

Dynamical Systems and Control
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Lecture - 07
Solution of Linear Systems – I

Hello friends, welcome to this lecture. In this lecture we continue our study of dynamical system and control. So in previous class we have discussed Existences and Uniqueness for say non-autonomous system of linear equation and autonomous system of linear equation. Let us recall what we have done in previous class. So in previous class we have discussed is linear system $\dot{x} = A(t)x + f(t)$.

(Refer Slide Time: 00:56)

Existence and Uniqueness for Linear Systems

Consider the linear system

$$\dot{x} = A(t)x + F(t), \quad x(t_0) = \eta, \quad (1)$$

where

$$A(t) = \begin{pmatrix} a_{11}(t) & \dots & a_{1n}(t) \\ \vdots & \ddots & \vdots \\ a_{m1}(t) & \dots & a_{nn}(t) \end{pmatrix}, \text{ and } F(t) = \begin{pmatrix} f_1 \\ \vdots \\ f_n \end{pmatrix},$$

and let the matrix $A(t)$ and the vector $F(t)$ are continuous on an interval I .

Theorem 1

If $A(t)$, $F(t)$ are continuous on some interval $a \leq t \leq b$, If $a \leq t_0 \leq b$, and if $|\eta| < \infty$, then the system (1) has a unique solution $\phi(t)$ satisfying the initial condition $\phi(t_0) = \eta$ and existing on the interval $a \leq t \leq b$.

Where $A(t)$ is given by this coefficient matrix $a_{11}(t)$ to $a_{1n}(t)$; $a_{n1}(t)$ to $a_{nn}(t)$. And $F(t)$ is given by this f_1 to f_n is column vector And here we assume that $A(t)$ and this vector function $F(t)$ are continuous on a given interval. And we have discussed the following Existence and Uniqueness theorem and it says that if $A(t)$ and $F(t)$ are continuous n some close interval a to b and $F(t_0)$ is some initial point line in this interval a to b and if norm of η is $< \infty$ then this system 1 has a unique solution $\Phi(t)$ satisfying the initial condition $\Phi(t_0)=\eta$.

So it means that if we assign a initial condition that $\Phi(t_0)=\eta$ then we have a solution, a unique solution of this initial value problem exist $\dot{x} = A(t)x + F(t)$ and $x(t_0)=\eta$, here it is x

rather than x , so it is your $x(t_0) = \theta$. So this, and this solution exist on the entire interval a to b . And so this is the result for non-autonomous system of linear equation.

(Refer Slide Time: 02:17)

For autonomous systems, we have the following existence and uniqueness theorem.

Theorem 2

(Existence-uniqueness theorem) The initial-value problem

$$\frac{dx}{dt} = Ax, \quad x(t_0) = x^0 = \begin{pmatrix} x_1^0 \\ \vdots \\ x_n^0 \end{pmatrix}. \quad (2)$$

has a unique solution for every $x^0 \in \mathbb{R}^n$, and all $t \in \mathbb{R}$.

$\dim(V) = n$
 ✓ Solution set
 $V = \{x(t) \in \mathbb{R}^{n \times 1} : x' = Ax\}$
 ✓ $Cx(t), x(t_1) + y(t_1)$
 $\subseteq V$

And similarly for autonomous system we have the following Existence and Uniqueness theorem. The initial value problem $dx/dt = Ax$, where A is independent of t and $x(t_0) = x_0$ where x_0 is initial condition given at $t=(t_0)$ and it is like $x_1(0)$ to $x_n(0)$. Then, this initial value problem as a unique solution for every x_0 belong to \mathbb{R} and for all t in \mathbb{R} . So this we have discussed. So it means also we have shown apart from this Existence and Uniqueness theorem that once we have Existence and Uniqueness theorem then we try to find out this solution set.

So solution set is basically, solution set is defined as a set v such that it is basically $x(t)$ belong to say $\mathbb{R}^{n \times 1}$ such that $x' = Ax$. So if we define this and we can check that that if $x(t)$ is a solution then c of $x(t)$ is also a solution and if $x_1; y_1$ are two solution then $x(t) + y(t)$ also be a solution or we can say that this set v from say vector space because 0 is already there because 0 belongs to v because $x' = Ax$ has a trivial solution.

So it means that this solution set from say vector sub-space and we have also checked that the dimension of this vector space is given as n , so it means that if we want to, so this is up to what we have discussed in previous class and we can say that once dimension of v is known to us then

we can find out n linearly independent solutions of $\dot{x} = Ax$ and any solution of $\dot{x} = Ax$ can be written as a linear combination of the L linearly independent solutions.

So it means that this much we have done. But the problem is that once we have some solutions $\dot{x} = Ax$ then it is quite difficult, how do we know that this solution is actually a linearly independent solution. So here we discuss next result by which we can check that a given set of solutions is actually a linearly independent solution here. So let us consider the following theorem.

(Refer Slide Time: 04:50)

Theorem 3

(Test for linear independence) Let x_1, x_2, \dots, x_n be n solutions of



$$\dot{x} = Ax. \quad (3)$$

Let $t_0 \in \mathbb{R}$. Then x_1, x_2, \dots, x_n are linearly independent solutions if and only if, $x_1(t_0), x_2(t_0), \dots, x_n(t_0)$ are linearly independent vectors in \mathbb{R}^n .

Proof: Assume that $x_1, x_2, x_3, \dots, x_n$ are solutions of (3) but not linearly independent. Then, there exist constants c_1, c_2, \dots, c_n not all zero, such that

$$c_1 x_1 + c_2 x_2 + \dots + c_n x_n = 0.$$

\checkmark $x_1(t_0), \dots, x_n(t_0)$ are LI vectors


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4

Which says that, let X_1 to X_n be n solutions of $\dot{x} = Ax$ and let t_0 is some real number in \mathbb{R} . Then X_1 to X_n are linearly independent solutions if and only if the $x_1(t_0)$ and $x_2(t_0)$ and $x_n(t_0)$ are linearly independent vectors in \mathbb{R}^n . So please try to understand here this x_1 to x_n are n linearly independent solutions of a $\dot{x} = Ax$ so it means that these are functions basically. So the linear independence of these functions can be obtained by linear independence of these vectors. So these are constant vectors in \mathbb{R}^n and these are functions of t .

