

Dynamical Systems and Control
Prof. N. Sukavanam
Department of Mathematics
Indian Institute of Technology – Roorkee

Lecture - 46
Introduction to Discrete Systems - II

Hello viewers. Welcome to this lecture on discrete systems. So in this lecture, we will see some results on the state transition matrix, solution of discrete system and the controllability of discrete systems which is these results are analogous to the continuous dynamical systems and control systems.

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$$\frac{dx}{dt} = Ax + Bu \quad (1)$$

$$x(t) = \phi(t, t_0)x_0 + \int_{t_0}^t \phi(t, s)Bu(s)ds$$

$$\frac{dx}{dt} = Ax$$
 Let $t_0 = k_0 h$: k_0 is an integer
 h is interval
 $t_i = (k_0 + i)h$: $i = 0, 1, 2, \dots$

If $u(t) = u(t_i)$ $t_i \leq t < t_{i+1}$
 then the solution at $t = t_1$
 $x(t_1) = \phi(t_1, t_0)x_0 + \int_{t_0}^{t_1} \phi(t_1, s)Bu(s)ds$
 $= e^{A(t_1 - t_0)}x_0 + \int_{t_0}^{t_1} e^{A(t_1 - s)}Bu(s)ds$
 $= e^{Ah}x_0 + \int_{t_0}^{t_0+h} e^{A(h-\theta)}Bu(\theta)ds$

$x(k+1) = A(k)x(k) + B(k)u(k) \quad (2)$
 $k = k_0, k_0+1, \dots, k_0+N$
 $x(k+1) = d \cdot x(k)$
 $= e^{Ah}x_0 + \left(\int_0^h e^{A(h-\theta)}B d\theta \right) u(t_0)$
 $x(t_1) = E x_0 + F \cdot u(t_0)$
 $x(t_2) = \phi(t_2, t_1)x(t_1) + \int_{t_1}^{t_2} \phi(t_2, s)Bu(s)ds$
 $= e^{Ah}x(t_1) + \left(\int_{t_1}^{t_1+h} e^{A(h-\theta)}Bu(\theta)ds \right)$
 $= E \cdot x(t_1) + F \cdot u(t_1)$
 $x(k+1) = E \cdot x(k) + F \cdot u(k)$

So earlier we have seen the continuous control system of the form $dx/dt = Ax + Bu$, so here A maybe a constant matrix or time variant matrix, similarly, B maybe constant or time variant. So this continuous control system has solution x of $t = \phi(t, t_0)x_0 + \int_{t_0}^t \phi(t, s)Bu(s)ds$. So this is the solution of the control system for the initial condition x of $t_0 = x_0$.

Now we have also introduced some discrete control system or discrete dynamical system which is of the form $x(k+1) = A(k)x(k) + B(k)u(k)$ is a discrete control system where k is varying from k_0, k_0+1, \dots, k_0+n . So we have the discrete control system in this form. So we call it as 1 and this as 2. Now how to write the solution of the discrete system and we can also see what is the connection between a continuous control system and a discrete control system.

So we can formulate various discrete control system on its own, various problem like population dynamics can be for example written as $x_{k+1} = \text{some constant times } x_k$. So this maybe representing or a mathematical model for a population dynamics or some such similar problems as we have seen population dynamics for a simple model of population dynamics in terms of a continuous dynamical system can be written like this where A is the rate of increase of the population.

And here also α is the rate of increase of the population and so it gives a simple mathematical model whether it is continuous or discrete system but for certain problems discrete systems are more suitable and certain problems continuous models are more suitable. Similarly, if we have a continuous model and if you want to solve it using a digital computer, in that case we have to convert this continuous system into a discrete system.

So we can also find such relation in the following way. So if we consider let t_0 that is the initial time instant which we have taken if you write it as $k_0 \text{ times } h$ where h is a small time instant which we take and k_0 is an integer such that k_0 is an integer and h is increment in time and if you are interested in the values of x of t at various instances of time, let us say let t_i it is $k_0 + i \cdot h$ for i is $0, 1, 2, 3$, etc.

So when we substitute i values differently, we will get the discrete points from t_0, t_1, t_2 , etc and the time increment is h here this. So now if u of t because while doing or while applying a control in a practical problem, it is not possible to give the control in a continuous fashion but we can give it in a discrete time instance. So u of t_i we assume that it is u of t is u of t_i for the interval $t_i \leq t < t_{i+1}$.

