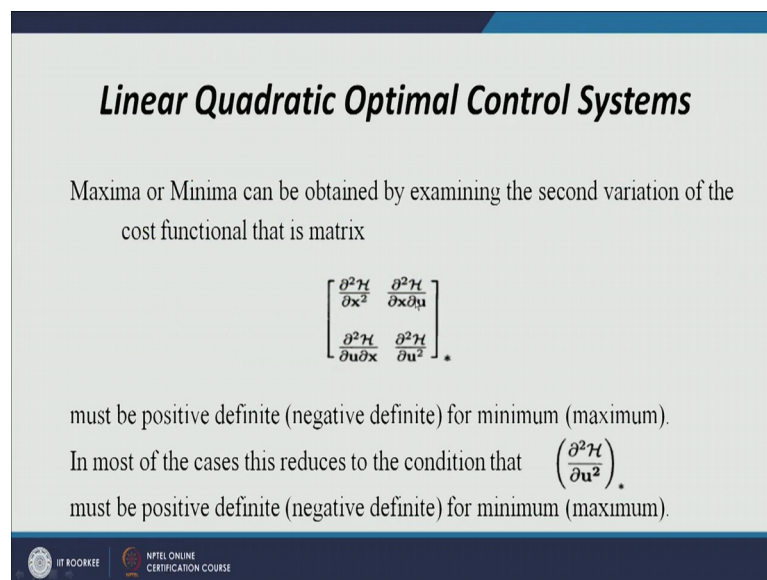


Optimal Control
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Lecture – 19
Linear Quadratic Optimal Control Systems
(Optimal Value of Performance Index)

Welcome class. In this class we will discuss about the how to determine the optimal value of a performance index. In the previous class we have discuss about the salient feature of matrix Riccati equation, as well as we have also discussed what is a sufficient condition in which we find if $\frac{\partial^2 H}{\partial u^2}$ is R and R is positive definite.

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Linear Quadratic Optimal Control Systems

Maxima or Minima can be obtained by examining the second variation of the cost functional that is matrix

$$\begin{bmatrix} \frac{\partial^2 \mathcal{H}}{\partial x^2} & \frac{\partial^2 \mathcal{H}}{\partial x \partial u} \\ \frac{\partial^2 \mathcal{H}}{\partial u \partial x} & \frac{\partial^2 \mathcal{H}}{\partial u^2} \end{bmatrix}_*$$

must be positive definite (negative definite) for minimum (maximum).

In most of the cases this reduces to the condition that $\left(\frac{\partial^2 \mathcal{H}}{\partial u^2}\right)_*$ must be positive definite (negative definite) for minimum (maximum).

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So, my matrix $\frac{\partial^2 H}{\partial x^2}$ $\frac{\partial^2 H}{\partial x \partial u}$, $\frac{\partial^2 H}{\partial u \partial x}$, and $\frac{\partial^2 H}{\partial u^2}$ must be a positive definite. With respect to a linear system if we will see my, this value is Q this half diagonal values are 0 and this value is r.

So my, this matrix is positive definite as Q is a positive semi definite and R is positive semi definite. So, we can say if we have $\frac{\partial^2 H}{\partial u^2}$ a positive definite, this means my we are minimizing the performance index in solving the optimal problem.

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$$\dot{X}(t) = A(t)X(t) + B(t)U(t)$$

Determine $J(x(t_0), t_0) = \frac{1}{2} x'(t_f) F(t_f) x(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [x'(t) B(t) x(t) + u'(t) R(t) u(t)] dt$

$$U^*(t) = -R^{-1}(t) B'(t) P(t) x(t)$$

MDRE
$$\dot{P}(t) + A'(t)P(t) + P(t)A(t) - P(t)B(t)R^{-1}(t)B'(t)P(t) + Q(t) = 0$$

$$\int_{t_0}^{t_f} \frac{d}{dt} (x'(t) P(t) x(t)) dt = [x'(t) P(t) x(t)]_{t_0}^{t_f} = x'(t_f) P(t_f) x(t_f) - x'(t_0) P(t_0) x(t_0)$$

$$J(x(t_0), t_0) = \frac{1}{2} x'(t_0) P(t_0) x(t_0) + \frac{1}{2} \int_{t_0}^{t_f} \left\{ x'(t) B(t) x(t) + u'(t) R(t) u(t) + \frac{d}{dt} (x'(t) P(t) x(t)) \right\} dt$$

We will see how to find out the optimal value of the performance index. My problem here is I have a system $\dot{x} = A x + B u$, and I have to find out the optimal u which will minimize my performance index, which will be defined as half of $x' F x$ plus integral from t_0 to t_f of $x' Q x + u' R u$. It is my finite time regulator problem, and this is $x' P x$, plus $u' R u$ and half we use here for our simplification.