So how the linear independence of functions can be obtained with the help of linear independence of constant vectors in \mathbb{R}^n . So that is a content of this theorem. And it says that, assume that x_1 to x_n are solutions of (3). So it means let us assume that x_1 to x_n are

solutions but not linearly independent. So we assume that these are solutions but may not be linearly independent.

So if they are not linearly independent then there exists constant c_1 to c_n not all 0 such that this happens. So we have to say that there exists constant c_1 to c_n not all 0 such that $C_1 X_1 + C_2 X_2 + \dots + C_n X_n = 0$. So here we are assuming that the solution to X_1 to X_n are not linearly independent or we can say that x_1 to x_n are linearly independent. So we can find out, so it means that they exist some constant C_1 to C_n not all 0 such that this linear combination equal to 0. Now this is true for all t , right. So if it is true for all t then in particular we evaluate at $t=t_0$ also.

(Refer Slide Time: 06:48)


At $t = t_0$, we have

$$c_1 x_1(t_0) + c_2 x_2(t_0) + \dots + c_n x_n(t_0) = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

and not all $c_i, i = 1, 2, \dots, n$ are zero. Which implies that $x_1(t_0), x_2(t_0), \dots, x_n(t_0)$ are linearly dependent vectors in \mathbb{R}^n , which is a contradiction. Hence the result holds.

Conversely, suppose that the values of x_1, x_2, \dots, x_n at some time t_0 are linearly dependent vectors in \mathbb{R}^n . Then, there exist constants c_1, c_2, \dots, c_n not all zero

Handwritten notes:
 $x_1(t), \dots, x_n(t)$ are LI soln of $x' = Ax$
 $\Rightarrow x_1(t_0), \dots, x_n(t_0)$ are LI vectors in \mathbb{R}^n
 $x_1(t), \dots, x_n(t)$ are also LI soln of $x' = Ax$



So, at $t=t_0$ we have $C_1 X_1(t_0) + C_2 X_2(t_0) + \dots + C_n X_n(t_0) = 0, 0, \dots, 0$. Now, so this implies what, that here not all C_i 's are 0 and still this linear combination is equal to 0. So this implies that, that $X_1(t_0), X_2(t_0)$ and $X_n(t_0)$ are linearly dependent vectors in \mathbb{R}^n . But here we have already assumed that this $X_1(t_0)$ to $X_n(t_0)$ are linearly independent. So here before proceeding this we are assuming that $X_1(t_0)$ and $X_n(t_0)$ are LI vectors.

So here if we assume that these $X_1(t_0)$ to $X_n(t_0)$ are LI vectors then we have a contradiction. It means that, the assumption that X_1 to X_n are linearly dependent solution may not be correct. If we assume that $X_1(t_0)$ to $X_n(t_0)$ are LI vectors. So here it means that if we assume that X_1 to X_n are linearly dependent solution then $X_1(t_0)$ to $X_n(t_0)$ has to be linearly dependent.

So this implies that if X_1 to X_n are linearly independent then $X_1(t_0)$ to $X_n(t_0)$ are also linearly independent vectors. So we have proved that, that if $X_1(t)$ to $X_n(t)$ are solutions which are dependent linearly dependent solution then at $t=t_0$ also $X_1(t_0)$ to $X_n(t_0)$ are also linearly dependent vectors in R^n . So if solutions, so let me write it here that if $X_1(t)$ to $X_n(t)$ are LD solutions of $\dot{x} = Ax$ then implies that $X_1(t_0)$ to $X_n(t_0)$ are LD vectors in R^n , so that is what we have shown.

Now we want to show the reverse part of this. So it means, let us assume that the X_1 to X_n at t_0 are linearly dependent vectors in R^n . So we want to now go back to, so we want to show reverse thing. That let us X_1 to X_n at t_0 are linearly dependent vectors in R^n . Now we want to show that our claim is that $X_1(t)$ to $X_n(t)$ are also linearly dependent solutions of $\dot{x} = Ax$, that is what we want to show it here. So if we assume that, that $X_1(t_0)$ to $X_n(t_0)$ are linearly dependent vectors in R^n it means that they exist constant C_1 to C_n , not all 0.

(Refer Slide Time: 09:46)

such that

$$c_1 x_1(t_0) + c_2 x_2(t_0) + \dots + c_n x_n(t_0) = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \mathbf{0}.$$

With this choice of constants c_1, c_2, \dots, c_n construct the vector valued function

$$\phi(t) = c_1 x_1(t) + c_2 x_2(t) + \dots + c_n x_n(t).$$

This function satisfies (3) since it is a linear combination of solutions. Moreover, $\phi(t_0) = \mathbf{0}$. Therefore by existence and uniqueness theorem 2, $\phi(t) \equiv \mathbf{0}$, that is

$$c_1 x_1(t) + c_2 x_2(t) + \dots + c_n x_n(t) = \mathbf{0}, \forall t$$

for all t . This implies that x_1, x_2, \dots, x_n are linearly dependent solutions.

Such that $C_1 X_1(t_0) + C_2 X_2(t_0) + C_n X_n(t_0) = 0$. So it means that linearly combination is equal to 0 and not all the C_i 's are 0. So taking these combinations of C_i 's let us form this function that is $\Phi(t) = C_1 X_1(t) + C_2 X_2(t) + C_n X_n(t)$. So once we have such a pair of constants C_1 to C_n then with the help of these C_i 's form a function like this $C_1 x_1(t) + C_2 C_2 (t) + C_n X_n(t)$. Now here please observe that each $X_i(t)$ is a solution of $\dot{x} = Ax$.

