaged is being shifted by 1 position. So, you get this moving average filter. More precisely this is called a first order moving average filter, if you want to know why the fact is that the order of a moving average filter, which we will simply call an MA filter, is decided by the number of delays or D blocks that it contains. In this case it contains only one. So, it is a first order filter.

If you wish to generalize this to a second order filter all you would have is an additional D block and therefore, a second delayed version of $x[n]$ namely $x[n-2]$, which would also be scaled by some factor and would get added to $x[n]$ and a scaled version of $x[n-1]$ and so on to get $y[n]$.

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So, let us write the equation for this I will not draw the block diagram of this instead I will go ahead and define the n th order moving average filters or the M th order moving order filter directly. If you wish to see how this is implemented well, there is a naïve way of implementing it and the other slightly smarter way of implementing it. You will apply the naïve approach for a second order implementation that is to say for a second order filter M equal to 2 and then we are going to see how we can make it slightly smarter.

We have $x[n]$ as before and we need both $x[n-1]$ and $x[n-2]$. So, 1 way would be to take $x[n-1]$ or to take $x[n]$ and delay it by 1 unit of time to get $x[n-1]$ this x

$x[n-1]$ would then be scaled by b_1 . $x[n]$ itself would be scaled by b_0 according to the equation and you also need $x[n-2]$. So, you could start with $x[n]$ right away put 1 delay over here and yet another delay over here. And get $x[n-2]$ to which you would attach the appropriate scale factor b_2 . Now, that you have all these you could take a summing block over here add this, add this and add this and all this together would give you $y[n]$.

Now, this kind of a diagram is right for just a few very very straight forward simplifications. First of all let us say, that we want to keep the number of blocks of whatever kind to the minimum be the delay blocks or gain blocks such as b_0 , b_1 , b_2 or the summing blocks that we have over here. In the case of the summing block, which is the circle with the plus inside, we want to ensure that no summing block requires is at any time involved with summing more than 2 numbers, 2 quantities, 2 functions whatever.

So, keeping these constraints in mind, we can make a simplified version of this diagram which does; however, exactly the same job and that is as follows. Now, we have the 3 delayed and appropriately scaled versions of $x[n]$. $x[n]$ times b_0 , $x[n-1]$ times b_1 and $x[n-2]$ times b_2 with this. We now, add these 2 at a time first we add $x[n-2]$ b_2 and $x[n-1]$ b_1 to get this, to this sum in turn, we add $b_0 x[n]$ to get this and this of course, is now $y[n]$.

The second block is much more regular in appearance has 1 delay block less and has 2 summing blocks instead of one, but each summing block have only 2 inputs and 1 output. So, this is the more standardized form of a second order moving average filter which could easily be generalized in the manner that I am about to proceed to do.

All you do to generalize this further is to have n delays instead of just two. So, that you go on like this here of course, you will have b_2 and you will add this through another summing block you go down like this, and you have the last delay and the output of this delay would be $x[n-M]$. This would go to b_M and would go to a summing block which attach which adds it to the previous 1 this now, is an M th order moving average filter.

Now, you see moving average filters, while they are nice to look at are not really related to our differential equations. In short they are a different class together and their general

form is summarized by the expression that we had earlier over here in this page namely, this is the general form of the Mth order moving average filter of the discrete kind of course.

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The image shows a whiteboard with handwritten mathematical derivations. The first part shows the derivation of the output $y[n]$ for an M-th order moving average filter. It starts with $y[n] = \frac{1}{\alpha} [x[n+1] - \frac{1}{\alpha} [x[n+2] - y[n+2]]]$, then expands it to $y[n] = \frac{1}{\alpha} [x[n+1] - \frac{1}{\alpha} [x[n+2] - \frac{1}{\alpha} [x[n+3] - y[n+3]]]]$, and continues this pattern. The final result is $y[n] = \sum_{k=1}^{\infty} \left(\frac{1}{\alpha}\right)^k (-1)^{k-1} x[n+k]$, labeled as an "Anticausal solution".