So if we assume this way, the control is applied in a discrete fashion for this continuous control system. Then, the solution of the system can be written in this following way. Solution for example at u of t_1 , the solution at $t=t_1$ that is x of t_1 is written as ϕ of $t_1, t_0 \cdot x_0 + \int_{t_0}^{t_1} \phi$ of t_1, s Bu s ds which is same as e to the power $A \cdot t_1 - t_0 \cdot x_0 + \int_{t_0}^{t_0+h}$.

This is e to the power $A \cdot t_0 + h - s$ B . Now u s in the interval t_0 to t_1 is u of t_0 because we are assuming that u of t is measured in a discrete fashion. So u of $t = u$ of t_i in this interval, so we substitute u of s to be like this in the interval t_0 to t_0+h and this is nothing but e to the power

$Ah^{t_1-t_0}$ is h^{x_0} here now we can convert this into the following way, s is between t_0 to t_1 so if you substitute $s = t_0 + h\theta$ where θ is between $0 \leq \theta < 1$.

Because t is or s is between t_0 to t_1 , so we can substitute this way. So if we substitute $s = t_0 + h\theta$ here, t_0 gets canceled so this will be $e^{A(t_1-t_0)\theta}$ and this is varying from t_0 to t_0+h $B ds$ of t_0 this expression. So this further can be written as the right hand side is $e^{A(t_1-t_0)\theta}$ now if we convert this integral in terms of $d\theta$ here, so $ds = h d\theta$ and when $s = t_0$ θ is 0.

And here it is 0 to h $e^{A(t_1-t_0)\theta} B d\theta$ of t_0 . So similarly we can so we can call this as the matrix E , $e^{A(t_1-t_0)}$ can be called as a matrix E^{x_0} and integral 0 to h of this expression we can call it as F of t_0 . So in the similar way, we can calculate x of so this is nothing but x of t_1 is given by this expression. Similarly, we can write x of t_2 in the place of t_1 if you put t_2 , here it will be ϕ of t_2, t_1 .

Because initial position is t_1 , initial time is t_1 , final time is t_2 x of $t_1 + \int_{t_1}^{t_2}$ of this similar expression ϕ of t_2, s B ds . So here u value is u of t_1 that is a constant throughout the interval t_1 to t_2 . So that will be out of the integration. So again we can see the same thing, in the place of $e^{A(t_1-t_0)}$, so it will also be $e^{A(t_2-t_1)}$ x of t_1 here also we will get the same thing $e^{A(t_2-t_1)}$ $B d\theta$ and u of s is u of t_1 .

So we can see that the expression is again E matrix x of t_1 this matrix is same F of t_1 . So if you replace this t_1, t_2, t_3 , etc by some suffix itself. So this implies that in general we can write x of t suffix $k+1$ if you write but we can replace it by x of $k+1$ just as a notation x of $k+1$ it means x of t suffix $k+1$ that is E^{x} of k this matrix x at t suffix $k+1$ F of k . So we will get a discrete control system in this particular fashion.

And this h is the time instant at which we find the values of x as well as use the control at these time instances for this control system. So when we solve a continuous control system using a computer, we can convert that system into a discrete system and then solve in a discrete fashion. So this will give various simpler results for solving a continuous control system in the form of discrete system as follows. So here we will see a various analogous results as in the case of continuous system.

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Autonomous System

Consider the discrete autonomous system

$$x(k+1) = Ax(k), \quad k = k_0, k_0 + 1, \dots, k_0 + N \quad (1)$$

Here A is $n \times n$ constant matrices, $x(k) \in \mathbb{R}^n$ for all k .
If the initial condition is $x(k_0) = x_0$ then we get

$$\begin{aligned} x(k_0 + 1) &= Ax(k_0) \\ x(k_0 + 2) &= Ax(k_0 + 1) \\ &= A^2x(k_0) \end{aligned}$$

For example, the state transition matrix we have defined for a continuous system earlier, so similarly we can define it here. So consider the simple system $x(k+1) = Ax(k)$ where the time instances are $k = k_0, k_0 + 1, \dots$. Now repeatedly applying the equation (1) except $k_0 + 1$ is $Ax(k_0)$ and $x(k_0 + 2)$ is $A^2x(k_0)$.

