

**Dynamical Systems and Control**  
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**Lecture - 55**  
**Optimal Control for Discrete Systems - II**

Dear students. Welcome to this lecture on optimal control for discrete systems.

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$$J(k_0) = \sum_{k=k_0}^{k_0+N-1} F(x(k), x(k+1), k), \quad x(k) : k = k_0+1, \dots, k_0+N$$

$$x(k_0) = x_0, \quad x(k_0+N) = x_1$$

$$\frac{\partial F(k)}{\partial x(k)} + \frac{\partial F(k-1)}{\partial x(k)} = 0 \quad \text{subject to the boundary conditions.}$$

$k = k_0+1, k_0+2, \dots, k_0+N-1$

is the necessary condition for optimal solution

$x(k_0) = x_0$  is fixed  
 $x(k_0+N)$  is free.

Then the final boundary condition is  $\left. \frac{\partial F(k-1)}{\partial x(k)} \right|_{k=k_0+N} = 0$

So in the previous lecture, if you recall that we have studied the necessary condition for finding the optimal control of discrete performance index which is given by a functional  $J$  of  $k_0$  summation  $k$  is  $=k_0, k_0+N-1$  of a function of  $x$  of  $k, x$  of  $k+1$  and  $k$ . So we want to minimize this functional  $J$  such that the sequence  $x$  of  $k, k$  is  $=k_0+1, \dots, k_0+N$ . So this finite sequence minimizes this expression.

So if you are given the initial value of  $x$  of  $k$  as  $x_0$  and the final position  $x$  of  $k_0+N$  as  $x_1$  at initial instant  $k_0$  and the final instant  $k_0+N$  if the value of the vector function  $x$  is given as  $x_0$  and  $x_1$ , then how to minimize the functional  $J$  of  $k_0$  given in this expression so where  $F$  is a suitable smooth function. So we have seen that the necessary condition for solving this problem is the following.

The  $\frac{\partial F}{\partial x}$  of  $k$  /  $\frac{\partial F}{\partial x}$  of  $k+1$  /  $\frac{\partial F}{\partial x}$  of  $k$ . Here  $F$  of  $k$  means this one  $F$  of  $x$   $k+1$  and  $k$  and  $F$  of  $k-1$  it denotes wherever  $k$  is there in this expression we will put  $k-1$  that is  $F$  of  $x$  of  $k-1, x$  of  $k, k-1$  that expression is the second one. So this is  $=0$  and the boundary conditions

are given subject to the condition which are already given at  $k_0$  and  $k_0+N$  the conditions are given, so this is solved for various values of  $k$ ,  $k$  is  $k_0+1$  and  $k_0+2$  etc  $k_0+N-1$ .

So if you substitute  $k$  value all these values we will get a set of equations algebraic equation and we can solve it using these two conditions. So this gives the necessary condition for optimal solution for this problem. It is also called the Euler's equation. Now we will see if the boundary is also free if  $x$  of  $k_0=x_0$  is fixed as in the previous case and the final value  $x$  of  $k_0+N$  is free which is not specified.

In that case, we have seen that the condition is then the final boundary condition, it is  $\frac{\partial F}{\partial x}$  of  $k$  evaluated at  $k$  is  $k_0+N$  should be  $=0$ . So this derivation we have seen in the last lecture for the final boundary condition and the equation is the same. The equation with initial boundary condition is already given, final boundary condition is given by this expression so we can solve the system of equation and get the optimal solution.

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**Functional with Terminal Cost**

Let us formulate the cost functional with terminal cost as

$$J = J(x(k_0), k_0) = S(x(k_0 + N), k_0 + N) + \sum_{k=k_0}^{k_0+N-1} F(x(k), x(k+1), k) \quad (1)$$

given the initial condition  $x(k_0)$  and the final time  $k_0 + N$  as fixed, and the final state  $x(k_0 + N)$  as free. First, assume optimal (\*) condition and then consider the variational as

$$\begin{aligned} x(k) &= x^*(k) + \delta x(k) \\ x(k+1) &= x^*(k+1) + \delta x(k+1) \end{aligned} \quad (2)$$

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Now we will see that functional with terminal cost. So if the expression instead of the functional  $J$  only with the summation if you have an additional function, function of the final position. So if the final position is fixed, there is, it does not have any meaning because if  $x$  of  $k_0+N$  is a fixed value, then the function  $S$  of that expression will also be a fixed one. So minimizing  $J$  without the first expression  $S$  or with the first expression, both will give the same optimal solution.

But in case the  $x$  of  $k_0+N$  is a free boundary condition. So then the minimization of this expression has a meaning. So in this case, if you follow the similar procedure as we did in the previous lecture that is the optimal solution for this is  $x^*$  of  $k$  and the increment in  $x^*$  of  $k$  is  $\delta x$  of  $k$  that is  $x^*$  of  $k+\delta$  a variation in the function  $x$  of  $k$  and when we put  $k=k+1$  the next position  $x$  of  $k+1$  is the increment from  $x^*$  of  $k+1$  which is given by this.

