

Digital Speech Processing
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Lecture - 24
Lattice Formulations of Linear Prediction

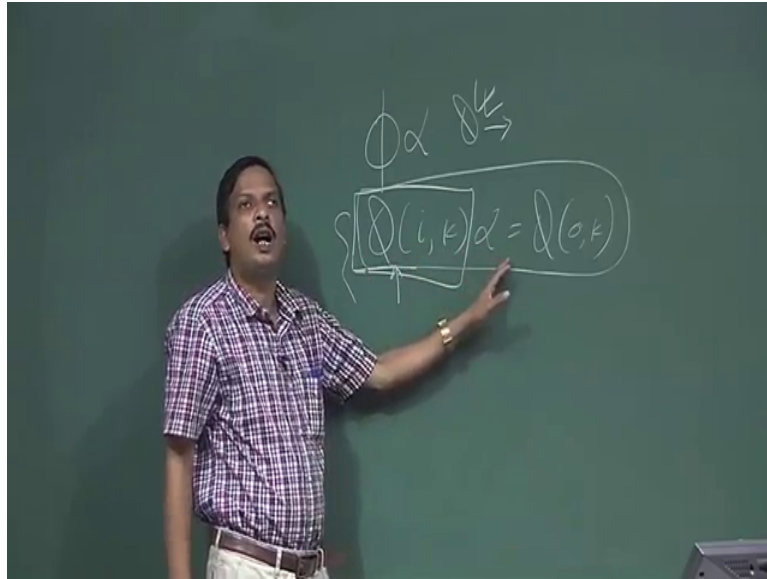
So, in last class, we have talk about that covariance method.

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For linear predictions and there first will explain that auto-correlation methods for linear predictions. So, now we are discussing about the lattices formulation of linear predictions. So, if you see in whether it is auto-correlation or covariance, we try to solve that matrix Φ into α or I can say $\Phi_{i,k}$ into α is equal to $\Phi_{0,k}$. We try to solve that matrix.

(Refer Slide Time: 00:43)



So, how do you solve that? We have calculated that phi. Phi is nothing but a correlation value in case of convolving. In case of covariance matrix, we calculate the correlation value of phi and then, we try to solve this linear equation. So, whether it is your auto-correlation methods or coral covariance methods, then a two step. First step is to compute this phi using correlation for covariance method using auto-correlation for auto-correlation methods. Next try to solve this peak linear equation p number of linear equation to find out the value of alpha.

So, I can say auto correlation methods and covariance methods both are two step solutions.

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Both covariance and autocorrelation methods of LP use two step solutions

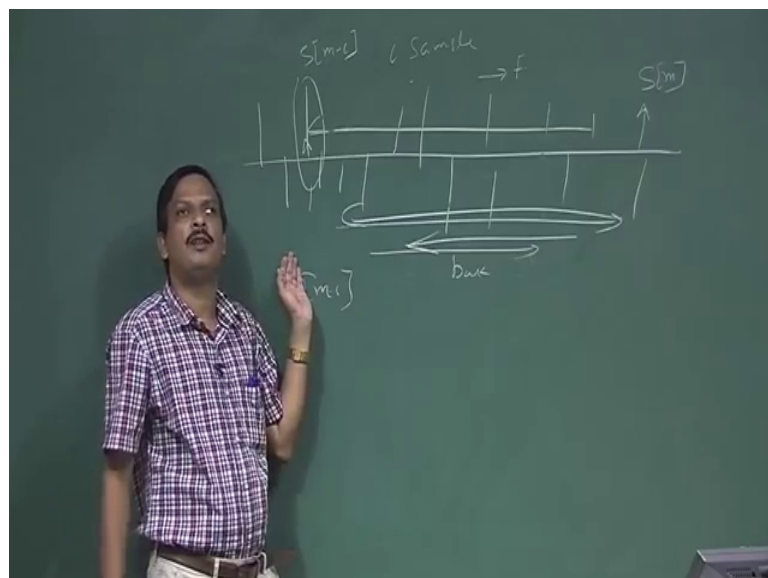
- Step-1: computation of a matrix of correlation values
- Step-2: efficient solution of a set of linear equations

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

Step 1, computation of matrix of correlation value and then, efficient solution of set of linear equation. Now, this lattices method says that I can combine these two steps that says that I do not want to use that auto correlation, compute correlation and solve the linear equation in two separate steps. So, I want that. Can I combine these two steps? Yes, it is possible in case of lattices formulation combining these two steps and use a simple step. How it is possible?

