

**Dynamical Systems and Control**  
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**Lecture – 37**  
**Results on Controllability and Observability**

Hello viewers, welcome to this lecture on the controllability and observability of the continuous control system, so in this lecture we will see some simple results on the controllability and observability of the given system and the corresponding canonical form system. So, let us first recall some simple results from the linear algebra.

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**Theorem 1:**

Let

$$P(t) = \begin{bmatrix} r_1(t) \rightarrow \\ r_2(t) \rightarrow \\ \vdots \\ r_n(t) \rightarrow \end{bmatrix}_{n \times n} \quad \text{and} \quad M = \int_{t_0}^T P(t)P'(t)dt$$

the rows of  $P(t)$  are Linearly Independent iff the matrix  $M$  has rank  $n$  ( $M$  is nonsingular).

So, if  $P$  of  $t$  is a matrix  $n$  cross  $n$  matrix with rows  $r_1, r_2, r_n$  as functions of  $t$  and  $M$  is a constant matrix given by integral  $t_0$  to capital  $T$   $P$  of  $t$  \*  $P$  dashed  $t$ , so this also a  $n$  cross  $n$  matrix and then the rows of the matrix  $P$  of  $t$  are linearly independent if and only if the matrix  $M$  has rank  $n$  or in other words, the matrix  $M$  is non-singular. So, here the each row is a function of  $t$ , so here we say that the rows of the matrix as functions of  $t$  are linearly independent.

And here the constant matrix is non-singular; capital  $M$  is constant matrix, so the relation between the  $P$  of  $t$  and  $M$  is given by this expression.

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**Proof:**

Let  $M$  be singular then  $\exists \alpha \in \mathbb{R}^n$  such that  $\alpha \neq 0$  and  $\alpha' M \alpha = 0$ .

$$\Rightarrow \int_{t_0}^T \alpha' P(t) P'(t) \alpha dt = 0$$

$$\Rightarrow \int_{t_0}^T \|\alpha' P(t)\|^2 dt = 0$$

$$\Rightarrow \alpha' P(t) = 0 \text{ for all } t \in [t_0, T]$$

Handwritten notes:

$$\alpha' = (\alpha_1, \alpha_2, \dots, \alpha_n)$$

$$P(t) = \begin{pmatrix} r_1(t) \\ \vdots \\ r_n(t) \end{pmatrix}$$

$\Rightarrow$  rows of  $P(t)$  are linearly dependent.

as  $\alpha_1 r_1(t) + \alpha_2 r_2(t) + \dots + \alpha_n r_n(t) = 0$  for all  $t \in [t_0, T]$  and some  $\alpha_j \neq 0$ .

Similarly if rows of  $P(t)$  are linearly dependent then  $M$  is singular. Hence the result.

So, the proof is like this, if  $M$  is singular matrix, then we can find a vector  $\alpha$  in  $\mathbb{R}^n$ , so that  $\alpha$  is non-zero and  $\alpha' M \alpha = 0$ , it is a standard result and so that implies by substituting the definition of  $M$  that is integral  $t_0$  to capital  $T$ ,  $P$  of  $t * P$  dashed  $t$  gives the  $M$  definition and then we multiply; premultiply with  $\alpha$  dashed and post multiply with  $\alpha$ , so that  $= 0$ .

But this nothing but norm of  $\alpha$  dashed  $* P$  of  $t$  square  $dt = 0$  and the integral of the positive quantity is 0, it implies that function inside the integral  $\alpha$  dashed  $* P$  of  $t = 0$  for all  $t$ . So, this implies the rows of the matrix  $P$  of  $t$  are linearly independent because  $\alpha$ ; if you write  $\alpha$  dashed to be  $\alpha_1, \alpha_2, \dots, \alpha_n$ , a row vector and if you multiply the matrix  $P$  of  $t$ ;  $P$  of  $t$  is first row is  $r_1$  of  $t$ , second row is  $r_2$  of  $t$  etc.

So, if you multiply  $\alpha$  dashed  $* P$  of  $t$ , we get this expression,  $\alpha_1 * \text{the first row} + \alpha_2 * \text{second row} \text{ etc.}, = 0$  for all  $t$  value and here,  $\alpha$  is a non-0 vector, so some of the  $\alpha_i$ 's will be non-0, so this implies that the rows of  $P$  of  $t$  are linearly dependent here. So if  $M$  is singular, the rows of  $P$  of  $t$  are linearly dependent and similarly, we can go back, if the rows are linearly dependent for  $P$  of  $t$ , then we can; the reverse order if we reach, then we will conclude that  $\alpha$  dashed  $* M * \alpha = 0$ .

