

Dynamical Systems and Control
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Lecture - 45
Introduction to Discrete Systems - I

Hello viewers. Welcome to the first lecture on discrete dynamical systems. In this lecture, we will define the state transition matrix of the discrete system and how to write the solution of the discrete system in terms of state transition matrix. Then, we will see the definition and conditions on the controllability of the discrete systems.

So in previous lectures, we have seen several examples of continuous dynamical systems in terms of the differential equations or the control systems in the form of a differential equations. So in this now we will see some examples of discrete dynamical systems. So for example if you consider prey-predator model, earlier we have considered as a continuous system, now we can also consider it as the following system.

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$x_1(t) = \text{prey population}$, $x_2(t) = \text{predator population}$ at t
 $t = k_0, k_0+1, k_0+2, \dots$
 $x_1(k) = x_1(k-1) + \alpha x_1(k-1) - (\beta x_2(k-1)) \cdot x_1(k-1)$
 $x_2(k) = x_2(k-1) - \gamma x_2(k-1) + (\delta x_1(k-1)) \cdot x_2(k-1)$
 $x_1(k+1) = (1 + \alpha) x_1(k) - \beta x_1(k) x_2(k)$
 $x_2(k+1) = (1 - \gamma) x_2(k) + \delta x_1(k) x_2(k)$
 $k = k_0, k_0+1, \dots$

Banking System
 initial investment $x(0)$
 rate of interest $= r\%$ per year
 yearly investment $= u(k)$ $k = 1, 2, \dots$

So let x_1 of t it represent the prey population and x_2 of t is the predator population at time t . So if the time is measured in discrete intervals, so if T is considered as say k_0, k_0+1, k_0+2 , etc in discrete time intervals. Then, the prey predator model can be written in this particular form that is the prey population at the time instant k that will be depending on the population of the previous time instant+some increase in the previous time instant population where α is the ratio of increment.

So this is our original population of the previous instant x_1 of $k-1$ + now α times x_1 of $k-1$ population increases. So this is about the increase and then it will decrease with respect to the predator population. So β times x_2 of $k-1$, so in the previous instant how much predator was available that is the ratio and then we have to multiply by the previous year population of the prey, so this amount will be decreasing.

So we can say that at time instant k , it is all depending on the population of prey and predator the previous time instant. Similarly, the population of the predator, it is the previous instance population but it will decrease by some amount let us say γ . So γ times its population in the previous time instant, so it will decrease but it will increase with a population δ times x_1 of $k-1$ times its own population at time $k-1$.

So this gives a prey predator model in the discrete case or we can write in another form. If you write $k+1$ as the current instant and k as the previous instant, we will write this as $1+\alpha$ times x_1 of k - β times x_1 of k x_2 of k . Similarly, x_2 of $k+1$ of $1-\gamma$ times x_2 of k + δ times x_1 of k * x_2 of k . So this gives a prey-predator model for k starting from k_0 and k_0+1 , etc. We can give different values for this k value.

Another model which is in the banking sector, we can see the following equation. So for example let us consider in the banking system, so if we consider the initial investment in a bank as x_0 of 0 at the time instant 0 and let say the rate of interest is $r\%$ per year and the yearly investment it is let say u of k for various k values, let us say $1, 2, 3$ etc. Every year the investment is given like this.

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$$x(k+1) = x(k) \left(1 + \frac{r}{100}\right) + u(k), \quad u(0) = 0$$

$$x(0) = x_0$$

$$x(k+1) = A(k)x(k) + B(k)u(k)$$

$$x(k) = \begin{bmatrix} x_1(k) \\ x_2(k) \\ \vdots \\ x_n(k) \end{bmatrix} \in \mathbb{R}^n, \quad u(k) = \begin{bmatrix} u_1(k) \\ u_2(k) \\ \vdots \\ u_m(k) \end{bmatrix}$$

$A(k)$ is $n \times n$, $B(k)$ is $n \times m$.