And when we substitute this to  $x$  of  $k$ ,  $x$  of  $k+1$  in the expression of  $J$  and find the first variation of  $J$ .

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Then, the corresponding functionals  $J$  and  $J^\delta$  become

$$J = S(x^*(k_0 + N), k_0 + N) + \sum_{k=k_0}^{k_0+N-1} F(x^*(k), x^*(k+1), k) \quad (3)$$

$$\text{and } J^\delta = S(x^*(k_0 + N) + \delta x(k_0 + N), k_0 + N) + \sum_{k=k_0}^{k_0+N-1} F(x^*(k) + \delta x(k), x^*(k+1) + \delta x(k+1), k). \quad (4)$$



So when we substitute  $x$  of  $k$  that is the first one when we substitute  $x$  of  $k+1$  that is second one and if you take the difference between these two, we will get the variation in  $J$ .

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Following the same procedure as given previously for a functional without terminal cost, we get the first variation as

$$\begin{aligned} \delta J = & \sum_{k=k_0}^{k_0+N-1} \left[ \frac{\partial F(x^*(k), x^*(k+1), k)}{\partial x^*(k)} + \frac{\partial F(x^*(k-1), x^*(k), k-1)}{\partial x^*(k)} \right]' \delta x(k) \\ & + \left[ \frac{\partial F(x^*(k-1), x^*(k), k-1)}{\partial x^*(k)} \delta x(k) \right]_{k=k_0}^{k=k_0+N} \\ & + \frac{\partial S(x^*(k_0 + N), k_0 + N)}{\partial x^*(k_0 + N)} \delta x(k_0 + N). \end{aligned} \quad (5)$$



And using the Taylor series and neglecting the higher order terms, we get this expression. So this can be easily seen from the previous lecture in the similar manner. So we get this expression for the first variation and equating the first variation to 0 and observing that the variation  $\delta x_k$  is arbitrary, so we have to get this bracket should be=0 and the boundary conditions are given by this expression.

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For extremization, the first variation  $\delta J$  must be zero. Hence, from (5) the Euler-Lagrange equation becomes

$$\frac{\partial F(x^*(k), x^*(k+1), k)}{\partial x^*(k)} + \frac{\partial F(x^*(k-1), x^*(k), k-1)}{\partial x^*(k)} = 0 \quad (6)$$

and the transversality condition for the free-final point becomes

$$\left[ \frac{\partial F(x^*(k-1), x^*(k), k-1)}{\partial x^*(k)} + \frac{\partial S(x^*(k_0+N), k_0+N)}{\partial x^*(k_0+N)} \right]_{k=k_0+N} = 0. \quad (7)$$

$$\lambda(k_0) = \lambda(k)$$

So equation 6 is nothing but the square bracket within the summation sign that should be=0 because the variation is arbitrary and then this 2 expression should be=0. Now because at  $k_0$ , the  $x$  of  $k_0$  is given so the variation at the initial position is 0 and  $x$  of  $k_0+N$  is free, so the variation at  $k_0+N$  is arbitrary. Therefore, the expression will give  $\frac{\partial F}{\partial x}$  of  $k-1/\frac{\partial x}{\partial k}$  at  $k_0+N$ +this expression should be=0.

So we get the boundary condition to be like this at the point at the final instant  $k_0+N$ , the condition is this. So we have to solve 6 and 7 together along with the initial condition  $x$  of  $k_0=x_0$ , so we will get the solution of this thing optimal control provided it exist because these conditions are only necessary conditions and if the optimal control exist for this problem it can be solved through these equations. For example, we will consider the following.

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$$\text{minimize } J(k) = x^2(3) + \sum_{k=0}^2 (x(k)x(k+1) + x^2(k+1))$$

$$x(0) = 1, \quad x(3) \text{ is free.}$$
 Find the values of  $x(1), x(2), x(3)$  which minimize  $J(k)$

Here  $F(k) = x(k)x(k+1) + x^2(k+1)$ ,  $S(x(3)) = x^2(3)$

Soln:  $\frac{\partial F(k)}{\partial x(k)} + \frac{\partial F(k-1)}{\partial x(k)} = 0 \Rightarrow x(k+1) + x(k-1) + 2x(k) = 0$   
 $k = 1, 2$