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**Theorem 2:**

Let

$$P(t) = \begin{bmatrix} r_1(t) \rightarrow \\ r_2(t) \rightarrow \\ \vdots \\ r_n(t) \rightarrow \end{bmatrix}_{n \times n}$$

and  $L(P(t)) = \bar{P}(s)$  which is  $n \times n$  matrix in  $s$  variable. Rows of  $P(t)$  are linearly independent for  $t > 0$  iff rows of  $\bar{P}(s)$  are linearly independent (as vector functions of  $s$ ).

That implies  $M$  itself is singular, so it proves the theorem that if  $M$  is non-singular, then the rows of  $P$  of  $t$  also should be linearly independent, vice versa, if row of  $P$  of  $t$  are linearly independent, then  $M$  is non-singular. So, this result follows

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**Theorem 3:**

Consider the autonomous system  $\dot{x} = Ax + Bu$  then the following are equivalent

- 1 the system is controllable.
- 2  $\text{rank}[B \ AB \ A^2B \ \dots \ A^{n-1}B] = n$ .
- 3 rows of  $e^{At}B$  are LI.
- 4 rows of  $(sI - A)^{-1}B$  are LI.

Handwritten notes:  $\Rightarrow M = \int_{t_0}^T \begin{pmatrix} A(t_0^i) \\ e^{A(t-t_0)} \end{pmatrix} B^T e^{A(t_0-t)} dt$  is non singular  $p(t) = e^{A(t-t_0)} B$

**Proof:** By taking  $P(t) = e^{At}B$  in the above theorems (1)-(2) the results (3) and (4) follows.

Handwritten equation:  $L(e^{At}B) = (sI - A)^{-1}B$

Now, we use that result for the controllability problem, so if we consider the system  $\dot{x} = Ax + Bu$ , where  $A$  is  $n$  cross  $n$  matrix and  $B$  is  $n$  cross  $m$  matrix, then the conditions; 4 conditions are equivalent; 4 statements, so the system is controllable and the second one is the rank of  $B, AB, A^2B$  etc.,  $A, n - 1B$  with the rank =  $n$ . So this we already proved, the equivalence of 1 and 2 has been already proved.

Then the controllability also for this constant matrix case, if  $A$  and  $B$  are constant matrix, we have also proved that the matrix  $0; t_0$  to  $T$  of  $\int_{t_0}^T \phi(t-t_0) B B^T \phi^T(t-t_0) dt$

\* dt, so this is non-singular. We have shown that this controllability Grammian is non-singular. So, now we can easily prove that the second implies third because 2 implies this matrix is non-singular. Now if you take  $e^{-t} A^{-1} B$  as the  $P$  of  $t$  matrix, we can assume that  $P$  of  $t$  is  $e^{-t} A^{-1} B$ .

Then, we can show that if this matrix  $M$  is non-singular, previously we have proved in the previous theorem that  $P$  of  $t$  is non-singular, the rows of  $P$  of  $t$ 's are linearly independent, then the matrix  $M$  is non-singular and vice versa, so by taking  $P$  of  $t$  to be  $e^{-t} A^{-1} B$ , we can show that the rows of  $e^{-t} A^{-1} B$  are linearly independent because  $e^{-t} A^{-1} B$  is a constant matrix which can be; which is non-singular, so we can prove 2 implies 3. Now 3 implies 4, also can be proved in the following way,

So, now if we take this theorem, let us assume that  $P$  of  $t$  is  $r_1(t), r_2(t), \dots, r_n(t)$  as earlier. Then the Laplace transform of this matrix  $P$  of  $t$ , we denote it by  $\bar{P}(s)$ , where  $s$  is the Laplace transform variable from  $t$  variable is transformed to the  $s$  variable here, and  $\bar{P}(s)$  is the Laplace transform of  $P$  of  $t$ , so if you assume that the rows of  $P$  of  $t$  are linearly independent, then the rows of  $\bar{P}(s)$  are also linearly independent and vice versa, that can be shown because the Laplace transform is a linear operator.

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Proof:

If rows of  $P(t)$  are linearly dependent then  $\exists \alpha \in \mathbb{R}^n, \alpha \neq 0$  such that  $\alpha' P(t) = 0$  for  $t > 0$  then

$$\begin{aligned} L(\alpha' P(t)) &= 0 \\ \alpha' \bar{P}(s) &= 0 \end{aligned}$$

$\Rightarrow$  rows of  $\bar{P}(s)$  are linearly dependent similarly if rows of  $\bar{P}(s)$  are LD then rows of  $P(t)$  are also LD.