So it means that each X_i is a solution of $\dot{x} = Ax$. So in particular linear combination of X_i will also be a solution of $\dot{x} = Ax$. So it means that this $\Phi(t)$ is also a solution of $\dot{x} = Ax$ because of the linear combination solution is again a solution. So; and not only this, we have $\Phi(t_0)=0$. Say t is at, if we evaluate $\Phi(t)$ at $t=t_0$ then we have already seen that it is equal 0, so it means that your $\Phi(t)$ is a solution of this provided that $\Phi(t_0)=0$. Now we already know that, that this system $\dot{x} = Ax$ has a 0 solution.

So 0 solution is what, 0 is already a solution such that 0 solution at any point is again a 0 vector. So it means that by Existence and Uniqueness theorem it can have only a unique solution. So it means that, that $\dot{x} = Ax$ with the condition that $x(t_0)=0$, it has it can have only a unique solution but if you look at we have two solution given as $x(t)$ ideally equal to 0 and $\Phi(t)$ given as this. So it means that, that these solutions has to be at in the equal to same. So it means that your $\Phi(t)$ has to be ideally equal to 0.

So it means that this linear combination that is $C_1 X_1(t) + C_2 X_2(t) + \dots + C_n X_n(t)$ has to be ideally equal to 0 for all t . So it means that, that these C_i 's are not all 0 so it; but C_i 's are not all 0 but it still this linear combination is equal to 0 proves that X_1 to X_n are linearly dependent solution. So it means that what we have proved here that in first part we have shown that $X_1(t)$ to $X_n(t)$ are LD. This implies that $X_1(t_0)$ to $X_n(t_0)$ are LD and in a reverse way we have shown that if $X_1(t)$ to $X_n(t)$ are LD vectors in \mathbb{R}^n then $X_1(t)$ to $X_n(t)$ are LD solutions $\dot{x} = Ax$.

So it means that this $X_1(t)$ to $X_n(t)$ are LD solutions of $\dot{x} = Ax$ if and only if $X_1(t_0)$ to $X_n(t_0)$ are LD vectors in \mathbb{R}^n . And we can say that not that this implies that P if and only if Q. So we can also show that not Q if and only if not P. So it means, so negotiation of this we can say that, that if $X_1(t)$ to $X_n(t)$ are LI solutions of $\dot{x} = Ax$ if and only if $X_1(t_0)$ to $X_n(t_0)$ are LI vectors in \mathbb{R}^n . So with the help of this we have shown that, the solutions X_1 to X_n the n solutions of this.

And we can easily check the linearly independent of these vectors if we can evaluate at any given point t_0 and if the vectors $X_1(t_0)$ to $X_n(t_0)$ are linearly independent then we are done. So

it means that here you need not to check for all t rather than you can check at one single point t_0 . And if it come out to be linearly interdependent vectors we can say that the our solutions are also linearly independent solutions.

So here the problem of checking the linearly independent functions now reduced to checking the linearly independent vectors in R^n because it is quite difficult to check that given set of functions are linearly independent functions.

(Refer Slide Time: 14:15)

$$\begin{array}{c} \checkmark f_1(t) \dots f_n(t) \text{ LI} \\ \hline \left| \begin{array}{c} \\ \\ \end{array} \right| \neq 0 \end{array}$$

So what I am trying to say is that given any say function $f_1(t)$ to say $f_n(t)$ it is quite difficult to check whether these are LI or LD. So it is quite difficult because here we cannot say that we can write, we can have a determinant if it is coming out to be non-zero then it is LI it may not be to at all. Or we can say that there are say LI functions for which determinant is 0 but still we cannot say anything whether it is LD or not.

So here we simply say that the problem of finding the LI, linearly independence of functions is now reduced to the problem of checking linearly independence of vectors in R^n and checking LI linearly independence of vectors is relatively easy then finding the independence of functions. Now here in the particular case when all these f_i 's are solutions of $x' = Ax$ this is quite relax.

And we can say that, if we simply calculate the X_1 to X_n at any given point t_0 , if they are coming out to be linearly independent so we can say that our solutions a function X_1 to X_n 's are linearly independent functions and they are solution of $\dot{x} = Ax$. So now it means that we can easily check that given set of solutions are linearly independent or not.

(Refer Slide Time: 15:45)

Example 1

Consider the following system of first order linear differential equations

$$\frac{dx_1}{dt} = x_2, \quad \frac{dx_2}{dt} = -x_1 - 2x_2 \Rightarrow \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

or $\frac{dx}{dt} = \begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix} x, \quad x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

This system of equations is equivalent to the following single second-order equation

$$\frac{d^2y}{dt^2} + 2\frac{dy}{dt} + y = 0 \quad (5)$$

We may re-write the system of equations (4) from (5) by setting $x_1 = y$ and $x_2 = \frac{dy}{dt}$.

Handwritten notes:
 $\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$
 $\frac{d^2y}{dt^2} + 2\frac{dy}{dt} + y = 0$
 $y(t) = e^{mt} \quad (m^2 + 2m + 1)e^{mt} = 0 \Rightarrow (m+1)^2 = 0$
 $m = -1, -1$

Now, let us take one simple example and proceed. So here consider the following system of first order linear differential equations that is $dx_1/dt = x_2$ and $dx_2/dt = -x_1 - 2x_2$. So we can easily write down this as a system of linear equation, we can write down $\dot{x}_1, \dot{x}_2 = x_1, x_2$ so it is 0 1, let me write it here x_1, x_2 . So $\dot{x}_1 = x_2$ and $\dot{x}_2 = -x_1 - 2x_2$ so -1 and here it is -2. So we can write down $\dot{x}/dt = \begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix} x$ where x is this vector x_1, x_2 .

