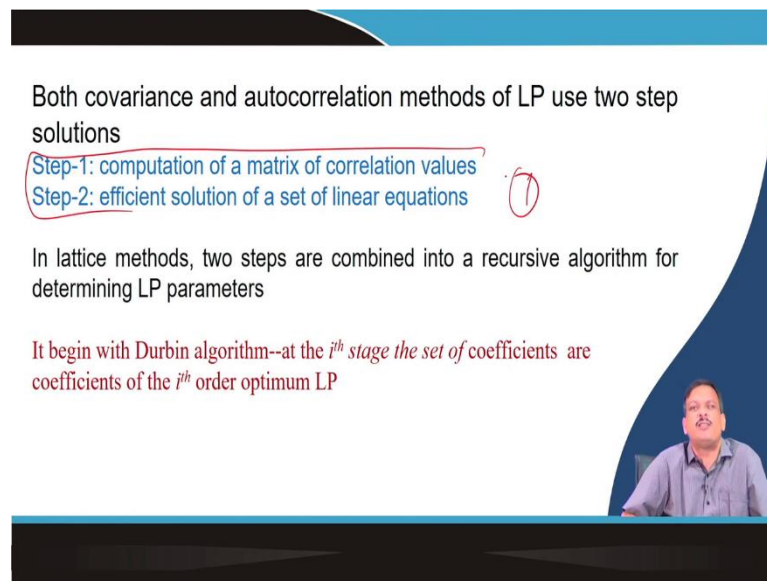


Signal Processing Techniques and Its Applications
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Lecture - 48
Lattice Formulations of Linear Prediction

So, now we go for that lattice formulation of linear lead prediction. So, what is there?

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Both covariance and autocorrelation methods of LP use two step solutions

Step-1: computation of a matrix of correlation values
Step-2: efficient solution of a set of linear equations

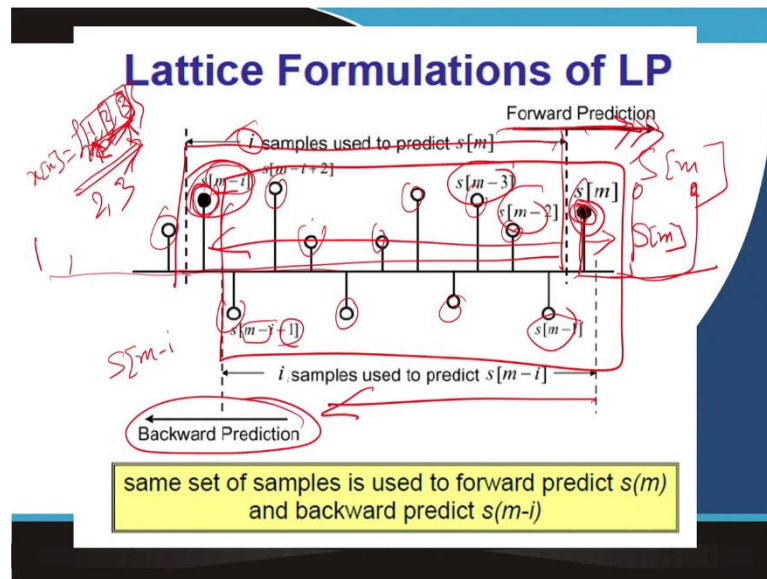
In lattice methods, two steps are combined into a recursive algorithm for determining LP parameters

It begin with Durbin algorithm--at the i^{th} stage the set of coefficients are coefficients of the i^{th} order optimum LP

The slide features a blue header and footer. The main content is on a white background. A red box highlights the two steps. A red circle with the number 1 is next to Step-2. A video inset of a man in a blue shirt is in the bottom right corner.

So, as I discussed, the computation of the matrix of correlation value or autocorrelation, whatever the covariance or correlation, does not matter. So, that requires a complex and efficient solution set of linear equations; that means an iterative method is the requirement. So, can I combine these two steps into one step? So, can I combine this step into one step? So, that method is called the lattice formulation method. Let us go for that. What is that?