So first if we assume  $P$  of  $t$ , has linearly independent rows, then there exists a  $\alpha$  non-0 vector such that  $\alpha' P(t) = 0$  for all  $t$ , then by taking Laplace transform of  $\alpha' P(t)$  that will be  $= 0$ ; that implies that because  $\alpha$  is a constant vector,

alpha dashed \* P bar of s is also = 0, so this implies that the rows of P bar a are also linearly independent.

So, similarly, the inverse Laplace transform is a linear operator, so we can easily proved in the reverse way, if row of P bar s are linearly dependent, we can show that the rows of P of t are also linearly dependent, so we proved. If rows of P of t are linearly independent, then the rows of P bar of s are also linearly independent. So using that result, we can prove that the condition; the statement 3 implies the statement 4.

And statement 4 implies the statement 3, both can be proved by taking P of t = e to the power At \* B. In the previous 2 theorems, we can easily do the statement, one we have proved here. The statement 2 implies the statement 3 and now we can prove the statement 3 implies the statement 4 by taking simply the Laplace transform of e power At \* B, so it will be nothing but SI - A inverse B, the Laplace transform.

So, if the rows of e power At \* B are linearly independent, then the rows of the SI - A inverse B are also linearly independent. So this proves the result, so in order to prove the controllability of the system,  $\dot{x} = Ax + Bu$ , we can prove any one of the statements either 2, 3, or 4, it will automatically imply the controllability.

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**Theorem:**

Let  $P^{-1}AP = J$  be the Jordan canonical form of  $A$  and let  $P^{-1}B = D$ . Then the system  $\dot{x}(t) = Ax(t) + Bu(t)$  is controllable iff the system  $\dot{z}(t) = Jz + Du$  ( $z = P^{-1}x$ ) is controllable.

**Proof:**

$$\text{rank}[B \quad AB \quad A^2B \quad \dots \quad A^{n-1}B] = \text{rank}[PD \quad PJD \quad \dots \quad PJ^{n-1}D]$$

$$= \text{rank}[D \quad JD \quad \dots \quad J^{n-1}D]$$

(since  $P$  is nonsingular).

$\text{rank} \left[ \begin{matrix} P \\ \vdots \\ P \end{matrix} \right] [D \quad JD \quad \dots \quad J^{n-1}D]$

Now, we will prove some result for the Jordan Canonical form, so if  $A$  is the matrix, whose Jordan Canonical form is denoted by  $J$ , then there exist a non-singular matrix  $P$  such that  $P$  inverse  $A$ ,  $P$  is the Jordan form,  $J$ . So, now let us assume that  $P$  inverse  $B$  is the matrix  $D$  and

now, the control system  $\dot{x} = Ax + Bu$ , this system is controllable, if and only if the system  $\dot{z} = Jz + Du$  is controllable, where the transformed variable  $z$  is related with  $x/z = P$  inverse of  $X$ .

So if  $x$  is the state variable; original state variable, then the transformed state variable is  $z$  here which is related by  $z = P$  inverse of  $x$  and the  $A$  matrix is transformed from to the  $J$  matrix and  $B$  is transformed to  $D$ , where  $D$  is  $P$  inverse  $B$  here. So, the controllability of the system  $\dot{x} = Ax + Bu$  is equivalent to the controllability of the canonical system,  $\dot{z} = Jz + Du$ , so that can be easily proved by the controllability of  $\dot{x} = x + Bu$ , condition is the rank of  $B, AB, A^2B, \dots, A^{n-1}B$  is  $n$ .

Now  $A$ ; this matrix  $B$  can be written as  $P * D$  and  $A$  can be written as  $PJ P$  inverse, so by substituting in the place of  $B$ , we get  $PD$ , in the place of  $AB$ , we get  $PJ * D$  etc.  $A^2B$ , we get  $PJ^2 * D$ , so the rank of this is same as the rank of  $D, JD, J^2D, \dots, J^{n-1}D$  because we can take this out from here, this is = this one, is the rank of the matrix  $P * D JD$  etc., because  $P$  is the non-singular matrix, so it can be written like this.

And the rank of this matrix will be same as the rank of  $D, JD, J^2D, \dots, J^{n-1}D$ , so if the  $AB$  system is controllable, then the  $JD$  system is also controllable from this expression. Similarly, we can go in the reverse order, if rank of  $D, JD, J^2D, \dots, J^{n-1}D$  is  $n$ , then by multiplying with  $P$  matrix, we can conclude that the rank of  $B, AB, A^2B, \dots, A^{n-1}B$  etc. that is also  $n$ , so in both the ways we can show.

The controllability of the given system is equivalent to the controllability of the canonical system. So, now we will prove a useful result only in the form of the Jordan canonical form because we have already shown that the controllability of the given system and the Jordan canonical system both are equivalent. So, if you prove the result for the Jordan canonical form, then it will be the; it will be applicable to the given system  $\dot{x} = Ax + Bu$  also.