And here if we simply that if you can look at carefully then this system of linear equation can be reduced to high order differential equation cost of coefficient. So here we can simply say that if $dx_1/dt = x_2$ and $dx_2/dt = -x_1 - 2x_2$. So here we simply differentiate this. And we can write down this as d , so it is what you can simply write say $x_1=y$ you can simply write and $x_2=dy/dt$. If you write x_2 as dy/dt then \dot{x}_2/dt will be what it is simply d^2y/dt^2 square equal to $-x_1 - 2x_2$ is $-y - 2, x_2$ we are writing dy/dt and it is this.

So we can simply write this equation, the system of linear equation is reduced to this single second order equation that is $d^2y/dt^2 + 2dy/dt + y=0$ and we can simply solve this

equation. And this can be solved using the method we have already say assume that this can be solved our solution is given by e to the power -t and t e to the power -t. So it is not very difficult.

Here we just assume that $y = e$ to the power $m t$ is a solution and we can simply say that it is what you can simply write $m^2 + 2m + 1 * e$ to the power $m t = 0$ or e to the power $m t$ is not cannot be 0 so this implies that $m+1$ whole square = 0 so $m=-1$ and -1 repeated we can consider.

(Refer Slide Time: 18:09)

Since $y_1(t) = e^{-t}$ and $y_2(t) = te^{-t}$ are two solutions of (5), we can see that

$$x_1(t) = \begin{pmatrix} e^{-t} \\ -e^{-t} \end{pmatrix} \quad \text{and} \quad x_2(t) = \begin{pmatrix} te^{-t} \\ (1-t)e^{-t} \end{pmatrix} \quad X = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix}$$

are two solutions of (4). Also, observe that $x_1(0)$ and $x_2(0)$ are linearly independent vectors in \mathbb{R}^2 .

Consequently by Theorem 3, $x_1(0)$ and $x_2(0)$ are linearly independent solutions of (4), and the general solution $x(t)$ of (4) can be written as follows:

$$x(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = c_1 \begin{pmatrix} e^{-t} \\ -e^{-t} \end{pmatrix} + c_2 \begin{pmatrix} te^{-t} \\ (1-t)e^{-t} \end{pmatrix}$$

$$= \begin{pmatrix} (c_1 + c_2 t)e^{-t} \\ (c_2 - c_1 - c_2 t)e^{-t} \end{pmatrix}$$

Handwritten notes:
 $X_1 = \begin{pmatrix} e^{-t} \\ -e^{-t} \end{pmatrix}$
 $X_2 = \begin{pmatrix} te^{-t} \\ e^{-t} + te^{-t} \end{pmatrix}$
 $t=0$
 $X_1(0) = \begin{pmatrix} 1 \\ -1 \end{pmatrix}, X_2(0) = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

And we can check that our solutions $y_1(t) = e$ to the power $-t$ is one solution and second solution you can find out using variation of parameter method or the alternative method available to find out the second linearly independent solution. So here in this case we can say that $y_1(t) = e$ to the power $-t$ is one solution and $y_2(t) = te$ to the power $-t$ is the second solution. Now with the help of these two solutions.

Now let us find out two solution of the system of linear equation here. Now here let us take the solution is basically what, here our solution is basically x_1 and x_2 and x_1 we have assume it is y and it is y dash. So we can find out one solution x_1 as you take y as a e to the power $-t$. So if you take y as e to the power t then it is e to the power $-t$ and thereby it is this thing. So t means that x_1 is one solution of this system given as equation number 4.

Now if we take y as this solution t to the power $-t$ then your second solution is given as $t e$ to the power $-t$ and thereby this and this you can find out the e to the power $-t + t e$ to the power $-t$ and -1 will come out, so we can simply say this is nothing but t to the power $-t$, e to the power $-t$ you can take it out and it is $1-t$ here. So it means that first solution is given by e to the power $-t - e$ to the power $-t$, a second solution is given by $t e$ to the power $-t$, e to the power $-t$ into $-1-t$.

And, so we have two solutions and now we want to check that these two solutions are linearly independent or not. So for that you can take any point and you check linearly independence of the corresponding vectors. So here for the simplicity I am taking $t=0$ then your $x_1 0 =$ what it is $1-1$ and $x_2 0$ is what, it is your say 0 and this is what, you write it 1 . So it means that you can check that $1-1$ and 0 and 1 are say linearly independent. That you can check very easily.

(Refer Slide Time: 20:22)

$$c_1 \begin{pmatrix} 1 \\ -1 \end{pmatrix} + c_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} c_1 \\ -c_1 + c_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\checkmark \quad \begin{cases} c_1 + 0c_2 = 0 \\ -c_1 + c_2 = 0 \end{cases} \Rightarrow \begin{cases} c_1 = 0 \\ c_2 = 0 \end{cases}$$

You write $C_1 * 1 - 1 + C_2 * 0 1$ and you simply write it, it is $0 0$ and we will not, that this implies what, this implies at C_1 and this is $C_1 - C_1 + C_2$ and this equal to $0 0$. So if you check that we have $C_1 = 0$ and $C_2 = 0$ the only solution. So this system of linear equation that is $C_1 + 0C_2 = 0$ and $-C_1 + C_2 = 0$. This system of linear equation has only trivial solution that is $C_1 = 0$ and $C_2 = 0$, so this implies that the solution we have obtained $X_1(t)$ and $X_2(t)$ are linearly independent solutions, linearly independent solution of $x \text{ dash} = Ax$.

So once we have linearly independent solution then we can write down the general solution using these x_1 and x_2 . So we can write down the solution $x(t)$ as say that is general solution I am writing as C_1 times $X_1(t)$ + C_2 times $X_2(t)$. So $X_1(t)$ is given by e to the power $-t$ – e to the power $-t$ and $X_2(t)$ is given by te to the power $-t$ and $1-t$ e to the power $-t$.