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Let us say. So, when I say the prediction, let us say I have a sample. So, let us say I can say this is a signal which is the many samples are there, many samples are there. So, this is my most number of samples. So, if I say I want to, I will use the sample to predict $s[m]$ sample. So, I want to predict this sample from the previous i^{th} sample. What is the meaning? Let us give an example; suppose I have a signal $x[n]$ 1 2 3. Let us say I want to predict 3 based on our previous 2 samples.

So, here also I am predicting $s[m]$ based on the previous i^{th} sample. So, what is the $s[m-1]$ $s[m-2]$ $s[m-3]$? Ultimately, $s[m-i]$ was the previous i^{th} sample. So, when I predict that way, it is called forward prediction. So I can, I can say that I am predicting 3 from the previous 2 samples. Also, I can say I can predict one from the past, I can say that from this 2 and 3.

So, prediction can go this way, once it goes from 1 2 3, that is called forward prediction. This way, I can predict the next sample. So, this is a forward prediction. I can predict the backwards. also, I can predict 1 from 2 and 3 because $x[n]$ is known. So, once the $x[n]$ is known, I can predict both ways. So, when I predict this way, suppose I want to predict $s[m]$ I minus 1 sample from the previous, or I can say this side with the sample. Understand or not?

So, I am predicting this sample from this previous sample, or I can predict this sample from this previous sample. So, when I predict $s[m]$ from a previous i^{th} sample, that is forward prediction; when I predict this side, this is called backward prediction. So,

prediction can be forward, and prediction can be backward. I can predict 3 from 1 and 2, I can predict 1 from 2 and 3. When I predict 3 from 1 and 2, it is a forward prediction; when I predict 1 from 2 and 3, it is a backward prediction.

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$$H(z) = \frac{A}{1 - \sum_{k=1}^p \alpha_k z^{-k}}$$

$$A(z) = 1 - \sum_{k=1}^p \alpha_k z^{-k}$$

i^{th} order prediction error filter

$$A^i(z) = 1 - \sum_{k=1}^i \alpha_k z^{-k}$$

$$e^i[m] = s[m] - \sum_{k=1}^i \alpha_k^i s[m-k]$$

Forward prediction error

$$b^i[m] = s[m-i] - \sum_{k=1}^i \alpha_k^i s[m-i+k]$$

$$B^i(z) = z^{-i} S(z) - \sum_{k=1}^i \alpha_k^i z^{-i+k} S(z) = z^{-i} S(z) \left[1 - \sum_{k=1}^i \alpha_k z^{-k} \right]$$

$$B^i(z) = z^{-i} S(z) A^i(z^{-1})$$

$$E^i(z) = A^i(z) S(z)$$

So, there is a forward prediction and a backward prediction. Ok, do you understand that part? Now I come to that: what is the equation of the forward prediction, and what is the equation for backward prediction? So, what is my prediction filter? This one. So, if I say i^{th} order prediction error filter. So, this is nothing but a prediction error filter $s[n]$ minus the previous sample. This is the previous sample sum. So, let us say i^{th} order prediction error filter.

So, $A_i(z)$ is equal to 1 minus k , equal to 1. So, I can say i^{th} iterative. So, this I represent the iteration, and this I represent the order. As I said, suppose there is a sample. So, the prediction error is maximum when I am predicting this sample from the previous 14th 0 sample. So, in the first iteration, I am predicting this sample: the second iteration, this one; the third iteration, this one; and the fourth iteration, this one.

So, once I go to the iteration, that error will be minimized. Ok or not? So, let us say i^{th} iterative iteration process is this one. So, what are the prediction errors? $E_i(z)$ is equal to $s[m]$ signal minus predicted signal. So, $s[m]$ is my predicted sample. So, in the z transform, I can say $E_i(z)$ is nothing but an $A_i(z)$ multiplied by $S(z)$, and $S(z)$ is the input signal. Now, what is backward prediction? I am predicting m . So, what am I predicting forward? In

forward prediction, I am predicting $s[m]$; in backward prediction, I am predicting s of m minus I , if the i^{th} order prediction is used.