So, my objective is to determine the optimal value of the u which will minimize the performance index, given by $J(x(t_0), t_0)$. So, my optimal u which we have found out that is $u = -R^{-1} B' P x$, where P is the solution of my matrix differential Riccati equation, which is so matrix, differential Riccati equation is what that is $\dot{P} = -P \dot{A} - A' P - P B R^{-1} B' P + Q = 0$. So, we write it in a different way. We try to simplify equation directly, $\dot{P} + A' P + P A - P B R^{-1} B' P + Q = 0$. This is my matrix differential equation. So, this is my optimal u to determine this u , I require P and P can be found out by the solution of given matrix differential Riccati equation.

So, in today's class my objective is if this value is what is the; I want to determine J . So, to find the solution of this problem, we just consider $\frac{d}{dt} (x' P x)$, into dt . What we can write for this. We can write for this because dt will be cancelled out. So, this is $\frac{d}{dt} (x' P x)$ with the limit t_0 to t_f .

So, this is $x^T P x$, which can be written as here sorry this is t_0 to t_f of $t_f P$ of $t_f x$ transpose, of $t_f P$ of $t_f x$ of t_f minus $x^T P x$ of t_0 . So, this term will have these 2 values. So, in the given performance index; so this J , I am writing in a different manner I am writing, J of t_0 to t_f as I am adding this term $x^T P x$ of t_0 of t_0 . And see I am writing this half, plus 1 by 2 t_0 to t_f , this term I will write $x^T Q x$, plus $u^T R u$. And I am adding this term d by $d t$ in this relation.

So, I am writing this and in this we are adding d by $d t$ of $x^T P x$, plus d by $d t$ of $x^T P x$ $d t$. Now, what this will have if I will explain this integral. So, this will give me x^T of P x of t_f P x of t_f . This term will be cancelled out and P t_f is nothing but my f of t_f . So, J written in this form or written in this form is the same value. This J of this J written in terms of the x prime f x t_f I can also write in this given form.

So, performance index we can write in this form here. So, if I will explain the integral of d by $d t$ of $x^T P x$, my this term will cancelled out and I will left only with the x^T of P x of t_f where P t_f is nothing but my f of t_f . Now to determine the value of the J we have to evaluate what we will have in this integral. So, we will separate out that term of the integral.

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$$\begin{aligned}
 & x'(t) R(t) x(t) + u'(t) R(t) u(t) \\
 &= x'(t) [R(t) + P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t) \\
 &= x'(t) [R(t) + P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t) \\
 & \left. \begin{aligned}
 u(t) &= -\bar{R}^{-1}(t) B'(t) P(t) x(t) \\
 \dot{x}(t) &= A(t) x(t) + B(t) u(t) \\
 &= [A(t) - B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t)
 \end{aligned} \right\} \\
 & \frac{d}{dt} [x'(t) P(t) x(t)] = x'(t) \dot{P}(t) x(t) + x'(t) P(t) \dot{x}(t) + \dot{x}'(t) P(t) x(t) \\
 &= x'(t) [A'(t) - P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t) + x'(t) \dot{P}(t) x(t) \\
 & \quad + x'(t) P(t) [A(t) - B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t) \\
 &= x'(t) [A'(t) P(t) - P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t) + \dot{P}(t) + P(t) A(t) - P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t) \\
 & J(x(t_0), t_0) = \frac{1}{2} x'(t_0) P(t_0) x(t_0) + \int_{t_0}^{t_f} x'(t) [R(t) - P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t) + A'(t) P(t) + P(t) A(t) - P(t) B(t) \bar{R}^{-1}(t) B'(t) P(t)] x(t) dt \\
 & \cdot J(x(t_0), t_0) = \frac{1}{2} x'(t_0) P(t_0) x(t_0)
 \end{aligned}$$

Let us first see; what is the value of $x^T Q x$ plus $u^T R u$.

So, here u we can write as, so we can take u^T as $-R^{-1}B^T P^T x^T$. So, if I will take u this, so value of this I can write as x^T transpose of t , Q^T plus if I have to take the transpose of this so x^T transpose will be the first term then $P^T R^{-1}$. So, this will be my x^T transpose I have taken outside. So, this will be nothing but $P^T B^T R^{-1}$ for the u^T transpose x . We have taken then this side this R^T , then u^T again directly I will write this so $R^{-1}B^T P^T$, and this I will write within this bracket.