$\frac{\partial F(k-1)}{\partial x(k)} \Big|_{k=3} + \frac{\partial S(x)}{\partial x(k)} \Big|_{k=3} = 0$   
 $x(k-1) + 2x(k) \Big|_{k=3} + 2x(k) \Big|_{k=3} = 0$   
 $x(2) + 2x(3) + 2x(3) = 0 \Rightarrow x(2) + 4x(3) = 0 \quad \dots \textcircled{1}$   
 $x(2) + x(0) + 2x(1) = 0 \quad \dots \textcircled{2}$   
 $x(3) + x(1) + 2x(2) = 0 \quad \dots \textcircled{3}$



So example so we will say minimize J of k which is given by x square 3+summation k varies from 0 to 2 of x of k\*x of k+1+x square of k+1. So here x of 0 is 1 and x of 3 is not specified. It is a free expression, free value here. So we have to find the sequence, find the values, the finite sequence. We want to find, x of 0 is already given, we want to find x of 1, x of 2 and x of 3 which minimizes the expression J of k whatever is given in the summation.

So for this we can adopt the procedure. The function capital F in the previous slide is here capital F of k is this expression. It is x of k, x of k+1+x square of k+1 and S function S of x of 3, it is=x square of 3. So that if you compare the expression in the previous J, we want to minimize J in this, so S is a function of the final position and final time. So in that particular example, it is x square of 3 where 3 is k0+N.

So in that problem k0 is 0 and N is capital N is 3, so we have the function S to be x square of 3. So now if you apply the condition, two conditions, so we have to solve the problem del F k/del x of k+del F k-1/del x of k=0. So this is the necessary condition. So del F/del x of k that gives from this expression, it is x of k+1 and if you put k=k-1 in this we get x of k-1, x of k and here x square of k.

So when we differentiate with respect to x of k, we will get + this will be x of k-1 and here + 2 times x of k from when we differentiate this. So this should be=0 for the values k is=1, 2. So we will get the expression and the boundary condition here is del F k-1/del x of k that is at k is=3 final condition+del S/del x of k that is also at k is=3, this should be=0. So that is this boundary condition.

$\frac{\partial F}{\partial x^k}$  when you put  $k-1$  in the expression  $\frac{\partial}{\partial x^k} \left( \frac{\partial S}{\partial x^k} \right)$  at the final position that should be  $=0$ . So if you take this expression  $F$  of  $k-1$  when we differentiate with respect to  $x^k$ , we will get  $x^{k-1}$  first and here  $+2$  times  $x^k$  at  $k=3$   $\frac{\partial S}{\partial x^k}$  that is  $2$  times  $x^k$  at  $k=3$  that is  $=0$ . So this condition will give when we put  $k=3$  we get  $x^{2+2}$  times  $x^3$  and here also we will get  $2$  times  $x^3$  that is  $=0$ .

So this implies  $x^{2+4}$  times  $x^3=0$ . So this is one equation we get, other equations are from here. When we put  $k=1$  in this expression, we will get  $x^2$  the first one, first  $k=1$  this will imply  $x^{2+x^0+2}$  times  $x^1$  that is  $=0$ . This is one equation. When we put  $k=2$  here  $x^3+x^{1+2}$  times  $x^2=0$ . This is equation 3. So if you solve this, now the unknowns are we want to find  $x_1, x_2, x_3$  and we have the 3 equations with 3 unknowns  $x_2, x_3$ , etc.

But  $x^0$  is known to us,  $x^0$  is 1 here, so here the second equation we get  $x^{2+2}$  times  $x^3=-1$  in that case. So we have 3 equations with 3 unknowns. By solving, we will get the sequence. So similarly, we can solve for more general expression of the summation. Here we note that this function  $x^k$ , it is a real-valued function but it can be a vector-valued function also. The  $x^1, x^2$ , all these functions may be a vector.

In that case, the derivatives will be with respect to the components. So if  $x$  is so if we have the component  $x$  is  $x_1, x_2, x_n$  the vector, in that case the equation will be in the form, this equations we will have if  $x$  is a vector  $\frac{\partial F}{\partial x^k}$  suffix  $i$  where  $i=1, 2, 3$ , up to  $N$ . So in this wherever  $x$  is there in the denominator, we have to write the  $x$  suffix  $i$  here and then we will get so many number of equations and it has to be solved in that corresponding manner.

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## Discrete Control Systems

Consider a linear, time-varying, discrete-time control system described by

$$x(k+1) = A(k)x(k) + B(k)u(k) \quad (8)$$

where,  $k = k_0, k_1, \dots, k_0 + N - 1$ .

Let the initial condition be

$$x(k = k_0) = x(k_0) = x_0 \quad (9)$$

So now we will consider the optimal control problem. Earlier, we have seen only the optimizing a cost function  $J$  but we want to optimize a cost function  $J$  under the constraint. So if we have the constraint as the control system, discrete control system like this  $x$  of  $k+1$  is  $A$  of  $k$  \*  $x$  of  $k$  +  $B$  of  $k$  \*  $u$  of  $k$  where  $u$  is the control function and  $x$  of  $k$  is the state of the system and the time instants are given as  $k$  is  $k_0, k_0+1, \dots, k_0+N-1$ .