So, I am predicting this sample forward prediction, I am predicting this sample. So, this is minus the estimated sample. So, backward prediction similarly this is my signal minus estimated signal $s[m]$ I minus so that one plus 1 plus 2 plus 3. So, that is why plus k if you see. This is $s[m-i]$. This is $s[m-i+1]$. Suppose I represent by s of. So, suppose this is the third, so I am predicting 1. So, this is nothing but a first sample. So, I can say the 2 minus. So, this one is the next part of this.

So, 1 will be added to this. So, that is why this is added k is added. Now, if I take the z transform, so, this will be $B(z)$ this will be z^{-1} multiplied by $S(z)$ m minus i minus k equal to 1 to k equal to 1 to i $\alpha_k i z^i$, so z minus i plus k into $S(z)$. Now, if you look at this, k is equal to 1 to i . So, I can say that z^{-i} is independent of this sum. So, z^{-i} is taken out, and $S(z)$ is taken out. So, this is 1 minus this one.

Now, if you see this one and this one are almost the same; the only difference is that here z is z^{-k} , here it is z^k . So, I can say that instead of $A(z)$, this can be $A(z)^{-1}$. So, I can say $B(z)$ is nothing but $A(z)^{-1}$ multiplied by $S(z)$ into $A(z)^{-1}$. So, $E(z)$ is equal to $A_i(z) S(z)$. So, it is $A_i(z) z^{-1} S(z)$.

So, this portion, this portion, except this error filter, z^1 , this z^{-1} and multiplied by z^{-i} .

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Levinson Recursion

Step-1

$$E^{(0)} = r(0)$$

$$\alpha_0^{(0)} = 0$$

Step-2 *Weighting factor of i^{th} pole model*

$$k_i = \left\{ r(i) - \sum_{j=1}^{i-1} \alpha_j^{(i-1)} r(i-j) \right\} / E^{(i-1)}, \quad 1 \leq i \leq p$$


Step-3

$$\alpha_i^{(i)} = k_i$$

$$\alpha_j^{(i)} = \alpha_j^{(i-1)} - k_i \alpha_{i-j}^{(i-1)}$$

Step-4 Update the mean square prediction error

$$E^{(i)} = (1 - k_i^2) E^{(i-1)}$$



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$$\alpha_j = \alpha_j^{(i-1)} - k_i \alpha_{j-i}^{(i-1)}$$

$$A^i(z) = 1 - \sum_{k=1}^i \alpha_k^{(i)} z^{-k}$$

$$A^i(z) = 1 - \sum_{k=1}^i (\alpha_k^{(i-1)} - k_i \alpha_{i-k}^{(i-1)}) z^{-k}$$

$$= 1 - \sum_{k=1}^{i-1} [\alpha_k^{(i-1)} z^{-k} + k_i \alpha_{i-k}^{(i-1)} z^{-k}] - (\alpha_i^{(i-1)} - \alpha_{i-i}^{(i-1)}) z^{-i}$$

$$= [1 - \sum_{k=1}^{i-1} \alpha_k^{(i-1)} z^{-k}] + k_i \sum_{k=1}^{i-1} \alpha_{i-k}^{(i-1)} z^{-k} - k_i z^{-i}$$

$$= [1 - \sum_{k=1}^{i-1} \alpha_k^{(i-1)} z^{-k}] + k_i \sum_{k'=i-1}^0 \alpha_{k'+1}^{(i-1)} z^{-k'} - k_i z^{-i}$$

put $k' = i - k$

$$= [1 - \sum_{k=1}^{i-1} \alpha_k^{(i-1)} z^{-k}] - k_i z^{-i} [1 - \sum_{k=1}^{i-1} \alpha_k^{(i-1)} z^{-k}]$$

$$= A^{i-1}(z) [1 - \alpha_i A^{i-1}(z^{-1})]$$

So, now, if I say that I want to write down this one, how can I write down $A_i(z)$ in terms of the previous error? So, what Levinson said is that recursion is possible. What recursion is possible? I can say that $A_i(z)$ can be predicted from $A_{i-1}(z)$. So, I want to write down that part. How can I include recursion Levinson recursion so that I can calculate $A_i(z)$ from the previous value? So, I know $A_i(z)$ is this one. Now I know α_k ; this is the Levinson recursion equation. α_j is equal to the previous iteration result minus k_i with the Parcor coefficient and α previous iteration.