So general solution you can write it down $C_1(t)$ to $C_2(t)$ * e to the power $-t$ and $C_2-C_1-C_2(t)$ e to the power $-t$. So general solution you can obtain using this. So it means that here in this particular example we have shown that how to find out a solutions of the system of linearly and how to write down, how to check that the obtained solutions are LI solutions and we can write down the general solutions once we have the basis of one basis of your solutions.

So here general solution is given by this. Now what happen, here we have solved by converting the system into this single second order equation and we know how to solve the single second order equation. But in general it may not be say may not be convertibles system of differential equation into a single second order equation then how to find out the solutions n linearly independence solution of $\dot{x} = Ax$, that is what we wanted to know now.

(Refer Slide Time: 22:40)



Linear homogeneous autonomous system

Consider first-order linear homogeneous differential equation

$$\dot{x} = \frac{dx}{dt} = Ax, \quad x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix}. \quad (6)$$

As we know that the space of solutions of (6) forms a vector space of dimension n .

So, now let us find out a basis of the solution space and for that we need to find n linearly independent solutions $x_1(t), x_2(t), \dots, x_n(t)$.



9

So now let us proceed further and we say that consider first order linear homogenous equation, $\dot{x} = dx/dt = Ax$, where x is this n cross 1 vector and A is given by e_{ij} . Here this A is a constant

matrix and X is a of course functions in functions, okay. So we know that the space of solution (6) forms a vector space of dimension n , that we have already discussed.

Now we want to find out a basis, a basis of the solution space and for that we need to find out n linearly independent solutions $x_1(t)$ to $x_n(t)$. We have seen that in particular case when a is a kind of a simpler matrix we can convert this system into a single first order differential equation and we can solve that. But in general it may not be true. So in general in; for general matrix a how to find out n linearly independent solutions that is the matter of this next few slides.

(Refer Slide Time: 23:37)

Consider the scalar differential equation

$$x' = ax, a \in \mathbb{R}.$$

The solution of the differential equation may be written in exponential form as

$$x(t) = e^{at}c,$$

where c is an arbitrary real constant.

In a similar way, as a solution of (6), we may try exponential function $x(t) = e^{\lambda t}v$, where λ is a constant scalar and v is a constant vector.

Handwritten notes in red ink:
 $X' = AX$
 $x' = ax$
 $X = e^{At}v$
 $x = e^{at}c$

So here, let us take the idea of scalar differential equation $x' = ax$, where a is some constant in real value constant. Now we know that the solution of this $x' = ax$, why we are considering this, because this $x' = ax$ and $x' = ax$, this is a vector value and this is a scalar value, the only difference is that here this can be written as one cross one matrix. So here we already know the solution of this scalar value the $x' = ax$ is given by $x(t) = e^{at}c$, where c is some real constant.

The solution of this system, this scalar value differential equation is given $x(t) = e^{at}c$. Now from this can we get some idea to find out the solution of this, one idea is that can we write this similar kind of instruction like let us $e^{at}c$. Another way is can we write down the solution sorry this should be written as in a different manner. Here we simply write

that, can we consider this candidate $x=e$ to the power e at*some kind of v as a candidate of solution.

Or another way is can we consider this that $x=e$ to the power some constant into v as a solution. So let us try this first, because it is a quite simpler one. In a similar way as a solution of (6) we may try exponential function that is $x(t)=e$ to the power $\lambda t v$ as a solution where λ is a some constant and v is a constant vector. So v is a constant vector means m cross one size, so this will be n cross one vector and we try to see whether this will be a solution of $x \text{ dash} = ax$ or not. So if we say that let us first try this and if it will not work then we will try this.

(Refer Slide Time: 25:30)

If $x(t) = e^{\lambda t} v$ is a solution of the system (6), then

$$\frac{d}{dt} e^{\lambda t} v = \lambda e^{\lambda t} v = A e^{\lambda t} v$$

and

$$A(e^{\lambda t} v) = e^{\lambda t} Av.$$

Hence $x(t) = e^{\lambda t} v$ is a solution of (6) if, and only if $\lambda e^{\lambda t} v = e^{\lambda t} Av$. Since $e^{\lambda t} \neq 0$, so divide by $e^{\lambda t}$ to have

$$Av = \lambda v, \quad (7)$$

which shows that (λ, v) is an eigenpair of the matrix A .

So here let us assume that $x = e$ to the power $\lambda t v$ is a solution of the system (6), then if is a system it must satisfy the differential system of linear differential equation, so differentiate it. So if you differentiate it then we will get what, d/dt of e to the power $\lambda t v$ is nothing but λe to the power $\lambda t v$ and this should be equal to your A of this e to the power $\lambda t v$.

So this is; it means that A into e to the power $\lambda t v = e$ to the power λt into $A v$. Now e to the power λt is simply a constant which we can take it out, so it means that if $x(t) = e$ to the power $\lambda t v$ is a solution, if and only if this condition hold, that is λe to the power $\lambda t v = e$ to the power $\lambda t A v$. Now e to the power λt is nearby 0 so we

can simply write that $x(t) = e^{\lambda t} v$ is a solution of (6), (6) means $\dot{x} = Ax$ provided that we have the following relation $Av = \lambda v$.

So it means that, that this $x(t) = e^{\lambda t} v$ is a solution if this λv forms an Eigen pair of the matrix A . It means that the following relation $Av = \lambda v$ holds. So we have a proper definition of Eigen vectors.

(Refer Slide Time: 27:02)

Definition 4 A v = lambda v

A nonzero vector v satisfying (7) is called an eigenvector of A with eigenvalue λ and the pair (λ, v) is called an eigenpair of the matrix A .

Thus, $x(t) = e^{\lambda t} v$ is a solution of (6) if, and only if, (λ, v) is an eigenpair of the matrix A . x' = Ax

Thus, for each eigenvector v_j of A with eigenvalue λ_j , we have a solution

$x_j(t) = e^{\lambda_j t} v_j$

of (6).