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$$\begin{aligned} x(k_0 + N) &= A^N x(k_0) \\ &= A^{(k_0 + N - k_0)} x(k_0) \end{aligned} \quad (2)$$

Put $k = k_0 + N$ and $\Phi(k, l) = A^{k-l}$ in (2) we get

$$x(k) = \Phi(k, k_0)x(k_0) \quad (3)$$

And similarly $x(k_0 + N)$, it can be easily seen that it is A to the power N $x(k_0)$, N can be written as $k_0 + N - k_0$. So $x(k_0 + N) = A^N x(k_0)$. N can be written as $k_0 + N - k_0$. So we get and if you substitute $k = k_0 + N$ and the notation $\Phi(k, l)$ means A to the power $k-l$. If we use this notation, then we can write the solution $x(k) = \Phi(k, k_0)x(k_0)$. So it looks similar to the continuous system state transition matrix. So we will see the analogous result of that one.

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State Transition Matrix

The state transition matrix is defined by

$$\Phi(k, k_0) = A^{k-k_0} \quad (4)$$

It is easy to verify that

$$\begin{aligned} \Phi(k+1, k_0) &= A\Phi(k, k_0) \\ \Phi(k, k) &= I \\ \Phi(k_0, k) &= \Phi^{-1}(k, k_0), \text{ provided } A \text{ is nonsingular} \\ \Phi(k, k_0) &= \Phi(k, k_1)\Phi(k_1, k_0), \quad k \geq k_1 \geq k_0 \end{aligned}$$

$$\phi(t, t_0)$$

$$\frac{d}{dt} \phi(t, t_0) = A \phi(t, t_0)$$

$$\left(A^{k-k_0} \right)^{-1} = A^{k_0-k}$$

Earlier, we have seen $\phi(t, t_0)$ is the state transition matrix for a continuous system and if it is a constant matrix we have seen it is equal to $e^{A(t-t_0)}$, otherwise we use the Peano-Baker series to find the state transition matrix but here in the discrete system if A is a constant matrix, the state transition matrix is $\phi(k, k_0) = A^{k-k_0}$ and the property of continuous system we have seen d/dt of $\phi(t, t_0) = A \phi(t, t_0)$.

So similar result in the discrete is $\phi(k+1, k_0) = A \phi(k, k_0)$ that can be directly seen from here. If you put $k+1$ here, it will be A to the power $k+1-k_0$ which is A times A to the power $k-k_0$ and this first one and if you put both variables are same $k=k_0$ we will get A to the power 0 which is the identity matrix and if you reverse this A^{k_0-k} that is nothing but A^{k-k_0} its inverse is nothing but A^{k_0-k} .

So we get this third result and fourth can be easily verified. So these 4 properties are analogous to the continuous case.

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Autonomous Control System

Consider the discrete autonomous system

$$x(k+1) = Ax(k) + Bu(k), \quad k = k_0, k_0 + 1, \dots, k_0 + N \quad (5)$$

Here A and B are $n \times n$ and $n \times m$ constant matrices respectively, $x(k) \in \mathbb{R}^n$ and $u(k) \in \mathbb{R}^m$ for all k .

If the initial condition is $x(k_0) = x_0$ then we get

$$\begin{aligned} x(k_0 + 1) &= Ax(k_0) + Bu(k_0) \\ x(k_0 + 2) &= Ax(k_0 + 1) + Bu(k_0 + 1) \\ &= A(Ax(k_0) + Bu(k_0)) + Bu(k_0 + 1) \\ &= A^2x(k_0) + ABu(k_0) + Bu(k_0 + 1) \end{aligned}$$

Now if you consider the control system x of $k+1=Ax$ $k+B$ u of k for these time instances, it can be easily verified by substituting one by one starting from k_0 , so x of k_0+1 is Ax k_0+Bu k_0 and substituting k_0+2 here, here it will be k_0+1+Bu of k_0+1 , etc.