So, I just put this one instead of j ; I said k and i minus 1 would be there. So, I can say α_k i with order; that means i th iteration can be written as i minus 1 iteration k instead of j . I am k writing k . Now, if you see this is k equal to 1 to i . Let us say I said this summation will be i minus 1. So, k equal to one to i minus 1 α_k i minus 1 z^{-k} minus k_i minus.

So, this is nothing but an α that will be k_i . So, k_i into α_{i-k} z^{-k} multiply and what I have written one term I have not written. So, this is my last term. So, this is nothing but a k_i . So, in the second stage, I writing this one as a k_i , and this one if you see minus. So, I can say 1 minus this one and minus plus k_i into k equal to 1 to i minus 1 this one minus k_i into z^{-i} .

Now, I can say this is one term, and if I say k_i , it is because summation is not independent of i . So, k_i is this side k equal to i minus 1 to i . I can put this one, okay? So, I put k dash equal to i minus k i minus k is equal to k dash ok. So, when k is equal to 1, it is nothing

but I minus 1. When k is equal to I minus 1, it is nothing but i . So I can write down this one. Now I can see this one is nothing but an A i minus 1 z .

Because $A_i(z)$ is equal to this one, A i minus 1 z is nothing but a α k i minus 1. So, A i minus z , and if I say k i and z^{-1} is outside, then it becomes 1 minus this one. If you see this, it is z^k . So, if I make z minus k . So, it is nothing but an A i minus 1 z^{-1} .

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$$A^i(z) = A^{i-1}(z) - k_i z^{-1} A^{i-1}(z^{-1})$$

$$E^i(z) = A^i(z)S(z) = A^{i-1}(z)S(z) - k_i z^{-1} A^{i-1}(z^{-1})S(z)$$

$$= E^{i-1}(z) - k_i z^{-1} B^{i-1}(z)$$

$$e^i[m] = e^{i-1}[m] - k_i b^{i-1}[m-1]$$

$$B^i(z) = z^{-1}S(z)A^i(z^{-1}) = z^{-1}B^{i-1}(z) - k_i E^{i-1}(z)$$

$$b^i[m] = b^{i-1}[m-1] - k_i e^{i-1}[m]$$

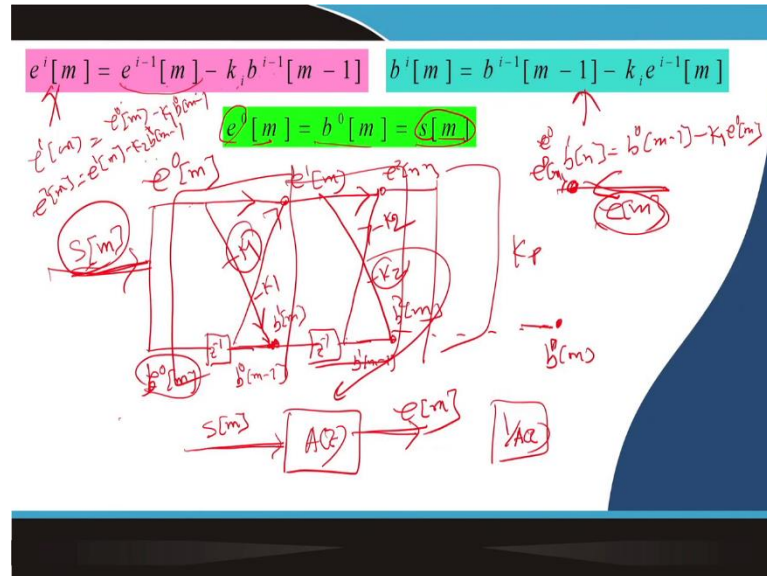
So, I can say that $A_i(z)$ is nothing but a A i minus 1 z minus k i into z^{-1} A i minus 1 z^{-1} . Now, if this is my error prediction, error filter, the prediction error is nothing but multiplied by the signal I put this $S(z)$. So, if you see A i minus 1 z $S(z)$, I can say it is nothing but an E i minus 1 z and A i minus 1 z to the power z minus 1 $S(z)$, which is nothing but a $B(z)$, $B(z)$ is this one only in this form.