The second part discusses the impulse response. It states "Impulse response: set $x[n] = \delta[n]$ ". Then it shows $h[n] = \sum_{k=1}^{\infty} \frac{(-1)^{k-1}}{\alpha^k} \delta[n+k]$. A small diagram shows the impulse response $h[n]$ as a sequence of impulses at $n=1, 2, 3, \dots$ with alternating signs and decreasing amplitudes. The text concludes: "Now it appears $h[n]$ is anticausal!..".

we are now, presently involved in only discussions about discrete systems moving average filters always have a finite impulse response, that is because the moving average filter simply sums up certain previous values of $x[n]$ with appropriate scales, that is all it does and the number of previous values of $x[n]$ that are added is simply M that many previous values of $x[n]$ are being added together. That is why the impulse response of this is very easily calculated $h[n]$, which is equal to $y[n]$. When $x[n]$ equals $\delta[n]$ is given by you just have to substitute $\delta[n]$ for $x[n]$ on the right side, k equals 0 to M , $b[k] \delta[n - \text{minus } k]$, this is the impulse response as simple as that and therefore.

It will have only M different non 0 points in the impulse response actually, M plus 1 because even at M equal to 0 you have. So, its M plus 1 at most it could have less than M non 0 values if any of the $b[k]$'s is 0 for k less than M . Then it would still be called as M th order filter, because there are M delays in the circuit M delay blocks in the circuit in the diagram, but it could have a smaller number of non 0 values alright.

Next, suppose we want to solve differential equations like, we hope we would be able to or as we are now, concerned with difference equations. If you wish to solve them let us see, how we can solve a difference equation of the simplest kind that is the study of what

are called auto regressive filters. Auto regressive filters are genuinely different from moving average filters. Moving average filters as I said have only a finite impulse response, in this sense that they have only finitely many non 0 values in the impulse response, that is why they are also called F I R filters finite impulse response filters. The auto regressive filters as it turns out will have an infinite number of non 0 values in the impulse response. An auto regressive filter the simplest case of, which would be first order difference equation. First order difference equation would be as follows an example, y_n plus say $a y_{n-1}$ equals x_n . So, here we have not cross values of x_n involved in the computation at least not explicitly, but certainly cross values of y_n inward in this computation.

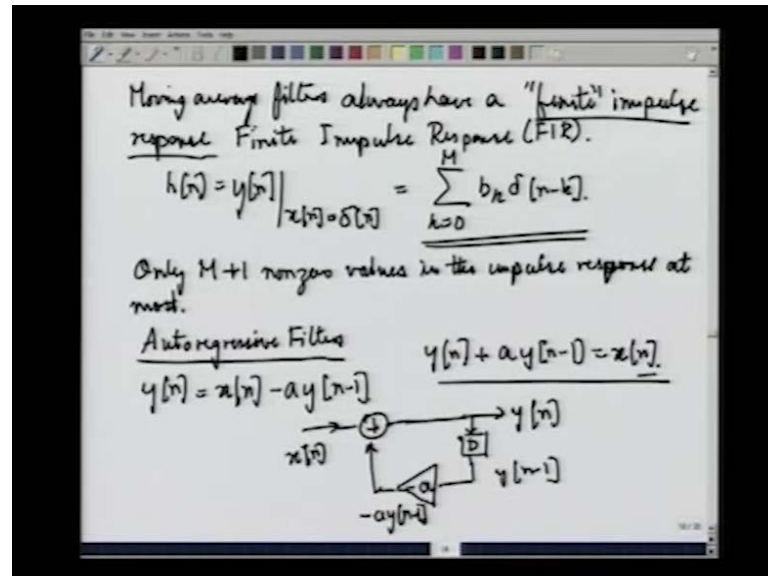
Now, one way to approach this problem would be to go along the lines that we followed, when we try to find the impulse response of a first order a system described by an equation. Such as, this that was there in some previous slide a little while ago let us just takes a look. Here is an example, there must be more examples, if we go further back here is the other example, this one is the causal version, causal solution and the next was the anti causal solution.