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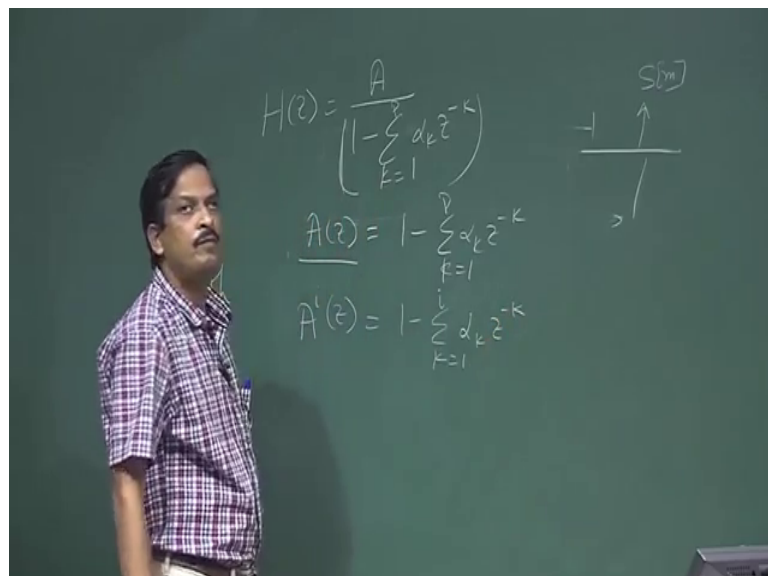


Now, first I have a speech signal m number of that is a speech signal that is a sample, lot of sample is there. So, there lot of sample are there in a equation. So, this is $A S m$ and this; this sample is let $S m$ minus i . So, i can say using this previous i sample, I want to predict this $S m$ sample. So, using this previous i sample, I want to predict this $S m$ sample. So, if I do that, then it is called forward prediction. I can say it is a forward prediction.

Similarly, I can predict less this $S m$ minus 1 sample this sample based on that this previous sample. So, I can predict $S m$ based on the i number of previous sample this side or I can predict this sample based on m number of sample i number of sample this side. So, either I can predict $S m$ cross from $S m$ minus i sample from this side or I can predict $S m$ minus i from i sample in this side. If I predict this one, then the prediction is this direction. This is called backward prediction.

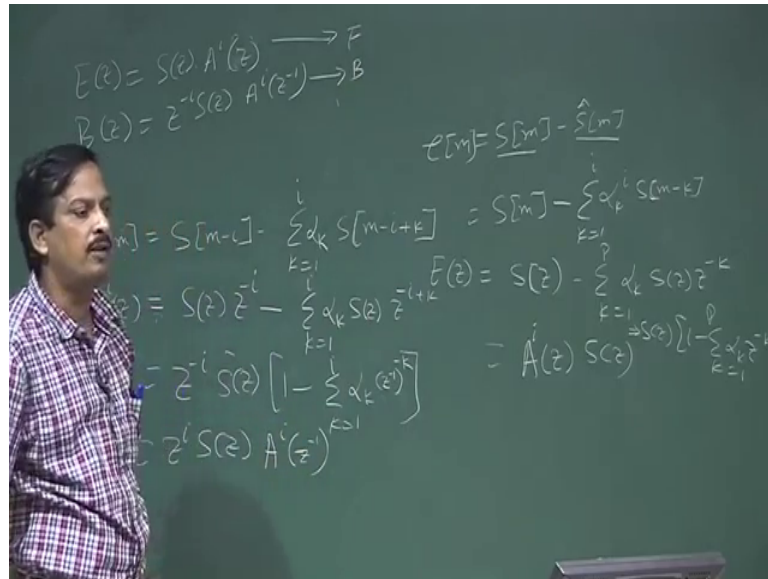
So, there is a forward prediction; there is a backward prediction. I can do both ways. So, next how we can do that I am not explaining the slide again. So, what is that linear prediction equation? If the system is linearly predictable, then equation of the system that $H z$.