Consider $x(k+1) = A x(k)$ $x(k_0) = x_0$

$x(k_0+1) = A x(k_0)$ put $k_0 + N = k$

$x(k_0+2) = A^2 x(k_0)$ Then $x(k) = A^{k-k_0} x(k_0)$

$x(k_0+N) = A^N x(k_0)$ put $\phi(k, k_0) = A^{k-k_0}$

So then the amount which will be available at a particular instant of time that is k+1th year how much amount will be available that is written as using a simple interest, the principal interest and r/100. So the previous year the amount was x of k and +x of k r/100 will be the interest earned by x of k+the investment of that year is u of k and we assume that u of 0=0 because the initial investment was already x0.

So this expression gives it is a discrete dynamical system and if this u of k can be considered as a control for achieving certain particular goal that is if you want to achieve that after certain years what should be the amount available in the bank. So for that purpose, we can adjust this u of k suitably so that we can achieve the goal. So it is a discrete control system in this form.

So in general a discrete control system is written in the form x of $k+1=A$ of $k*x$ of $k+B$ of $k*u$ of k where x of k is a vector $x_1 k, x_2 k,$ etc, $x_n k$ belongs to \mathbb{R}^n . So it is the state variable and the control variable u of k will be u_1 of k, u_2 of k, u_n of k . So it is analogous to the continuous control system and A of k is a $n \times n$ matrix, B of k means $n \times m$ matrix, so almost similar to that one except that the time is measured in a discrete manner.

So if we consider only the dynamical system without the control term, so we can say x of $k+1=A$ of $k*x$ of k . If you consider this equation for starting from let us say x of $k_0=x_0$ initial time instant, we take it as k_0 and then at every increment by 1 we will measure the state variable x of $k+1$. First, let us consider the case in which A is a constant matrix instead of A of k let us consider A itself and then x of $k+1=A*x$ of k we consider.

So x of k_0+1 is $A \cdot x$ of k_0 and x of k_0+2 is A times A of x of k_0 . So we will get this one etc. So we will get the expression k_0+n it is A to the power $N \cdot x$ of k_0 . So if you substitute k_0+N as the notation k then we will get the expression x of k is A to the power $k-k_0 \cdot x$ of k_0 . So this is a very simple expression we can easily see from step-by-step. So this A of $k-k_0$ we will call it as the state transition matrix, $\phi(k, k_0) = A^{k-k_0}$.

So if you give this notation $\phi(k, k_0) = A$ to the power $k-k_0$, we call it as the state transition matrix and this satisfy the various properties.

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The slide is titled "State Transition Matrix". It contains the following text and equations:

The state transition matrix is defined by

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

It is easy to verify that

$$\Phi(k+1, k_0) = A\Phi(k, k_0)$$

$$\Phi(k, k) = I$$

$$\Phi(k_0, k) = \Phi^{-1}(k, k_0), \text{ provided } A \text{ is nonsingular}$$

$$\Phi(k, k_0) = \Phi(k, k_1)\Phi(k_1, k_0), \quad k \geq k_1 \geq k_0$$

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$$\begin{aligned} \phi(k_1, k) &= A^{k-k_1} \\ &= A^{k-k_0} A^{-(k_1-k_0)} \\ &= \left(A^{k-k_0} \right)^{-1} \\ &= \left(\phi(k, k_0) \right)^{-1} \end{aligned}$$

$$\begin{aligned} \frac{d}{dt} \phi(t, \tau) &= A \phi(t, \tau) \end{aligned}$$

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So now from here we will see that $\phi(k, k_0) = A$ power $k-k_0$, it is easy to verify that these properties are satisfied, $\phi(k+1, k_0)$ if you substitute here A to the power $k+1-k_0$, so we can take one A out, so we will get $A \cdot A$ power $k-k_0$ so this is satisfied. Now $\phi(k, k)$ if you put the same thing you will get identity from this expression and $\phi(k_0, k)$ will be it is A to the power k_0-k which can be written as A to the power $-(k-k_0)$.

