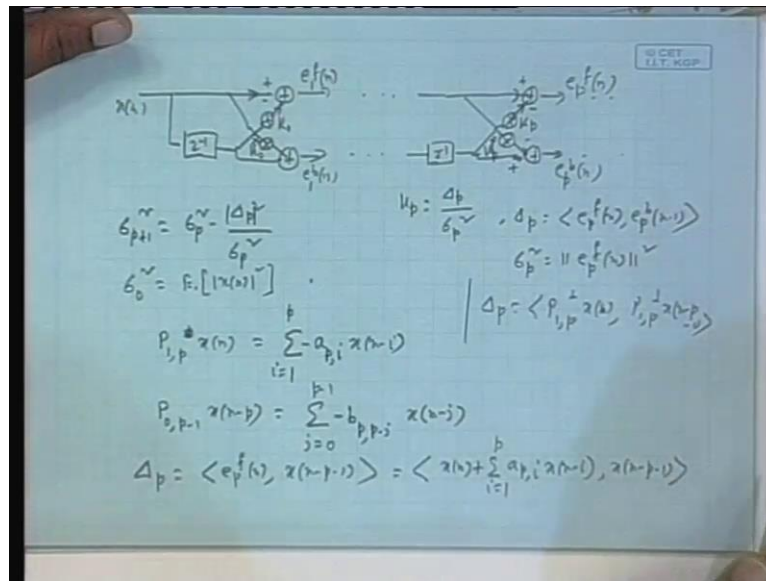


Adaptive Signal Processing
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Lecture - 24
Lattice as Optimal Filter

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Let us start with what we did last time. We had the lattice filter. It is had stage. This is just recap part actually. This has the coefficients K_0 , both the coefficients are same; one is the conjugate of the other; is it not? K_0, K_0^* . This is your $e_1^f(n)$; this is your $e_1^b(n)$, dot, dot, dot; dot, dot, dot, in general. Dot, dot, dot; this is your $e_p^f(n)$, $e_p^b(n)$, K_p, K_p^* , where K_p is Δ_p by σ_p^2 because both the forward and backward prediction error are same; variances we have seen.

So, I am not writing σ_p^2 and σ_p^2 ; no point, I told you that I will drop that index f for a superscript f (b). But, Δ_p was inner product between $e_p^f(n)$, $e_p^b(n)$ these are all standard stuff; I mean, what we have discussed. σ_p^2 is norm square of any of the 2. So, once the multiplied coefficients which are also called, either also called reflection coefficients.

K_p is the p th stage reflection coefficients. Once these are known you can always construct the lattice and go on filtering it; $x(n)$ you really get the prediction error for varies orders, for both forward and backward. But before that you have to obtain these multipliers and all that; what will be giving is the statistics of the process; in the sense the, of course, $x(n)$

is 0 means it will give you only the correlation, auto correlation function; or atleast many values of the auto correlation functions. Or, I mean, many values means auto correlation functions for many lack values; out of using that only you have to construct these.

Then, for σ_p^2 you have found out; this also that was the recursion that σ_{p+1}^2 was $\sigma_p^2 - \frac{\Delta_p^2}{\sigma_p^2}$, right. This is the one we did, is it not? But then, that time I told you; and, of course, how to start the recursion, σ_0^2 is nothing but $e^T e$ because $e^T e$ is $x^T x$. So, therefore, $e^T e$ is $x^T x$. So, just norm square of that, means, variances of that, but how about Δ_p ; that is the question? Because, you can go into the recursion, provided you also know Δ_p , the update Δ_p , because this σ_{p+1}^2 depend not only σ_p^2 , it depends upon the Δ_p also. So, I have to update Δ_p also. From Δ_p to Δ_{p+1} , then only the recursion can be proceed. So, that is what we will be doing now.

Now, I define this. So long we have been discussing the prediction errors. Also remember, we have been doing, I mean this prediction and all that, but what is the utility of the lattice and all, that still has not got. We have got a structure. What is the purpose of structure and all, we will let you know all. I will study its property and all. I mean its utility in various forms; will leave on that with you.

Now, so long we have been discussing when we have only the prediction errors, our projection errors. For a while, I will now considering not the error, but the projection only. That is x_n , as I earlier I projected x_n on the space span by x_{n-1} to x_{n-p} to the error, now I will not take error for a while, and that is the projection I mean for a while. So, that means, I will be considering things like this; $p-1$ not perpendicular, $p-1$ x_n ; getting the error is very simple, just subtracting from x_n ; is it not?

This will be nothing but a linear combination of those elements, x_{n-1} to x_{n-p} . So, it is a linear combination will be like this, $i=1$ to p . Then the coefficients, for some reasons I will introduce a minus sign in these coefficients; it is I just put a minus also. And coefficients will be called, denoted by a_i , or subscript will be i , but also with the p th order prediction. Suppose, instead of p , I use p th order prediction I will get a new set of coefficients; is it not?

So, that is why I am putting p subscripts, p comma i . These are also; I mean, you see, these are independent of n ; left hand side is function of n . You can say that how about n ? n goes simply because of our logic of stationarity; remember this are nothing but optimal filter coefficients after all. When you now give x_n as linear response, and x_{n-1} sequence of the input; obviously, the filter coefficients will be given by this combiner coefficients, because of stationarity.