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$$\begin{aligned} x(k_0 + 3) &= A^3x(k_0) + A^2Bu(k_0) + ABu(k_0 + 1) + Bu(k_0 + 2) \\ x(k_0 + N) &= A^Nx(k_0) + A^{N-1}Bu(k_0) + A^{N-2}Bu(k_0 + 1) \\ &\quad + \dots + Bu(k_0 + N - 1) \\ &= A^{(k_0+N-k_0)}x(k_0) + A^{(k_0+N)-(k_0+1)}Bu(k_0) \\ &\quad + A^{(k_0+N)-(k_0+2)}Bu(k_0 + 1) + \dots + Bu(k_0 + N - 1) \quad (6) \end{aligned}$$

Put $k = k_0 + N$ and $\Phi(k, l) = A^{k-l}$ in (6) we get

$$x(k) = \Phi(k, k_0)x(k_0) + \sum_{i=k_0}^{k-1} \Phi(k, i+1)Bu(i) \quad (7)$$

$$x(t) = \phi(t, t_0)x_0 + \int_{t_0}^t \phi(t, \tau)Bu(\tau)d\tau$$

So we will get the formula easily Ax of k_0+N we will get A to the power N x of k_0 +this quantity which can be written in the similar as we have seen in the previous case. N can be written as k_0+N-k_0 , so by using the notation ϕ of k, l is A to the power $k-l$ we get the formula x of k where k is k_0+N is=this is ϕ of k, k_0 this into x of k_0 +the remaining portion can be written in the summation i is= k_0 to $k-1$ of ϕ of $k, i+1$ * Bu i .

So this is the formula and it is analogous to the formula for the continuous case x of t is ϕ of t , t_0 $x_0 + \int_{t_0}^t \phi$ of t, s $Bu s ds$. So instead of integral, we have summation and the discrete points we are using for the formula for the solution of a discrete system.

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Nonautonomous System

Consider the discrete system

$$\begin{aligned} x(k+1) &= A(k)x(k), \quad k = k_0, k_0 + 1, \dots, k_0 + N \\ x(k_0) &= x_0 \end{aligned} \quad (8)$$

then

$$\begin{aligned} x(k_0 + 1) &= A(k_0)x(k_0) \\ x(k_0 + 2) &= A(k_0 + 1)x(k_0 + 1) \\ &= A(k_0 + 1)A(k_0)x(k_0) \\ &= A(k_0 + 1)A(k_0)x(k_0) \end{aligned}$$



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Now if you take non-autonomous system where A is not a constant, we can derive in a similar fashion, only thing is A of k should be replaced at every place. So x of k_0+1 is A of $k_0 * x$ of k_0 , x of k_0+2 we can get A of k_0+1 A of $k_0 * x$ of k_0 . So for k_0+N it is clear that the right hand side we will start with A of k_0+N-1 and then $N-2$, etc up to the initial point k_0 .

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$\phi(k_0, k)$
 $\phi(k_0, k)$



$$x(k_0 + N) = \Phi(k_0 + N, k_0)x(k_0) \quad (9)$$

where

$$\Phi(k_0 + N, M) = A(k_0 + N - 1)A(k_0 + N - 2) \dots A(M), \quad M \leq k_0 + N$$

Put $k = k_0 + N$, we get

$$\begin{aligned} \Phi(k, M) &= A(k-1)A(k-2) \dots A(M) \\ x(k) &= \Phi(k, k_0)x(k_0) \end{aligned} \quad \begin{matrix} k > M \\ \phi(M, k) \end{matrix}$$



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So we will get this result x of $k_0 + N$ will be A of k_0+N-1 and k_0+N-2 , etc up to the initial point that is k_0 and then multiplied by x of k_0 . So in general we can give the notation ϕ of k_0+N, M it means we will start the A of k_0+N-1 then end with the last integer M here.

So this gives the state transition matrix for the time variant matrix here and if you give the notation $k=k_0+N$ we get the formula ϕ of k, M is A^{k-1} up to A of M and the solution 9 is written as x of $k=\phi$ of $k, k_0 * x$ of k_0 .

So now we have to note here if you write ϕ of M, k . Here k is $>M$ when we write like this. So $k-1, k-2$ up to M it can decrease but when we write ϕ of M, k we cannot make use of a similar formula, we cannot start with $M-1, M-2$ and we cannot end at the point k here because M is $<k$. So we have to make how to write analogous to the continuous system, continuous system we always write ϕ of t, s inverse it is $=\phi$ of s, t .

So we do not consider, we do not bother about the $t>s$ or $t<s$. In all cases, we can write in this particular fashion but here it is not possible to use this notation as it is here.