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Controllability of system in Jordan form:

$$[J]_{n \times n} = \begin{bmatrix} J_1 & & & \\ & J_2 & & \\ & & \ddots & \\ & & & J_k \end{bmatrix}_{n \times n} \quad \lambda_1, \lambda_2, \dots, \lambda_k$$

$$[J]_{n_i \times n_i} = \begin{bmatrix} J_{i1} & & & \\ & J_{i2} & & \\ & & \ddots & \\ & & & J_{ir_i} \end{bmatrix}_{n_i \times n_i}$$

*Handwritten notes:*  
 $\lambda_i$  has geometric multiplicity  $r_i$   
 $\lambda_i$  has algebraic multiplicity  $n_i$

So, instead of the A matrix, let us assume its canonical form J, so J is an n cross n matrix in which we have the Jordan blocks given like this, J1 J2 and Jk. So, here the eigenvalues are lambda 1, lambda 2, lambda k, these are the k distinct eigenvalues and for each eigenvalue, there is a Jordan canonical block for the ith eigenvalue lambda i, we get Ji is the ith canonical; the Jordan block.

So, the Jordan block Ji is given by Ji1, Ji2 etc., Ji ri, so this J lambda i has geometric multiplicity Ri that means, for the eigenvalue lambda i, there are Ri linearly independent eigenvectors available but the algebraic multiplicity of that is given by ni, so in the characteristic polynomial, the root lambda i is repeated n suffix i times but it has r suffix i linearly independent eigenvectors, so it will have ri Jordan blocks like this

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$$[J_{ij}]_{n_j \times n_j} = \begin{bmatrix} \lambda_j & 1 & 0 & \dots & \\ 0 & \lambda_j & 1 & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & \dots & \lambda_j \end{bmatrix}_{n_j \times n_j}$$

$i = 1, 2, 3, \dots, k; j = 1, 2, \dots, r_i$  and  $\sum_{i=1}^k r_i = n, \sum_{j=1}^{r_i} n_{ij} = n_i.$

Now for each  $J$ , this  $J_{ij}$ , capital  $J$ , the Jordan form,  $i$  suffix small  $j$ , it is a  $n_{ij} \times n_{ij}$  matrix, so the Jordan block for that eigenvalue, the  $J$ th Jordan block will look like this,  $\lambda_i \ 1 \ 0 \ 0$   
 $\lambda_i \ 1 \ 0$  etc. The half-diagonal elements are 1 here and so this is repeated,  $n_{ij}$  times, this the size of the block is this one.

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$$B = \begin{bmatrix} B_1 \\ B_2 \\ \vdots \\ B_k \end{bmatrix}, \quad B_i = \begin{bmatrix} B_{i1} \\ B_{i2} \\ \vdots \\ B_{in_i} \end{bmatrix}, \quad B_{ij} = \begin{bmatrix} b_{11}^{ij} \rightarrow \\ b_{22}^{ij} \rightarrow \\ \vdots \\ b_{n_{ij}n_{ij}}^{ij} \rightarrow \end{bmatrix}$$

$\beta_{i,1}$   
 $\beta_{i,2}$

Consider

$$B^i = \begin{bmatrix} b_{n_{i1}n_{i1}}^{ij} \rightarrow \\ b_{n_{i2}n_{i2}}^{ij} \rightarrow \\ \vdots \\ b_{n_{in_i}n_{in_i}}^{ij} \rightarrow \end{bmatrix} \quad \left[ \begin{array}{l} b_{n_{i1}n_{i1}}^{i1} \rightarrow \\ b_{n_{i2}n_{i2}}^{i2} \rightarrow \\ \vdots \\ b_{n_{in_i}n_{in_i}}^{in_i} \rightarrow \end{array} \right]$$

Now the matrix corresponding to  $B$ , it can be split into  $B_1, B_2$  etc., whatever is the number of eigenvalues that is we have taken  $k, J_1, J_2, J_k$ , so correspondingly we have to divide this  $B$  matrix into that much form, so it should be  $B_k$  and for each  $i$ , the  $B_i$  block is having  $B_{i1}$ , is one block,  $B_{i2}$  is the second block and  $B_{iri}$  is the last block of this. So, corresponding to Jordan canonical form, we can split the  $B$  matrix in this form.

And then for each  $B_{ij}$ , we can write the matrix in this form, so this superfix  $ij$  represents it is corresponding to the matrix  $B_{ij}$ . So, the superfix  $ij$  and it has the rows,  $B_1$  is the first row,  $B_2$  is the second row and  $B$  suffix  $n_{ij}$  is the last row of the matrix  $B_{ij}$ . Now, we take the new matrix, it is called  $B$  superfix  $i$ , a matrix which we define, it is like this, for each block here,  $B_{i1}, B_{i2}, B_{iri}$ , the last row looks like this.