So, this is equal to this one in this one R will be i . So, if I will write this is nothing but my identity matrix i . So, I can write x^T transpose $t^T Q^T$ plus $P^T B^T$ only one R^{-1} is left out, $R^{-1}B^T P^T x^T$. So, the first 2 term x^T prime, $Q^T x^T$ prime $R^T u$, I can write in this form Q^T plus $P^T B^T R^{-1} B^T$ prime p .

Now, if I will explain d by d^T of x^T transpose P^T , x^T this is nothing but my x^T dot transpose t , sorry $P^T x^T$ plus. So, first term I am taking the derivative of this second taking the derivative of p , this is $P^T \dot{x}^T$ and the third will be plus x^T transpose $t^T P^T \dot{x}^T$.

So, d by d^T of this can be written in this form, and here what is my x^T dot. If we will see x^T dot of t , is $A^T x^T$ plus $B^T u^T$ where u^T is $R^{-1}B^T P^T x^T$. So, this is nothing but A^T minus $B^T R^{-1}B^T P^T x^T$ of t . So, this is my x^T dot t . So, I will place my x^T dot t in the first and the last term.

So, this I can simply write as in place of x^T dot. So, if I am taking the x^T dot transpose. So, this I can write as x^T transpose of t because I have to take the transpose of the whole term this whole term means x^T transpose plus sorry multiplied with the transpose of second term and the transpose of this term is a transpose t minus the transpose of this term will be once you will transpose this so my first term comes out to be $P^T P^T$ is symmetric. So, is transpose remains the same B^T transpose will become $B^T R$ again in symmetric remains same and B will become the B^T transpose.

So, this is my x^T dot transpose multiplied with $P^T x^T$ second term, I left it same x^T transpose $t^T P^T \dot{x}^T$ plus, the third term x^T transpose of $t^T P^T$ and in place of x^T dot t I will place the value of this x^T dot which is A^T minus $B^T R^{-1}B^T P^T$ and x^T .

Now, in the next term we will write the whole expression in between x^T transpose of t and last x^T . So, I will take x^T transpose of t from here. So, this is left by and t we are

multiplying into this in the first equation, this is a transpose $t^T P t$ minus, $P^T B, R$ inverse B transpose p . So, by expanding this I will get this; this x we will take some outside the bracket. So, by expanding the first term we got this from here we got only the P dot of t , from the second term plus from the third term x transpose taken outside P , I will multiply plus $P a$ and minus P , sorry $P B R$ inverse B transpose $P t$, and last term will be $my x t$.

So, if I will see my first 2 term are in terms of the x transpose x , and d by $d t$ term I am also writing in terms of x transpose of x ; so to get the integral second term. Now, what I can write for this j . Now what I have to do? I have to at this term plus this term with $d t$. So, once we will add these 2 terms. So, we will get j as my first term was one by $2 x^T P t^T x^T$ transpose of the first term, plus integral t_0 to t_f and these 2 term I have to add. And this I am adding in this way x transpose of t , I will write here from the first 2 term I will have, $Q t$ plus $P B R$ inverse B transpose p .

So, I am getting this term from here and from here if I will see plus a transpose P once I will write. So, this will be the 2 times twice of P plus I will write first this as the P dot. Next term will be a P which I am getting from sorry this will be $P^T A t P a$ and minus $2 P B R$ inverse B transpose $P t$ and $x t d t$.

So, if I will further explain this. Now, you will see this is the $2 P B, R$ inverse B prime $P B R$ inverse B prime p . So, in a way what I can write I can remove this, and this I can make as minus, and I can write this into $d t$. So, now, what you see what I will have in this, P transpose sorry P dot t plus a transpose $P a$ minus $P B R$ inverse B prime P plus $Q t$ and this is nothing but my matrix differential equation which value is 0. So, all this value is nothing but 0; so half $P t^T x^T t_0$.

So, I will get j optimal value of performance index is nothing but half of x transpose $t_0^T P t_0^T x^T t_0$. So, what we have done once we will see.

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Optimal Performance Index

First let us note that

$$\int_{t_0}^{t_f} \frac{d}{dt} (\mathbf{x}^{*'}(t)\mathbf{P}(t)\mathbf{x}^*(t)) dt = -\frac{1}{2}\mathbf{x}^{*'}(t_0)\mathbf{P}(t_0)\mathbf{x}^*(t_0) + \frac{1}{2}\mathbf{x}^{*'}(t_f)\mathbf{P}(t_f)\mathbf{x}^*(t_f)$$

Substituting $\frac{1}{2}\mathbf{x}^{*'}(t_f)\mathbf{P}(t_f)\mathbf{x}^*(t_f)$


into the **PI** noting that $P(t_f) = F(t_f)$



We have taken the d by d t of this expanding as x transpose t 0 P t 0 x t 0, with condition that P of t f equal to f of t f. So, this means if I will replace this so this is showing me nothing but my terminal cost.