Now, for that causal solution or for that anti causal solution, this should be delta over here and likewise let us go to the previous one and see what we have done h_n equals this is. So, we could keep recursively using older and older values of x_n and not using the past values of y_n , but as you saw in the process by which we obtained the impulse response of solutions for causal and non causal, anti causal solutions for difference equations. You still require some one past value of y_n . So, the live way of going about solving an equation such as, this would be to express y_n in terms of y_{n-1} and x_n and then y_{n-1} in terms of x_{n-1} and x_n and y_{n-2} and so on.

So, if we keep doing this we will get an expression for y_n that involves an infinite number of terms. Such as, this expression has an infinite number of terms because k equals 1 to infinity. It goes from 1 to infinity; this also has an infinite number of terms. Now, this is not convenient, we if it has an infinite number of terms; that means, it requires an infinite number of delay blocks, D blocks in that very very clumsy sense of the term. Any auto regressive filter would be an infinite order filter, but that is not, how we are going to look at an auto regressive filter. We are not going to treat it as some kind of a distorted or mutated F I R filter with an infinite number of delay elements. There is a

much nicer and simpler way of doing it, but in order to understand this one has to temporarily live with a paradox, what I am going to assume in this case well.

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Let me first rearrange the equation. So, that it is in the convenient form of $y[n]$ equal to something. So, you can write this as $y[n]$ equals $x[n]$ minus $a y[n-1]$ suppose, this is the equation, you have then what is the best way of implementing this.

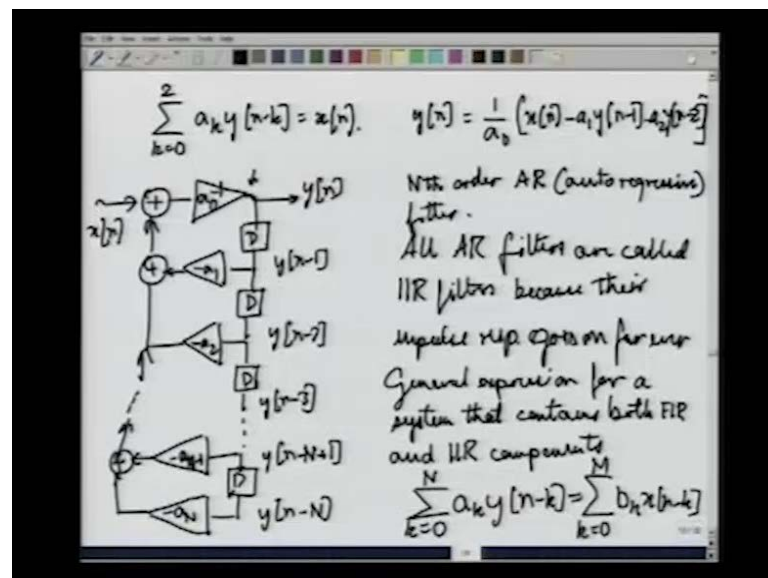
You need to know $y[n]$. You have to compute $y[n]$, but unfortunately to compute $y[n]$, you need to know both $x[n]$ as well as $y[n-1]$. a is known now. So, you need $y[n-1]$ to find out $y[n]$ alright, but $y[n-1]$ is not known either $y[n-1]$ is the $x[n-1]$ minus $a y[n-2]$. So, you would have to know $y[n-2]$. So, you have to keep going further and further back, because there is always one term of $y[n]$, which you do not know, such as $y[n-2]$ or $y[n-3]$ or $y[n-4]$. The problem only goes further and further away, but it does not go away altogether. So, what is the solution? The solution is to pretend temporarily that you have somehow solved this equation and have actually evaluated $y[n]$, suppose you make that temporary hypothetical belief, assumption that $y[n]$ has somehow been found.

Now, if $y[n]$ has been found then you know the value of the sequence y for the point $n-1$ and $n-2$, and everything only then you will say that. You know the entire sequence n . So, well take this and put a delay on it, the sequence that appears at this point will give $y[n-1]$ this of course, is a delay block as I said it is a delay block.