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It is easy to verify that

$$\begin{aligned} \Phi(k+1, M) &= A(k)A(k-1)\cdots A(M) \\ &= A(k)\Phi(k, M) \end{aligned} \tag{10}$$

$$\Phi^{-1}(k, M) = A^{-1}(M)A^{-1}(M+1)\cdots A^{-1}(k-1) \tag{11}$$

which is denoted by $\Phi(M, k)$.



So we have to consider like this, so ϕ inverse of k, M is from the definition of ϕ of k, M if you take inverse both sides, we will A inverse M first A inverse $M-1$, etc the last will be A inverse $k-1$. So that is the straight forward formula ϕ inverse of k, M is given by equation 11 but we will give a notation for this to be ϕ of M, k . So it is not directly using the definition of ϕ but it is only a notation for ϕ inverse k, M we will write it as ϕ of M, k .

And the property similar to the continuous case ϕ of $k+1, M$ is given by this expression which is same as A of $k * \phi$ of k, M . So we can prove all the 4 properties of the state transition matrix using these notations. So whether it is an autonomous case or non-

autonomous case before properties of the state transition matrix is proved in this way using this particular notation.

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Example
Solve

$$\begin{aligned}x_1(k+1) &= -x_1(k) + Kx_2(k) \\x_2(k+1) &= 2x_2(k) \\x_1(0) &= x_1, \quad x_2(0) = x_2\end{aligned}$$

Solution: Given $x_2(k+1) = 2x_2(k)$. Taking Z-transformation

$$\begin{aligned}z(\bar{x}_2(z) - x_2(0)) &= 2\bar{x}_2(z) \\ \Rightarrow \bar{x}_2(z) &= \frac{z}{z-2}x_2(0).\end{aligned}$$

$A(k) = \begin{bmatrix} -1 & K \\ 0 & 2 \end{bmatrix}$
 $x(0) = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

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So now let us illustrate with an example how to solve a discrete dynamical system. So let us consider a second-order system x_1 of $k+1$ is $-x_1$ $k+k$ times x_2 of k , the x_2 of $k+1$ is 2 times x_2 of k , initial condition is $x_1(0)=x_1$ and $x_2(0) = x_2$, two numbers. So now here the matrix A is a time variant matrix because A is -1 k and 0 2 . So k is varying it is A of k is=this. So it is a dynamical system and the initial condition x_0 it is nothing but x_1 x_2 .

The vector is given x at 0 is given by this expression. Now we can solve the equation directly by using this z transform, taking z transform both sides of the first equation, sorry of the second equation, so we will get z transform of x_2 of $k+1$ that is= 2 times z transform of x_2 of k and using the formulas of z transform, we will get z *the z transform of x_2-x_2 at $0=2$ times z transform of x_2 and so the z transform of x_2 is given by $z/z-2*x_2$ of 0 .

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Using inverse Z–transformation

$$x_2(k) = 2^k x_2(0), \quad k = 0, 1, 2, \dots$$

Now,

$$\begin{aligned} x_1(k+1) &= -x_1(k) + kx_2(k) \\ \Rightarrow x_1(k+1) &= -x_1(k) + k2^k x_2(0). \end{aligned}$$

Taking Z–transformation

$$\begin{aligned} z(\bar{x}_1(z) - x_1(0)) &= -\bar{x}_1(z) - z \frac{d}{dz} \left(\frac{z}{z-2} \right) x_2(0) \\ \Rightarrow (z+1)\bar{x}_1(z) - zx_1(0) &= \frac{2z}{(z-2)^2} x_2(0). \end{aligned}$$

So by using the inverse z transform, we will get x_2 of k to be 2 to the power k x_2 at 0 . So that can be easily verified. Now substituting x_2 of k in the first equation which is x_1 of $k+1$ is $-x_1$ of $k+k$ times x_2 of k is given by this expression, so that is substituted 2 to the power k x_2 of 0 . Now again taking z transform both sides of this equation, we will get z^*z transform of $x_1 - x_1(0)$ is $-z$ transform of x_1 .

And z transform for this expression is $-z$ times d/dz of the z transform of 2 power k that is $z/z-2^*$ the initial condition.