So, I can say I write down the B z , but one thing, this is z i minus 1. So, I required z^{-1} because I required z i minus 1 previous one. So, 1 minus z to the minus 1 will be written. So, it can be i minus 1 plus 1, or I can say if i is common j i minus 1. So, that is why 1 z^{-1} will remain, and k i will remain, and this B i will be minus z . Now, from this equation, I can say that $E_i(z)$ is in the time domain.

So, this is in z domain i minus 1 E i minus 1 m minus k i into B i minus 1 m minus 1. Understand or not? Similarly, if I put $B_i(z)$, so $z^{-1} S(z) A(z)$ to the power this one. Again, I can say that this is nothing but this one. So, which is nothing, but a b i m is equal to b i

minus 1 minus k_1 into e_1 minus 1. So, now say that this is my equation forward prediction error. This is my backward prediction error.

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Now what? So, I can say that i^{th} iteration is nothing but i minus 1 iteration, so if I say 0th iteration. So, i equal to 0 $e_0[m]$ is nothing but a $s[m]$. There is no iteration. So, the error is nothing but a signal itself; the signal itself is the error. So, it is nothing but a $s[m]$; similarly, $b_0[m]$ is nothing but $s[m]$. So, I can say that when I do the signal flow diagram, I can say let us know if this is my speech signal s or a signal $s[m]$.

So, I can say this is nothing but an $e_0[m]$, this is nothing but a sorry, this is nothing but a $b_0[m]$, this is nothing but an $e_0[m]$. So, this is the 0th-order prediction and 0th iteration. So, this is nothing but a signal itself. Then what is e_1 m ? So, I know this equation e_1 m is nothing, but an $e_0[m]$ will be there minus k_1 . So, i equal to 1 minus k_1 equal to b_0 . So, $e_0[m-k_1]$ into; k_1 into $b_0[m-1]$.

So, this is $b_0[m]$. So, how do I synthesise $b_0[m-1]$? I put $A(z)^{-1}$ delay. So, this is nothing but a $b_0[m-1]$. So, this will be added up with this one with a minus k_1 coefficient. What is k ? k is nothing but a partial reflection coefficient or Parcor, a partial correlation coefficient. If it is on the go for tube model, it is called a partial reflection coefficient.

So, it acts as a partial reflection coefficient. Parcor partial correlation coefficient is. Now, what is $b_1[m]$? $b_1[m]$ is equal to $b_0[m-1]$ minus k_1 $e_0[m]$. So, I can know that b_0 . So,

this is my $b_1[m]$. So, $b_1[m]$ will be this one will be multiplied minus k_1 and added up with this one, I get $b_1[m]$. Now, when I say $e_2[m]$ is nothing but a $e_1[m]$ minus k_2 into $b_2[m]$ minus 1. So, $b_1[m]$, I know.

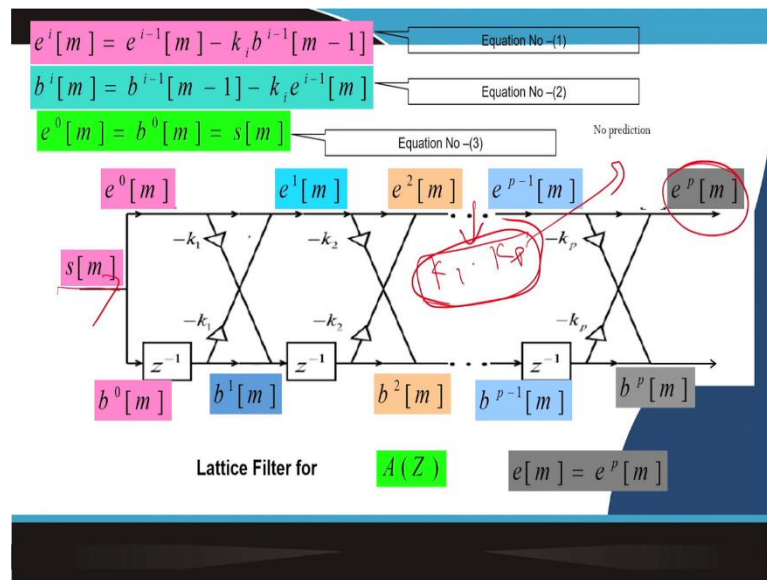
So, that will be delayed by 1 sample, $b_1[m]$ minus 1 I get. And this is my $e_2[m]$, $e_2[m]$ is this signal multiplied by this signal with a minus k_1 k_2 and $b_2[m]$ is nothing but a this will be a multiplied by minus k_2 . So, that way, I can say this is nothing but a, let us say, p th order. So, $e_p[m]$ I will get, and I will get $b_p[m]$. So, on this side, I will get $b_p[m]$, and on this side, I will get $e_p[m]$. So, what is $b_p[m]$ and $e_p[m]$? $b_p[m]$ is nothing but a final error.