For example, for  $B_{i1}$ , the last row is small  $b_{i1}$  small  $b_{ij} / n_{j1}$ , so the matrix  $b_{ij}$  is written as  $b_{ij}$ , 1 is the first row,  $b_{ij2}$  is the second row,  $b_{ij}, n_{ij}$  is the last row of the matrix  $b_{ij}$ . Now, we collect the last row of each such matrix that is  $B$  superfix  $i$  is the matrix defined as follows, so from the matrix  $B_{i1}$ , the last row of  $B_{i1}$  is given by  $B_{i1}$  as superfix and  $n_{i1}$  that is the last row like this.



And when we take the second block  $B_{i2}$  here, its last row is  $B_{i2}$  and the last row is given by  $n_{i2}$ , this one, so in the place of  $j$ , if you put 1, 2, 3 etc., we will get all the rows of  $B$  superfix  $i$ , so it is  $B_{i1}$   $n_{i1}$  is the first row,  $B_{i2}$   $n_{i2}$  is the second row and  $B_{i r_i}$  and  $n$  suffix  $i r_i$  is the last row of the matrix  $B$  superfix  $i$ . So, that means it is the last row of each such block, this last row is collected for each  $J$ , we get this one. So now, the result is as follows

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We know that the system

$$\dot{x} = Jx + Bu \quad (1)$$

is controllable iff rows of  $(sI - J)^{-1} B$  are linearly independent.  
Now

$$(sI - J)^{-1} = \begin{bmatrix} (sI - J_1)^{-1} & 0 & 0 \\ 0 & (sI - J_2)^{-1} & 0 \\ \vdots & \vdots & \vdots \\ 0 & 0 & (sI - J_k)^{-1} \end{bmatrix}, \quad B = \begin{bmatrix} B_1 \\ B_2 \\ \vdots \\ B_k \end{bmatrix}$$

If the row of this  $B$  superfix  $i$ , if they are linearly independent, this is a constant matrix and there are  $r_i$  rows here,  $n_{i1}$ , so here we see that 1, 2, up to  $r_i$  rows, so if this  $r$  suffix  $i$  rows are linearly independent, then we say that the system is controllable, that is the theorem to be proved. So, here we consider the system  $\dot{x} = Jx + Bu$ , where  $B$  is a matrix we can consider, it is a  $n$  cross  $n$  matrix and it is  $n \times m$  matrix.

$B$  is a  $n$  cross  $m$  matrix,  $u$  is  $m$  cross 1, so we know that this system is controllable if  $sI - J$  inverse  $B$  is linearly independent; if the rows of this matrix are linearly independent, that was the theorem proved earlier. So, we see under what condition this rows of  $sI - J$  Inverse  $B$  are linearly independent, so  $sI - J$  inverse can be written like this, because it has Jordan block  $J_1, J_2, J_k$ , so other blocks are zero blocks here we can write, they are all 0, depending on the size, we have to write the blocks here, okay.

And  $B$  has blocks  $B_1, B_2$ , as we have seen earlier  $B_k$ , now, we have to see if the rows of  $sI - J$  inverse  $B$  are linearly independent, then the system is controllable, so we will see under what condition the rows of  $sI - J$  inverse  $B$  are linearly independent. So if you multiply  $sI - J$  Inverse \* the  $B$  matrix, we get this expression as given here.

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$$\therefore (sI - J)^{-1} B = \begin{bmatrix} (sI - J_1)^{-1} B_1 \\ (sI - J_2)^{-1} B_2 \\ \vdots \\ (sI - J_k)^{-1} B_k \end{bmatrix}$$

$$(sI - J_i)^{-1} B_i = \begin{bmatrix} (sI - J_{i1})^{-1} B_{i1} \\ (sI - J_{i2})^{-1} B_{i2} \\ \vdots \\ (sI - J_{iri})^{-1} B_{iri} \end{bmatrix}$$

$sI - J$ , here  $I$  is the identity matrix,  $n \times n$  identity matrix,  $s$  is a variable; real variable and so we can write it like this, by multiplying these 2 matrices. Now each block  $sI - J_i$  Inverse \*  $B_i$ , so if you take this one, it can also be written in the similar form, only thing is for each  $I$ , we take the corresponding blocks,  $J_{i1}$ ,  $J_{i2}$ ,  $J_{in}$  suffix  $i$  are taken these blocks and similarly,  $B_{i1}$ ,  $B_{i2}$ ,  $B_{iri}$ , so here  $sI - J_i$  suffix  $i$   $ri$  is there not  $ni$ , there is a correction here.