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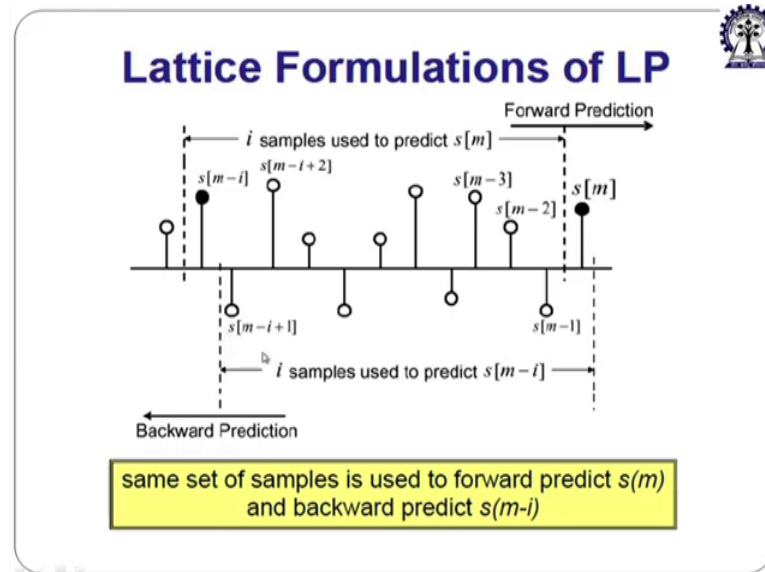
Transformation of the system is gain is A minus K equal to 1 to P A K or α K i can say z to the power minus K . I can say A K or α K whatever now if I say that what is the predic. So, this one is the error filter. So, this is $A z$. So, I can say $A z$ let us this one is

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So, I can say forward prediction error E_z is nothing but s_z into $A_i z$. Now, I want to say the backward prediction error. So, I want to estimate s_{m-i} from previous i sample. So, I can say s_{m-i} is the original sample minus estimated sample k equal to 1 to i $\alpha_k s_{m-i+k}$. I want to predict.

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If I see this, I want to predict this $s[m-i]$ sample from the previous sample. So, if it is that, then I can say this is nothing but b_i or b_m , sorry b_m . So, b_m that original sample minus estimated sample value is the prediction error. So, I can say b_i if it is i th order predictor, so I can say $b_i z$ is equal to i , take the z domain $s z$ into z to the power minus i minus k equal to 1 to i alpha k $s z$ into z to the power minus i plus k z domain.

So, I can say it is nothing but a z to the power minus i if I take and $s z$ if I take out there, it is 1 minus k equal to 1 to i alpha k z to the power k . If I write z to the power k is nothing but a z to the power minus if I say z to the power k is positive. So, I can say z to the power k is nothing but a z to the power minus 1 to minus k . So, in that case I can say this is nothing but a z to the power i $s z$ into $b_i z$ to the power minus 1 or I can say it is nothing but i , sorry $a_i z$ is nothing but z . So, it is 1 minus k equal to 1 to i alpha k z to the power minus k . So, here instead of z to the power minus k a minus $k i$ z to the power minus 1 inside, I can say it is nothing but a $i z$ to the power minus 1 . So, I can say $b z$ is nothing but a z to the power minus i $s z$ into $a_i z$ to the power minus 1 .

So, this is forward prediction error and this is backward prediction error. If it is that, then forward prediction error, backward prediction error, you now think about the Levinson recursion. What are the Levinson recursion said in, Levinson recursion I am not explaining in the board. It is written in the slides if you see.