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$$\begin{aligned} \therefore \bar{x}_1(z) &= \frac{2z}{(z-2)^2(z+1)} x_2(0) + \frac{z}{z+1} x_1(0) \\ Z^{-1} \left(\frac{2z}{(z-2)^2} \right) &= k \cdot 2^k = f_k. \\ Z^{-1} \left(\frac{1}{z+1} \right) &= (-1)^{k+1} = g_k \quad k = 1, 2, \dots \end{aligned} \quad (12)$$

Now collecting the z transforms in one side and taking the inverse z transform, we get the z transform of 1 is given by this expression and the inverse z transform gives the equation 12. So the solution the inverse z transform of $1/z+1$ is given by this. So this expression is the

product of the two z transforms, f suffix k and g suffix k are the inverse z transform of this expression. So we can use the convolution theorem to find the inverse z transform of the expression $2z/(z-2)^2(z+1)$.

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$$\begin{aligned} \therefore Z^{-1}\left(\frac{2z}{(z-2)^2(z+1)}\right) &= \sum_{i=0}^k f_i g_{k-i} \\ &= \sum_{i=0}^k \left(i2^i(-1)^{k-i+1}\right) \\ \therefore x_1(k) &= (-1)^k x_1(0) + \sum_{i=0}^k \left[i2^i(-1)^{k-i+1}\right] x_2(0) \end{aligned}$$

So the convolution formula is given in the right hand side here and the expression is given by this one. So the solution x_1 of k is given in the last line here.

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Non-autonomous Control System

Consider the discrete system

$$\begin{aligned} x(k+1) &= A(k)x(k) + B(k)u(k), \quad k = k_0, k_0 + 1, \dots, k_0 + N \quad (13) \\ x(k_0) &= x_0 \end{aligned}$$

then

$$\begin{aligned} x(k_0+1) &= A(k_0)x(k_0) + B(k_0)u(k_0) \\ x(k_0+2) &= A(k_0+1)x(k_0+1) + B(k_0+1)u(k_0+1) \\ &= A(k_0+1)(A(k_0)x(k_0) + B(k_0)u(k_0)) + B(k_0+1)u(k_0+1) \\ &= A(k_0+1)A(k_0)x(k_0) + A(k_0+1)B(k_0)u(k_0) + B(k_0+1)u(k_0+1) \end{aligned}$$

Now we come to the control systems. Consider the non-autonomous control system in which the matrix A and B are functions of k , the time instances and initial time instant is k_0 and initial condition is x of x_0 . Now making use of a similar step-by-step method, so if you calculate x at k_0+2 , we will get this expression given in the last step.

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$$\begin{aligned}
 x(k_0 + N) &= \Phi(k_0 + N, k_0)x(k_0) + \Phi(k_0 + N, k_0 + 1)B(k_0)u(k_0) \\
 &+ \Phi(k_0 + N, k_0 + 2)B(k_0 + 1)u(k_0 + 1) \\
 &+ \dots + B(k_0 + N - 1)u(k_0 + N - 1)
 \end{aligned}$$

where

$$\Phi(k_0 + N, M) = A(k_0 + N - 1)A(k_0 + N - 2) \dots A(M), \quad M \leq k_0 + N$$

So this is only substituting step-by-step, so except $k_0 + N$ that is at the N th time instant is given by the formula as shown in the right hand side and here the state transition matrix Φ of $k_0 + N, M$ is given by this expression where M is \leq it should be strictly $<$ okay. M is strictly $< k_0 + N$ because the right hand side it starts with $k_0 + N - 1$ and decreasing up to capital M , so M cannot be equal to this expression okay.

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Put $k = k_0 + N$, we get

$$\Phi(k, M) = A(k - 1)A(k - 2) \dots A(M) \quad (14)$$

$$x(k) = \underbrace{\Phi(k, k_0)}_{A} \left[x(k_0) + \sum_{i=k_0}^{k-1} \underbrace{\Phi(k_0, i+1)}_{A} B(i)u(i) \right] \quad (15)$$

So by substituting $k = k_0 + N$ we can see that the state transition matrix Φ of k, M is A^{k-1} up to A^M the product in the right hand side and the solution x of k is given by the formula as shown in the expression 15 here, the solution in the last step x of $k_0 + N$ is given in this right hand side so we can easily compare it with the compact expression which is given in the equation 15 as the solution of the problem.

So now so this expression is common to the time variant case or time invariant case. So in the time invariant case autonomous case, phi is replaced by simply A to the power this expression k_0-i+1 , so that is the only difference wherever phi is there we will replace it with A to the power $k-k_0$ here and here we will replace it, so that is for the time invariant case and this is 15 is the general formula for the time variant case.