You know there exact values also, but I will not go for that calculation that R inverse p and that here; I will recursion updating that. So, they are independent of n . Similarly, this is for forward prediction; for the backward prediction we have p_0 p minus 1, and x_n minus p was projected; x_n minus p was projected on the space span by p future terms; p future terms are standing at n minus 0 that is x_n ; x_n , x_{n-1} , upto x_{n-p+1} .

So, that projection, instead of the projection error again I am taking the projection; that again will be a linear combination of those elements which has erring the space. So, introduce j ; the range I will write later. Suppose, you write n minus j , so obviously, j has to go from 0 to p minus 1; a minus sign again I put deliberately.

Coefficient side denoted by b , but I will just, you know, to confirm to notation of the thing that is used, what I will do, first co index will be of course p because it is p th order prediction; instead of having second index as j I will call it p minus j ; that is I am just naming this, this way. When j equal to 0 that is x_n , this should be called $b_{p,p}$; while j equal to 1, x_{n-1} , this should be called $b_{p,p-1}$.

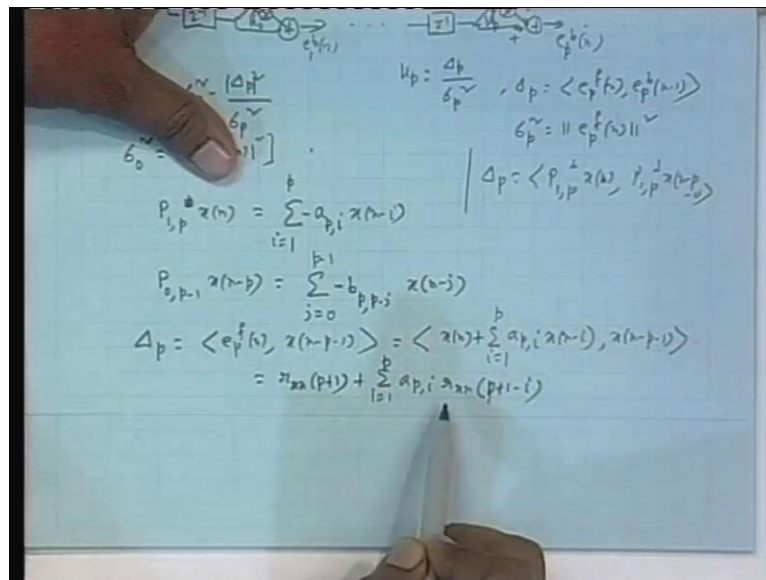
Unlike here, a p_1 for x_{n-1} , a p_2 x_{n-2} , a p_3 x_{n-3} , here, my, the person I am projecting is n minus p . But I am not taking the coefficient with n minus p plus 1; there is the immediate neighbor as $b_{p,1}$. I am taking that to be $e_{p,1}$, infact. You understood that clearly; x_n projected on space span by x_{n-1} , n minus 2, upto n minus p , the nearest 1 n minus 1, put the index 1 for that, next one put the index 2 for that, similarly n minus p .

So, when it is n minus p plus 1, that is when j equal to p minus 1 here, that time I want 1 to be here; project this p minus 2 that time I want 2 to be here. It is just a notation that is a p minus j . Suppose, these are known, the p f stage we are known, then obviously, Δp , what is Δp ? Δp we have seen already; but Δp can also be written as; can I? I am being doing this again and again.

What is e_p ? You understand how it comes? Those who are forgotten, for them, I am saying, I am writing here. Δp is nothing but, if you, you can write this e_p as $p \perp x$; and this is as $p \perp x$, is it not? I am being doing it again, again, again. And then, I will just drop this. Logic, you all know; several times I have explained. And then, again you call it e_p , that is what I have done here. And, e_p is what? X_n minus the projection.

What is e_p ? X_n minus this quantity. This I would write here. X_n and then it becomes plus, that is why I put a minus here; x_n minus this; and, minus and minus plus comma; and what is inner product? Inner product is correlation. So, what you can do? You take out this term; take out correlation of this term; forgot to, take the correlation; and what will you get in the first term? X_n into x_n minus $p \perp x$; that gives what?

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In terms of correlation, what is it? Using inner product, inner product means what, $e(x_n)$ into x_{n+1} ; that is the difference of the inner product; that is correlation. What is the lag? $p+1$. And then, other ones, a_p remain as it is; this and this, this will be R_{xx} , what, $n-i$ into x_{n+1} . So, you know the lag that will be $p+1-i$.

So, if you know this coefficients a_p 's, and if you know all the correlation values of the process x_n , then you can find out Δp . So, that is how Δp is to be computed, mind it. There is no past algorithm, recursive way of updating Δp directly; recursive way means efficient, computationally efficient.

Here you see, so much of computational is equal; as the value of p increases because you go higher stage number of operations in this because more and more, but this is something have to be I mean you cannot help. There are ways to paralyses this computational that I am not going to, but surely I will not go into recursive algorithm; but you understand, this computational I mean depends on p , and this you have to carry out. Σp^2 could be updated recursively, very simply. You do not want to really carry out Σp^2 computation elaborately and explicitly all terms; just recursively.