(Refer Slide Time: 12:04)

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

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

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

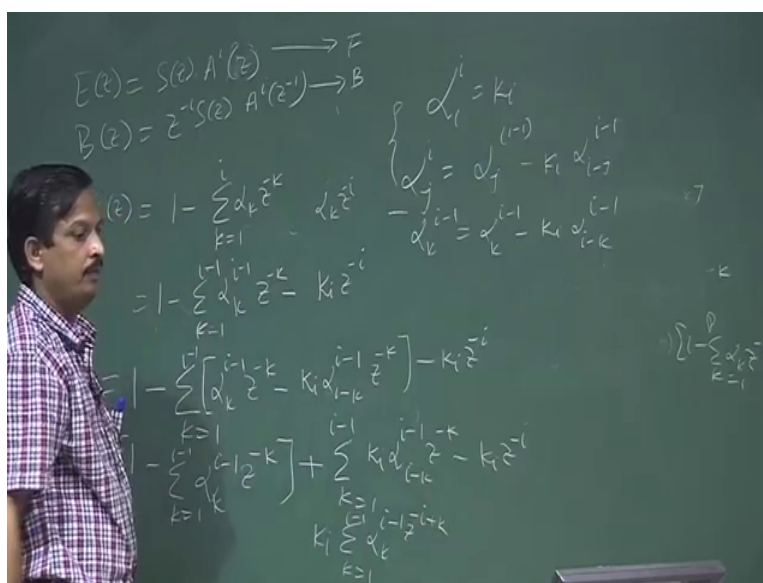
Step-4 Update the mean square prediction error

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

Step 1, E to the power zero order predictor is nothing but a r_0 . So, $0.0 \alpha_0 0.0$ is equal to 0, then you know that there are three iteration and one is weight factor with pole model k_i , then $\alpha_i k_i$ is $\alpha_i i$ equal to k_i , then $\alpha_j i$ is equal to iteration and then, update the mean square error.

So, those are the fourth prediction is used in levinson recursion. So, I can say that let levinson recursion take this $1/\alpha_{i,i}$ is equal to k_i .

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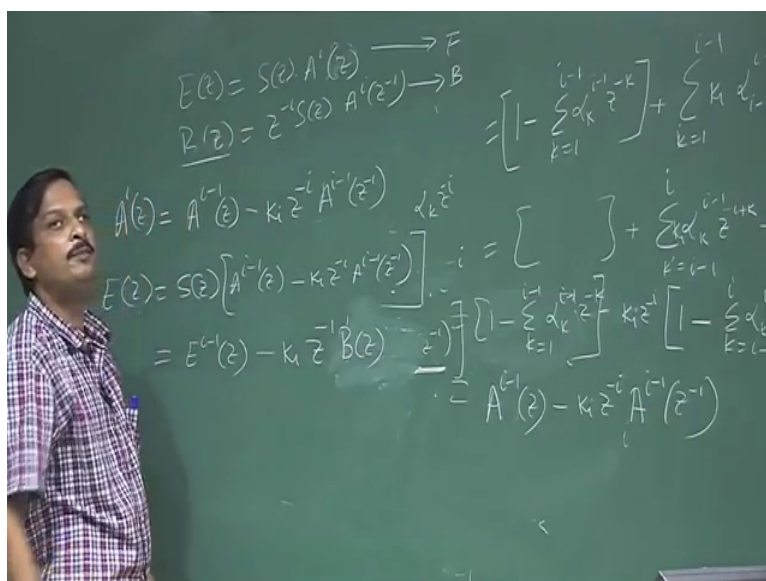
Similarly, α_j is equal to α_{j-1} . Levinson recursion said that current state can be estimated from the previous iteration. So, it is previous iteration alpha value minus k into α_{i-j} to the power minus 1, ok.

So, that is Levinson recursion. Now, think about in here what $A_i z$ is. $A_i z$ is nothing but $1 - k$ equal to $1 - \alpha_k z$ to the power minus k . I can write this one as $A_{i-1} - k$ equal to $1 - \alpha_k \alpha_{k-1}$ previous iteration alpha value of previous iteration z to the power minus k minus k into z to the power minus i . This is $i-1$. So, I can say k equal to $1 - i$, I can say k equal to $1 - i - 1$ α_k $i-1$ predictority. So, it is not i th iteration z to the power minus k minus k into z to the power minus i . So, z if I do it, the sum k equal to $1 - i$, then this will be sum together. So, it will be same as this one, ok.