So, the final error is nothing but a $b_p[m]$ or $a_p[m]$. Whatever I can say, that is nothing but a final error. So I can get the final error. So, what I said? I have a filter. If I apply $s[m]$, and this is what I implemented here, I will get $e[m]$. So, what do I have to know? I have to know k_1 k_2 k_p . So, let us say all k values. Once I know all k values, I can implement that signal flow diagram on a computer, and if I apply $s[m]$, I can calculate $e[m]$.

So that is called lattice implementation because if you see one block. This block, there is another block, there will be another block. So, each block is simple, only k_1 . So, I can say the one for loop only if you say the for loop is repeated for k_1 k_2 k_3 k_4 k_p . So, the p stage will be there for p th order prediction. So, I can implement $A(z)$ immediately if I know k_1 k_2 k_p . So when I say synthesis, let us say I want to implement a synthesizer; that means which is nothing but a one by $A(z)$.

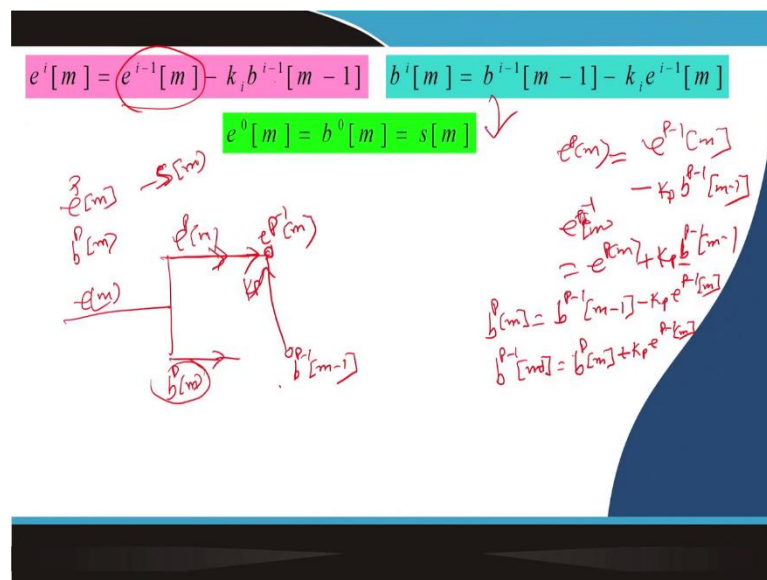
So, here I am, applying for $s[m]$ and getting m . Now, if I see the signal flow is reversed. So, I apply $e[m]$. I should get back my $s[m]$, but the signal flow will be in the reverse direction. So, how do we do that how do I do the reverse direction signal flow? This is the diagram that I did.

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So, I want to draw the reverse direction signal flow. Let us say I just deleted this one, and again, I want to draw it. So, what I said is if I apply $s[m]$ input, I get $e^p[m]$.