To find n linearly independent solutions we need to have n linearly independent eigenvectors v_1, \dots, v_n with respective eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ (may not be distinct), then $x_j = e^{\lambda_j t} v_j, j = 1, \dots, n$ are n linearly independent solutions of (6).

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We simply say that a non-zero vector v satisfying $Av = \lambda v$ is called an Eigenvector of A with Eigenvalue λ . And this pair λ and v is called an eigenpair of the matrix here. So what we have discussed here is a following that this candidate $e^{\lambda t} v$ is a solution of $\dot{x} = Ax$ if and only if the pair λ, v , this λ and this v is an eigenpair of the matrix here. So it means that if λ, v is an eigenpair then we can write $x(t)$ as $e^{\lambda t} v$ as a solution of this.

So it means that for each eigenvector corresponding to eigenvalue λ_j this candidate $x_j(t) = e^{\lambda_j t} v_j$ is a candidate of a solution. So it means that $x_j(t) = e^{\lambda_j t} v_j$ is a solution of (6) that is $\dot{x} = Ax$. Now we need to find out n linearly independent solution so it means that we need to find out n linearly independent eigenvectors v_1 to v_n . And corresponding to some eigenvalue say λ_1 to λ_n . So idea is to find out n linearly independent eigenvectors.

And once we have n linearly independent eigenvectors then we can write down our solution $x_j = e^{\lambda_j t} v_j$ where j is from 1 to n are n linearly independent solutions of (6). So it means that the problem of finding n linearly independent solutions is now reduced to finding n linearly independent eigenvectors v_1 to v_n . So that is what we wanted to now find out.

(Refer Slide Time: 28:47)

*By using LI vectors $j=1, \dots, n$
 $e^{\lambda_j t} v_j$ are LI solutions*

Linear independent-ness of the solutions are guaranteed from Theorem 3 and the fact that $x_j(0) = v_j$, $j = 1, \dots, n$ are n linearly independent vectors. In this case the general solution $x(t)$ of (6) is given by


$$x(t) = c_1 e^{\lambda_1 t} v_1 + c_2 e^{\lambda_2 t} v_2 + \dots + c_n e^{\lambda_n t} v_n \quad (8)$$

where c_i , $i = 1, \dots, n$ are n constants.

Hence any solution of the system (6) can be written as linear combination of x_j , $j = 1, \dots, n$.

So, to find n linearly independent solutions of (6) we need to find only n linearly independent eigenvectors of the coefficient matrix A .

*$X' = A X$
 $A v_j = \lambda_j v_j$ has n LI soln
 n LI eigenvectors of A
 $j=1, \dots, n$*



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13

Now, once we, now we want to check whether these are linearly independent or not, so that can be check very easily because since these are solutions. So $X_j t$ are solutions, now we need to check the values at some points. So let us check the values at 0 so $x_j(0)$ is v_j . So it means that the linearly independence of the solution is depending on the linearly dependence of v_j . So if v_j are n linearly independent vectors then the solutions are also n linearly independent solutions.

So it means that in this case, so it means that if we v_j are linearly independent LI vectors for $j=1, \dots, n$ then we can write that $e^{\lambda_j t} v_j$ are also LI solutions, right so this implies. And once we have LI solutions, any LI solutions then we can write down the general solution as follows $x(t) = C_1 e^{\lambda_1 t} v_1 + C_2 e^{\lambda_2 t} v_2$ and so on. So this is your solution, right where C_i 's are some constants and real constant.

And hence; by this discussion we can say that any solution of this system (6) can be written as linear combination of x_j and for this to find out this x_j n linearly independent solution of (6) we

need to find only n linearly independent eigenvectors of the coefficient matrix A . So it means that if you want to find out n linearly independent solution of this we need to find out n linearly independent eigenvectors of A .

So it means that given a matrix A once we have say, v_1 to v_n , n LI eigenvectors then $x' = Ax$ has n LI solutions and it is given as $e^{\lambda_j t} v_j$ for $j=1$ to n . So our now, we will focus on how to find out n linearly independent eigenvectors. But finding n linearly independent eigenvectors may not be easy task, and we have only one result which says that under a special condition. It may have, we have a guarantee that we have n linearly independent eigenvectors and that is the following result.

(Refer Slide Time: 31:41)

In this regard, we have the following theorem, which says that if we have n distinct eigenvalues of A , then our job is done.

Theorem 5

Let $\lambda_1, \lambda_2, \dots, \lambda_k$ be k distinct eigenvalues of a matrix A , then the corresponding eigenvectors v_1, \dots, v_k of the matrix A are linearly independent.

Handwritten notes:
 $n=1$ (λ, v)
 $n=2$ λ_1, λ_2
 v_1 v_2
 $n=k-1$
 $k=k$

That let λ_1 to λ_k be k distinct eigenvalues of a matrix A , then the corresponding eigenvectors v_1 to v_k are linearly independent. So it means that the crux of this theorem is that corresponding to distinct eigenvalues we have linearly independent eigenvectors. So this can be proved using mathematical induction. So let us; I am giving you the info that $n=1$ it is quite obvious because for $n=1$ we have λv and eigenpair and by definition of eigenvalue this has to be a nonzero vector.

So it has to be linearly independent. Now for $n=2$ let us say that we have λ_1 and λ_2 eigenvalues corresponding to which λ_1 and we have v_1 as an eigenvector

corresponding to lambda 2 we have v2 as an eigenvector. And we want to show that v1 and v2 are LI vectors.