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Controllability

Definition
The linear system

$$x(k+1) = Ax(k) + Bu(k) \quad (16)$$

Handwritten notes:
 A is $n \times n$
 B is $n \times m$
 $x_0, x_f \in \mathbb{R}^n$

is controllable if for any two arbitrary states x_0 and x_f there exists an integer $N > 0$ and a control sequence $u(0), u(1), \dots, u(N-1)$ such that $x(0) = x_0$ and $x(N) = x_f$. Here A is nonsingular.

Theorem 1
The system (16) is controllable if and only if

$$\text{rank}[B \ AB \ \dots \ A^{n-1}B] = n.$$

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Now we define the controllability of the non-autonomous system, sorry one minute so now we will see the controllability of the autonomous system x of $k+1 = Ax$ of $k + Bu$ of k . So this system is said to be controllable if for any arbitrary state x_0 and x_1 , two vectors in \mathbb{R}^n , so here we will consider A is a $n \times n$ matrix, B is a $n \times m$ matrix. So two arbitrary vectors x_0 and x_1 both belong to the state space \mathbb{R}^n .

There exists an integer capital N , that is the final time capital N and a sequence of control u_0, u_1 up to u_{n-1} such that the solution of the system 16 satisfies the condition x of $0 = x_0$ and x of capital N is x_f , x of f is the final. So the system is controllable for any two arbitrary vectors x_0 and x_f given in the state space \mathbb{R}^n there exists an integer N as final instant and a sequence of control u_1, u_2 up to u_{n-1} such that the solution satisfies the initial and the final conditions.

So here in all these problems, we note that A should be a nonsingular matrix because we have seen that the state transition matrix for satisfying the properties 4 properties it should be a nonsingular or invertible matrix. So now we prove the theorem which is analogous to the continuous case where we have defined the controllability Gramian matrix, so the system where we have proved the Kalman theorem earlier.

So the system (16) is controllable if and only if the Kalman matrix $B, AB, A^2B, \dots, A^{n-1}B$ has rank n . So the similar theorem we have proved it for the continuous case earlier.



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Proof: Solving (16) by successive substitution, we get

$$\begin{aligned} x(1) &= Ax(0) + Bu(0) \\ x(2) &= Ax(1) + Bu(1) \\ &= A^2x(0) + ABu(0) + Bu(1) \\ &\vdots \\ x(N) &= A^Nx(0) + A^{N-1}Bu(0) + \dots + ABu(N-2) + Bu(N-1) \quad (17) \end{aligned}$$

According to the definition of controllability, system (16) is controllable if there exists a sequence $u(0), u(1), \dots, u(N-1)$, which transfer an arbitrary state $x(0)$ to an arbitrary state x_f in N sampling period where N is finite positive integer. Equation (17) can be written as



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So the proof is a quite straight forward here. So now the time instances are 0, 1, 2, 3 up to capital N . If you substitute each instant one by one, so x of 1 is $Ax(0) + Bu(0)$. Similarly, x of capital N is $A^N x(0) + A^{N-1}Bu(0) + \dots + ABu(N-2) + Bu(N-1)$ and we get the equation (17) by substituting step-by-step. Now the definition of controllability is we have to find an integer capital N so that the solution satisfies.

We have to find an integer capital N as well as a sequence of control that is u_0, u_1 up to u_{N-1} so that the solution satisfies the condition $x(0) = x_0$ and $x(N) = x_f$. So we will see under the condition the rank condition given in the last line, we can prove that the system is controllable.

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$$x_f - A^N x(0) = A^{N-1} B u(0) + \dots + A B u(N-2) + B u(N-1) \quad \begin{matrix} \beta_{n \times m} \\ u_{m \times 1} \end{matrix}$$

$$\text{or } \hat{x}_{n \times 1} = [B \quad AB \quad \dots \quad A^{N-1} B] \begin{bmatrix} u(N-1) \\ u(N-2) \\ \vdots \\ u(0) \end{bmatrix} \quad (18)$$

$$\text{or } \hat{x} = [U_N]_{n \times Nm} [u]_{Nm \times 1} \quad (19)$$

for (19) to be satisfied for arbitrary $x(0)$ and x_f , it is necessary that

$$\text{rank}\{U_N\} = \text{rank}\{U_N \quad \hat{x}\} \quad \begin{matrix} \text{exists} \\ \text{solution } [u]_{Nm \times 1} \end{matrix}$$

So now if we substitute x of capital $N=x_f$ in the equation 17 and taking this A to the power N of 0 in the left hand side, we will get the first line here $x_f - A$ to the power N of 0 it is equal to the remaining terms which can be denoted in this following way. So let us write the left hand side as \hat{x} which is a vector in R^n and because $x(0)$ and x_f both are arbitrary, this \hat{x} is also an arbitrary vector in R^n .