Now, that you have $y[n-1]$. Let us, try to proceed remember this entire thing is founded on the belief that we have $y[n]$, somehow and we will just have to live with that belief for a little longer. Since, we have $y[n-1]$. Now, we will make the appropriately scaled version of $y[n-1]$, which is $-a_1 y[n-1]$. So, what you have, here is $-a_1 y[n-1]$. So, if you have $-a_1 y[n-1]$ and you have $x[n]$. All you have to do is to add these 2 and the sum will itself be equal to $y[n]$. So, all I do is put up a summing block like this, and let $x[n]$ going to it and $-a_1 y[n-1]$ go into it, what should I get over here according to the equation, I should get $y[n]$ over here and well though I got $y[n]$. Assuming that I already had it I am happy to see that I do have it. So, I just go ahead and connect these 2 points. So, it produces $y[n]$ and feeds upon itself to produce $y[n]$ continually in future in perpetuity. This is how a first order auto regressive filter functions you are able to get $y[n]$, if you start, if you start by pretending that you have $y[n]$.

A second order auto regressive filter will be 1, which can solve a second order difference equation. Second order difference equation would be of the form.

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Let me put it in the more general form with all kinds of coefficients in it summation k equals 0 to 2 in this case a $k y[n-k]$ equals $x[n]$. You just have to extend the previous model, a little further and write $y[n]$ equals $\frac{1}{a_0}$ times $x[n]$ minus $a_1 y[n-1]$ minus $a_2 y[n-2]$ this would give me $y[n]$.

Now, this also can be implemented, if you follow the same trick that we followed a little while ago, which is to pretend that you have y_n , then you put 1 delay on this and get y_{n-1} . Put another delay on it and get y_{n-2} . So, you have these, all these hypothetical for the time being because y_n itself is not known, but now that you have these or in a position to put in some of these coefficients this will be minus a_2 this will be minus a_1 , and you have to add all these things.

So, you put a summing point over here now you have these 2 things. These 2 things plus x_n complete the picture in some sense x_n plus minus a_1 times y_{n-1} plus minus a_2 times y_{n-2} . What is all this equal to this is, just a_0 times y_n , but we do not want $a_0 y_n$. We want y_n . So, we just now, here put a forward looking scaling block and write here a_0 to the power minus 1 that is $1/a_0$.

What is do what do you find at this point you just find y_n . So, you have got y_n , once again. The more general equation for the n th order autoregressive filter, which we just write as AR filter, this found by just generalizing this diagram, a little bit which I will do without rewriting it all over again, I have this already. So, I will put another delay block over here. So, I will get y_{n-3} and so on ... until, I come to the last delay block before, which I have y_{n-n+1} and with this delay finally, you get y_{n-n} , and you take this and scale it with minus a_N , this will be minus a_{N-1} , minus a_{N-2} , and then you add all these using the text for shortcuts, that you are already familiar with.

So, this is it you now have an, n th order autoregressive filter, which will give the output y_n , if you have x_n and these may be delays scale multiplier scalars like a_0 inverse a_1 and a_2 minus a_1 minus a_2 and so on until the last summation minus a_n , this gives you an n th order AR filter.

Now, see the point within AR filter, it does not even have to be n th order even a first order AR filter is full of surprises, what we want to see is whether, if the impulse response is really infinite, it is not hard to see suppose you apply x_n as just δ_n . Then what would happen in a system like this, δ_n would add to something and come over here, and a delayed δ_n would be δ_{n-1} . It would get scaled and it would come back here and even though input is no longer being given this would now, act as the input δ_n or rather a minus a_1 minus a_2 would appear as the next output and

this would go on and on, on and on, on and on. So, that you would get an infinite impulse response. An impulse response with a support that is infinite.

So, these filters are called I I R filters, all AR filters are called I I R filters, because their impulse response goes on forever that is why it is called an infinity impulse response filter or I I R filter. Now, you could in principle have a filter, which contains both F I R and I I R elements. It would of course, jointly be I I R only it would not be F I R under any circumstances, but we will see how one can combine these 2 and make a block diagram for a more general filter, which contains both an I I R and an F I R component.