So these are the time instant at which we evaluate this state of the system  $x$  of  $k$  and the initial condition is given as  $x$  at  $k_0 = x$  suffix 0, we can say like this.

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Let a general performance index with terminal cost be

$$\begin{aligned}
 J &= J(x(k_0), u(k_0), k_0) \\
 &= \frac{1}{2} x'(k_0 + N) \mathbf{F}(k_0 + N) x(k_0 + N) \\
 &\quad + \frac{1}{2} \sum_{k=k_0}^{k_0+N-1} \left[ x'(k) \mathbf{Q}(k) x(k) + u'(k) \mathbf{R}(k) u(k) \right] \quad (10)
 \end{aligned}$$

where,  $\mathbf{F}(k_0 + N)$  and  $\mathbf{Q}(k)$  are each  $n \times n$  order symmetric, positive semidefinite matrices, and  $\mathbf{R}(k)$  is  $m \times m$  symmetric, positive definite matrix.

$$\begin{aligned}
 &S(x(k_0+N)) \\
 &x(k) \in \mathbb{R}^n \\
 &u(k) \in \mathbb{R}^m
 \end{aligned}$$

So with this constraint if you want to minimize the expression  $J$ , the performance index  $J$  which is given by this expression, so what we observe here is earlier we have written a general function  $S$   $x$  of  $k_0+N$  a function of the final position. So say instead of a general

function here we are considering a linear expression of the function  $S$ . So  $S$  is given by  $x^T F x$  at the final time  $k_0 + N$ , so  $k_0 + N$  is the final instant.

So here  $F$  is a  $n \times n$  matrix and  $x$  is the column vector and  $x^T$  is the row vector. So when we multiply all this three we will get a scalar expression at the final instant  $k_0 + N$  this one+1/2 times summation  $k$  varies from  $k_0$  to  $k_0 + N - 1$  of this expression. Here also  $Q$  is a  $n \times n$  matrix and  $x$  is the vector in  $R^n$  and  $u$  is a vector in  $R^n$ . We have already introduced that  $x$  of  $k$  is a vector in  $R^n$ , it has  $N$  components.

And the control  $u$  of  $k$  it belongs to  $R^m$ , so the matrix  $R$  here is  $m \times m$  matrix and we assume that it is positive definite matrix and  $Q$  is a positive semi definite matrix,  $F$  is also a positive semi definite matrix, symmetric matrix. So these conditions are required in some practical situations, practical problems. Otherwise, we can replace them with a general function also as we have seen in the earlier expression.

Here we are seeing that there is no restriction on the function  $S$  and  $F$ , only thing is they have to be differentiable functions and we solve the necessary condition this Euler-Lagrange condition, so same thing applies to these equations also. Here it is a specific expression, we are taking the particular expression for the linear discrete control systems and we can solve in the similar manner.

So  $S$  function is replaced by the first expression given in this equation 10 and the capital  $F$  function in the previous problem is replaced by the linear expressions here,  $x^T Q x + u^T R u$  is a quadratic expression it is called. So for example if we replace  $Q$  with identity matrix, this is nothing but the norm of the vector  $x$  square and similarly  $R$  is replaced with identity we will get this as  $u^T u$  is nothing but the norm of  $u$  square  $u^T u$  of  $k$  square.

So it is a quadratic expression. So generally it is called a quadratic optimization problem for the discrete system.

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The methodology for linear quadratic optimal control problem.

**Augmented Performance Index:** The augmented cost functional using Lagrange multiplier  $\lambda(k+1)$  is

$$\begin{aligned}
 J_a = & \frac{1}{2} x'(k_0 + N) F(k_0 + N) x(k_0 + N) \\
 & + \frac{1}{2} \sum_{k=k_0}^{k_0+N-1} \left[ x'(k) Q(k) x(k) + u'(k) R(k) u(k) \right] \\
 & + \lambda(k+1) \left[ A(k) x(k) + B(k) u(k) - x(k+1) \right] \quad (11)
 \end{aligned}$$

Minimization of the augmented cost functional (11) is the same as that of the original cost functional (10), since  $J = J_a$ .

So if we adopt the same procedure because only the functions, expressions are written as a quadratic and some linear expression, other things are same. We can adopt the same procedure, only difference here is instead of just finding the optimum value of J, we also have the constraint given by the control system. So when there is a constraint, we adopt the Lagrange's multiplier method.

So along with the J function, we add plus a Lagrange's multiplier lambda at k+1\*the constraint that is  $x$  of  $k+1=A*x+B*u$  because that is 0 we are writing in any order, we are writing it as  $A*x+B*u-x$  of  $k+1$ . So when the optimal function  $x$  of  $k$  is minimizing the J as well as satisfying the constraint then the final term this last bracket is anyway going to be 0 because it will satisfy the constraint.