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Now, my target is to know $e^p[m]$, and $b^p[m]$. I have to get back the signal $s[m]$. So; that means, I know $e^p[m]$, I know $b^p[m]$. So, all are $e^p[m]$ is equal to $b^p[m]$ at the final stage. So, let us say I know $e^p[m]$. So, which is nothing but an $e^p[m]$, here is nothing but a $b^p[m]$. Now, what do I want to know? So, what is known? I know $e^p[m]$. What do I want

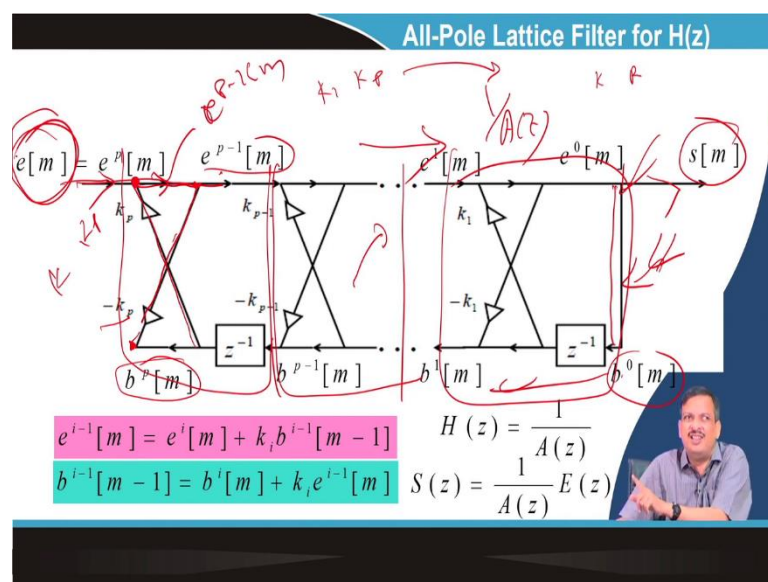
to calculate? $e^{p-1}[m]$. So, I can say that $e^p[m]$ is equal to $e^{p-1}[m]$ plus k_p into $b^{p-1}[m]$.

Now, what do I want? I want $e^p[m]$. So, I said the $e^{p-1}[m]$ is nothing but an $e^{p-2}[m]$ plus k_{p-1} into $b^{p-2}[m]$. So, let us say this one is my point minus 1 m at this point. So, how do I know? e^{p-1} is nothing but the e^{p-2} . So, this signal directly goes here, and I required k_{p-1} multiplied by b^{p-2} . So, let us say this one is b^{p-2} , this one is b^{p-3} , let us say this one.

So, this one will be multiplied by plus k_p , and I will get up minus 1 m. So, how do I get b^{p-1} m? So, from this equation, I know $b^p[m]$ is equal to $b^{p-1}[m]$ minus k_p into $e^{p-1}[m]$. Now, I say $b^p[m]$ is equal to $b^{p-1}[m]$ minus k_p into $e^{p-1}[m]$.

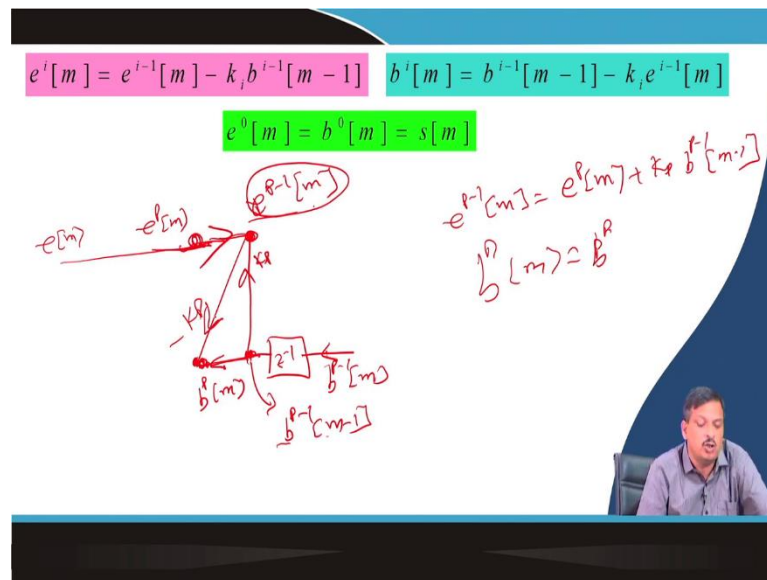
So, if I know $e^{p-1}[m]$, if I know b^p . So, this one $b^p[m]$, I know $b^p[m]$. So, if I want to know if the BP is minus 1, I have to delay the signal. So, what is required? So, what can I say? I am okay. Let us say the $b^p[m]$ I want to generate.