So, let us look at the equation of a system, which contains both I I R and F I R components. Such a system would have a general expression like this, on the left side. We would have the part that is a difference equation that is like; the left side of a difference equation k equals 0 to N . $a_k y[n-k]$ and this would be equal to a similar sum weighted sum of past values, and the present value of $x[n]$ that is to say k equals 0 to M , $b_k x[n-k]$, this is what you have in the right side.

So, alright suppose, you have this, then M in general could be different from N , but it would be convenient for us to keep both numbers the same. And in order to keep both numbers the same there is a very simple trick I can use.

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Choose the larger of M, N , $\max(M, N)$ and call it N . Suppose $N=5$, $M=3$.

$a_5 \neq 0$
 $b_3 \neq 0$.
 $\rightarrow b_4 = b_5 = 0$.

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^{M+2} b_k x[n-k]$$

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^N b_k x[n-k]$$

$$y[n] = \frac{1}{a_0} \left[\sum_{k=0}^N b_k x[n-k] - \sum_{k=1}^N a_k y[n-k] \right]$$

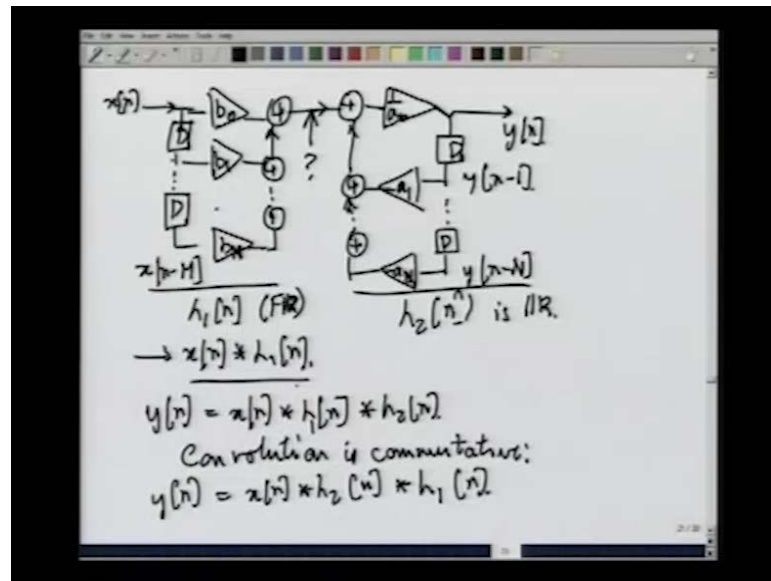
The trick 1 uses is to choose the larger of M and N that is to say $\max M, N$. Choose the larger of the 2 and call it N , then we rewrite the equation. So, that there are exactly n terms on either side. Now, the extra terms that need to be added on the side, that was earlier possess of a smaller number of terms will essentially be terms. Where the coefficients are 0. So, suppose N were equal to 5 and M were equal to 3 then. Since N is equal to 5. I would have a 5 not equal to 0 and I cannot comment on a_4 to a 0 sum of those terms might possibly be 0 on the other hand for the right side. Since M is equal to 3, we can write that b_3 is not equal to 0.

There is no higher coefficient on the right side than b_3 , because M equals 3, but that does not stop us from simply saying that b_4 equals b_5 equals 0. Thus, we would write summation keeping this in mind, we would write summation $a_k y_{n-k}$, k equals 0 to n , which is equal to 5 equals k equals 0 to $M+2$, $b_k x_{n-k}$. Now since, m equals $M+2$, we can call both of them n and simply write summation k equals 0 to n $a_k y_{n-k}$ equals k equals 0 to n , $b_k x_{n-k}$ keeping in mind that this is true.

So, we now have a more convenient form of this kind of an equation which we want to solve to find y_n given x_n . So, how do we do it, we have to rewrite this equation. So, that we have an explicit expression for y_n and that is not hard to do, we can write y_n equals 1 by a_0 , which is the coefficient of y_n times b_k summation k equals 0 to N , x_{n-k} minus a_k summation k equals 1 to n , y_{n-k} note that the y_n minus 0 term is not incorporated on the right side, but on the left side namely y_n itself.