Or it is nothing but A to the power $k-k_0$ inverse okay because first we can write it as A inverse whole power $k-k_0$ which is same as this expression. So this is $\phi(k, k_0)$ inverse so that can be easily verified from this expression and the last one by directly substituting this we can verify. So these 4 conditions they are much similar to the conditions of the state transition matrix of continuous dynamical systems.

So there the first condition is replaced by d/dt of phi t, s is A times phi t, s. So this condition is replaced by the first one here.

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Autonomous System

Consider the discrete autonomous system

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

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

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



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Now using this state transition matrix, we can write the solution of the equation in the following way. So x of $k+1$ will consider the discrete control system. So ϕ of $k+1 = A * x$ of $k + B u$ of k and initial time we consider as k_0 . So by substituting the values one by one, x of k_0+1 is $A x$ of $k_0 + B u$ of k_0 and x of k_0+2 is $A x$ of k_0+1 . In that place, we substitute in the next step the previous equation that is A of x $k_0 + B u$ x k_0 .

And then the sum last term is there and simplifying we will get A square x of k_0 + etc. The last line is there.

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Autonomous System



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

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

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

$x(k) = \phi(k, k_0) \left[x(k_0) + \sum_{i=k_0}^{k-1} \phi(k, i+1) Bu(i) \right]$

 $k = k_0, k_0+1, \dots$

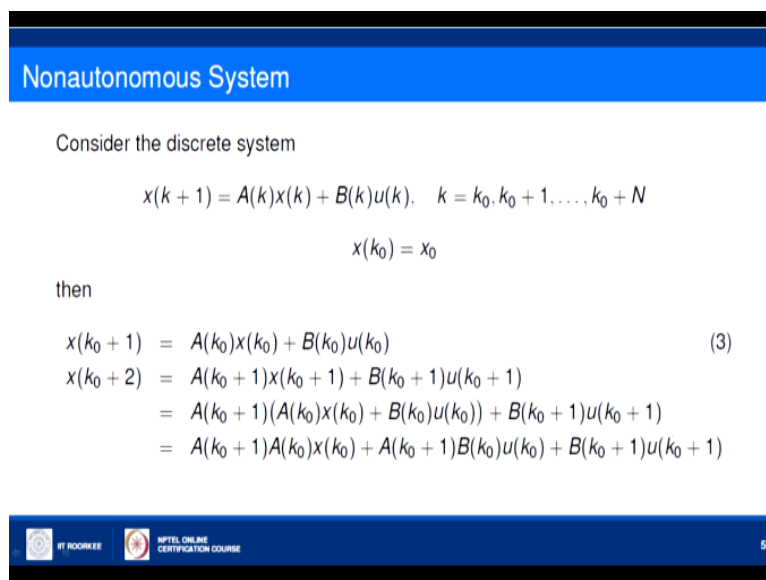


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Now using the similar step, continuing in a similar manner we will get x of $k_0 + N$, we get the last expression as A to the power $k_0 + N - k_0$ because N is replaced by this and A to the power $k_0 + N - k_0 + 1$ etc. So this expression by replacing this as the state transition matrix, we will get the expression as x of $k = \phi$ of k , k_0 x of $k_0 + \sum_{i=k_0}^{k-1}$ okay.

Substituting $k = k_0 + N$ and calling ϕ of k , l as A to the power $k - l$ as defined earlier we get the expression which can be verified in the previous step directly by substitution. So this expression can further be simplified as this following. This can be written as ϕ of k , k_0 x of $k_0 + \sum_{i=k_0}^{k-1}$. So this ϕ of k_{i+1} can be written as ϕ of k , k_0 ϕ of k_0 , $i+1$. So ϕ k , k_0 we have taken out of the bracket.

And the remaining we write it this way B of u of i . So this we call it as the solution of the system for various values of k , k_0 , $k_0 + 1$, etc for the time invariant system.