So the optimum value of  $J_a$ , the augmented function is same as the optimum value of the performance index J without this last term. So it is first we will optimize  $J_a$  and whatever solution we obtain that is the solution for the original problem of optimizing J along with this constraint.

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**Lagrangian:**

$$\begin{aligned} \mathcal{L}(x(k), u(k), x(k+1), \lambda(k+1)) \\ = \frac{1}{2} x'(k) \mathbf{Q}(k) x(k) + \frac{1}{2} u'(k) \mathbf{R}(k) u(k) \\ + \lambda'(k+1) \left[ A(k)x(k) + B(k)u(k) - x(k+1) \right]. \end{aligned} \quad (12)$$

So the Lagrangian is defined as this expression. The Lagrangian is the expression given here. Except the first term, this term and the next term the second and third term constitute the Lagrangian of the problem.

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**Euler-Lagrange Equations:** We now apply the Euler-Lagrange equation (6) to this new function  $\mathcal{L}$  with respect to the variable  $x(k)$ ,  $u(k)$ , and  $\lambda(k+1)$ . Thus, we get

$$\begin{aligned} \frac{\partial \mathcal{L}(x^*(k), x^*(k+1), u^*(k), \lambda^*(k+1))}{\partial x^*(k)} \\ + \frac{\partial \mathcal{L}(x^*(k-1), x^*(k), u^*(k-1), \lambda^*(k))}{\partial x^*(k)} = 0 \end{aligned} \quad (13)$$

$$\begin{aligned} \frac{\partial \mathcal{L}(x^*(k), x^*(k+1), u^*(k), \lambda^*(k+1))}{\partial u^*(k)} \\ + \frac{\partial \mathcal{L}(x^*(k-1), x^*(k), u^*(k-1), \lambda^*(k))}{\partial u^*(k)} = 0 \end{aligned} \quad (14)$$

And the necessary condition for the optimum value is  $\frac{\partial L}{\partial x_k} + \frac{\partial L}{\partial x_{k-1}} = 0$ . It is exactly similar to what we have seen earlier. Only thing is the L is replaced by capital F in the previous case because we have S function here+inside summation we have the capital F function if you compare it with the previous expression. So the necessary condition is in terms of the capital F function which comes inside the summation.

In the place of the capital F function, we are defining the Lagrangian capital L. So the condition is written in terms of L here,  $\frac{\partial L}{\partial x_k} + \frac{\partial L}{\partial x_{k-1}} = 0$ . Similarly, the

boundary condition is here instead of one variable in the previous problem now we have 3 variables. L is a function of not only, so L is not only a function of x alone like the previous one, it is also a function of u of k and it is also a function of lambda of k+1.

So there are 3 different functions involved here. So the procedure adopted in the previous one can be extended for all these 3 variables. So L is differentiated with respect to x of k. Similarly, L is differentiated with respect to u of k these two terms and L should be differentiated partially with respect to lambda of k also, is not it?

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$$\frac{\partial \mathcal{L}(x^*(k), x^*(k+1), u^*(k), \lambda^*(k+1))}{\partial \lambda^*(k)} + \frac{\partial \mathcal{L}(x^*(k-1), x^*(k), u^*(k-1), \lambda^*(k))}{\partial \lambda^*(k)} = 0 \quad (15)$$

and the boundary (final) condition (7) becomes

$$\left[ \frac{\partial \mathcal{L}(x(k-1), x(k), u(k-1), \lambda(k))}{\partial x(k)} + \frac{\partial S(x(k), k)}{\partial x(k)} \right]'_{k=k_0} \delta x(k) \Big|_{k=k_0}^{k=k_0+N} = 0 \quad (16)$$

where, from (10),

$$S(x(k_0+N), k_0+N) = \frac{1}{2} x'(k_0+N) \mathbf{F}(k_0+N) x(k_0+N). \quad (17)$$



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So with these 3 necessary conditions, we get the boundary condition in the following way. Exactly similar to the previous one, L at k-1 is partially differentiated with respect to del x k and the S function del S/del x of k at the final position. So this will give the boundary condition for because S is only a function of x of k not u of k or lambda of k, etc.

So only the boundary condition involves the derivative with respect to x of k so where the function S is given by the expression the linear expression which we have seen. So these equations if you see the equation 13, 14 and 15, 16, so they constitute the necessary condition for getting the optimal control problem. So a simpler way or simplifying the expression we can write or introduce the Hamiltonian function.

Hamiltonian function is defined by this expression. It is nothing but the Lagrangian function if you observe except the k+1 x of k+1 terms all the remaining terms are called the

Hamiltonian and so the Lagrangian is nothing but the Hamiltonian function-lambda dashed  $x$  of  $k+1$ .