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$$\text{Now } (sI - J_{ij})_{n_j \times n_j} = \begin{bmatrix} s - \lambda_i & 1 & 0 & 0 & \dots & \dots \\ 0 & s - \lambda_i & 1 & 0 & \dots & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \vdots & s - \lambda_i & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & s - \lambda_i \end{bmatrix}$$

$$(sI - J_{ij})_{n_j \times n_j}^{-1} = \begin{bmatrix} \frac{1}{(s - \lambda_i)} & \frac{1}{(s - \lambda_i)^2} & \dots & \dots & \frac{1}{(s - \lambda_i)^{n_j}} \\ 0 & \frac{1}{(s - \lambda_i)} & \dots & \dots & \dots \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \frac{1}{(s - \lambda_i)} \\ 0 & 0 & 0 & 0 & 0 & \frac{1}{(s - \lambda_i)} \end{bmatrix}$$

Now we take one such block,  $sI - J_{ij}$  which is  $n_j$  cross  $n_j$  size matrix, which can be written in this particular way because the eigenvalues are  $\lambda_i$  for this and they are repeated  $n_j$  times and so,  $sI - J_{ij}$  is given by this matrix, now it is inverse, if you directly calculate, we can find the inverse of the matrix,  $sI - J_{ij}$  inverse is given by this expression.

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$$(sI - J_i)^{-1}_{n_i \times n_i} B_{ij} = \begin{bmatrix} b_1^j \rightarrow \\ b_2^j \rightarrow \\ \vdots \\ b_{n_i}^j \rightarrow \end{bmatrix}$$

Now  $(sI - J_i)^{-1} B_i$  (order  $n_i \times m$ ) has rows.

Now if we multiple with the matrix  $B_{ij}$  with the inverse, we will get the following,  $sI - J_i$  inverse  $B_i$ , it has  $n_i$  cross  $m$  rows and that can be written like this.

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$$(sI - J_i)^{-1} B_i = \begin{bmatrix} \frac{1}{(s-\lambda_i)} b_1^{i1} + \frac{1}{(s-\lambda_i)^2} b_2^{i1} + \dots + \frac{1}{(s-\lambda_i)^{n_i}} b_{n_i}^{i1} \\ \frac{1}{(s-\lambda_i)} b_1^{i2} + \dots + \frac{1}{(s-\lambda_i)^{n_i-1}} b_{n_i}^{i1} \\ \dots \\ \frac{1}{(s-\lambda_i)^{n_i}} b_{n_i}^{i1} \\ \dots \\ \frac{1}{(s-\lambda_i)} b_1^{i2} + \frac{1}{(s-\lambda_i)^2} b_2^{i2} + \dots + \frac{1}{(s-\lambda_i)^{n_i/2}} b_{n_i/2}^{i2} \\ \dots \\ \frac{1}{(s-\lambda_i)^{n_i/2}} b_{n_i/2}^{i2} \\ \dots \\ \frac{1}{(s-\lambda_i)} b_1^{i r_i} + \frac{1}{(s-\lambda_i)^2} b_2^{i r_i} + \dots + \frac{1}{(s-\lambda_i)^{n_i}} b_{n_i}^{i r_i} \\ \dots \\ \frac{1}{(s-\lambda_i)} b_{n_i}^{i r_i} \end{bmatrix} \begin{matrix} n_i \times n_i & n_i \times m \\ & n_i \times m \end{matrix}$$

The product directly will give this expression,  $1/s - \lambda_i * B_{i1}$  suffix 1 +  $1/s - \lambda_i I$  square  $B_{i1}$  suffix 2 etc. So, note that these are all row vectors, okay,  $B_{i1}$ ,  $1B_{i1}$ ,  $i12$ ,  $B_{i1}$ ,  $n_{i1}$ , so these are all row vectors and the  $1/s - \lambda_i$  etc., these are functions of  $s$  here. So, by adding all these terms, we will get a vector function in  $s$  variable. The second vector is also the linear combination of these vectors and it is a function of  $s$ .

So, the last one is the vector  $B_{i1}$  and  $n_{i1}$  vector multiplied by  $1/s - \lambda_i$ , so this expression, we get from this, from the previous slide, we multiply the last row with the matrix, we get this expression, I think here the correction is this one, there is no power here,

the powers will be decreasing. The first term is  $1/s - \lambda_i$ , the last term here is  $1 / s - \lambda_i$  power  $n_i$ .