(Refer Slide Time: 32:42)

$$\begin{aligned}
 & \checkmark c_1 u_1 + c_2 u_2 = 0 \text{ (1)} \Rightarrow \underline{c_1 = c_2 = 0} \\
 & A(c_1 u_1 + c_2 u_2) = 0 \\
 & \Rightarrow c_1 \lambda_1 u_1 + c_2 \lambda_2 u_2 = 0 \text{ (2)} \\
 & \lambda_1 \times \text{(1)} \Rightarrow c_1 \lambda_1 u_1 + c_2 \lambda_1 u_2 = 0 \\
 & \Rightarrow \underline{c_2 (\lambda_2 - \lambda_1) u_2 = 0} \quad \lambda_1 \neq \lambda_2 \\
 & \Rightarrow c_2 = 0 \quad c_1 u_1 = 0 \\
 & \Rightarrow c_1 = 0
 \end{aligned}$$

So here we want to check that $c_1 v_1 + c_2 v_2 = 0$; we want to show that this has only a trivial solution that is $C_1=C_2=0$. For that, let us some apply matrix A on this, so A of $c_1 v_1 + c_2 v_2 = 0$ so this implies at $c_1 \lambda_1 v_1 + c_2 \lambda_2 v_2 = 0$ because $A v_1$ is $\lambda_1 v_1 + c_2 A v_2$ is $\lambda_2 v_2 = 0$. So we have relation $c_1 \lambda_1 v_1 + c_2 \lambda_2 v_2 = 0$. Now here we have this equation number 1, this is equation number 2.

So multiply the equation number 1 by say λ_1 so we have this 1 implies this, so λ_1 times this will give you $c_1 \lambda_1 v_1 + c_2 \lambda_1 v_2 = 0$ so this implies that if you subtract these two then we have $c_2 (\lambda_2 - \lambda_1) v_2 = 0$. Now here $\lambda_2 - \lambda_1$ is nonzero because here we have assumed that λ_1 is not to λ_2 . And v_2 is nonzero eigenvector. So this implies that if it is nonzero eigenvector then the only possibility left out is that c_2 has to be 0.

Now if c_2 is 0 we can put it in value here in 1 and we can have $c_1 v_1 = 0$. And v_1 is nonzero vector so this implies at c_1 has to be 0. So it means that this $c_1 v_1 + c_2 v_2 = 0$ has the only solution that is $C_1=C_2=0$. So it means that for $n=2$ if λ_1 is not equal to λ_2 then v_1

and v_2 has to be linearly independent eigenvectors. And now you can prove; you assume for $n=k-1$ and you can prove for $n=k$ using the similar idea.

So this I am leaving it to you. And we simply say that, let us summarize it and we simply say that, let λ_1 to λ_k be k distinct eigenvalues of matrix A then the corresponding eigenvectors are also linearly independent vectors. So it means that if we take this theorem and the previous thing then we say that if you want to find out n linearly independent eigenvectors then it is quite easy that if we have n distinct eigenvalues of A then we can have n linearly independent eigenvectors.

So in this case finding n linearly independence solution is quite easy when we have n distinct eigenvalues of the matrix A .

(Refer Slide Time: 35:22)

Example 2

Find the general solution of the following system of differential equations

$$\dot{x} = \begin{pmatrix} 1 & -1 & 4 \\ 3 & 2 & -1 \\ 2 & 1 & -1 \end{pmatrix} x.$$

*3 LI soln
 x_1, x_2, x_3
 $\checkmark \begin{pmatrix} 1 & -1 & 4 \\ 3 & 2 & -1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} = \lambda \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$
 $\checkmark \begin{pmatrix} 1 & -1 & 4 \\ 3 & 2 & -1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} = \lambda \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$*

Solution The characteristic equation of the matrix A is $\det(A - \lambda I) = 0$, which gives $(1 - \lambda)(\lambda - 3)(\lambda + 2) = 0$.

Thus eigenvalues of A are given as $\lambda_1 = 1, \lambda_2 = 3, \lambda_3 = -2$, all are distinct.

*$x_1 = e^{1t} v_1$
 $x_2 = e^{3t} v_2$
 $x_3 = e^{-2t} v_3$*

$|A - \lambda I| = 0$

Okay, so let us take one example based on this theorem 5. Example is this, find the general solution of the following system of differential equations $\dot{x} = Ax$ where A is given as $\begin{pmatrix} 1 & -1 & 4 \\ 3 & 2 & -1 \\ 2 & 1 & -1 \end{pmatrix}$. So as we have pointed out the system, the general solution can be written as, once we have since it is 3×3 matrix we need to find out 3 LI solutions say X_1, X_2 and X_3 . So if we have three LI solutions then we can write down the general solution.

And also we have described that, that if we can find out say v_1 to v_3 LI eigenvectors then corresponding to λ_1 to λ_3 then the solution e to the power $\lambda_1 t v_1$ can be treated as x_1 , x_2 is e to the power $\lambda_2 t v_2$ and x_3 is written as e to the power $\lambda_3 t v_3$. Then, if v_1 to v_3 are LI eigenvectors then x_1 to x_3 are the LI solutions of this system. So once we have solutions we can write down the general solution.

So idea is to find out LI eigenvectors v_1 to v_3 . So for that let us simplify, let us first find out the eigenvalues of this. So here taking the liberty that finding the eigenvalues is quite easy and you can do that by writing $A - \lambda I$ determinant of this = 0. So as a root of $A - \lambda I = 0$ we can have this characteristic equation that is $1 - \lambda$ times $\lambda^2 - 3\lambda + 2 = 0$. So here I am leaving it to you that so that characteristic equation of this matrix A is given by this.

So once we have this then we can check that $\lambda_1 = 1$, $\lambda_2 = 3$, a $\lambda_3 = -2$ or 3. eigenvalue of this matrix is A . Now here we can see that they are distinct. So it means that now if they are all distinct so we have this assurance that the corresponding eigenvector are also linearly independent eigenvector. So let us find out these linearly independent eigenvectors.