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Let us say I want to generate $b^p[m]$ instead of $b^p[m]$, and I am generating $b^p[m]$, okay?

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So, what will I do? I will draw it fresh; what I know I know let us say I know $e^p[m]$ is known $e^p[m]$ is known. And this one is, let us say, $b^p[m]$. So, if this one is delayed by z^{-1} , then this one is $b^p[m]$. This one is nothing but a. So, this is delayed by z^{-1} . So, this one is nothing but a one-sample delay, ok? So, this is b^p minus 1 m delayed by one sample. So, at this point, I get b^p minus 1 m. This point lets us know this is $b^p[m]$.

So, let us say this is b^p minus 1 m. So, I know $e^p[m]$. Let us say this one is my e^p minus 1 m. So, e^p minus 1 m is nothing, but this sample multiplied by this will be multiplied by k_p . Understand or not? So, this point is my $e^p[m]$, and this point is my e^p minus 1 m. So, this point is b^p minus 1 m minus 1 multiplied by. So, I know that e^p minus 1 m is equal to $e^p[m]$ plus k_p into b^p minus 1 m minus 1.

So, I know this is b^p minus 1 m minus 1. This will be multiplied by k_p , and I get up $[m]$. Now, what is $b^p[m]$? It is nothing but a b^p minus 1. So, this is b^p minus 1, and this will go here minus. So, this is nothing but an e^p minus 1. So, I can say this will go here: multiply minus k_p . This is nothing but a $b^p[m]$. Understand? So, the ultimate signal flow diagram will be what? This is $e^p[m]$. So, this is my plus k_p .

So, I get e^p minus 1 m. After this point, it is e^p minus 1 m. So, from that, the only thing is the signal diagram. So, from here, if I take up minus 1, I am taking multiplied minus k_p . I get $b^p[m]$ in here. Understand? So, if I apply e^p here, I get output $s[m]$. Because I know e^0 and b^0 is the same.

So, $s[m]$, so this is backward, and this is forward signal flow. So, why am I implementing it? I am implementing one by $A(z)$. first, I am implementing $A(z)$; now, I am implementing one by $a(z)$. Understand? Again, it is also a lattice, 1 lattice. Understand?

So, now, if I apply the error signal if I know k_1 k_2 k_p when I say the voice is transmitted. So, I analyzed, and I said k_1 to k_p . I extracted and transmitted in the receiver side from a k_1 to k_p . I can generate this filter, and I can pass the $e[m]$. What is $e[m]$? $e[m]$ is nothing but a glottal excitation in the case of voice.

So, that is nothing but an impulse. So, if I pass an impulse, I can get the speech signal back again. Is it clear? So, that is called lattice formulation. Now, how do I get the k_1 value? Can I calculate the k_1 value using auto correlation? Yes, you can do that. But again, I have to do autocorrelation, and also I have to implement the filter.

So, on the synthesizer side, I know that I have to implement the filter only with these methods, but on the analyzer side, this is how I use it for the analyzer. So, how can I directly calculate the k value from $b_p[m]$ and $a_p[m]$ and how can I directly implement it? So, how can I directly calculate this k_1 to k_p -value? One solution is that if I know the k_1 and k_p values, I can synthesize the signal $s[m]$ from the error signal. In ideal cases, the error signal is 0, and the error is minimized. Understand? So, I can synthesize the signal from the error signal. So, I have to know k_1 k_2 k_3 k_p .

Now, when I am analyzing this diagram, I am inputting the $s[m]$ signal I can calculate the error signal if I know k_1 k_p . So, how do I extract the value of k_1 and k_p ? I can extract those values using auto correlation ok no problem, I can extract those things. But can I directly compute that k_1 and k_p value? So, in the next class, I will describe how I directly calculate the k_1 and k_p values, and I give an example. Let us give a signal, and I can give you an example of how it can be implemented in computer programming,. You should know, ok.

Thank you.