All other terms of the left hand side have been pushed to the right. So, that you now have an explicit expression for y_n with this explicit expression, we can see how this thing can be constructed, you have 1 by a_0 into b_k summation sorry there is a mistake here (Refer Slide Time: 45:55) this should be over here this also should be over here let us just erase this and make it more spacious this is the correct expression sorry for the mistake. Now, how do you implement this, we look at these 2 blocks as 2 separate blocks and try to find these 2 intermediate quantities the result of the summation of all the terms in the first summation and the result of the second summation.

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To get the first summation, we already have a certain mechanism namely, we take $x[n]$ and delay it a sufficient number of times to get the different delayed versions of the signal. So, you have here $x[n-M]$, here you have $x[n]$. So, you multiply all these by respective scales b_0 and this will be b_M .

Now, let us add all these things up by adding up over here, this another sum over here and. So, on till you have a summation over here, which gets an input from b_M , $x[n-M]$. So, you have all these things this is the summation of the first set of terms.

Now, we need the second set of terms after we add these 2 there is still a little more work to be done, but let us construct the second set of terms the second set of terms if you look back has $y[n-k]$ starting from k equal to 1 to k equals capital N . So, you have various delayed versions of $y[n]$ starting with delay by 1 delay by 2 and so on.

So, now I will assume that $y[n]$ is available over here the same trick that I played before $y[n]$ is available here. So, now, if I put a delay over here you find that $y[n-1]$ is available and. So, on add infinite term. So, you get a further delay and here you get $y[n-N]$ that is what we have over here.

Now, let us scale all each of these terms by the appropriate scale factor $y[n-N]$ gets scaled by $a[n-N]$ actually and $y[n]$. So, on up to here where $y[n-1]$ is getting scaled by $a[n-1]$. Lets now, add all these terms together separately like we are about

to do. So, we have added all the terms in the second summation we have added all the terms in the first summation. Now, we have to put all these together now, how do you put them together not hard you have got all these terms in the sum over here you put another summing block over here and you now have a y_n .

So, in order to get all this to evaluate to y_n , you just now multiply by 1 by a_0 . So, what you have is 1 by $a_0 y_n$ and that, when added to this or rather when connected to this will give you y_n , this is a completely self consistent set of blocks arranged set of blocks, which will generate y_n the solution to the differential equation, which had x_n as the input.

Now, we have this you see each of these blocks is essentially implementing some kind of a convolution, because it is a linear time invariant block having constants at all these summation points at before all these summation points, which are scaling the original signal and then we are adding up.

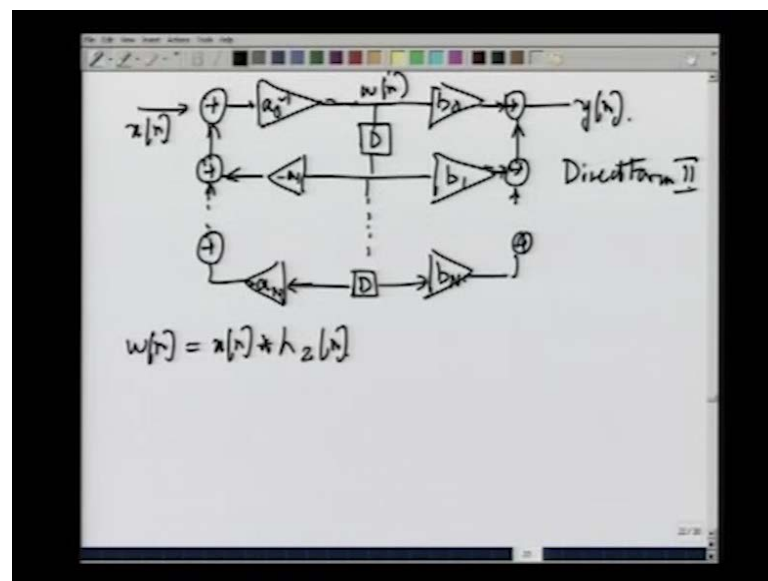
So, we are just generating a linear function or a quantity, which is linearly related to x_n . Now, keeping this in mind what can we do because this whole thing is linear, let me say that this block has an impulse response, which is I will something I will call h_1 of n , likewise this will have an impulse response which I will call h_2 of n . Now, h_1 of n is F I R as you know h_2 of n is I I R.