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**Hamiltonian:** Now we give the conditions in terms of the Hamiltonian which is defined as

$$\mathcal{H}(x^*(k), u^*(k), \lambda^*(k+1)) = \frac{1}{2} x^{*T}(k) \mathbf{Q}(k) x^*(k) + \frac{1}{2} u^{*T}(k) \mathbf{R}(k) u^*(k) + \lambda^{*T}(k+1) [A(k)x^*(k) + B(k)u^*(k)] \quad (18)$$

Then, the Lagrangian (12) is given by

$$\mathcal{L}(x^*(k), u^*(k), x^*(k+1), \lambda^*(k+1)) = \mathcal{H}(x^*(k), u^*(k), \lambda^*(k+1)) - \lambda^{*T}(k+1) x^*(k+1) \quad (19)$$



So we define the Hamiltonian in this manner. So Lagrangian  $L$  is written as Hamiltonian-lambda dashed  $x$  of  $k+1$ . So here the star everywhere it denotes the optimal solution. So these equations will be satisfied at the optimal values of  $x$ , optimal values of  $u$  and optimal value of lambda, etc are substituted. So but when we solve the problems, we generally differentiate it with respect to the variables and then solve.

So the meaning of putting star in each equation is that at the optimal values all these equations are satisfied okay. So now  $L$  is written in terms of  $H$  in this way and converting the equations 13, 14, 15, 16 in terms of  $H$ , directly we can verify  $\frac{\partial L}{\partial u}$  for example it means  $\frac{\partial H}{\partial u}$  because the next term is not there. So the Lagrangian  $L$  is written in terms of  $H$  as in the equation 19.

Now converting these equations 13, 14, 15, 16 in terms of the Hamiltonian, so we can easily see that for example we get  $\frac{\partial L}{\partial u}$  if you want to calculate. So that is in this equation 19  $\frac{\partial L}{\partial u}$  is nothing but  $\frac{\partial H}{\partial u}$  because other term there is no  $u$  involved in the second term. So similarly, all the terms in terms of all the derivatives of  $L$  can be replaced by the derivatives of  $H$  from the equation 19.

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Now, the conditions (13)-(16) can be written in the following form

$$\lambda^*(k) = \frac{\partial \mathcal{H}(x^*(k), u^*(k), \lambda^*(k+1))}{\partial x^*(k)}, \quad (20)$$

$$0 = \frac{\partial \mathcal{H}(x^*(k), u^*(k), \lambda^*(k+1))}{\partial u^*(k)}, \quad (21)$$

$$x^*(k) = \frac{\partial \mathcal{H}(x^*(k-1), u^*(k-1), \lambda^*(k))}{\partial \lambda^*(k)}. \quad (22)$$

o

So we will get the system of equation in this way in terms of H. So  $\frac{\partial H}{\partial x_k} = \lambda^*(k)$  and  $\frac{\partial H}{\partial u_k} = 0$  and similarly  $\frac{\partial H}{\partial \lambda^*(k)} = x^*(k)$ . So these are the 3 necessary conditions we get and the boundary condition written from this expression. Boundary condition is given in the equation 16.

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The relation (22) can also be written at the next stage as

$$x^*(k+1) = \frac{\partial \mathcal{H}(x^*(k), u^*(k), \lambda^*(k+1))}{\partial \lambda^*(k+1)}. \quad (23)$$

For the system described by the equation (8) and the performance index (10) we have the relations (22), (20), and (21) for the state, costate, and control respectively as

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$$\left\{ \begin{array}{l} x^*(k+1) = A(k)x^*(k) + B(k)u^*(k) \\ \lambda^*(k) = Q(k)x^*(k) + A'(k)\lambda^*(k+1) \\ 0 = R(k)u^*(k) + B'(k)\lambda^*(k+1) \end{array} \right. \quad \begin{array}{l} (24) \\ (25) \\ (26) \end{array}$$

Now if you solve this system of equations with the boundary condition, we get the optimal control value here okay. So now converting actually substituting the expression of H, S, etc, so when we see that H is involving the matrices Q, R, A, B, etc. So converting actually substituting H and then differentiating with respect to those variables, we will get the system of equation to be like this.

We have to solve these 3 equations along with the boundary condition so that we will get the optimal control for the system.

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**Optimal Control:** The optimal control is then given by (26) as

$$u^*(k) = -R^{-1}(k)B'(k)\lambda^*(k+1) \quad (27)$$

where, the positive definiteness of  $R(k)$  ensures its invertibility. Using the optimal control (27) in the state equation (24) we get

$$\begin{aligned} x^*(k+1) &= A(k)x^*(k) - B(k)R^{-1}(k)B'(k)\lambda^*(k+1) \\ &= A(k)x^*(k) - E(k)\lambda^*(k+1) \end{aligned} \quad (28)$$

where,  $E(k) = B(k)R^{-1}(k)B'(k)$ .