But so, here the first block will have the last term  $1/s - \lambda_i$  only multiplied by this row vector. Similarly, the second block if you take, the last row of the second block is given by  $1/s - \lambda_i$  multiplied by  $B_{i2}$  suffix  $n_i$  and the last block is  $1/s - \lambda_i$  multiplied by the row  $B_{i-r_i}$  suffix  $n_i$  block. So, this is the matrix which is of this size because this is of the size  $n_i$  cross  $n_i$  and this is of the size  $n_i$  cross  $m$ , so this is of the size  $n_i$  cross  $m$  matrix.

Now, we can see that this last rows of each block is made up of the rows of  $B$  superfix  $i$  matrix. So this matrix we have seen, this  $B$  superfix  $i$  matrix have this particular rows as given, so we have this matrix, the last rows like this.

**(Refer Slide Time: 28:03)**

$B_1 = \begin{bmatrix} B_{11} \\ B_{12} \end{bmatrix}$ ,  $B_2 = \begin{bmatrix} B_{21} \\ B_{22} \end{bmatrix}$ ,  $\lambda_1 = 2$ ,  $\lambda_2 = 3$ .  
 $B = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix}$ ,  $B^1 = \begin{bmatrix} \text{---} \\ \text{---} \end{bmatrix}_{2 \times 2}$ ,  $B^2 = \begin{bmatrix} \text{---} \\ \text{---} \end{bmatrix}_{2 \times 2}$ ,  $B_{7 \times 2}$ .

From this it is clear that if rows  $b_{n_1}^1, b_{n_2}^2, \dots, b_{n_i}^i$  are linearly independent, then all these rows of  $(sI - J_i)^{-1} B_i$  are LI.

Finally if the rows of  $(sI - J_i)^{-1} B_i$  are LI for each  $i$  then the rows of  $(sI - J)^{-1} B$  are LI (as function of  $s$ ).

$\dot{x} = Jx + Bu$  is controllable if the rows of  $B$  are LI.

$J = \begin{bmatrix} J_1 & & \\ & J_2 & \\ & & J_n \end{bmatrix}_{7 \times 7}$ ,  $J_1 = \begin{bmatrix} J_{11} & \\ & J_{12} \end{bmatrix}_{4 \times 4}$ ,  $J_2 = \begin{bmatrix} J_{21} & \\ & J_{22} \end{bmatrix}_{2 \times 2}$ ,  $J_n = \begin{bmatrix} 2 & 1 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{bmatrix}$ ,  $J_{n1} = \begin{bmatrix} 1 & 1 \\ 0 & 3 \end{bmatrix}$ ,  $J_{n2} = \begin{bmatrix} 3 \end{bmatrix}$ .

So, if we observe that if these rows  $b_{i1}$  suffix  $n_{i1}$ ,  $b_{i2}$  suffix  $n_{i2}$  and  $b_{i-r_i}$  suffix  $n_{i-r_i}$ , if they are linearly independent, then all the rows of  $(sI - J)$  suffix  $i$  inverse \*  $B_i$  are linearly independent, so this matrix is what is given in the previous, so this is the matrix, so if this rows are linearly independent, then automatically, all the rows of this matrix are linearly independent because we can see that none of the rows can be written as product of any of these rows.

But if any of this rows, the last row of the each block, they become 0, then obviously if any one of the row is 0 in a matrix, that matrix has linearly dependent row, so first thing is they should not be; should be non-zero, these rows,  $b_{i1}$  suffix  $n_{i1}$ ,  $b_{i2}$  suffix,  $n_{i2}$ , so these rows

should be first of all non-zero, then if they are also linearly independent, then the entire matrix  $n \times m$  matrix, they are all linearly independent.

And if any of them become 0, then we can see that the entire matrix has 0, the linearly dependent rows and the condition, similarly, if for each  $i$ ,  $(sI - J_i)^{-1} b_i$  are linearly independent, then the total; the entire matrix  $(sI - J)^{-1} b$  are also linearly independent, so that can be seen from here,  $(sI - J)^{-1} B$  is given by this one, so if the blocks, for each  $i$ , the rows of each block are linearly independent.

And since the  $J_1, J_2$  they are all made up of different eigenvalues, the first block is made up of eigenvalue  $\lambda_1$  and second block we will have  $(sI - \lambda_2)^{-1} b_2$ ; sorry,  $(sI - J_2)^{-1} b_2$  inverse, so this will contain the element like  $1/(s - \lambda_2)$  or its powers + etc. So, here we can see that each block contains different types of terms, like  $1/(s - \lambda_1)$  will be coming in the first block,  $1/(s - \lambda_2)$  will come in the second block and its powers etc.