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Nonautonomous System



Consider the discrete system

$$x(k+1) = A(k)x(k) + B(k)u(k), \quad k = k_0, k_0 + 1, \dots, k_0 + N$$

$$x(k_0) = x_0$$

then

$$\begin{aligned} x(k_0 + 1) &= A(k_0)x(k_0) + B(k_0)u(k_0) \\ x(k_0 + 2) &= A(k_0 + 1)x(k_0 + 1) + B(k_0 + 1)u(k_0 + 1) \\ &= A(k_0 + 1)(A(k_0)x(k_0) + B(k_0)u(k_0)) + B(k_0 + 1)u(k_0 + 1) \\ &= A(k_0 + 1)A(k_0)x(k_0) + A(k_0 + 1)B(k_0)u(k_0) + B(k_0 + 1)u(k_0 + 1) \end{aligned} \tag{3}$$

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Now if you consider the system which is time variant system nonautonomous system, x of $k+1$ is A of k x of k $+ B$ of k u of k for the various values of k initial condition is x of k_0 is x_0 . So proceeding in the similar manner substituting these values step-by-step, we will arrive at this one, x of $k_0 + 1$ is A of k_0 x of k_0 $+ B$ of k_0 u of k_0 that is the initial time and the second step we will get we will make use of the previous step and we arrive at the relation A of $k_0 + 1$ A of k_0 x of k_0 that is first term $+ A$ of $k_0 + 1$ B of k_0 u of k_0 second term etc.

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$$\begin{aligned}
x(k_0 + 3) &= A(k_0 + 2)A(k_0 + 1)A(k_0)x(k_0) \\
&+ A(k_0 + 2)A(k_0 + 1)B(k_0)u(k_0) \\
&+ A(k_0 + 2)B(k_0 + 1)u(k_0 + 1) + B(k_0 + 2)u(k_0 + 2) \\
x(k_0 + N) &= \Phi(k_0 + N, k_0)x(k_0) + \Phi(k_0 + N, k_0 + 1)B(k_0)u(k_0) \\
&+ \Phi(k_0 + N, k_0 + 2)B(k_0 + 1)u(k_0 + 1) \\
&+ \dots + B(k_0 + N - 1)u(k_0 + N - 1)
\end{aligned}$$

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 we can easily generalize it for $k_0 + N$, we will get the expression x of $k_0 + N$ is written in the form of $\Phi(k_0 + N, k_0)$. So we can notice that we define the state transition matrix in this case as $\Phi(k_0 + N, M)$ is written as $A(k_0 + N - 1)$. So we have to notice here, for example in the third step when x of $k_0 + 3$ is there, in the right hand side we have $A(k_0 + 2)$ and $k_0 + 1$. So it is in the decreasing order.

Similarly, when we have $k_0 + N$, it will start with $k_0 + N - 1$ and $k_0 + N - 2$ etc. So that is what we write here. We denote $\Phi(k_0 + N, M)$ as the starting will be $k_0 + N - 1$ and -2 , etc. It will stop at the stage capital M . So we can write the expression of the solution in this particular form. Now if you substitute $k_0 + N$ as k we get $\Phi(k, M)$ as $A(k - 1)$.

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Put $k = k_0 + N$, we get

$$\begin{aligned}
\Phi(k, M) &= A(k - 1)A(k - 2) \dots A(M) \\
x(k) &= \Phi(k, k_0) \left[x(k_0) + \sum_{i=k_0+1}^k \Phi(k_0, i)B(i - 1)u(i - 1) \right]
\end{aligned}$$

$k - M$
A

From the last step $k_0 + N$ is k , so we get here $k-1$ $k-2$ up to A of M . So this is called the state transition matrix for the time variant system and if A is a constant matrix we can easily see that this turns out to be the same ϕ of $k-M$ will be $=A$ to the power $k-M$ as in the previous case. So the solution of the time variant system can be written in this form x of $k = \phi$ of k , k_0 * this expression.

So here we can easily verify the 4 conditions of the state transition matrix. So in this lecture we have seen for the time invariant system how to write the state transition matrix etc. So in the next lecture, we will see about the controllability of the time varying system, similarly the conditions for the observability and stability of all these systems. Thank you.