So the optimal control from the equation we can see for example the equation 26 we can directly get the control optimal control u by taking the remaining terms to other side and writing u star of k in this manner. So we get the optimal control expression in terms of

lambda. Then, substituting this u expression in equation 24, then we will get a coupled equation in x and lambda.

So we will get the equation like this substituting u value from 27, we will get the equations like this.

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From the we get

$$\begin{bmatrix} x^*(k+1) \\ \lambda^*(k) \end{bmatrix} = \begin{bmatrix} A(k) & -E(k) \\ Q(k) & A'(k) \end{bmatrix} \begin{bmatrix} x^*(k) \\ \lambda^*(k+1) \end{bmatrix} \quad (29)$$

The slide shows a handwritten derivation of the coupled equations. It starts with the state equation  $x(k+1) = 2x(k) + u(k)$  and the cost function  $J = x^2(3) + \sum_{k=0}^2 (x^2(k) + u^2(k))$ . The Lagrangian is defined as  $L(k) = x^2(k) + u^2(k) + \lambda(k+1)[2x(k) + u(k) - x(k+1)]$ . The Hamiltonian is  $H(k) = x^2(k) + u^2(k) + \lambda(k+1)[2x(k) + u(k)]$ . The costate equation is  $S(k) = x^2(k)$ . The Euler-Lagrange equations are derived as  $\frac{\partial H(k)}{\partial x(k)} = \lambda(k) \Rightarrow \lambda(k) = 2x(k) + 2\lambda(k+1)$  and  $\frac{\partial H(k)}{\partial \lambda(k)} = x(k) \Rightarrow x(k) = 2x(k+1) + u(k+1)$ . The boundary condition is  $\frac{\partial L(k+1)}{\partial x(k)} \Big|_{k=3} + \frac{\partial S(k)}{\partial x(k)} = 0$ , which leads to  $\lambda(k) + 2x(k) = 0$  at  $k=3$ , and  $\lambda(3) + 2x(3) = 0$ . For  $k=1$ ,  $\lambda(1) = -\frac{\lambda(2)}{2}$ , and for  $k=2$ ,  $\lambda(2) = 2x(2) + 2\lambda(3)$  and  $x(2) = 2x(3) - \frac{\lambda(2)}{2}$ .

And finally we will get a coupled equation of this type and we can solve this equation along with the boundary condition we get the optimal control solution.

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The slide shows a handwritten derivation of the coupled equations. It starts with the state equation  $x(k+1) = 2x(k) + u(k)$  and the cost function  $J = x^2(3) + \sum_{k=0}^2 (x^2(k) + u^2(k))$ . The Lagrangian is defined as  $L(k) = x^2(k) + u^2(k) + \lambda(k+1)[2x(k) + u(k) - x(k+1)]$ . The Hamiltonian is  $H(k) = x^2(k) + u^2(k) + \lambda(k+1)[2x(k) + u(k)]$ . The costate equation is  $S(k) = x^2(k)$ . The Euler-Lagrange equations are derived as  $\frac{\partial H(k)}{\partial x(k)} = \lambda(k) \Rightarrow \lambda(k) = 2x(k) + 2\lambda(k+1)$  and  $\frac{\partial H(k)}{\partial \lambda(k)} = x(k) \Rightarrow x(k) = 2x(k+1) + u(k+1)$ . The boundary condition is  $\frac{\partial L(k+1)}{\partial x(k)} \Big|_{k=3} + \frac{\partial S(k)}{\partial x(k)} = 0$ , which leads to  $\lambda(k) + 2x(k) = 0$  at  $k=3$ , and  $\lambda(3) + 2x(3) = 0$ . For  $k=1$ ,  $\lambda(1) = -\frac{\lambda(2)}{2}$ , and for  $k=2$ ,  $\lambda(2) = 2x(2) + 2\lambda(3)$  and  $x(2) = 2x(3) - \frac{\lambda(2)}{2}$ .

So this can be demonstrated by a simple example for the equation x of k+1=2 times x of k+u of k, here k is=0, 1, 2 and we take the similar to the previous problem x of 0=1 and we want to find x of 1, x of 2 and x of 3 which minimizes the expression J which is given by x square

$3 + \sum_{k=0}^2 x_k^2 + u^2$ . So the performance index is given by this expression and the initial condition.

We have to find the finite sequence  $x_1, x_2, x_3$  for minimizing this under the constraint. So we can formulate the Lagrangian directly. The function  $L$  of  $k$  at the point  $k$  is so that is given by  $x_k^2 + u^2 + \lambda_{k+1} (x_{k+1} - x_k)$ . So this is Lagrangian expression we have. Then, we can write it in the form of the Hamiltonian. Hamiltonian at  $k$ , this is Lagrangian when we write  $k$  first.