So, I just from the summation term only $1 - \alpha_k z$ to the power minus i term i th term, I have taken out. So, this is nothing but $A_{i-1} - k$ equal to $1 - i - 1$ α_k $i-1$ z to the power minus k minus k α_{i-k} $i-1$ z to the power minus k . I can write this as $1 - k_i z$ to the power minus i . Why I write this one? I write α_k $i-1$ from here α_k $i-1$, I can write in term of α_k $i-1$ minus k into k $i-1$. Sorry, this is α_k $i-1$ into $i-1$ z to the power minus k . So, α_k instead of here j , I put instead of j . So, α_k $i-1$ equal to α_k $i-1$ minus k into α_k $i-1$ minus k $i-1$. That is why I write $i-1$ k into z to the power minus k , sorry z to the power minus k minus k is here.

So, it is now if I do that $1 -$ I can put that value $1 - k$ equal to $1 - i - 1$ α_k $i-1$ z to the power minus k . This is one term. I can write another term is if I take that this sign, this is minus minus 1 is a plus. So, I can say plus k equal to $1 - i - 1$ k , k α_{i-k} $i-1$ z to the power minus k minus k z to the power minus i . If I see this term $1 - k$ equal to $1 - i - 1$ α_k $i-1$ z to the power minus k , it is nothing but A_{i-1} or naught A_{i-1} and then, if I push same thing in here also, if I say this is one term. Second term is this one. So, if the second term if I say k k and k . So, I can say k is here, then I can say k equal to $1 - i - 1$ α_k $i-1$ z to the power minus i plus k or I will do it here.

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So, I can write from here that is equal to 1 minus k equal to 1 to i minus 1 alpha k i minus 1 z to the power minus k. This is one equation. This part, this part and another equation is minus minus plus plus ki ki. I put the sum also here. So, summation of i is equal to k equal to 1 to i minus 1. This sum minus minus plus ki alpha i minus k i minus 1 z to the power minus k minus ki z to the power minus i ki z to the power minus i.

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

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

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

$A^i(z)$ (orange box) $A^i(z^{-1})$ (green box)

Now, if this here if I say that k dash is equal to i minus k, k dust is equal to i minus k. So, if I put k dust is equal to i minus k, so I can say this is k dust is equal to i minus k ki k

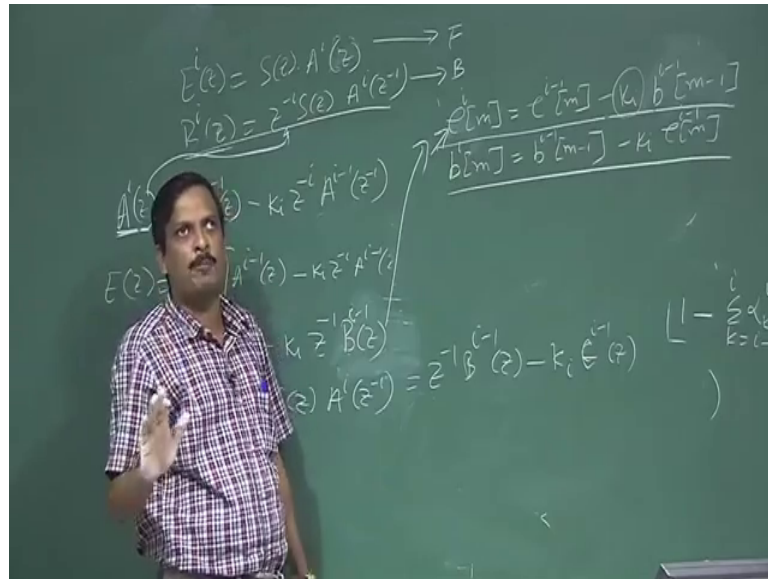
equal to 1 to i minus 1 and k dust is equal to i minus k . So, k dust is equal to i minus 1 to i if you see here. So, k dust is equal to i minus 1 to i alpha k i minus 1 . So, i minus k is equal to since this there is A i minus k , so I can say k dash is equal to i minus k if it is k dust is i minus k . So, what is k ? K is nothing but A k dust. So, if it is k varies from 1 to i minus 1 , k varies from 1 to i minus 1 . So, if I instead of k if I put it, this term is there class instead of k i , put k dust.