More importantly since, these two diagrams have a lot in common we can rearrange these two diagrams, but before we before we rearrange we should be very clear about not just the fact that this is F I R and this is I I R. But, the fact that this block is essentially implementing a linear time invariant system and therefore, can be equivalent to the computation of convolution, in short if this is what we call h_1 of n and this is what we call x_n , what signal do we have at this point. The answer we have is that is x_n , convolved with h_1 of n that is what you get over here. Now, this is another block, which is also linear time invariant and this block is now, processing this signal and producing y_n .

In short, I can express y_n also as a convolution of h_2 of n and the input to the second block therefore, I can write y_n equals x_n convolved with h_1 of n , this in turn convolved with this was h_2 of n , this in turn convolved with h_2 of n this is a total convolution of the F I R part followed by the I I R part with the input signal to get y_n the output signals.

An important thing that we can now do is remember that convolution is commutative. So, we can as well write $y[n]$ as equal to $x[n]$, what we want is to reverse the order of the 2 convolutions and write $h_2[n]$ over here and write $h_1[n]$ over here. What does this serve, what help does this give us what do we get out of doing this. That is what we will now, see I have a block over here and I have a second block over here this is h_1 and this is h_2 . We are now going to exchange their positions you apply $x[n]$ and now you have to process $x[n]$ with the h_2 block.

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First and the h_2 block consisted of a summation and here, you had a_0 inverse here after this was actually some output from here you had a delay and so on. Until you had the last delay over here, you add a set of delays. Now, this delay was scaled by a or rather minus a_N . This similarly would be scaled by minus a_1 and this would get into this and so on. This is just the same block that we had before except that now, the input is $x[n]$ and the output is something that we call say $u[n]$ or $w[n]$ still better.

Now, $w[n]$ is $x[n]$ convolved with $h_2[n]$, what do we do with $w[n]$, we now put the moving average part the FIR part on this side. So, that we take this over like this and scale it by b_0 take it down through a delay and scale this by b_1 take it further down until you finally, get b_n .

Now, remember that the number of coefficients in both is same in both sides are same for both the set of terms is same except that we are always free to assume as per convenient

to us, whether some of the higher end b_n 's are 0 or some of the higher end a_n 's are zero. So, now we have this and then again construct the intermediate sums and get since, we have just exchanged the blocks, the final result of all these must still be y_n , which indeed it. So, you have y_n over there.

But now, let us look at these blocks, you have something called w_n over here and w_n has been delayed n different times for the moving average part and n different times for the auto regressive part, but it is the same signal which is getting delayed 2 times through two sets of parallel delay blocks we need only one set of parallel delay blocks. So, we could actually simplify this diagram by combining all these delays together if this is w_n this will be w_{n-1} , but w_{n-1} is required for this as well and so on. Until here, you will get w delayed by n . You do not get w_{n-1} over here, what you would get is w_n minus 1, which is required by both sides. Finally, here you would get w_n minus n that is what you would get over here.

So, let us just knock off two parallel sets of delay blocks and replace it by one set of delay blocks. So, I am going to do that by removing these parts and putting a single signal this directly goes in over here. This gets delayed by 1 time this is w_n actually and this goes to both sides. So, on until you finally, have again a delay 1 input going over here and of course, the other input going over here this is a further simplification of making block diagrams for systems described by differential equations the general systems that we had auto regressive as well as moving average systems.

So, this is the general block diagram this is called a direct form two implementation and the implementation you had before, where you had two sets of blocks 2 complete independent set of delays etcetera is called a direct form one realization.

So, which is better direct form two is defiantly better, because it does the same job as direct form one, but it does it using half the number of delays that the direct form one used .