(Refer Slide Time: 37:46)

(i) For $\lambda_1 = 1$, an eigenvector $U = \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix}$ may be found by solving the system of equation

$$(A - I)U = \begin{pmatrix} 0 & -1 & 4 \\ 3 & 1 & -1 \\ 2 & 1 & -2 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

or

$$\begin{aligned} 0 \cdot u_1 - u_2 + 4u_3 &= 0 \\ 3u_1 + u_2 - u_3 &= 0 \\ 2u_1 + u_2 - 2u_3 &= 0. \end{aligned}$$

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So corresponding to $\lambda_1 = 1$, let us eigen; assume that eigenvector is of this form $u_1 u_2 u_3$ and we can find out by solving this system of linear equation $(A - I)U = 0$. So here again I am

putting $\lambda = 1$ and just writing $-I*U$ so we have this system of linear equation and this is to this. And here again I am assuming that we can easily solve this system of linear equation.

(Refer Slide Time: 38:16)

On solving this we get $U = \begin{pmatrix} -1 \\ 4 \\ 1 \end{pmatrix}$ and hence $x_1(t) = e^{t} \begin{pmatrix} -1 \\ 4 \\ 1 \end{pmatrix}$

$$x_1(t) = c_1 e^t \begin{pmatrix} -1 \\ 4 \\ 1 \end{pmatrix}$$

is the corresponding solution for any constant c_1

(ii) For $\lambda_2 = 3$, an eigenvector $V = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix}$ may be found by solving the system of equation

$$(A - 3I)V = \begin{pmatrix} -2 & -1 & 4 \\ 3 & -1 & -1 \\ 2 & 1 & -4 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

And we can have solution as $-1 \ 4 \ 1$ as a solution. So once we have eigenvectors and solutions can be written as $C_1 e^{\lambda_1 t}$ so solution is what $x_1(t) = e^{\lambda_1 t} \begin{pmatrix} -1 \\ 4 \\ 1 \end{pmatrix}$ is 1 here; $e^{\lambda_1 t} v_1$, so v_1 is $-1 \ 4 \ 1$. So the solution $x_1(t)$ as written as $e^{\lambda_1 t} \begin{pmatrix} -1 \\ 4 \\ 1 \end{pmatrix}$, so that is what we have written. And C_1 may be considered as a constant C_1 . So if $x_1(t)$ is a solution then any cost multiple will also be a solution of this.

Now let us look at for $\lambda_2 = 3$. Let us assume that the eigenvector is given by $v_1 \ v_2 \ v_3$ and we can find out $v_1 \ v_2 \ v_3$ by solving the system of linear equation $(A - 3I)v = 0$.

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On solving the above system of linear equations we have $V = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$. And hence the second linearly independent solution can be given as

$$x_2(t) = c_2 e^{3t} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} \quad x_2(t) = e^{3t} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$$

where c_2 is an arbitrary constant.

(iii) Similarly for $\lambda_3 = -2$, an eigenvector $W = \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix}$ can be found by solving the system of equation

$$(A + 2I)W = \begin{pmatrix} 3 & -1 & 4 \\ 3 & 4 & -1 \\ 2 & 1 & 1 \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

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And after simplification we have the solution given as $\begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$ and solution is given by $x_2(t) = e^{3t} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$. So this is a solution and any constant multiple of this will also be a solution. So $x_2(t)$ can be written as $c_2 e^{3t} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$. And similarly for $\lambda_3 = -2$, let us assume eigenvector is taking this W_1, W_2, W_3 kind of format. And this we can find out by solving this system of linear equation $(A+2I)W=0$. Now A is known to you, you can find out $(A+2I)W$ and you can find out this.

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On solving the above system of linear equations we have $W = \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$. And hence the third linearly independent solution can be given as

$$x_3(t) = c_3 e^{-2t} \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix} \quad x_3(t) = e^{-2t} \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$

for any constant c_3 . Therefore, every solution $x(t)$ must be the form

$$x(t) = c_1 \begin{pmatrix} -1 \\ 4 \\ 1 \end{pmatrix} e^t + c_2 e^{3t} \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} + c_3 e^{-2t} \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$

$$x(t) = \begin{pmatrix} -c_1 e^t + c_2 e^{3t} - c_3 e^{-2t} \\ 4c_1 e^t + 2c_2 e^{3t} + c_3 e^{-2t} \\ c_1 e^t + c_2 e^{3t} + c_3 e^{-2t} \end{pmatrix}$$

v_1, v_2, v_3 are LI

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As the solution is given by $\begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$. So t means that the solution is given by $X_3(t) = e^{-2t} \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$. So here solution $x_3(t)$ is given to us. We can write down the general solution $x_3(t) = c_3 e^{-2t} \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$. Now, we already know that v_1, v_2, v_3 are LI.

How we know? We know by the result but you can still check whether you are getting a LI solution or not, so you can check that $\begin{bmatrix} -1 & 4 & 1 \\ 1 & 2 & 1 \end{bmatrix}$ and $\begin{bmatrix} -1 & 1 & 1 \end{bmatrix}$ is basically are LI eigenvectors.

So once they are LI eigenvectors then the general solution can be written as linearly combination of $x_1(t)$ and $x_2(t)$ and $x_3(t)$. So $C_1 \begin{bmatrix} -1 & 4 & 1 \end{bmatrix} e^{3t} + C_2 \begin{bmatrix} 1 & 2 & 1 \end{bmatrix} e^{-2t} + C_3 \begin{bmatrix} -1 & 1 & 1 \end{bmatrix} e^{-1t}$. Then general solution is given by $x(t) =$ this, the following thing will serve as a general solution of this system. So it means that if we have distinct eigenvalue then we have linearly independent eigenvectors and we can find out the n linearly independence solutions. And; but it may not always be true.

So we will discuss more cases of this and how we can find out n linearly independence solution in say coming lectures. So here we stop, we will continue this study in next lecture also. Thank you very much for listening us, thank you.