So, I can say k dust is equal to k minus 1 or k dust, sorry k dust is equal to i minus 1 to i alpha k . K i will be there, alpha k i minus 1 , k into alpha k i minus 1 into z to the power minus k ; so z to the power minus i plus k minus k i z to the power minus i . So, if I say this term, this term will be same. Then, this term will be k i . If I take minus sign here, then this will become beside k i z to the power minus i , if I take common, then what will happen is, 1 minus this term 1 minus k equal to i minus 1 i alpha k i minus 1 z to the power plus and this term. So, finally this will be 1 minus k equal to 1 to i minus 1 alpha k i minus 1 z to the power minus k minus this term.

So, the two term if you see the slides, there are two terms. One is this one and another is this one. So, if I say this is i minus 1 th order predictor, so I can say this is nothing but A i minus 1 z minus k i z to the power minus i into since it is z to the power plus k , then I can say it is nothing but A i minus 1 z to the power minus 1 A i minus 1 z to the power minus 1 . So, I can say that prediction predicted the error filter A i z can be expressed as A i minus 1 z previous iteration i minus 1 z minus k i z to the power minus i , A i minus 1 z to the power minus 1 . If it is that, then what is forward prediction errors? So, I can say e z or I can say e z is nothing but A i . So, instead of A i , I can put this one. So, sz into A i minus 1 z minus k i z to the power minus i A i minus 1 z to the power minus 1 . I just put the value of A i z . So, I can say it is nothing but sz into A i minus 1 z . So, it is e i minus 1 z minus k i into z to the power z to the power minus 1 A i minus 1 or I can say z to the power minus 1 digit. What is digit? It is z to the power minus i sz into A i z to the power minus 1 .

So, if it is c k i z to the power minus i , z to the power if I want to write in term of b z , I have to write z to the power minus 1 into z to the power i minus 1 . I take one outside. So, i minus 1 and minus 1 , then I can write A i minus 1 j to the power minus 1 . So, if you see these term, they are nothing but b z . So, I can say it is nothing but A z to the power minus 1 if it is b z . So, I can say the forward predict foreigner in time domain.

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If I express in time domain this terminology, I can say e_m or I can write in time domain e_m is e_{m-1} . So, I can say it is e_{m-1} forward prediction error minus k_i into z to the power minus 1. It is there to be e_{m-1} . So, it is e_{m-1} minus z . So, it is b_{m-1} minus z to the power minus 1 is here, I can write minus 1 here. So, I can say here or I can write clearly b_{m-1} minus 1 because z to the power minus 1 is there. So, b_{m-1} minus 1 sample delay b_{m-1} sample delay $m-1$. So, this is the forward prediction error.

Similarly, if you said B_i if it is B_i , B_i is equal to if I want to replace this z in here, then what I will get B_i is equal to z to the power minus i is z into A_i to the power minus 1. So, now I can put some A_i value instead of z to the power minus 1. I can put here. If I put that, then I get z to the power minus 1 b_{m-1} minus $k_i e_{m-1}$.

If I put this B_i to value this, this A_i value in here instead of z to the power minus 1. So, from basic question I can get b_m is nothing but $A b_m$ is nothing but $A b_{m-1}$. So, B_{m-1} , I have order prediction $m-1$ z to the power minus 1 is there minus k_i into e , that means e_{m-1} . So, this is called forward prediction error; this is called backward prediction error. So, I can easily say forward prediction error i th iteration, forward prediction error is equal to $i-1$ th iteration forward prediction error minus k_i into b_{m-1} minus 1, where k is that partial reflection coefficient, ok.

So, I can say the forward prediction error can be expressed in top of previous forward prediction error and backward prediction error. Similarly, backward prediction error can be expressed in top of previous backward prediction error and forward previous forward prediction error.

So, next class we want to discuss about how we draw the signal flow diagram, what should be the pictographic or signal flow diagram of these two equations, ok.

Thank you.