So, none of them can be linearly dependent on any of these blocks because of the nature of the different eigenvalues, so if separately if they have linearly independent rows, then the total rows; every row is linearly independent. The set of all rows of this matrix are linearly independent, so that is what is given here. For each  $i$ , if  $(sI - J_i)^{-1} b_i$  are linearly independent, then  $(sI - J)^{-1} B$  also will have linearly independent rows.

And this implies the controllability of the system, so the condition for controllability is given by this matrix, ultimately this  $(sI - J)^{-1} B$  matrix is an important one, if the rows of this matrix are linearly independent, then these given system is controllable the system  $\dot{x} = Jx + Bu$  is controllable and if the Jordan system is linearly independent, then the original system will also be linearly independent by the previous theorem.

So here, we have seen the relation between the Jordan canonical form of a system and the system itself, how they are related by the controllability property, so now we have shown that the system is controllable, the system  $\dot{x} = Jx + Bu$  is controllable, if the rows of the  $(sI - J)^{-1} B$  are linearly independent, so this what we have shown. So now, we will see an example in the following way.

So, let us consider the Jordan canonical form like this,  $J_1$  and  $J_2$ , okay, so where  $J_1$  is having  $J_{11}$  and  $J_{12}$  and  $J_2$  is  $J_{21}$ ,  $J_{22}$ , so in the previous notation as we have seen, so where  $J_{11}$  matrix is  $\begin{bmatrix} 2 & 1 & 0 & 0 \\ 0 & 2 & 1 & 0 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 2 \end{bmatrix}$  and  $J_{12}$  is simply the 1 cross 1 matrix 2, and  $J_{21}$  is  $\begin{bmatrix} 3 & 1 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}$  okay  $J_{22}$  is matrix 3, so this total size of the matrix  $J$  is 4 6 7, so it is a 7 cross 7 matrix and  $J_1$ ; it is 4 cross 4 matrix and  $J_2$  is 3 cross 3 matrix and the corresponding matrices are given here, so we can divide now the matrix  $B$  into 4 parts.

It will contain, first it is  $B_1$ ,  $B_2$ , corresponding to  $J_1$  and  $J_2$ , we have  $B_1$ ,  $B_2$  and  $B_1$  has 2 blocks,  $B_{11}$ ,  $B_{12}$  and  $B_2$  have 2 blocks;  $B_{21}$ ,  $B_{22}$ . Now, the matrix corresponding to the eigenvalues, so you can easily see that the eigenvalues are  $\lambda_1 = 2$  and  $\lambda_2 = 3$  and its geometric multiplicity is 2 for each one of them. Algebraic multiplicity of  $\lambda_1 = 2$ ; it is repeated 4 times.

Algebraic multiplicity is 4, but geometric multiplicity is 2 here and for  $\lambda_2 = 3$ , algebraic multiplicity is 3 and geometric multiplicity is 2, the linearly independent eigenvectors will be there. Now, so, this will have 4 blocks, the matrix  $B$  finally has 4 blocks that is  $B_{11}$ ,  $B_{12}$ ,  $B_{21}$ ,  $B_{22}$ . So according to the theorem, this theorem statement we have seen there, so this matrix  $B$  superfix  $i$ , for each eigenvalue  $\lambda_i$ , the last row of these blocks corresponding to that  $B_i$ , whatever is the last row of each block, they are linearly independent.

Then the system is controllable, so if we have this following example, this example, for  $B_1$ , we have 2 blocks, so the last row of each block should be linearly independent, the collection of the last row. So for example, if we take  $B_1$ , there are 2 blocks, so it will have first row and second row. The first row is the last row of  $B_{11}$  and the second row is last row of  $B_{12}$ , similarly,  $B$  superfix 2, it means the last row of  $B_{21}$  will be taken.

And the last row of  $B_{22}$  is taken, so it is a matrix with 2 rows, so for example if we take 2 rows with 2 columns, we can say that the 2 rows can be linearly independent, so the theorem is, for each  $i$ ,  $B$  superfix  $i$  has linearly independent rows, so minimum row we required here is; number of rows we required is 2 because there are 2 blocks, so the matrix  $B$  should be of the size, it should at least have 2 columns.

So, it will be 7 cross 2 at least, so for this example, if we take a matrix  $B$  as a column matrix simply, that is 7 cross 1 matrix, then the system will not be controllable, that is clear because

of this Jordan canonical forms and the theorem that  $B$  superfix  $i$  should be having linearly independent rows, so we can see that if you select 2 rows, linearly independent for  $B$  superfix 1 and  $B$  superfix 2.

We can assure the controllability of the system given by this example with this Jordan canonical form. So this, in this lecture we have seen various results on the controllability of the system and its related Jordan canonical form, so similar result can be extended for the observability of the system also. Thank you.