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Optimal Performance Index

$$\begin{aligned} J^*(\mathbf{x}^*(t_0), t_0) &= \frac{1}{2}\mathbf{x}^{*'}(t_0)\mathbf{P}(t_0)\mathbf{x}^*(t_0) \\ &+ \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}^{*'}(t)\mathbf{Q}(t)\mathbf{x}^*(t) + \mathbf{u}^{*'}(t)\mathbf{R}(t)\mathbf{u}^*(t) \\ &+ \frac{d}{dt} (\mathbf{x}^{*'}(t)\mathbf{P}(t)\mathbf{x}^*(t))] dt \\ &= \frac{1}{2}\mathbf{x}^{*'}(t_0)\mathbf{P}(t_0)\mathbf{x}^*(t_0) \\ &+ \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}^{*'}(t)\mathbf{Q}(t)\mathbf{x}^*(t) + \mathbf{u}^{*'}(t)\mathbf{R}(t)\mathbf{u}^*(t) \\ &+ \dot{\mathbf{x}}^{*'}(t)\mathbf{P}(t)\mathbf{x}^*(t) + \mathbf{x}^{*'}(t)\dot{\mathbf{P}}(t)\mathbf{x}^*(t) \\ &+ \mathbf{x}^{*'}(t)\mathbf{P}(t)\dot{\mathbf{x}}^*(t)] dt. \end{aligned}$$


In my performance index, I am substituting I am adding this half of x transpose t 0 P t 0 x t 0 and including d by d t of x transpose P x into the integral part.

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Optimal Performance Index

Now, using the state equation

$$J^*(\mathbf{x}^*(t_0), t_0) = \frac{1}{2} \mathbf{x}^{*T}(t_0) \mathbf{P}(t_0) \mathbf{x}^*(t_0) + \frac{1}{2} \int_{t_0}^{t_f} \mathbf{x}^{*T}(t) [\mathbf{Q}(t) + \mathbf{A}'(t) \mathbf{P}(t) + \mathbf{P}(t) \mathbf{A}(t) - \mathbf{P}(t) \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}'(t) \mathbf{P}(t) + \dot{\mathbf{P}}(t)] \mathbf{x}^*(t) dt$$

Finally, using the matrix DRE in this relations, the integral part becomes zero

$$J^*(\mathbf{x}(t_0), t_0) = \frac{1}{2} \mathbf{x}^{*T}(t_0) \mathbf{P}(t_0) \mathbf{x}^*(t_0)$$

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By expanding this as we have seen here we got nothing but half of $\mathbf{x}^T(0) \mathbf{P}(0) \mathbf{x}(0)$ plus half of $\mathbf{x}^T(0) \mathbf{Q}(0) \mathbf{x}(0)$ plus a transpose $\mathbf{P}(0) \mathbf{A}(0) \mathbf{x}(0)$ minus $\mathbf{P}(0) \mathbf{B}(0) \mathbf{R}^{-1}(0) \mathbf{B}'(0) \mathbf{P}(0) \mathbf{x}(0)$ plus $\dot{\mathbf{P}}(0) \mathbf{x}(0)$. And if you will see this is nothing but my matrix differential Riccati equation which value is 0. So, my this whole term will become 0 some left with the initial cost as $\frac{1}{2} \mathbf{x}^T(0) \mathbf{P}(0) \mathbf{x}(0)$.

So, this expression gives me nothing but my optimal value of the objective function. Now as we have solve this from t_0 to t_f . Similarly, I can solve it from t to t_f for any intermediate point.

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Optimal Performance Index

Now, the previous relation is also valid for any $x^*(t)$. Thus,

$$J^*(\mathbf{x}^*(t), t) = \frac{1}{2} \mathbf{x}^{*T}(t) \mathbf{P}(t) \mathbf{x}^*(t)$$

In terms of the final time t_f the previous optimal cost becomes

$$J^*(\mathbf{x}(t_f), t_f) = \frac{1}{2} \mathbf{x}^{*T}(t_f) \mathbf{P}(t_f) \mathbf{x}^*(t_f)$$

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So, if I am solving it for t to t_f my optimal cost will be $\mathbf{x}^T \mathbf{P} \mathbf{x}$, because t is varying from t_0 to t_f even I can also express my optimal cost as half of $\mathbf{x}^T \mathbf{P} \mathbf{x}$. So, by this we can determine the optimal value of my objective function, and as my second variation is 0, R is positive definite. So, this means this optimal cost is nothing but the minimal cost of my performance index.

In the next class we will start with the infinite time lQR system. So, I stop my discussion for this class here.

Thank you very much.