So Hamiltonian is given by  $x_k^2 + u^2 + \lambda_{k+1} (x_{k+1} - x_k)$  except the  $x_{k+1}$  term others are Lagrangian. So now the function  $S$ ,  $S$  function is given by this thing.  $S$  of  $k$  we can write it as  $x_k^2$  that first term here. So the equations are given by so the necessary condition the Euler-Lagrange equation. So they are given by  $\frac{\partial H}{\partial u} = 0$  that is the first equation.

So  $\frac{\partial H}{\partial u}$  this implies  $2u$  and here sorry this is  $+$  here differentiate with respect to  $u$ ,  $u$  of  $k$  we will get  $\lambda_{k+1} = 0$ . So this is one equation. We can get  $u$  of  $k$  from in terms of  $\lambda_{k+1}$ . Then, the second equation is  $\frac{\partial H}{\partial x_k} = \lambda_k$ . If we recall that is the other equation. So this implies  $\lambda_k$  is  $\frac{\partial H}{\partial x_k}$  so that is nothing but differentiate with respect to  $x_k$ , it is  $2x_k$  here also it is  $2\lambda_{k+1}$  from the derivative.

So this is the second equation and the other equation will automatically give the same equation. We have  $\frac{\partial H}{\partial \lambda_k} = x_{k+1} - x_k$ . So this is the third equation. This implies  $x_{k+1} = x_k$  so when we write  $H_{k-1}$  everywhere, so we will get when we write  $k-1$  we will get here  $\lambda_k$  and the remaining things are  $2x_{k-1} + u^2$ . When we differentiate that with respect to  $\lambda_k$ , we will get  $2x_{k-1} + u^2$ .

So it is nothing but the same equation whatever constraint already given, the control system, it is just one step before. Instead of  $k+1$  we have  $k$  and the right hand side  $k$  is replaced by  $k-1$ . So it is the same equation, equation 3 and the boundary conditions can be written as given in the previous slides. So these are the equations which we wrote just now and the boundary condition we can have from this expression.



$\frac{\partial L}{\partial x_k} + \frac{\partial S}{\partial x_k}$ , so when in the L function we replace it with  $k-1$ . So first term is  $x^2$ , then  $u^2$ , etc and we have to differentiate with respect to  $x_k$ , so that will come only here. So the boundary condition is  $\frac{\partial L}{\partial x_k}$  at  $k=3$  that is the final time similarly  $\frac{\partial S}{\partial x_k}$  at  $k=3$  that will be 0. So  $\frac{\partial L}{\partial x_k}$  that is nothing but  $\lambda_k$ , only this term.

And  $\frac{\partial S}{\partial x_k}$  that is 2 times  $x$  of  $k=0$  at  $k=3$ , this is only at  $k=3$ . So this implies we get  $\lambda$  of  $3+2$  times  $x$  of  $3=0$ , this is the fourth equation. So now we can solve this equation 1, 2, 3, 4 and we will get the values of them, so we will write one by one. If you write  $k=1$  in all these equations, we will get a set like this. Similarly, so for example if you put  $k=1$  in all these equations that will give  $u$ .

So wherever  $u$  comes we can replace it with the  $\lambda$  value okay. So we will get  $u$  of  $1 = -\lambda/2$  so that can be substituted in the next equation and then  $\lambda$  of 1 that is given by  $2$  times  $x$  of  $1+2$  times  $\lambda$  of 2 and the next equation is  $x$  of  $1=2$  times  $x$  of 0 and here  $u$  of 0. So  $u$  of 0 is nothing but  $-\lambda$  of  $1/2$  from the equation 1. So we get a set like this. Similarly, we can write at  $k=2$  similar expression.

So by writing system of equation, we will get exactly 6 unknowns that is we want to find  $x_1$ ,  $x_2$ ,  $x_3$  and  $u_1$ ,  $u_2$  these 5 unknowns we want to find  $x_1$ ,  $x_2$ ,  $x_3$  and we want to find the control for this expression is  $u_0$ ,  $u_1$ ,  $u_2$ . These are the control because we do not want to find  $u_3$  because  $u_2$  the control given at the instant 2 will take the system to the step 3 from the given system of equations.

So we need to find  $u_0$ . At initial instant, the control will take the system to the first instant and control at first instant will take it to the second and second instant will take it to the third. So we want to find this and we want to find the state values  $x$  of 1,  $x$  of 2 and  $x$  of 3 which will minimize the expression  $J$ . So we have 6 unknowns and we will get 6 equations which can be solved in a usual manner and then get the optimal control for the problem okay. Thank you for listening.