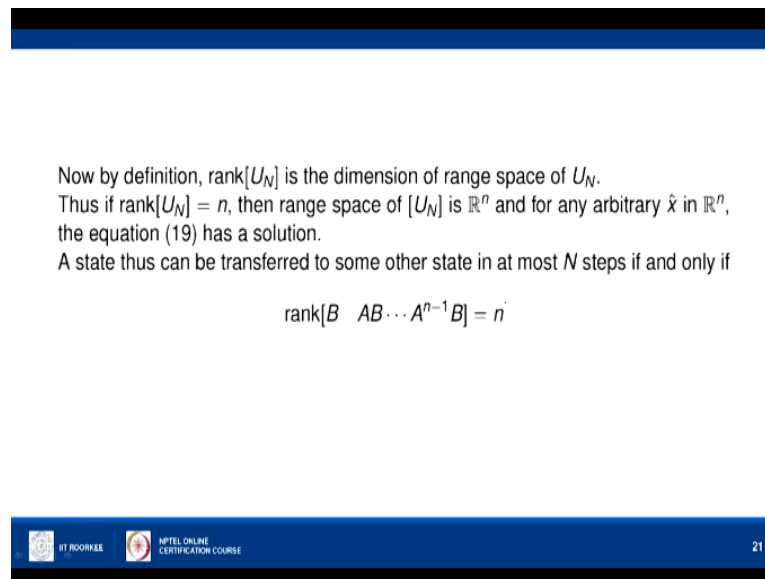
So the remaining terms in the right hand side can be written as $B \quad AB \quad A$ to the power $N-1$ B here u of $N-1$. We can note that the B matrix is multiplied by u of $N-1$, so that is here. AB matrix is multiplied by u of $N-2$, etc. A to the power $N-1$ is multiplied by B matrix okay. A to the power $N-1$ B is multiplied by u of 0 so this expression can be written in the form of equation 18.

Now the size of the matrix if you call this as the U matrix capital U , its size is $n \times N \times m$ because B is a $n \times m$ matrix and AB is a $n \times m$ matrix, etc. So the size of this matrix capital U is $n \times$ capital $N \times m$. Now u is a vector which is $m \times 1$ vector. So each u is $m \times 1$ vector, so totally you have $N \times m \times 1$ vector here in the equation 18. So equation 19 it is simply a system of algebraic equation where \hat{x} is an arbitrary vector, U is the matrix given here and small u is the vector of size capital $Nm \times 1$.

So if the rank of this matrix U is N then it is clear that there exist a solution for this system because for existence of the solution of 19, the rank of u and rank of the augmented matrix U and \hat{x} should be equal and if the rank of u itself is N then the system 19 will have a

solution. There will be infinitely many solutions because of the size of this matrix. U is not a square matrix. So it will have a solution means it will have infinitely many solutions.

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Now by definition, $\text{rank}\{U_N\}$ is the dimension of range space of U_N .
Thus if $\text{rank}\{U_N\} = n$, then range space of $\{U_N\}$ is \mathbb{R}^n and for any arbitrary \hat{x} in \mathbb{R}^n , the equation (19) has a solution.
A state thus can be transferred to some other state in at most N steps if and only if

$$\text{rank}[B \quad AB \cdots A^{n-1}B] = n$$

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So we get if the rank of this matrix is $=N$ then the system will have a solution that is U . So this sequence is defined from the equation 9. So we will get the solution u $N \times 1$ exist. From this, we can get the solution of the discrete control for the discrete system. So the converse part also can be proved analogous to the continuous case. That is if the system is controllable, then the rank of this matrix should be equal to n .

So it is exactly similar analogous to the continuous control system, so we need not repeat the same thing. So with this, I will complete the lecture for the discrete autonomous control systems. So in the next lecture, we will see the controllability of the discrete non-autonomous control systems. Thank you.